



ALLEGATI
alla
Relazione scientifica finale
Linee di ricerca A1-A2

Consiglio Nazionale delle Ricerche
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Allegato 1.1 – Tabella soglie di controllo dati meteo

MASTER	posizione file	dato aberrante	allarme
year anno	1	--	--
doym giorno dell'anno	2	--	--
Tempmedia Temperatura media dell'aria (°C)	3	da -20 a +50	da 0 a +40
Tmax Temp massima dell'aria (°C)	4	da -10 a +60	da 0 a +50
Tmin Temp minima dell'aria (°C)	5	da -20 a +50	da -20 a +30
Tdevstd deviazione standard temperatura media	6	--	--
Tultimo ultimo rilievo temperatura (°C)	7	da -20 a +50	da -20 a +50
Umidmedia umidità relativa media (%)	8	da 0 a 110	da 10 a 100
URmax umidità massima (%)	9	da 0 a 110	da 20 a 100
URmin umidità minima (%)	10	da 0 a 110	da 10 a 80
URdevstd deviazione standard umidità	11	--	--
URultimo ultimo rilievo umidità (%)	12	da 0 a 110	da 10 a 100
Presmedia pressione (hPa)	13	da 700 a 1400	da 850 a 1200
Pmax pressione massima (hPa)	14	da 700 a 1400	da 850 a 1200
Pmin pressione minima (hPa)	15	da 700 a 1400	da 850 a 1200
Pdevstd deviazione standard umidità	16	--	--
Pultimo ultimo rilievo pressione (hPa)	17	da 700 a 1400	da 850 a 1200
RadGmedia radiazione globale media (W m-2)	18	da 0 a 2000	da 0 a 2000
Rmax radiazione massima (W m-2)	19	da 0 a 2000	da 0 a 2000
Rmin radiazione minima (W m-2)	20	da 0 a 2000	da 0 a 2000
Rdevstd deviazione standard radiazione	21	--	--
Rultimo ultimo rilievo radiazione (W m-2)	22	da 0 a 2000	da 0 a 2000
Piogsomma precipitazioni (mm)	23	da 0 a 2000	da 0 a 2000
VVent1Vmed velocità media del vento (m s-1)	24	da 0 a 10000	da 0 a 10000
Vdevstd dev standard velocità vento	25	--	--
Vraf vento di raffica (m s-1)	26	da 0 a 10000	da 0 a 10000
Vmin velocità minima vento (m s-1)	27	da 0 a 10000	da 0 a 10000

Vultimo	28	da 0 a 10000	da 0 a 10000
ultimo rilievo vel. vento (m s-1)			
DVent1Dprev	29	da 0 a 360	da 0 a 360
direzione prevalente del vento (°)			
Draf	30	da 0 a 360	da 0 a 360
direzione del vento di raffica (°)			
Dultimo	31	--	--
deviazione standard direzione vento			
numcamp	32	--	--
numero di acquisizioni (#)			
Batteria	33	da 0 a 15000	da 0 a 15000
livello della batteria (mV)			

SLAVE	dati ogni ora	dato aberrante	allarme
year	1	--	--
anno			
DOY_hh	2	--	--
giorno dell'anno			
Rad	3	da +25 a +1500	da +25 a +800
radiazione globale (W m-2)			
Tair	4	da -20 a +50	da -20 a +40
temperatura dell'aria media (°C)			
Tgrap	5	da -20 a +50	da -20 a +40
temperatura del grappolo media (°C)			
Tleaf	6	da -20 a +50	da -20 a +40
temperatura (infrarosso) della chioma (°C)			
Tsoil1	7	da -20 a +50	da -20 a +40
temperatura del suolo 30cm (°C)			
Tsoil2	8	da -20 a +50	da -20 a +40
temperatura del suolo 60cm (°C)			
potenziale idrico 30	9	da +0.01 a +2.75	da +0.01 a +2.75
riscaldatore sensore di temperatura del suolo 30 cm (Delta T °C)			
potenziale idrico 60	10	da +0.01 a +2.75	da +0.01 a +2.75
riscaldatore sensore di temperatura del suolo 60 cm (Delta T °C)			
VV	11	da 0 a 5000	da 0 a 5000
velocità del vento (m s-1)			
Bat	12	da 0 a 15000	da 0 a 15000
livello della batteria (mV)			
Bagn_leaf	13	da 0 a 60	da 0 a 60
bagnatura foglie (#)			

Allegato 1.2 – Protocollo di installazione sensori delle stazioni Slave

Le stazioni periferiche (Slave) una volta posizionate all'interno del vigneto seguendo il disegno sperimentale, sono state dotate di un codice identificativo relativo al blocco di vigoria e alla parcella d'appartenenza. Nell'allegato X è mostrata la posizione di tutte le Slave all'interno di ciascun vigneto. I sensori collegati a ciascuna Slave sono stati installati secondo la metodologia seguente:

- **TEMPERATURA GRAPPOLO:** viene individuato ad inizio stagione un grappolo campione e su di esso viene posizionato il sensore di temperatura sul rachide a circa metà della lunghezza dello stesso fissato con un piccolo spezzone di filo da potatura in modo da non creare una strozzatura sul rachide. La scelta di questo posizionamento è stata valutata in base alla necessità di ottenere una misura della temperatura che fosse rappresentativa delle condizioni termiche medie all'interno del grappolo.
- **BAGNATURA FOGLIARE:** il sensore verrà posizionato vicino al grappolo campione, all'interno della cortina, alla stessa altezza del grappolo, fissato al palo di legno adiacente al grappolo stesso, parallelo alla linea del cordone, in modo da evidenziare le condizioni di umidità e bagnatura rappresentative del grappolo e del microclima che lo circonda.
- **RADIAZIONE SOLARE:** il sensore di radiazione solare è da posizionare vicino al grappolo campione sopra all'attacco del "rachide", simulando in maniera ottimale la superficie esposta.
- **POTENZIALE IDRICO DEL SUOLO:** i sensori verranno installati a coppie di 2, a 30 e 60cm di profondità in quanto tale posizionamento permette di capire oltre al contenuto anche le dinamiche dell'acqua nel suolo, in prossimità della pianta campione e sulle stazioni Slave considerate rappresentative delle aree, nella vigna Cacciagrande nelle Slave 1, 2, 4, 8, 9, nella vigna Cortigliano 1, 5, 8, 9, 10, nella vigna Donna Olimpia 1, 4, 6, 9, 10, nella vigna Brolio 1, 2, 3, 4, 11 (i numeri abbinati alle Slave sono mostrati nell'allegato X).
- **TEMPERATURA FOGLIARE AD INFRAROSSO:** i sensori verranno posizionati sul lato esposto a sud a circa 20 cm dalla cortina con angolo di vista ellittico orientato sulla spalliera verso il basso (evitando esposizione alla pioggia), così da raccogliere

informazioni su un'area più ampia e rappresentativa. Considerando l'angolo di vista (FOV) di 45° del sensore sarà posizionato evitando di puntare il terreno o oggetti di disturbo (sostegni, paleria ecc.). Nei casi in cui l'orientamento dei filari sia nord/sud il sensore sarà posizionato ad ovest. Essendo i box delle stazioni posizionati in modo asimmetrico rispetto ad i 2 pali il sensore IR verrà posizionato dal lato vicino al palo con un sistema di blocco di sicurezza sul palo stesso, per aumentarne la stabilità e mantenerne inalterata la posizione. La misura della temperatura fogliare puntuale non è rappresentativa della cortina in quanto, la singola foglia è soggetta ad alta variabilità di temperatura per diversi motivi (ombreggiamento, stato fisiologico della foglia, carenze nutrizionali e attacchi parassitari etc..). Pertanto è stato scelto un sensore che integrasse la misura della temperatura delle foglie presenti nell'area del campo di monitoraggio del sensore.

- TEMPERATURA ARIA: il sensore di temperatura aria verrà posizionato sul secondo filo a circa la metà della cortina ed in posizione il più possibile indisturbata per essere rappresentativa della spalliera.
- VELOCITA' VENTO: il sensore di velocità del vento verrà installato su un supporto che lo colloca al centro della zona interfilare a circa 50cm dal suolo. La posizione di questo sensore ci permette di valutare i differenti campi di vento sulle diverse tecniche di gestione della chioma (tesi defogliante e non).

Allegato 1.3 – Tabelle di calibrazione sensori

		2008							
BROLIO	Tair		Tgrape		Tleaf		vigorìa	tesi	
	span	offset	span	offset	span	offset			
A	0.9768	-0.3841	0.9947	-0.4849	X	X	6	5	
B	0.9687	-0.4947	1.0013	-0.5348	X	X	3	5	
C	uguale H	uguale H	uguale H	uguale H	X	X	4	5	
D	0.9805	0.3615	1	0.3224	X	X	1	5	
E	0.9963	0.9952	0.9986	1.2254	X	X	1	3	
F	0.9936	-0.1988	0.9515	0.1085	X	X	1	7	
G	0.9879	0.7426	0.9832	1.2596	X	X	1	4	
H	0.9896	0.3281	0.998	-0.515	X	X	1	6	
I	1	-0.0471	0.9995	-0.2466	X	X	1	2	
L	1.0245	-1.0027	1.0293	-1.0655	X	X	1	8	
M	0.9871	0.5001	0.9659	0.5802	X	X	1	1	
OLIMPIA	Tair		Tgrape		Tleaf		vigorìa	tesi	
span	offset	span	offset	span	offset				
A	0.9635	0.2372	1.0487	-0.7762	X	X	1	2	
B	1.1082	-1.8737	1.1019	-1.9487	X	X	1	6	
C	1.0333	0.0884	0.9895	0.8975	X	X	1	1	
D	1.0157	0.9189	1.0048	-0.2804	X	X	5	5	
E	1.0389	0.017	0.9617	-0.2618	X	X	1	8	
F	1.0085	-0.5595	1.0022	-0.5959	X	X	1	3	
G	0.9971	0.0606	0.9724	0.6029	X	X	1	5	
H	0.9553	-0.3388	0.9918	-0.4012	X	X	1	4	
I	0.9532	0.7286	1.0005	0.2987	X	X	1	7	
L	0.9822	-0.7899	1.0031	0.9814	X	X	4	5	
MORTELLE	Tair		Tgrape		Tleaf		vigorìa	tesi	
span	offset	span	offset	span	offset				
A	0.976	-0.4343	0.9842	0.2374	1.0493	-1.0401	1	6	
B	NO	NO	NO	NO	NO	NO	1	7	
C	1.0248	-0.9627	0.9926	0.1723	1.1018	-2.2474	1	1	
D	0.9842	0.8467	0.9853	0.9958	1.0705	-1.2487	1	3	
E	1.0053	-0.1367	0.9859	0.2215	NO	NO	1	5	
F	0.9795	0.5064	0.9912	0.1312	1.0877	-0.9331	1	4	
G	1.0171	-0.2492	0.9774	0.9222	1.113	-1.5984	1	8	
H	0.9686	0.4459	1.0022	-0.6727	ERROR	ERROR	1	2	
I	NO	NO	NO	NO	NO	NO	2	5	
L	0.9805	0.8548	1.0192	-0.1189	ERROR	ERROR	4	5	
M	NO	NO	NO	NO	NO	NO	6	5	
N	NO	NO	NO	NO	NO	NO	3	5	
O	0.9922	-0.2944	0.9882	0.1157	0.9509	0.2133	1	7	
P	0.9844	0.1739	0.9745	0.5809	0.9724	0.3896	1	3	
Q	0.9936	0.9748	0.9238	2.1741	1.0826	-1.0441	1	4	
R	1.0149	-0.2097	0.9713	0.499	1.0143	-0.1543	1	2	
S	0.9786	0.2287	0.9783	0.1835	ERROR	ERROR	1	5	
T	0.998	0.3717	0.9825	0.8863	0.9769	1.2074	1	8	
U	0.9737	0.4536	0.9768	0.2278	0.9545	1.1554	1	6	
V	0.9881	-0.0971	0.9955	-0.4314	0.9725	-0.0696	1	1	

2010	Tair				Tgrape			vigorìa	tesi
	SPAN	OFFSET	R	SPAN	OFFSET	R			
A	1.0188	-0.8384	0.9994	1.0079	-0.9532	0.9999	6	5	
B	0.9942	-0.9244	0.9999	1.0204	-1.2992	0.9997	3	5	
C	1.0091	0.1344	0.9996	1.003	0.0801	0.9999	4	5	
D	1.0004	0.2262	0.9998	0.9799	0.1633	0.9991	1	5	
E	1.0225	0.6878	0.9989	1.0057	0.748	0.9997	1	3	
F	0.9317	0.4141	0.9945	1.0054	-0.4873	0.9994	1	7	
G	1.011	0.6059	0.9996	0.9953	0.7672	0.9999	1	4	
H	1.0208	0.0152	0.9994	0.9991	-0.8051	0.9999	1	6	
I	0.9662	0.5053	0.9988	0.9697	0.2543	0.9995	1	2	
L	1.0142	-0.7781	0.9998	1.0202	-1.3212	0.9998	1	8	
M	1.0082	0.0379	0.9999	1.0095	-0.5879	0.9997	1	1	
DO	SPAN	OFFSET	R	SPAN	OFFSET	R	vigorìa	tesi	
A	0.99	0.1699	0.9985	0.9987	0.1038	0.9990	1	2	
B	0.9585	0.1919	0.9961	0.9685	-0.1121	0.9973	1	6	
C	0.9829	1.1405	0.9998	1.0146	0.8983	0.9996	1	1	
D	0.9743	1.677	0.9945	0.9892	0.4551	0.9955	5	5	
E	1.0262	0.3749	0.9965	1.1276	-98.3750	0.9765	1	8	
F	1.0091	-0.9814	0.9994	1.1609	-2.9836	0.9304	1	3	
G	1.0105	-0.3057	0.9973	1.0123	-0.1987	0.9993	1	5	
H	1.0116	-1.1616	0.9986	0.9790	-0.5475	0.9997	1	4	
I	1.0067	-0.1615	0.9982	1.0087	-0.0768	0.9997	1	7	
L	1.0122	-0.6744	0.9991	0.9654	1.4403	0.9969	4	5	
CO	SPAN	OFFSET	R2	SPAN	OFFSET	R2	vigorìa	tesi	
A	1.0044	-0.5906	0.9995	1.0169	-0.7412	0.9963	1	6	
B	1.0043	0.4623	0.9997	0.9938	0.4298	0.9992	1	7	
C	1.002	-0.4945	0.9997	0.9976	-0.3232	0.9993	1	1	
D	1.1139	-10.123	0.9565	0.9957	-0.3357	0.9994	1	3	
E	0.9987	-0.0766	0.9996	1.0861	-13.036	0.9704	1	5	
F	1.0048	-0.147	0.999	1.0075	-0.93	0.9987	1	4	
G	0.9812	-25.632	0.9686	1.0095	-0.0173	0.9991	1	8	
H	0.9993	-0.6601	0.9994	0.9941	-1.0343	0.9994	1	2	
I	0.952	4.665	0.9949	1.0485	-10.462	0.9834	2	5	
L	1.0006	0.1833	0.9995	0.9953	-0.1521	0.9996	4	5	
CA	SPAN	OFFSET	R2	SPAN	OFFSET	R2	vigorìa	tesi	
M	0.9942	0.4249	0.9996	1.0018	0.3671	0.9996	6	5	
N	1.0048	0.1023	0.9995	1.0104	0.0218	0.9994	3	5	
O	0.9954	-0.3013	0.9997	1.0101	-0.3623	0.9994	1	7	
P	1.0569	-10	0.9949	0.9996	-0.125	0.9994	1	3	
Q	0.9949	1.0266	0.9994	0.9989	0.894	0.9993	1	4	
R	0.9974	0.3551	0.9997	1.0021	-0.2703	0.9993	1	2	
S	0.9873	-0.1147	0.999	0.9878	-0.3561	0.9992	1	5	
T	1.0001	0.2877	0.9995	1.0104	-2.1315	0.9967	1	8	
U	1.0007	-0.0536	0.9994	0.992	-0.4066	0.9993	1	6	
V	1.0029	-0.3098	0.9995	0.8269	-6.8881	0.9853	1	1	

		2009							
BROLIO	Tair			Tgrape			Tleaf		
	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²
A	0.9986	-0.4787	0.996	1.0105	-0.8012	0.9985	1.0048	-0.4084	0.9984
B	0.9998	-0.7786	0.998	1.0132	-0.8762	0.9983	0.9911	-0.9066	0.9967
C	0.9873	0.5256	0.998	1.0178	0.2164	0.9988	1.0001	0.1508	0.9979
D	1.0031	0.4673	0.997	1.0109	0.0853	0.9961	0.9926	0.3013	0.9992
E	1.0014	1.0731	0.9921	1.0083	1.0467	0.9946	0.9925	1.0134	0.994
F	1.0062	-0.1523	0.9969	1.0184	-0.4123	0.9959	1.0035	-0.0218	0.9982
G	0.9983	0.9217	0.9995	0.9929	-4.754	0.984	X	X	X
H	0.9871	-0.632	0.9996	0.9664	-0.0009	0.9973	1.0863	-7.6104	0.9405
I	0.9932	0.2644	0.9995	0.9633	0.6149	0.9974	1.192	-17.116	0.9445
L	1.0162	-1.2181	0.9995	0.9542	-12.097	0.9986	1.0324	-89.01	0.9632
M	0.984	0.1937	0.9984	0.9707	0.2857	0.998	1.0709	-8.7856	0.9464
OLIMPIA	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²
A	0.9788	0.2369	0.9921	0.999	0.3749	0.9798	0.9919	-0.153	0.9947
B	0.9795	0.0239	0.9923	0.998	-0.1538	0.983	0.9839	-0.2263	0.997
C	0.9864	0.8506	0.9884	0.993	1.2598	0.9891	0.9904	1.2385	0.9972
D	0.9944	0.8263	0.9873	1.003	-0.2013	0.9847	0.9424	-8.3202	0.9975
E	1.0424	-0.2877	0.9883	1.0056	-86.608	0.9791	0.9688	-12.345	0.9985
F	1.0026	-0.792	0.9896	1.0163	-0.5571	0.9918	0.9909	-0.4957	0.9682
G	0.9639	0.0876	0.966	0.9145	0.7891	0.9398	1.1171	-10.776	0.9726
H	0.9725	-0.732	0.9987	0.9513	-0.3176	0.9965	1.0556	-6.6562	0.9349
I	0.9605	0.1879	0.9556	0.9471	0.2983	0.9536	1.1161	-10.505	0.9707
L	0.9794	0.7034	0.9983	0.9141	-11.397	0.9968	1.0356	-89.537	0.9684
MORTELLE	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²	SPAN	OFFSET	R ²
A	0.9548	-0.0942	0.9353	0.9938	-0.0294	0.9949	0.9901	0.0339	0.9961
B	1.0117	0.6779	0.9939	0.9847	1.1542	0.9947	0.9974	0.3704	0.998
C	0.9956	-0.0444	0.9938	0.9859	0.2331	0.9937	0.9971	0.0087	0.9966
D	1.0159	0.1027	0.9554	0.9984	0.034	0.9959	0.9974	0.1513	0.9972
E	0.99	0.0005	0.9936	0.9935	0.0323	0.9967	0.992	0.0754	0.9976
F	0.8773	1.5153	0.7814	0.7297	2.8069	0.7859	0.9953	-88.51	0.9319
G	0.9179	-4.7264	0.9722	0.9906	0.4958	0.9941	0.9917	0.5024	0.9946
H	0.9964	-0.1811	0.995	0.9929	-0.3474	0.9975	1.0197	-0.1468	0.9741
I	0.9323	5.1852	0.9833	0.9971	0.0039	0.9932	0.9854	-0.4012	0.9798
L	1.0101	0.213	0.9877	0.9771	0.4906	0.9732	0.9977	0.0628	0.9922
M	0.9923	0.7539	0.9955	0.9899	0.8492	0.997	0.9925	0.6657	0.9984
N	1.0085	0.4724	0.9961	1.0018	0.5539	0.9976	0.9902	0.5189	0.998
O	0.9997	0.0667	0.9941	0.9996	0.2551	0.9973	0.9928	0.0485	0.9983
P	0.9598	-7.0232	0.9366	0.9854	0.5774	0.9971	0.99	0.5852	0.9981
Q	1	1.2136	0.9973	0.9865	1.3162	0.9977	1.0008	-2.9023	0.9571
R	0.9934	0.5465	0.9952	0.9954	-5.7291	0.9866	0.9873	0.3903	0.9982
S	0.9921	0.3158	0.9836	0.9796	0.5696	0.9845	0.9846	0.4683	0.9849
T	0.9906	0.6995	0.9824	0.9758	0.9061	0.9872	1.0151	-2.6994	0.9618
U	0.9771	0.6183	0.9919	1.0209	0.2218	0.9823	0.996	0.1953	0.9907
V	0.9925	0.2132	0.9882	0.9916	0.1567	0.9778	0.98		

Allegati 1.4 – 1.5 – 1.6 – Listati delle
procedure matlab di despiking dati slave
(2008–2009–2010)

```

clear all;
close all;
clc;

disp('          CI SONO I COEFFICIENTI CALIBRAZIONE le soglie e il
despiking e anche media mobile          ')

st=input('ID station (LOC=Master;S=Slave): ','s');
stt=input('Vigna (BR=Brolio;MA=Mortelle alta Cacciagrande;MB=Mortelle bassa
Cortigliano;DO=Donna Olimpia): ','s');
styear=input('Anno: ');
safety=input('Vuoi salvare il file output ? Si(1) No(2); ');
File_save='C:\TUSCANIA\DATI_2008\Scheduler\OUTPUT';
File_save_desp='C:\TUSCANIA\DATI_2008\Scheduler\DESPIKING';
File_path_master='C:\TUSCANIA\DATI_2008\Scheduler\Master';

if stt == 'BR'
    File_path='C:\TUSCANIA\DATI_2008\Scheduler\Brolio\OUTPUT';
    File_path_loc='C:\TUSCANIA\DATI_2008\Scheduler\Brolio\OUTPUT';
    %tolto coef Tgrape di G ed L
    slavenumb=[65 66 67 68 69 70 71 72 73 76 77];
    span_air=[0.9768      0.9687  0.9896  0.9805  0.9963  0.9936
0.9879  0.9896  1      1.0245  0.9871];
    offset_air=[-0.3841  -0.4947  0.3281  0.3615  0.9952  -0.1988
0.7426  0.3281  -0.0471  -1.0027  0.5001];
    span_grape=[0.9947      1.0013  0.998  1      0.9986  0.9515  1
0.998  0.9995  1      0.9659];
    offset_grape=[-0.4849  -0.5348  -0.515  0.3224  1.2254  0.1085  0
-0.515  -0.2466  0      0.5802];
    %
    span_grape=[0.9947      1.0013  0.998  1      0.9986  0.9515
0.9832  0.998  0.9995  1.0293  0.9659];
    %
    offset_grape=[-0.4849  -0.5348  -0.515  0.3224  1.2254  0.1085
1.2596  -0.515  -0.2466  -1.0655  0.5802];
    span_leaf=[1      1      1      1      1      1      1      1
1      1      1];
    offset_leaf=[0  0      0      0      0      0      0      0
0      0];
    offset_rad=[-13.1      5.2      8.1      -5.6      -13.4      9.3
-2.0  12.6      -6.6      -1.3      -5.6];
    doy_cal=9400;
    doy_rad=15900;
    doy_cal2=13900;
    doy_delta=10600;

    %slavenumb=char('A','B','C','D','E','F','G','H','I','L','M');
elseif stt == 'DO'
    File_path='C:\TUSCANIA\DATI_2008\Scheduler\Donna_Olimpia\OUTPUT';
    File_path_loc='C:\TUSCANIA\DATI_2008\Scheduler\Donna_Olimpia\OUTPUT';
    slavenumb=[65 66 67 68 69 70 71 72 73 76];
    span_air=[0.9635      1.1082  1.0333  1.0157  1.0389  1.0085
0.9971  0.9553  0.9532  0.9822];
    offset_air=[0.2372      -1.8737  0.0884  0.9189  0.017  -0.5595
0.0606  -0.3388  0.7286  -0.7899];
    span_grape=[1.0497      1.1019  0.9895  1.0048  0.9617  1.0022
0.9724  0.9918  1.0005  1.0031];
    offset_grape=[-0.7762  -1.9487  0.8975  -0.2804  -0.2618  -0.5959
0.6029  -0.4012  0.2987  0.9814];
    span_leaf=[1      1      1      1      1      1      1      1
1      1];
    offset_leaf=[0  0      0      0      0      0      0      0
0      0];
    offset_rad=[19.1      14.7      14.7      14.7      10.2      -4.3
-17.9  -5.2      -2.8      -13.7];
    doy_cal=10200;
    doy_rad=16500;
    doy_cal2=14700;
    doy_delta=11400;

    %slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MB'
    File_path='C:\TUSCANIA\DATI_2008\Scheduler\Mortelle\OUTPUT';
    File_path_loc='C:\TUSCANIA\DATI_2008\Scheduler\Mortelle_Bassa\OUTPUT';
    slavenumb=[65 66 67 68 69 70 71 72 73 76];
    span_air=[0.976  1      1.0248  0.9842  1.0053  0.9795  1.0171

```

```

0.9686 1 0.9805];
offset_air=[-0.4343 0 -0.9627 0.8467 -0.1367 0.5064
-0.2492 0.4459 0 0.8548];
span_grape=[0.9842 1 0.9926 0.9853 0.9859 0.9912
0.9774 1.0022 1 1.0192];
offset_grape=[0.2374 0 0.1723 0.3958 0.2215 0.1312
0.9222 -0.6727 0 -0.1189];
span_leaf=[1.0493 1 1.1018 1.0705 1 1.0877
1.113 1 1 1];
offset_leaf=[-1.0401 0 -2.2474 -1.2487 0 -0.9331
-1.5984 0 0 0];
offset_rad=[-20.0 7.3 -18.5 9.6 11.5 -9.7 0.0
19.7 2.0 3.0];
doy_cal=10200;
doy_rad=15000;
doy_cal2=14700;
doy_delta=11400;
%slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MA'
File_path='C:\TUSCANIA\DATI_2008\Scheduler\Mortelle\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2008\Scheduler\Mortelle_Alta\OUTPUT';
slavenumb=[77 78 79 80 81 82 83 84 85 86];
span_air=[1 1 0.9922 0.9844 0.9936 1.0149 0.9786
0.998 0.9737 0.9881];
offset_air=[0 0 -0.2944 0.1739 0.9748 -0.2097 0.2287
0.3717 0.4536 -0.0971];
span_grape=[1 1 0.9882 0.9745 0.9238 0.9713 0.9783
0.9825 0.9768 0.9955];
offset_grape=[0 0 0.1157 0.5809 2.1741 0.499 0.1835
0.8863 0.2278 -0.4314];
span_leaf=[1 1 0.9509 0.9724 1.0826 1.0143 1
0.9769 0.9545 0.9725];
offset_leaf=[0 0 0.2133 0.3896 -1.0441 -0.1543 0
1.2074 1.1554 -0.0696];
offset_rad=[-13.7 12.9 7.7 16.6 -16.3 9.6 0.0
-10.4 9.6 -17.4];
doy_cal=10200;
doy_rad=15000;
doy_cal2=14700;
doy_delta=11400;
%slavenumb=char('M','N','O','P','Q','R','S','T','U','V');
end
if stt=='LOC'
File_wrk=[File_path_loc '\' st '.dat'];
N=dload(File_wrk);
% MASTER
% year doym Tempmedia Tmax Tmin Tdevstd Tultimo Umidmedia
URmax URmin URdevstd URultimo Presmedia
% 1 2 3 4 5 6 7 8 9 10
11 12 13
% Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
Rultimo Piogsomma VVentlVmed Vdevstd Vraf
% 14 15 16 17 18 19 20 21 22 23
24 25 26
% Vmin Vultimo DVentlDprev Draf Dultimo numcamp Batteria
% 27 28 29 30 31 32 33
%step 1 ordina ed elimina doppie
[Y,I]=sort(N.doyM(:,1));
datatemp=[N.year(I,1) N.doyM(I,1) N.Tempmedia(I,1) N.Tmax(I,1)
N.Tmin(I,1) N.Tdevstd(I,1) N.Tultimo(I,1) N.Umidmedia(I,1)
N.URmax(I,1) N.URmin(I,1) N.URdevstd(I,1) N.URultimo(I,1)
N.Presmedia(I,1) N.Pmax(I,1) N.Pmin(I,1) N.Pdevstd(I,1)
N.Pultimo(I,1) N.RadGmedia(I,1) N.Rmax(I,1) N.Rmin(I,1)
N.Rdevstd(I,1) N.Rultimo(I,1) N.Piogsomma(I,1) N.VVentlVmed(I,1)
N.Vdevstd(I,1) N.Vraf(I,1) N.Vmin(I,1) N.Vultimo(I,1)
N.DVentlDprev(I,1) N.Draf(I,1) N.Dultimo(I,1) N.numcamp(I,1)
N.Batteria(I,1)];
[I,J]=unique(datatemp(:,2));
data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
datatemp(J,13) datatemp(J,14) datatemp(J,15) datatemp(J,16)

```

```

datatemp(J,17) datatemp(J,18) datatemp(J,19) datatemp(J,20)
datatemp(J,21) datatemp(J,22) datatemp(J,23) datatemp(J,24)
datatemp(J,25) datatemp(J,26) datatemp(J,27) datatemp(J,28)
datatemp(J,29) datatemp(J,30) datatemp(J,31) datatemp(J,32)
datatemp(J,33)];
F = data(:, :);
F(F==-999) = NaN;
data(:, :)=F;

% elaborazioni giornaliere
appog=data(1,2);
app2=num2str(data(1,2));
data(1,2)=1111111;
prova=num2str(data(:,2));
D=str2num(prova(:,1:3));
D(1)=appog;
doyi=D(2);
doyf=D(end);
%elaborazioni orarie
prova2=num2str(data(:,2));%
D2=str2num(prova2(:,4:7));%
%D2(1)=str2num(app2(:,4:7));
D2=D2/100;%
D3=fix((D2+0.1)*24/100) ; %
D3(1)=D3(2);
ii=1;
i2=1;%
data(1,2)=appog;

for i=doyi:doyf
    ijj = find(D==i);
    if size(ijj,1)>1
        daily(ii,:) = nanmean(data(ijj,:)); % per le medie giornaliere
        daily(ii,2)=i;
        daily(ii,23)= nansum(data(ijj,23)); % cumulate giornaliere
        precipitazione mm
        daily(ii,4)= nanmax(data(ijj,4));
        daily(ii,5)= nanmin(data(ijj,5));
        daily(ii,9)= nanmax(data(ijj,9));
        daily(ii,10)= nanmin(data(ijj,10));
        daily(ii,19)= nanmax(data(ijj,19));
        daily(ii,20)= nanmin(data(ijj,20));

        if styear>2007
            daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
            sfilato km giorno rilev 5 min
            daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
            radiazione Mj /m2
        else
            if (stt=='DO' & i>=222 | stt=='BR' & i>=194 | stt=='MB' &
                i>=261 | stt=='MA' & i>=261)
                daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
                sfilato km giorno rilev 5 min
                daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
                radiazione Mj /m2
            else
                daily(ii,24)= nansum(data(ijj,24)*900/1000); % vento
                sfilato km giorno rilev 15 min
                daily(ii,18)= nansum(data(ijj,18)*900/1000000); % cumulate
                radiazione Mj /m2
            end
        end
        ii=ii+1;
    else
        daily(ii,:) = NaN;
        daily(ii,2)=i;
        ii=ii+1;
    end
end

%elaborazione oraria
    if size(ijj,1)>1

```

```

zz1=ijj(1);
for iii=D3(ijj(1)):D3(ijj(end))
    ij2=find(D3(ijj)==iii);
    if size(ij2,1)>1
        zz=zz1+length(ij2);
        azz=data(zz1:(zz-1),:);
        zz1=zz;
        hourly(i2,:)=nanmean(azz(:,,:));
        hourly(i2,2)=i+double(iii*100/24)/100;
        hhour(i2,1)=i;
        hhour(i2,2)=iii;
        hourly(i2,23)= nansum(azz(:,23)); %
        cumulate giornaliera precipitazione mm
        hourly(i2,4)= nanmax(azz(:,4));
        hourly(i2,5)= nanmin(azz(:,5));
        hourly(i2,9)= nanmax(azz(:,9));
        hourly(i2,10)= nanmin(azz(:,10));
        hourly(i2,19)= nanmax(azz(:,19));
        hourly(i2,20)= nanmin(azz(:,20));

        if iii==24
            hourly(i2,2)=i;
        end
        i2=i2+1;
    else
        hourly(i2,:) = NaN;
        hourly(i2,2)=i+double(iii*100/24)/100;
        i2=i2+1;
    end
end
end
else
    hourly(i2,:) = NaN;
    hourly(i2,2)=i+double(iii*100/24)/100;
    i2=i2+1;
end
end
%fine elab oraria

end

    %inserisci giorni mancanti
hourly(:,2)=hourly(:,2)*100;
startday=fix(hourly(1,2)/100);
endday=fix(hourly(end,2)/100);
ndays=endday-startday+1;
ncol=33;
A=NaN*ones(ndays*24,ncol);
A(:,1)=styear;
for i=0:ndays-1
    for j=0:23
        A(i*24+j+1,2)=
            str2num([num2str(startday+i,'%3d')
                num2str(j,'%2d')]);
    end
end

    newdoy1=hourly(:,2)/100-fix(hourly(:,2)
/100);
newdoy2=int16(newdoy1*24);
appdoy=hourly(:,2);
hourly(:,2)=str2num([num2str(fix(hourly(:,2)
/100),'%3d') num2str(newdoy2,'%2d')]);;
B=hourly(:,2);

    for i=1:size(A,1)
        index=find(B(:,2)==A(i,2));
        if length(index)>=1
            A(i,:)=B(index(1),:);
        end
    end

hourly=A;
% salvataggio file .mat
eval(['M_' stt '_RT = data(:,,:);']);
eval(['M_' stt '_DY = daily(:,,:);']);
eval(['M_' stt '_HY = hourly(:,,:);']);

```



```

%           subplot(4,2,4),plot(hourly(:,2),hourly(:,18));
%           hold on
%           ylabel('Radiation (W/m2)Medie orarie non SUM')
%           subplot(4,2,5),plot(hourly(:,2),hourly(:,23));
%           hold on
%           ylabel('Precipitation (mm)')
% %           VVent1Vmed Vdevstd Vraf Vmin Vultimo DVent1Dprev Draf Dultimo
numcamp Batteria
%           subplot(4,2,6),plot(hourly(:,2),hourly(:,24));
%           hold on
%           ylabel('Wind (Km giorno)')
%           subplot(4,2,7),plot(hourly(:,2),hourly(:,32));
%           hold on
%           ylabel('numpack (#)')
%           subplot(4,2,8),plot(hourly(:,2),hourly(:,33));
%           hold on
%           ylabel('Battery (mV)')
%
%           figure('Name','Master Acquisition-time 5min');
% %           doym Tempmedia Tmax Tmin Tdevstd Tultimo Umidmedia URmax URmin
URdevstd URultimo
%           subplot(4,2,1),plot(data(:,2),data(:,3));
%           hold on
%           ylabel('Temperature')
%           subplot(4,2,1),plot(data(:,2),data(:,4),'r');
%           hold on
%           ylabel('Tmax')
%           subplot(4,2,1),plot(data(:,2),data(:,5),'r');
%           hold on
%           ylabel('Tmin')
%           %axis([600 900 15 28])
%           subplot(4,2,2),plot(data(:,2),data(:,8));
%           hold on
%           ylabel('Relative Humidity')
%           %axis([0 i -20 20])
%           subplot(4,2,2),plot(data(:,2),data(:,9),'r');%URmax
%           hold on
%
subplot(4,2,2),plot(data(:,2),data(:,10),'r');%URmin
%           hold on
% %           Presmedia Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
Rultimo Piogsomma
%           subplot(4,2,3),plot(data(:,2),data(:,13));
%           hold on
%           ylabel('Pressure (hPA)')
%           subplot(4,2,4),plot(data(:,2),data(:,18));
%           hold on
%           ylabel('Radiation (W/m2)')
%           subplot(4,2,5),plot(data(:,2),data(:,23));
%           hold on
%           ylabel('Precipitation (mm)')
% %           VVent1Vmed Vdevstd Vraf Vmin Vultimo DVent1Dprev Draf Dultimo
numcamp Batteria
%           subplot(4,2,6),plot(data(:,2),data(:,24));
%           hold on
%           ylabel('Wind (Km giorno)')
%           subplot(4,2,7),plot(data(:,2),data(:,32));
%           hold on
%           ylabel('numpack (#)')
%           subplot(4,2,8),plot(data(:,2),data(:,33));
%           hold on
%           ylabel('Battery (mV)')
%
%           figure('Name','Master Daily')
%           subplot(4,2,1),plot(daily(:,2),daily(:,3));
%           hold on
%           ylabel('Temperature')
%           subplot(4,2,1),plot(daily(:,2),daily(:,4),'r');
%           hold on
%           ylabel('Tmax')
%           subplot(4,2,1),plot(daily(:,2),daily(:,5),'r');
%           hold on
%

```

```

%                               %ylabel('Tmin')
%                               %axis([600 900 15 28])
%                               subplot(4,2,2),plot(daily(:,2),daily(:,8));
%                               hold on
%                               ylabel('Relative Humidity')
%                               %axis([0 i -20 20])
%
% subplot(4,2,2),plot(daily(:,2),daily(:,9),'r');%URmax
%                               hold on
%
% subplot(4,2,2),plot(daily(:,2),daily(:,10),'r');%URmin
%                               hold on
% %                               Presmedia Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
Rultimo Piosomma
%                               subplot(4,2,3),plot(daily(:,2),daily(:,13));
%                               hold on
%                               ylabel('Pressure (hPA)')
%                               subplot(4,2,4),plot(daily(:,2),daily(:,18));
%                               hold on
%                               ylabel('Radiation (MJ/m2)Sum Daily')
%                               subplot(4,2,5),plot(daily(:,2),daily(:,23));
%                               hold on
%                               ylabel('Precipitation (mm)')
% %                               VVent1Vmed Vdevstd Vraf Vmin Vultimo DVent1Dprev Draf Dultimo
numcamp Batteria
%                               subplot(4,2,6),plot(daily(:,2),daily(:,24));
%                               hold on
%                               ylabel('Wind (m/s)')
%                               subplot(4,2,7),plot(daily(:,2),daily(:,32));
%                               hold on
%                               ylabel('numpack (#)')
%                               subplot(4,2,8),plot(daily(:,2),daily(:,33));
%                               hold on
%                               ylabel('Battery (mV)')

else
s1=1;%inizializzo indici vettori calibrazione
f1=1;
for hh=slavenumb(1):slavenumb(end)
if (hh==74 | hh==75)
pippo='N';
else
File_wrk=[File_path '\' hh '.dat'];
M=dload(File_wrk);
File_master=[File_path_master '\M_' stt '.mat'];
load(File_master);
eval(['Master hy=M ' stt ' HY;'])
eval(['Master_dy=M_' stt '_DY;'])

% SLAVE
% year DOY_hh rad Tair Tgrap Tleaf Tsoil1 Tsoil2 T_heatsoil1
T_heatsoil2 VV Bat Bagn_leaf
% 1 2 3 4 5 6 7 8 9
10 11 12 13
%step 1 ordina ed elimina doppie
[Y,I]=sort(M.DOY_hh(:,1));
datatemp=[M.Year(I,1) M.DOY_hh(I,1) M.Tcjc(I,1) M.Tair(I,1)
M.Tgrap(I,1) M.Tleaf(I,1) M.Tsoil1(I,1) M.Tsoil2(I,1)
M.T_heatsoil1(I,1) M.T_heatsoil2(I,1) M.VV(I,1) M.Bat(I,1)
M.Bagn_leaf(I,1)];
[I,J]=unique(datatemp(:,2));
data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
datatemp(J,13)];
F = data(:,:);
F(F==-999) = NaN;
data(:,:)=F;

wind=data(:,11);
wind(wind==0.2)=0;
data(:,11)=wind;

```

```

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%%
%applico i coefficienti di calibrazione per le slave
data(data(:,2)<=doy_cal,4)=(data(data(:,2)<=doy_cal,4)
*span_air(s1))+offset_air(f1);
data(data(:,2)<=doy_cal,5)=(data(data(:,2)<=doy_cal,5)
*span_grape(s1))+offset_grape(f1);
data(data(:,2)<=doy_cal,6)=(data(data(:,2)<=doy_cal,6)
*span_leaf(s1))+offset_leaf(f1);

%dalla data di inserimento span-offset errore nel calcolo
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,4)=
((data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,4)-
offset_air(f1))*span_air(s1))+offset_air(f1);
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,5)=
((data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,5)-
offset_grape(f1))*span_grape(s1))+offset_grape(f1);
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,6)=
((data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,6)-
offset_leaf(f1))*span_leaf(s1))+offset_leaf(f1);
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,4)=
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,4)-
offset_air(f1);
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,5)=
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,5)-
offset_grape(f1);
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,6)=
data(data(:,2)>=doy_cal & data(:,2)<=doy_cal2,6)-
offset_leaf(f1);

%dopo aver ripristinato gli span-offset 1 e 0 il seguente
data(data(:,2)>=doy_cal2,4)=(data(data(:,2)>=doy_cal2,4)
*span_air(s1))+offset_air(f1);
data(data(:,2)>=doy_cal2,5)=(data(data(:,2)>=doy_cal2,5)
*span_grape(s1))+offset_grape(f1);
data(data(:,2)>=doy_cal2,6)=(data(data(:,2)>=doy_cal2,6)
*span_leaf(s1))+offset_leaf(f1);

%sensori radiazione grappolo
data(data(:,2)<doy_rad,3)=NaN;
%coefficienti calibrazione radiazione grappolo
data(data(:,2)>=doy_rad,3)=data(data(:,2)>=doy_rad,3)
+offset_rad(f1);
%elimino i negativi
apprad=data(:,3);
apprad(apprad<=25)=0;
data(:,3)=apprad;
s1=s1+1;
f1=f1+1;
%conversione T_heatsoil da temperatura a delta T
data(data(:,2)<=doy_delta,9)=(data(data(:,2)<=
doy_delta,9))-(data(data(:,2)<=doy_delta,7));
data(data(:,2)<=doy_delta,10)=(data(data(:,2)<=
doy_delta,10))-(data(data(:,2)<=doy_delta,8));

%inserimento soglie deltaT per il potenziale idrico
%considerare solo valori di delta tra 0.01<x<2.75
pot30=data(:,9);
pot30(pot30>=2.75)=2.75;
pot30(pot30<=0.01)=0.01;

pot60=data(:,10);
pot60(pot60>=2.75)=2.75;
pot60(pot60<=0.01)=0.01;

%RANGE DI POTENZIALE IDRICO
%
%
%
%
Tasterisco30=(3.15-pot30)/(3.15-0.75);
pot30=-0.056*(Tasterisco30.^(-1/0.45)-1);
Tasterisco60=(3.15-pot60)/(3.15-0.75);
pot60=-0.056*(Tasterisco60.^(-1/0.45)-1);
%da Mpa a pF (scala 1-7) %-LOG10(1/ASS(E12*10000)) %-
log10(1/abs(-2.95*10000))

```

```

%           pf_i_30=abs(pot30*10000);
%           pf_i_60=abs(pot60*10000);
%           pf30=-log10(1./pf_i_30);
%           pf60=-log10(1./pf_i_60);
%           rh30=-22.392*log(pf30-1)+42.554; curva ritenzione media
campioni donna olimpia
%           rh60=-22.392*log(pf60-1)+42.554;
%           data(:,9)=rh30;
%           data(:,10)=rh60;
%           %%%%%%%%%%
%           data(:,9)=pot30;
%           data(:,10)=pot60;

%inserimento soglie temperatura aria e grappolo
s_air=data(:,4);
s_air(s_air>=50 | s_air<=-20)=NaN;
data(:,4)=s_air;
s_grape=data(:,5);
s_grape(s_grape>=50 | s_grape<=-20)=NaN;
data(:,5)=s_grape;
s_leaf=data(:,6);
s_leaf(s_leaf>=60 | s_leaf<=-20)=NaN;
data(:,6)=s_leaf;

%despiking manuale ESEGUITO GUARDANDO I DATI GIORNALIERI

%filtro a 12 ore potenziale idrico
aaa=1;
bb12 = [1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12];
yy12_30 = filter(bb12,aaa,data(:,9));
yy12_60 = filter(bb12,aaa,data(:,10));
data(:,9)=yy12_30;
data(:,10)=yy12_60;

if stt == 'BR'
%slavenu= [65 66 67 68 69 70 71 72 73 76 77];
data(data(:,2)<9199,4)=NaN; %Tair dal 150 in poi
data(data(:,2)<15999,5)=NaN; %Tgrape dal 178 in poi
data(data(:,2)>28100,3:12)=NaN; %dati oltre vendemmia
data(data(:,2)<17099,6)=NaN; %Tleaf dal 170 in poi
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
if hh==slavenu(5) | hh==slavenu(6) | hh==slavenu(7) |
hh==slavenu(8) | hh==slavenu(9) | hh==slavenu(10)
%tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenu(5) | hh==slavenu(7) | hh==slavenu(9) |
hh==slavenu(11)%tolgo wind
data(data(:,2)>100,11)=NaN;
end
if hh==slavenu(7) % tgrape G tolgo le ore 8-9-10
orenul=num2str(data(:,2));%
nuoveore=str2num(orenul(:,4:5));%
nnore=find(nuoveore<43 & nuoveore>32);
data(nnore,5) = NaN;
end
if hh==slavenu(4)%tgrape D tolgo le ore dalle 12 alle 16
troppo alte
orenul=num2str(data(:,2));%
nuoveore=str2num(orenul(:,4:5));%
nnore=find(nuoveore<72 & nuoveore>49);
data(nnore,5) = NaN;
end
if hh==slavenu(10)%tgrape L tolgo le ore centrali troppo
alte
orenul=num2str(data(:,2));%
nuoveore=str2num(orenul(:,4:5));%
nnore=find(nuoveore<79 & nuoveore>59);
data(nnore,5) = NaN;
end
if hh==slavenu(8)%tgrape H

```

```

data((data(:,2))>=22100 & data(:,2)<=24899),5)=
data((data(:,2))>=22100 & data(:,2)<=24899),5)-11.8;
data((data(:,2))>=24900 & data(:,2)<=24999),5)=NaN;
data((data(:,2))>=22000 & data(:,2)<=22099),5)=NaN;
end
if hh==slavenumb(10)%tleaf L
data(:,6)=data(:,6)-10;
end
if hh==slavenumb(3)%tleaf C
data((data(:,2))>=20900 & data(:,2)<=20999),6)=NaN;
end
if hh==slavenumb(1) | hh==slavenumb(3) | hh==slavenumb(5) |
hh==slavenumb(6) | hh==slavenumb(9)
data((data(:,2))>=25700 & data(:,2)<=26099),6)=NaN;
data((data(:,2))>=26300 & data(:,2)<=26399),6)=NaN;
data((data(:,2))>=27700 & data(:,2)<=27799),6)=NaN;
end

elseif stt == 'DO'
%slavenumb=[65 66 67 68 69 70 71 72 73 76];
data(data(:,2)<9199,4)=NaN; %Tair dal 150 in poi
data(data(:,2)<15999,5)=NaN; %Tgrape dal 178 in poi
data(data(:,2)>28100,3:12)=NaN; %dati oltre vendemmia
data(data(:,2)<17099,6)=NaN; %Tleaf dal 170 in poi
data((data(:,2))>=11300 & data(:,2)<=11699),7:10)=NaN; %pot
idr di tutte
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
if hh==slavenumb(2) | hh==slavenumb(3) | hh==slavenumb(5) |
hh==slavenumb(7) | hh==slavenumb(8) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenumb(1) | hh==slavenumb(3) | hh==slavenumb(6) |
hh==slavenumb(8)%tolgo wind
data(data(:,2)>100,11)=NaN;
end

if hh==slavenumb(2)%tolgo valori della slave B
data((data(:,2))>=15000 & data(:,2)<=28000),4:12)=NaN;
end
if hh==slavenumb(3)%tleaf C
data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(5)%tleaf E
data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(7)%tleaf G
data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(4)%tleaf D
data((data(:,2))>=18900 & data(:,2)<=19299),7:10)=NaN;
data((data(:,2))>=22600 & data(:,2)<=28099),7:10)=NaN;
end

elseif stt == 'MB'
%slavenumb=[65 66 67 68 69 70 71 72 73 76];
data(data(:,2)<9199,4)=NaN; %Tair dal 150 in poi
data(data(:,2)<15999,5)=NaN; %Tgrape dal 178 in poi
data(data(:,2)>28100,3:12)=NaN; %dati oltre vendemmia
data(data(:,2)<17099,6)=NaN; %Tleaf dal 170 in poi
if hh==slavenumb(2) | hh==slavenumb(3) | hh==slavenumb(4) |
hh==slavenumb(6) | hh==slavenumb(7) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenumb(2) | hh==slavenumb(4) | hh==slavenumb(6) |
hh==slavenumb(8)%tolgo wind
data(data(:,2)>100,11)=NaN;
end
if hh==slavenumb(5)%tgrape E
data((data(:,2))>=18400 & data(:,2)<=18499),5)=NaN;
data((data(:,2))>=18700 & data(:,2)<=18799),5)=NaN;
data((data(:,2))>=19600 & data(:,2)<=20099),5)=NaN;
data((data(:,2))>=21000 & data(:,2)<=21099),5)=NaN;
data((data(:,2))>=21200 & data(:,2)<=21799),5)=NaN;

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data((data(:,2))>=22200 & data(:,2)<=22299),5)=NaN;
data((data(:,2))>=24600 & data(:,2)<=25299),5)=NaN;
data((data(:,2))>=25600 & data(:,2)<=25899),5)=NaN;
data((data(:,2))>=26700 & data(:,2)<=28099),5)=NaN;
end
if hh==slavenumb(4)%tgrape D
data((data(:,2))>=22800 & data(:,2)<=22899),5)=NaN;
data((data(:,2))>=26200 & data(:,2)<=26499),5)=NaN;
data((data(:,2))>=27400 & data(:,2)<=28099),5)=NaN;
end
if hh==slavenumb(6)%tgrape E
data((data(:,2))>=27000 & data(:,2)<=28099),5)=NaN;
end
%
16 troppo alte
%
%
%
%
end
if hh==slavenumb(2)%tgrape B senza coeff
data(:,5)=data(:,5)+1;
end
if hh==slavenumb(7)%tgrape G tolgo le ore dalle 12 alle 16
troppo alte
orennull=num2str(data(:,2));%
nuoveore=str2num(orennull(:,4:5));%
nnore=find(nuoveore<72 & nuoveore>49);
data(nnore,5) = NaN;
end
if hh==slavenumb(7)%tair G la tolgo
data((data(:,2))>=15000 & data(:,2)<=28099),4)=NaN;
end
if hh==slavenumb(3)%tair C tolgo le ore dalle 12 alle 16
troppo alte
orennull=num2str(data(:,2));%
nuoveore=str2num(orennull(:,4:5));%
nnore=find(nuoveore<72 & nuoveore>49);
data(nnore,4) = NaN;
end
if hh==slavenumb(4)%tleaf D
data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(9)%tleaf I
data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(2)%tleaf B
data((data(:,2))>=17000 & data(:,2)<=21999),6)=NaN;
end
if hh==slavenumb(7)%tleaf G
data((data(:,2))>=17000 & data(:,2)<=22699),6)=NaN;
data((data(:,2))>=23600 & data(:,2)<=28099),6)=NaN;
end
if hh==slavenumb(10)%tleaf L
data((data(:,2))>=22600 & data(:,2)<=23099),6)=NaN;
data((data(:,2))>=25700 & data(:,2)<=26099),6)=NaN;
end
if hh==slavenumb(1)%tleaf A
data((data(:,2))>=22800 & data(:,2)<=22999),6)=NaN;
data((data(:,2))>=25700 & data(:,2)<=25999),6)=NaN;
end

elseif stt == 'MA'
%slavenumb=[77 78 79 80 81 82 83 84 85 86];
data(data(:,2)<9199,4)=NaN; %Tair dal 150 in poi
data(data(:,2)<15999,5)=NaN; %Tgrape dal 178 in poi
data(data(:,2)>28100,3:12)=NaN; %dati oltre vendemmia
data(data(:,2)<17099,6)=NaN; %Tleaf dal 170 in poi
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
if hh==slavenumb(3) | hh==slavenumb(5) | hh==slavenumb(6) |
hh==slavenumb(7) | hh==slavenumb(10) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end

```

```

    if hh==slavenumb(5) | hh==slavenumb(6) | hh==
    slavenumb(7)%tolgo wind
    data(data(:,2)>100,11)=NaN;
    end

    if hh==slavenumb(6)%tgrape R troppo alta
    data((data(:,2)>=20400 & data(:,2)<=25699),5)=
    data((data(:,2)>=20400 & data(:,2)<=25699),5)-6;
    data((data(:,2)>=20300 & data(:,2)<=20399),5)=NaN;

    data((data(:,2)>=21750 & data(:,2)<=22200),5)=NaN;
    data((data(:,2)>=22400 & data(:,2)<=22600),5)=NaN;
    data((data(:,2)>=23000 & data(:,2)<=23550),5)=NaN;
    data((data(:,2)>=23600 & data(:,2)<=23950),5)=NaN;
    data((data(:,2)>=24600 & data(:,2)<=25700),5)=NaN;
    end
    if hh==slavenumb(3)%tgrape O tolgo le ore dalle 12 alle 15
    troppo alte
        orenull=num2str(data(:,2));%
        nuoveore=str2num(orenull(:,4:5));%
        nnore=find(nuoveore<69 & nuoveore>42);
        data(nnore,5) = NaN;
    end
    if hh==slavenumb(4)%tair P
    data((data(:,2)>=23600 & data(:,2)<=28099),4)=NaN;
    end
    if hh==slavenumb(5)%tleaf Q
    data((data(:,2)>=17000 & data(:,2)<=23099),6)=NaN;
    end
    if hh==slavenumb(7)%tleaf S
    data((data(:,2)>=17000 & data(:,2)<=28099),6)=NaN;
    end
    if hh==slavenumb(3)%tleaf O
    data((data(:,2)>=25700 & data(:,2)<=28099),6)=NaN;
    end
    if hh==slavenumb(6)%tleaf R
    data((data(:,2)>=25700 & data(:,2)<=26099),6)=NaN;
    data((data(:,2)>=26300 & data(:,2)<=26499),6)=NaN;
    end
end

    disp('          DESPIKE MANUALE INSERITO          ')

    %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

    %inserisci giorni mancanti
hourly=data;

    startday=fix(hourly(1,2)/100);
    endday=fix(hourly(end,2)/100);
    ndays=endday-startday+1;
    ncol=13;
    A=NaN*ones(ndays*24,ncol);
    A(:,1)=styear;
    for i=0:ndays-1
        for j=0:23
            A(i*24+j+1,2)=
                str2num([num2str(startday+i, '%.3d')
                    num2str(j, '%.2d')]);
        end
    end
    newdoy1=hourly(:,2)/100-fix(hourly(:,2)
    /100);
    newdoy2=int16(newdoy1*24);
    appdoy=hourly(:,2);
    hourly(:,2)=str2num([num2str(fix(hourly(:,2)
    /100), '%.3d') num2str(newdoy2, '%.2d')]);;
    B=hourly(:,2);

    for i=1:size(A,1)
        index=find(B(:,2)==A(i,2));
        if length(index)>=1
            A(i,:)=B(index(1),:);
        end
    end

```



```

clear all;
close all;
clc;

disp('          CI SONO I COEFFICIENTI CALIBRAZIONE le soglie e il
despiking e anche media mobile          ')

st=input('ID station (LOC=Master;S=Slave): ','s');
stt=input('Vigna (BR=Brolio;MA=Mortelle alta Cacciagrande;MB=Mortelle bassa
Cortigliano;DO=Donna Olimpia): ','s');
styear=input('Anno: ');
safety=input('Vuoi salvare il file output ? Si(1) No(2); ');
File_save='C:\TUSCANIA\DATI_2009\Scheduler\OUTPUT_ORARI_DESPIKING';
File_save_desp='C:\TUSCANIA\DATI_2009\Scheduler\OUTPUT_GIORNALIERI_DESPIKIN
G';
filemast=input('File MASTER: ','s');

%
%                                coefficienti calibrazione 2008
% if stt == 'BR'
%   File_path='C:\TUSCANIA\DATI\Scheduler\Brolio\OUTPUT';
%   File_path_loc='C:\TUSCANIA\DATI\Scheduler\Brolio\OUTPUT';
%   slavenumb=[65 66 67 68 69 70 71 72 73 76 77];
%   span_air=[0.9768      0.9687  0.9896  0.9805  0.9963  0.9936
0.9879  0.9896  1      1.0245  0.9871];
%   offset_air=[-0.3841  -0.4947  0.3281  0.3615  0.9952  -0.1988
0.7426  0.3281  -0.0471  -1.0027  0.5001];
%   span_grape=[0.9947      1.0013  0.998  1      0.9986  0.9515
0.9832  0.998  0.9995  1.0293  0.9659];
%   offset_grape=[-0.4849  -0.5348  -0.515  0.3224  1.2254  0.1085
1.2596  -0.515  -0.2466  -1.0655  0.5802];
%   span_leaf=[1  1      1      1      1      1      1
1      1      1];
%   offset_leaf=[0      0      0      0      0      0      0
0      0      0      0];
%   offset_rad=[-13.1  5.2  8.1  -5.6  -13.4  9.3
-2.0  12.6  -6.6  -1.3  -5.6];
%   doy_cal=9400;
%   doy_rad=15900;
%   doy_cal2=13900;
%   doy_delta=10600;
%
%   %slavenumb=char('A','B','C','D','E','F','G','H','I','L','M');
% elseif stt == 'DO'
%   File_path='C:\TUSCANIA\DATI\Scheduler\Donna_Olimpia\OUTPUT';
%   File_path_loc='C:\TUSCANIA\DATI\Scheduler\Donna_Olimpia\OUTPUT';
%   slavenumb=[65 66 67 68 69 70 71 72 73 76];
%   span_air=[0.9635      1.1082  1.0333  1.0157  1.0389  1.0085
0.9971  0.9553  0.9532  0.9822];
%   offset_air=[0.2372  -1.8737  0.0884  0.9189  0.017  -0.5595
0.0606  -0.3388  0.7286  -0.7899];
%   span_grape=[1.0497      1.1019  0.9895  1.0048  0.9617  1.0022
0.9724  0.9918  1.0005  1.0031];
%   offset_grape=[-0.7762  -1.9487  0.8975  -0.2804  -0.2618  -0.5959
0.6029  -0.4012  0.2987  0.9814];
%   span_leaf=[1  1      1      1      1      1      1
1      1];
%   offset_leaf=[0      0      0      0      0      0      0
0      0      0];
%   offset_rad=[19.1  14.7  14.7  14.7  10.2  -4.3
-17.9  -5.2  -2.8  -13.7];
%   doy_cal=10200;
%   doy_rad=16500;
%   doy_cal2=14700;
%   doy_delta=11400;
%
%   %slavenumb=char('A','B','C','D','E','F','G','H','I','L');
% elseif stt == 'MB'
%   File_path='C:\TUSCANIA\DATI\Scheduler\Mortelle\OUTPUT';
%   File_path_loc='C:\TUSCANIA\DATI\Scheduler\Mortelle_Bassa\OUTPUT';
%   slavenumb=[65 66 67 68 69 70 71 72 73 76];
%   span_air=[0.976      1      1.0248  0.9842  1.0053  0.9795
1.0171  0.9686  1      0.9805];
%   offset_air=[-0.4343  0      -0.9627  0.8467  -0.1367  0.5064

```

```

-0.2492 0.4459 0 0.8548];
% span_grape=[0.9842 1 0.9926 0.9853 0.9859 0.9912
0.9774 1.0022 1 1.0192];
% offset_grape=[0.2374 0 0.1723 0.3958 0.2215 0.1312
0.9222 -0.6727 0 -0.1189];
% span_leaf=[1.0493 1 1.1018 1.0705 1 1.0877
1.113 1 1 1];
% offset_leaf=[-1.0401 0 -2.2474 -1.2487 0 -0.9331
-1.5984 0 0 0];
% offset_rad=[-20.0 7.3 -18.5 9.6 11.5 -9.7 0.0
19.7 2.0 3.0];
% doy_cal=10200;
% doy_rad=15000;
% doy_cal2=14700;
% doy_delta=11400;
% %slavenumb=char('A','B','C','D','E','F','G','H','I','L');
% elseif stt == 'MA'
% File_path='C:\TUSCANIA\DATI\Scheduler\Mortelle\OUTPUT';
% File_path_loc='C:\TUSCANIA\DATI\Scheduler\Mortelle_Alta\OUTPUT';
% slavenumb=[77 78 79 80 81 82 83 84 85 86];
% span_air=[1 1 0.9922 0.9844 0.9936 1.0149 0.9786
0.998 0.9737 0.9881];
% offset_air=[0 0 -0.2944 0.1739 0.9748 -0.2097 0.2287
0.3717 0.4536 -0.0971];
% span_grape=[1 1 0.9882 0.9745 0.9238 0.9713 0.9783
0.9825 0.9768 0.9955];
% offset_grape=[0 0 0.1157 0.5809 2.1741 0.499
0.1835 0.8863 0.2278 -0.4314];
% span_leaf=[1 1 0.9509 0.9724 1.0826 1.0143 1
0.9769 0.9545 0.9725];
% offset_leaf=[0 0 0.2133 0.3896 -1.0441 -0.1543 0
1.2074 1.1554 -0.0696];
% offset_rad=[-13.7 12.9 7.7 16.6 -16.3 9.6 0.0
-10.4 9.6 -17.4];
% doy_cal=10200;
% doy_rad=15000;
% doy_cal2=14700;
% doy_delta=11400;
% %slavenumb=char('M','N','O','P','Q','R','S','T','U','V');
% end

if stt == 'BR'
File_path='C:\TUSCANIA\DATI_2009\Scheduler\Brolio\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2009\Scheduler\Brolio\OUTPUT';
slavenumb=[65 66 67 68 69 70 71 72 73 76 77];
span_air=[0.9986 0.9998 0.9873 1.0031 1.0014 1.0062
0.9983 0.9871 0.9932 1.0162 0.984];
offset_air=[-0.4787 -0.7786 0.5256 0.4673 1.0731 -0.1523
0.9217 -0.632 0.2644 -1.2181 0.1937];
span_grape=[1.0105 1.0132 1.0178 1.0109 1.0083 1.0184
1.0083 0.9664 0.9633 1.0132 0.9707];
offset_grape=[-0.8012 -0.8762 0.2164 0.0853 1.0467 -0.4123
1.0467 -0.0009 0.6149 -0.8762 0.2857];
span_leaf=[1.0048 0.9911 1.0001 0.9926 0.9925 1.0035 1
1.0863 1.192 1.0324 1.0709];
offset_leaf=[-0.4084 -0.9066 0.1508 0.3013 1.0134 -0.0218 0
-7.6104 -17.116 -89.01 -8.7856];
offset_rad=[-13.1 5.2 8.1 -5.6 -13.4 9.3
-2.0 12.6 -6.6 -1.3 -5.6];

span_grape2=[1 1 1 1 0.9479 0.9164 0.9586
0.923 0.5545 0.9891 1];
offset_grape2=[0 0 0 0 0.6468 0.947
0.6883 0.2833 17.756 0.3975 0];
span_grape3=[1 1 1 1 0.9419 0.9479 0.9718
0.9538 1.0063 0.9966 0.9664 1];
offset_grape3=[0 0 0 0 1.2228 0.6468 0.785
0.0037 0.4635 -0.5021 0];

doy_cal=6000;
doy_rad=6000;
doy_cal2=6000;
doy_delta=6000;

```

```

%slavenumb=char('A','B','C','D','E','F','G','H','I','L','M');
elseif stt == 'DO'
File_path='C:\TUSCANIA\DATI_2009\Scheduler\Donna Olimpia\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2009\Scheduler\Donna Olimpia\OUTPUT';
slavenumb=[65 66 67 68 69 70 71 72 73 76];
span_air=[0.9788      0.9795  0.9864  0.9944  1.0424  1.0026
0.9639  0.9725  0.9605  0.9794];
offset_air=[0.2369      0.0239  0.8506  0.8263  -0.2877  -0.792
0.0876  -0.732  0.1879  0.7034];
span_grape=[0.999      0.998  0.993  1.003  1      1.0163
0.9145  0.9513  0.9471  1];
offset_grape=[0.3749  -0.1538  1.2598  -0.2013  0      -0.5571
0.7891  -0.3176  0.2983  0];
span_leaf=[0.9919      0.9839  0.9904  0.9424  0.9688  0.9909
1.1171  1.0556  1.1161  1.0356];
offset_leaf=[-0.153  -0.2263  1.2385  -8.3202  -12.345  -0.4957
-10.776  -6.6562  -10.505  -89.537];
offset_rad=[19.1      14.7  14.7  14.7  10.2  -4.3
-17.9  -5.2  -2.8  -13.7];

span_grape2=[1  0.9894  0.9373  1  0.8827      1  0.9935
0.9619  1      1];
offset_grape2=[0      0.3922  0.6034  0  2.8871      0  -0.2301
0.6141  0      0];
span_grape3=[1  0.9971  0.9595  1      0.891  1  1.0009
0.9414  1  1];
offset_grape3=[0  -0.5111  0.5734  0      1.4996  0
-0.1732  1.3277  0  0];

doy_cal=6000;
doy_rad=6000;
doy_cal2=6000;
doy_delta=6000;
%slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MB'
File_path='C:\TUSCANIA\DATI_2009\Scheduler\Mortelle\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2009\Scheduler\Mortelle_Bassa\OUTPUT';
slavenumb=[65 66 67 68 69 70 71 72 73 76];
span_air=[0.9563      1.0174  1.0020  1      0.9957  1
0.9238  1.003  1      1.0152];
offset_air=[-0.5709  0.1590  -0.5750  -5      -0.5224  0
-5.2875  -0.7151  0      -0.3015];
span_grape=[0.9935      0.9845  0.9856  0.998  0.9932  1
0.9903  0.9926  1      0.9756];
offset_grape=[-0.3551  0.8275  -0.0938  -0.2911  -0.2935  -3
0.1701  -0.6737  0      0.1772];
span_leaf=[0.9901      0.9974  0.9971  0.9974  0.992  0.9953
0.9917  1.0197  0.9854  0.9977];
offset_leaf=[0.0339  0.3704  0.0087  0.1513  0.0754  -88.51
0.5024  -0.1468  -0.4012  0.0628];
offset_rad=[-20.0      7.3  -18.5  9.6  11.5  -9.7  0.0
19.7  2.0  3.0];
span_grape2=[1  1.0767  1.0134  1  1      0.7785  0.8568  1
1  1];
offset_grape2=[0      0.0343  -1.1855  0      0      3.0006
6.2336  0  0  0];
span_grape3=[1  0.9537  1.4806  1  1      0.7903  0.9374  1
1  1];
offset_grape3=[0      0.9909  -6.472  0      0      2.3206
1.5795  0  0  0];
doy_cal=6000;
doy_rad=6000;
doy_cal2=6000;
doy_delta=6000;
%slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MA'
File_path='C:\TUSCANIA\DATI_2009\Scheduler\Mortelle\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2009\Scheduler\Mortelle_Alta\OUTPUT';
slavenumb=[77 78 79 80 81 82 83 84 85 86];
span_air=[0.9990      1.0152  1.0063  0.9621  1.0064  0.9991
0.9991  0.9975  0.9838  0.9991];
offset_air=[0.2243      -0.0583  -0.4651  -7.5271  0.6906  0.0262

```

```

-0.2196 0.1673 0.0872 -0.3192];
span_grape=[0.9895 1.0015 0.9993 0.9851 0.9861 1
0.9794 0.9754 1.0208 0.9908];
offset_grape=[0.524 0.227 -0.0716 0.2514 0.9906 0
0.2423 0.5813 -0.1064 -0.1645];
span_leaf=[0.9925 0.9902 0.9928 0.99 1.0008 0.9873
0.9846 1.0151 0.996 0.9894];
offset_leaf=[0.6657 0.5189 0.0485 0.5852 -2.9023 0.3903
0.4683 -2.6994 0.1953 0.2872];
offset_rad=[-13.7 12.9 7.7 16.6 -16.3 9.6 0.0
-10.4 9.6 -17.4];

span_grape2=[1 1 0.8662 1 1 1.1 0.9969 1
1 0.9919];
offset_grape2=[0 0 2.6229 0 0 -0.7983 -0.4557 0
0 -0.1082];
span_grape3=[1 1 0.9581 1 1 0.9921 1.0114 1
1 1.0238];
offset_grape3=[0 0 0.8046 0 0 -0.0535 -0.105 0
0 -0.5867];

doy_cal=6000;
doy_rad=6000;
doy_cal2=6000;
doy_delta=6000;
%slavenumb=char('M','N','O','P','Q','R','S','T','U','V');
end

if st=='LOC'
File_wrk=[File_path_loc '\\' st '.dat'];
N=dload(File_wrk);
% MASTER
% year doym Tempmedia Tmax Tmin Tdevstd Tultimo Umidmedia
URmax URmin URdevstd URultimo Presmedia
% 1 2 3 4 5 6 7 8 9 10
11 12 13
% Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
Rultimo Piogsomma VVentlVmed Vdevstd Vraf
% 14 15 16 17 18 19 20 21 22 23
24 25 26
% Vmin Vultimo DVentlDprev Draf Dultimo numcamp Batteria
% 27 28 29 30 31 32 33
%step 1 ordina ed elimina doppie
[Y,I]=sort(N.doyM(:,1));
datatemp=[N.year(I,1) N.doyM(I,1) N.Tempmedia(I,1) N.Tmax(I,1)
N.Tmin(I,1) N.Tdevstd(I,1) N.Tultimo(I,1) N.Umidmedia(I,1)
N.URmax(I,1) N.URmin(I,1) N.URdevstd(I,1) N.URultimo(I,1)
N.Presmedia(I,1) N.Pmax(I,1) N.Pmin(I,1) N.Pdevstd(I,1)
N.Pultimo(I,1) N.RadGmedia(I,1) N.Rmax(I,1) N.Rmin(I,1)
N.Rdevstd(I,1) N.Rultimo(I,1) N.Piogsomma(I,1) N.VVentlVmed(I,1)
N.Vdevstd(I,1) N.Vraf(I,1) N.Vmin(I,1) N.Vultimo(I,1)
N.DVentlDprev(I,1) N.Draf(I,1) N.Dultimo(I,1) N.numcamp(I,1)
N.Batteria(I,1)];
[I,J]=unique(datatemp(:,2));
data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
datatemp(J,13) datatemp(J,14) datatemp(J,15) datatemp(J,16)
datatemp(J,17) datatemp(J,18) datatemp(J,19) datatemp(J,20)
datatemp(J,21) datatemp(J,22) datatemp(J,23) datatemp(J,24)
datatemp(J,25) datatemp(J,26) datatemp(J,27) datatemp(J,28)
datatemp(J,29) datatemp(J,30) datatemp(J,31) datatemp(J,32)
datatemp(J,33)];
F = data(:, :);
F(F== -999) = NaN;
data(:, :)=F;

% elaborazioni giornaliere
appog=data(1,2);
app2=num2str(data(1,2));
data(1,2)=1111111;

```

```

prova=num2str(data(:,2));
D=str2num(prova(:,1:3));
D(1)=appog;
doyi=D(2);
doyf=D(end);
%elaborazioni orarie
prova2=num2str(data(:,2));%
D2=str2num(prova2(:,4:7));%
%D2(1)=str2num(app2(:,4:7));
D2=D2/100;%
D3=fix((D2+0.1)*24/100) ; %
D3(1)=D3(2);
ii=1;
i2=1;%
data(1,2)=appog;

for i=doyi:doyf
    ijj = find(D==i);
    if size(ijj,1)>1
        daily(ii,:) = nanmean(data(ijj,:)); % per le medie giornaliere
        daily(ii,2)=i;
        daily(ii,23)= nansum(data(ijj,23)); % cumulate giornaliere
        precipitazione mm
        daily(ii,4)= nanmax(data(ijj,4));
        daily(ii,5)= nanmin(data(ijj,5));
        daily(ii,9)= nanmax(data(ijj,9));
        daily(ii,10)= nanmin(data(ijj,10));
        daily(ii,19)= nanmax(data(ijj,19));
        daily(ii,20)= nanmin(data(ijj,20));

        if styear>2007
            daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
            sfilato km giorno rilev 5 min
            daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
            radiazione Mj /m2
        else
            if (stt=='DO' & i>=222 | stt=='BR' & i>=194 | stt=='MB' &
                i>=261 | stt=='MA' & i>=261)
                daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
                sfilato km giorno rilev 5 min
                daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
                radiazione Mj /m2
            else
                daily(ii,24)= nansum(data(ijj,24)*900/1000); % vento
                sfilato km giorno rilev 15 min
                daily(ii,18)= nansum(data(ijj,18)*900/1000000); % cumulate
                radiazione Mj /m2
            end
        end
        ii=ii+1;
    else
        daily(ii,:) = NaN;
        daily(ii,2)=i;
        ii=ii+1;
    end
end

%elaborazione oraria
    if size(ijj,1)>1
        zz1=ijj(1);
        for iii=D3(ijj(1)):D3(ijj(end))
            ij2=find(D3(ijj)==iii);
            if size(ij2,1)>1
                zz=zz1+length(ij2);
                azz=data(zz1:(zz-1),:);
                zz1=zz;
                hourly(i2,:)=nanmean(azz(:,,:));
                hourly(i2,2)=i+double(iii*100/24)/100;
                hhour(i2,1)=i;
                hhour(i2,2)=iii;
                hourly(i2,23)= nansum(azz(:,23)); %
                cumulate giornaliere precipitazione mm
                hourly(i2,4)= nanmax(azz(:,4));
                hourly(i2,5)= nanmin(azz(:,5));
            end
        end
    end

```



```

        hourly(i2,9)= nanmax(azz(:,9));
        hourly(i2,10)= nanmin(azz(:,10));
        hourly(i2,19)= nanmax(azz(:,19));
        hourly(i2,20)= nanmin(azz(:,20));

        if iii==24
            hourly(i2,2)=i;
        end
        i2=i2+1;
    else
        hourly(i2,:) = NaN;
        hourly(i2,2)=i+double(iii*100/24)/100;
        i2=i2+1;
    end
end
else
    hourly(i2,:) = NaN;
    hourly(i2,2)=i+double(iii*100/24)/100;
    i2=i2+1;
end
end
%fine elab oraria

end

        %inserisci giorni mancanti
hourly(:,2)=hourly(:,2)*100;
startday=fix(hourly(1,2)/100);
endday=fix(hourly(end,2)/100);
ndays=endday-startday+1;
ncol=33;
A=NaN*ones(ndays*24,ncol);
A(:,1)=styear;
for i=0:ndays-1
    for j=0:23
        A(i*24+j+1,2)=
            str2num([num2str(startday+i,'%3d')
                num2str(j,'%2d')]);
    end
end

        newdoy1=hourly(:,2)/100-fix(hourly(:,2)
            /100);
        newdoy2=int16(newdoy1*24);
        appdoy=hourly(:,2);
        hourly(:,2)=str2num([num2str(fix(hourly(:,2)
            /100),'%.3d') num2str(newdoy2,'%2d')]);
        B=hourly(:,2);

        for i=1:size(A,1)
            index=find(B(:,2)==A(i,2));
            if length(index)>=1
                A(i,:)=B(index(1),:);
            end
        end

hourly=A;
%
% % salvataggio file .mat
% %
% % eval(['M_' stt '_RT = data(:,:);']);
% % eval(['M_' stt '_DY = daily(:,:);']);
% % eval(['M_' stt '_HY = hourly(:,:);']);
% % %eval(['M_' stt '_HY(:,2) = M_' stt '_HY(:,2)*100;']);
% % fname=['M_' stt];
% % eval(['save C:\TUSCANIA\DATI\Scheduler\Master\' fname ' M_*;']);

% salvataggio files REAL TIME MASTER

if safety == 1
%
% adesso=datestr(now, 'ddmmyyyy');
% file_out=[File_path '\' stt 'M_' adesso '.out'];
% fidl = fopen(file_out, 'w');
% fprintf(fidl,'data ora Tempmedia Tmax Tmin Tdevstd Tultimo
Umidmedia URmax URmin URdevstd URultimo Presmedia Pmax Pmin Pdevstd Pultimo
RadGmedia Rmax Rmin Rdevstd Rultimo Piogsomma VVentlVmed Vdevstd Vraf Vmin
Vultimo DVentlDprev Draf Dultimo numcamp Batteria\r');

```



```

%             hold on
%             ylabel('Temperature')
%             subplot(4,2,1),plot(data(:,2),data(:,4),'r');
%             hold on
%             ylabel('Tmax')
%             subplot(4,2,1),plot(data(:,2),data(:,5),'r');
%             hold on
%             ylabel('Tmin')
%             %axis([600 900 15 28])
%             subplot(4,2,2),plot(data(:,2),data(:,8));
%             hold on
%             ylabel('Relative Humidity')
%             %axis([0 i -20 20])
%             subplot(4,2,2),plot(data(:,2),data(:,9),'r');%URmax
%             hold on
%             subplot(4,2,2),plot(data(:,2),data(:,10),'r');%URmin
%             hold on
%             % Presmedia Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
%             Rultimo Piogsomma
%             subplot(4,2,3),plot(data(:,2),data(:,13));
%             hold on
%             ylabel('Pressure (hPA)')
%             subplot(4,2,4),plot(data(:,2),data(:,18));
%             hold on
%             ylabel('Radiation (W/m2)')
%             subplot(4,2,5),plot(data(:,2),data(:,23));
%             hold on
%             ylabel('Precipitation (mm)')
%             % VVent1Vmed Vdevstd Vraf Vmin Vultimo DVent1Dprev Draf Dultimo
%             numcamp Batteria
%             subplot(4,2,6),plot(data(:,2),data(:,24));
%             hold on
%             ylabel('Wind (Km giorno)')
%             subplot(4,2,7),plot(data(:,2),data(:,32));
%             hold on
%             ylabel('numpack (#)')
%             subplot(4,2,8),plot(data(:,2),data(:,33));
%             hold on
%             ylabel('Battery (mV)')
%
%             figure('Name','Master Daily')
%             subplot(4,2,1),plot(daily(:,2),daily(:,3));
%             hold on
%             ylabel('Temperature')
%             subplot(4,2,1),plot(daily(:,2),daily(:,4),'r');
%             hold on
%             ylabel('Tmax')
%             subplot(4,2,1),plot(daily(:,2),daily(:,5),'r');
%             hold on
%
%             ylabel('Tmin')
%             %axis([600 900 15 28])
%             subplot(4,2,2),plot(daily(:,2),daily(:,8));
%             hold on
%             ylabel('Relative Humidity')
%             %axis([0 i -20 20])
%             subplot(4,2,2),plot(daily(:,2),daily(:,9),'r');%URmax
%             hold on
%             subplot(4,2,2),plot(daily(:,2),daily(:,10),'r');%URmin
%             hold on
%             % Presmedia Pmax Pmin Pdevstd Pultimo RadGmedia Rmax Rmin Rdevstd
%             Rultimo Piogsomma
%             subplot(4,2,3),plot(daily(:,2),daily(:,13));
%             hold on
%             ylabel('Pressure (hPA)')
%             subplot(4,2,4),plot(daily(:,2),daily(:,18));
%             hold on
%             ylabel('Radiation (MJ/m2)Sum Daily')
%             subplot(4,2,5),plot(daily(:,2),daily(:,23));
%             hold on

```

```

%                               ylabel('Precipitation (mm)')
% %                               VVent1Vmed Vdevstd Vraf Vmin Vultimo DVent1Dprev Draf Dultimo
numcamp Batteria
%                               subplot(4,2,6),plot(daily(:,2),daily(:,24));
%                               hold on
%                               ylabel('Wind (m/s)')
%                               subplot(4,2,7),plot(daily(:,2),daily(:,32));
%                               hold on
%                               ylabel('numpack (#)')
%                               subplot(4,2,8),plot(daily(:,2),daily(:,33));
%                               hold on
%                               ylabel('Battery (mV)')
%
else
s1=1;%inizializzo indici vettori calibrazione
f1=1;
for hh=slavenumb(1):slavenumb(end)
if (hh==74 | hh==75)
pippo='N';
else
File_wrk=[File_path '\' hh '.dat'];
M=dload(File_wrk);
%                               File_master=[File_path_master '\M_' stt '.mat'];
%                               load(File_master);
%                               eval(['Master_hy=M_' stt '_HY;'])
%                               eval(['Master_dy=M_' stt '_DY;'])

% SLAVE
% year DOY_hh rad Tair Tgrap Tleaf Tsoil1 Tsoil2 T_heatsoil1
T_heatsoil2 VV Bat Bagn_leaf
% 1 2 3 4 5 6 7 8 9
10 11 12 13
%step 1 ordina ed elimina doppie
[Y,I]=sort(M.DOY_hh(:,1));
datatemp=[M.Year(I,1) M.DOY_hh(I,1) M.Tcjc(I,1) M.Tair(I,1)
M.Tgrap(I,1) M.Tleaf(I,1) M.Tsoil1(I,1) M.Tsoil2(I,1)
M.T_heatsoil1(I,1) M.T_heatsoil2(I,1) M.VV(I,1) M.Bat(I,1)
M.Bagn_leaf(I,1)];
[I,J]=unique(datatemp(:,2));
data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
datatemp(J,13)];
F = data(:, :);
F(F==-999) = NaN;
data(:, :)=F;

wind=data(:,11);
wind(wind==0.2)=0;
data(:,11)=wind;

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%%%
%applico i coefficienti di calibrazione per le slave
data(data(:,2)>=doy_cal,4)=(data(data(:,2)>=doy_cal,4)
*span_air(s1))+offset_air(f1);
data(data(:,2)>=doy_cal,5)=(data(data(:,2)>=doy_cal,5)
*span_grape(s1))+offset_grape(f1);
data(data(:,2)>=doy_cal,6)=(data(data(:,2)>=doy_cal,6)
*span_leaf(s1))+offset_leaf(f1);
data(data(:,2)>=doy_cal,7)=(data(data(:,2)>=doy_cal,7)
*span_grape2(s1))+offset_grape2(f1);
data(data(:,2)>=doy_cal,8)=(data(data(:,2)>=doy_cal,8)
*span_grape3(s1))+offset_grape3(f1);

%
%

%sensori radiazione grappolo
data(data(:,2)<doy_rad,3)=NaN;
%coefficienti calibrazione radiazione grappolo
data(data(:,2)>=doy_rad,3)=data(data(:,2)>=doy_rad,3)
+offset_rad(f1);
%elimino i negativi

```

```

    apprad=data(:,3);
    apprad(apprad<=25)=0;
    data(:,3)=apprad;
    s1=s1+1;
    f1=f1+1;
%
% conversione T_heatsoil da temperatura a delta T
% data(data(:,2)<=doy_delta,9)=(data(data(:,2)<=
doy_delta,9))-(data(data(:,2)<=doy_delta,7));
% data(data(:,2)<=doy_delta,10)=(data(data(:,2)<=
doy_delta,10))-(data(data(:,2)<=doy_delta,8));
%
%inserimento soglie deltaT per il potenziale idrico
%considerare solo valori di delta tra 0.01<x<2.75
pot30=data(:,9);
pot30(pot30>=2.75)=2.75;
pot30(pot30<=0.01)=0.01;

pot60=data(:,10);
pot60(pot60>=2.75)=2.75;
pot60(pot60<=0.01)=0.01;

%
% RANGE DI POTENZIALE IDRICO
% Tasterisco30=(3.15-pot30)/(3.15-0.75);
% pot30=-0.056*(Tasterisco30.^(-1/0.45)-1);
% Tasterisco60=(3.15-pot60)/(3.15-0.75);
% pot60=-0.056*(Tasterisco60.^(-1/0.45)-1);
% %da Mpa a pF (scala 1-7) %-LOG10(1/ASS(E12*10000)) %-
log10(1/abs(-2.95*10000))
% % pf_i_30=abs(pot30*10000);
% % pf_i_60=abs(pot60*10000);
% % pf30=-log10(1./pf_i_30);
% % pf60=-log10(1./pf_i_60);
% % rh30=-22.392*log(pf30-1)+42.554; curva ritenzione media
campioni donna olimpia
% % rh60=-22.392*log(pf60-1)+42.554;
% % %data(:,9)=rh30;
% % %data(:,10)=rh60;
% % %%%%%%%%%%
% data(:,9)=pot30;
% data(:,10)=pot60;

%
%inserimento soglie temperatura aria e grappolo
s_air=data(:,4);
s_air(s_air>=50 | s_air<=-20)=NaN;
data(:,4)=s_air;
s_grape=data(:,5);
s_grape(s_grape>=50 | s_grape<=-20)=NaN;
data(:,5)=s_grape;
s_leaf=data(:,6);
s_leaf(s_leaf>=60 | s_leaf<=-20)=NaN;
data(:,6)=s_leaf;

%
%
%
%despiking manuale ESEGUITO GUARDANDO I DATI GIORNALIERI

%filtro a 12 ore potenziale idrico
aaa=1;
bb12 = [1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12 1/12
1/12 1/12];
yy12_30 = filter(bb12,aaa,data(:,9));
yy12_60 = filter(bb12,aaa,data(:,10));
data(:,9)=yy12_30;
data(:,10)=yy12_60;

%
if stt == 'BR'
    %slavenumb=[65 66 67 68 69 70 71 72 73 76 77];
    data(data(:,2)<9099,4)=NaN; %Tair dal 91 in poi
    data(data(:,2)<15299,5)=NaN; %Tgrape dal 162 in poi
    data(data(:,2)>36500,3:12)=NaN; %dati oltre vendemmia
    data(data(:,2)<16299,6)=NaN; %Tleaf dal 162 in poi

    data(data(:,2)<16299,7)=NaN; %Tgrape2 dal 162 in poi
    data(data(:,2)<16299,8)=NaN; %Tgrape3 dal 162 in poi

```



```

%           if hh==slavenumb(5)%tgrape E
%           data((data(:,2))>=18400 & data(:,2)<=18499),5)=NaN;
%           data((data(:,2))>=18700 & data(:,2)<=18799),5)=NaN;
%           data((data(:,2))>=19600 & data(:,2)<=20099),5)=NaN;
%           data((data(:,2))>=21000 & data(:,2)<=21099),5)=NaN;
%           data((data(:,2))>=21200 & data(:,2)<=21799),5)=NaN;
%           data((data(:,2))>=22200 & data(:,2)<=22299),5)=NaN;
%           data((data(:,2))>=24600 & data(:,2)<=25299),5)=NaN;
%           data((data(:,2))>=25600 & data(:,2)<=25899),5)=NaN;
%           data((data(:,2))>=26700 & data(:,2)<=28099),5)=NaN;
%           end
%           if hh==slavenumb(4)%tgrape D
%           data((data(:,2))>=22800 & data(:,2)<=22899),5)=NaN;
%           data((data(:,2))>=26200 & data(:,2)<=26499),5)=NaN;
%           data((data(:,2))>=27400 & data(:,2)<=28099),5)=NaN;
%           end
%           if hh==slavenumb(6)%tgrape E
%           data((data(:,2))>=27000 & data(:,2)<=28099),5)=NaN;
%           end
%           if hh==slavenumb(4)%tgrape D tolgo le ore dalle 12 alle
16 troppo alte
%           orenull=num2str(data(:,2));%
%           nuoveore=str2num(orenull(:,4:5));%
%           nnore=find(nuoveore<72 & nuoveore>49);
%           data(nnore,5) = NaN;
%           end
%           if hh==slavenumb(2)%tgrape B senza coeff
%           data(:,5)=data(:,5)+1;
%           end
%           if hh==slavenumb(7)%tgrape G tolgo le ore dalle 12 alle
16 troppo alte
%           orenull=num2str(data(:,2));%
%           nuoveore=str2num(orenull(:,4:5));%
%           nnore=find(nuoveore<72 & nuoveore>49);
%           data(nnore,5) = NaN;
%           end
%           if hh==slavenumb(7)%tair G la tolgo
%           data((data(:,2))>=15000 & data(:,2)<=28099),4)=NaN;
%           end
%           if hh==slavenumb(3)%tair C tolgo le ore dalle 12 alle 16
troppo alte
%           orenull=num2str(data(:,2));%
%           nuoveore=str2num(orenull(:,4:5));%
%           nnore=find(nuoveore<72 & nuoveore>49);
%           data(nnore,4) = NaN;
%           end
%           if hh==slavenumb(4)%tleaf D
%           data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
%           end
%           if hh==slavenumb(9)%tleaf I
%           data((data(:,2))>=17000 & data(:,2)<=28099),6)=NaN;
%           end
%           if hh==slavenumb(2)%tleaf B
%           data((data(:,2))>=17000 & data(:,2)<=21999),6)=NaN;
%           end
%           if hh==slavenumb(7)%tleaf G
%           data((data(:,2))>=17000 & data(:,2)<=22699),6)=NaN;
%           data((data(:,2))>=23600 & data(:,2)<=28099),6)=NaN;
%           end
%           if hh==slavenumb(10)%tleaf L
%           data((data(:,2))>=22600 & data(:,2)<=23099),6)=NaN;
%           data((data(:,2))>=25700 & data(:,2)<=26099),6)=NaN;
%           end
%           if hh==slavenumb(1)%tleaf A
%           data((data(:,2))>=22800 & data(:,2)<=22999),6)=NaN;
%           data((data(:,2))>=25700 & data(:,2)<=25999),6)=NaN;
%           end
elseif stt == 'MA'
    %slavenumb=[77 78 79 80 81 82 83 84 85 86];
    data(data(:,2)<9099,4)=NaN; %Tair dal 148 in poi

```



```

        newdoy1=hourly(:,2)/100-fix(hourly(:,2)
        /100);
        newdoy2=int16(newdoy1*24);
        appdoy=hourly(:,2);
        hourly(:,2)=str2num([num2str(fix(hourly(:,2)
        /100), '%.3d') num2str(newdoy2, '%.2d')]);
        B=hourly(:,:);

        for i=1:size(A,1)
            index=find(B(:,2)==A(i,2));
            if length(index)>=1
                A(i,:)=B(index(1),:);
            end
        end

hourly=A;

appog=hourly(1,2);
hourly(1,2)=11111;

        doyi=int16(hourly(2,2)/100);
        doyf=int16(hourly(end,2)/100);
        prova=num2str(hourly(:,2));
        D=str2num(prova(:,1:3));
        ii=1;
hourly(1,2)=appog;
D(1)=D(2);
        for i=doyi:doyf
            ijj = find(D==i);
            if size(ijj,1)>1
                % controllo giorni parziali
                for iii=1:13
                    ina=isnan(hourly(ijj,iii));
                    if sum(ina) >= 7
                        hourly(ijj,iii) = NaN;
                    end
                end
                end
                %%%%%%%%%%%
                daily(ii,1:13) = nanmean(hourly(ijj,:)); % per le medie
                giornaliere
                daily(ii,2)=i;
                daily(ii,13)= sum(hourly(ijj,13)); % per le cumulate
                giornaliere es. pioggia
                daily(ii,11)= sum(hourly(ijj,11)*3.6); % vento sfilato
                daily(ii,3)= nansum(hourly(ijj,3)); %radiazione W m-2
                (integrale giornaliera)
                daily(ii,14)= nanmax(hourly(ijj,4)); %Tair max
                daily(ii,15)= nanmin(hourly(ijj,4)); %Tair min
                daily(ii,16)= nanmax(hourly(ijj,5)); %Tgrap max
                daily(ii,17)= nanmin(hourly(ijj,5)); %Tgrap min
                daily(ii,18)= nanmax(hourly(ijj,6)); %Tleaf max
                daily(ii,19)= nanmin(hourly(ijj,6)); %Tleaf min
                daily(ii,20)= nanmax(hourly(ijj,7)); %Tsoil1 max
                daily(ii,21)= nanmin(hourly(ijj,7)); %Tsoil1 min
                daily(ii,22)= nanmax(hourly(ijj,8)); %Tsoil2 max
                daily(ii,23)= nanmin(hourly(ijj,8)); %Tsoil2 min
                daily(ii,24)= nanmax(hourly(ijj,11)); %Wind max
                daily(ii,25)= nanmin(hourly(ijj,11)); %Wind min

                ii=ii+1;
            else
                daily(ii,:) = NaN;
                daily(ii,2)=i;
                ii=ii+1;
            end
        end

        % salvataggio file .mat per despiking
        %eval(['S_' hh '_' stt '_HY = hourly(:,:);']);

        if safety == 1
            %salvataggio file orario per elaborazioni

```



```

clear all;
close all;
clc;

disp('          CI SONO I COEFFICIENTI CALIBRAZIONE le soglie e anche media
mobile          ')

st=input('ID station (LOC=Master;S=Slave): ','s');
stt=input('Vigna (BR=Brolio;MA=Mortelle alta Cacciagrande;MB=Mortelle bassa
Cortigliano;DO=Donna Olimpia): ','s');
styear=input('Anno: ');
safety=input('Vuoi salvare il file output ? Si(1) No(2); ');
File_save='C:\TUSCANIA\DATI_2010\Scheduler\OUTPUT_ORARI_DESPIKING';
File_save_desp='C:\TUSCANIA\DATI_2010\Scheduler\OUTPUT_GIORNALIERI_DESPIKIN
G';
%File_path_master='C:\TUSCANIA\DATI_2009\Scheduler\Master';

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%CALIBRAZIONI
2010%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

if stt == 'BR'
    File_path='C:\TUSCANIA\DATI_2010\Scheduler\Brolio\OUTPUT';
    File_path_loc='C:\TUSCANIA\DATI_2010\Scheduler\Brolio\OUTPUT';
    slavenumb=[65 66 67 68 69 70 71 72 73 76 77 78];
        span_air=[1.0188          0.9942  1.0091  1.0004  1.0225  0.9317
1.011  1.0208  0.9662  1.0142  1.0082 1];
        offset_air=[-0.8384   -0.9244  0.1344  0.2262  0.6878  0.4141
0.6059  0.0152  0.5053  -0.7781  0.0379 0];
        span_grape=[1.0079          1.0204  1.003  0.9799  1.0057  1.0054
0.9953  0.9991  1  1.0202  1.0095 1];
        offset_grape=[-0.9532   -1.2992  0.0801  0.1633  0.748  -0.4873
0.7672  -0.8051  0  -1.3212  -0.5879 0];
        span_leaf=[1  1  1  1  1  1  1  1
1  1  1 1];
        offset_leaf=[0  0  0  0  0  0  0  0
0  0  0 0];
        offset_rad=[-13.1  5.2  8.1  -5.6  -13.4  9.3
-2.0  12.6  -6.6  -1.3  -5.6 0];

%
% span_grape2=[1  1  1  0.9479  0.9164
0.9586  0.923  0.5545  0.9891 1];
% offset_grape2=[0  0  0  0.6468  0.947
0.6883  0.2833  17.756  0.3975 0];
% span_grape3=[1  1  1  0.9419  0.9479
0.9718  0.9538  1.0063  0.9966  0.9664 1];
% offset_grape3=[0  0  0  1.2228  0.6468  0.785
0.0037  0.4635  -0.5021 0];

        doy_cal=6000;
        doy_rad=6000;
        doy_cal2=6000;
        doy_delta=6000;

    %slavenumb=char('A','B','C','D','E','F','G','H','I','L','M');
elseif stt == 'DO'
    File_path='C:\TUSCANIA\DATI_2010\Scheduler\Donna_Olimpia\OUTPUT';
    File_path_loc='C:\TUSCANIA\DATI_2010\Scheduler\Donna_Olimpia\OUTPUT';
    slavenumb=[65 66 67 68 69 70 71 72 73 76];
        span_air=[0.99  0.9585  0.9829  0.9743  1.0262  1.0091  1.0105
1.0116  1.0067  1.0122];
        offset_air=[0.1699          0.1919  1.1405  1.677  0.3749  -0.9814
-0.3057  -1.1616  -0.1615  -0.6744];
        span_grape=[0.9987          0.9685  1.0146  0.9892  1  1.1609
1.0123  0.979  1.0087  0.9654];
        offset_grape=[0.1038   -0.1121  0.8983  0.4551  0  -2.9836
-0.1987  -0.5475  -0.0768  1.4403];
        span_leaf=[1  1  1  1  1  1  1  1
1  1];
        offset_leaf=[0  0  0  0  0  0  0  0
0  0];
        offset_rad=[19.1  14.7  14.7  14.7  10.2  -4.3
-17.9  -5.2  -2.8  -13.7];

```

```

%      span_grape2=[1      0.9894  0.9373  1  0.8827      1  0.9935
0.9619  1      1];
%      offset_grape2=[0      0.3922  0.6034  0  2.8871      0  -0.2301
0.6141  0      0];
%      span_grape3=[1  0.9971      0.9595  1      0.891  1  1.0009
0.9414  1      1];
%      offset_grape3=[0  -0.5111  0.5734  0      1.4996  0
-0.1732  1.3277  0  0];

      doy_cal=6000;
      doy_rad=6000;
      doy_cal2=6000;
      doy_delta=6000;
      %slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MB'
File_path='C:\TUSCANIA\DATI_2010\Scheduler\Mortelle\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2010\Scheduler\Mortelle_Bassa\OUTPUT';
slavenumb=[65 66 67 68 69 70 71 72 73 76];
      span_air=[1.0044      1.0043  1.002  1.1139  0.9987  1.0048
0.9812  0.9993  0.952  1.0006];
      offset_air=[-0.5906  0.4623  -0.4945 -10.123 -0.0766 -0.147
-25.632 -0.6601  4.665  0.1833];
      span_grape=[1.0169  0.9938  0.9976  0.9957  1.0861  1.0075
1.0095  0.9941  1.0485  0.9953];
      offset_grape=[-0.7412  0.4298  -0.3232 -0.3357 -13.036 -0.93
-0.0173 -1.0343 -10.462 -0.1521];
      span_leaf=[1  1      1  1      1  1      1  1
1      1];
      offset_leaf=[0  0      0  0      0  0      0  0
0      0];
      offset_rad=[-20.0  7.3  -18.5  9.6  11.5  -9.7  0.0
19.7  2.0  3.0];

%      span_grape2=[1      1.0767  1.0134  1      1      0.7785
0.8568  1      1];
%      offset_grape2=[0      0.0343  -1.1855  0      0      3.0006
6.2336  0      0];
%      span_grape3=[1      0.9537  1.4806  1      1      0.7903
0.9374  1      1];
%      offset_grape3=[0      0.9909  -6.472  0      0      2.3206
1.5795  0      0];
      doy_cal=6000;
      doy_rad=6000;
      doy_cal2=6000;
      doy_delta=6000;
      %slavenumb=char('A','B','C','D','E','F','G','H','I','L');
elseif stt == 'MA'
File_path='C:\TUSCANIA\DATI_2010\Scheduler\Mortelle\OUTPUT';
File_path_loc='C:\TUSCANIA\DATI_2010\Scheduler\Mortelle_Alta\OUTPUT';
slavenumb=[77 78 79 80 81 82 83 84 85 86];
      span_air=[0.9942      1.0048  0.9954  1      0.9949  0.9974
0.9873  1.0001  1.0007  1.0029];
      offset_air=[0.4249  0.1023  -0.3013  0      1.0266  0.3551
-0.1147  0.2877  -0.0536 -0.3098];
      span_grape=[1.0018  1.0104  1.0101  0.9996  0.9989  1.0021
0.9878  1.0104  0.992  1];
      offset_grape=[0.3671  0.0218  -0.3623 -0.125  0.894  -0.2703
-0.3561 -2.1315 -0.4066 0];
      span_leaf=[1  1      1  1      1  1      1  1
1      1];
      offset_leaf=[0  0      0  0      0  0      0  0
0      0];
      offset_rad=[-13.7  12.9  7.7  16.6  -16.3  9.6  0.0
-10.4  9.6  -17.4];

%      span_grape2=[1  1  0.8662  1      1      1.1  0.9969  1
1  0.9919];
%      offset_grape2=[0  0  2.6229  0      0      -0.7983
-0.4557  0  0  -0.1082];
%      span_grape3=[1  1  0.9581  1      1      0.9921  1.0114  1
1  1.0238];
%      offset_grape3=[0  0  0.8046  0      0      -0.0535

```

```

-0.105  0      0      -0.5867];

    doy_cal=6000;
    doy_rad=6000;
    doy_cal2=6000;
    doy_delta=6000;
    %slavenumb=char('M','N','O','P','Q','R','S','T','U','V');
end

%%%%%
if st=='LOC'
    File_wrk=[File_path_loc '\ ' st '.dat'];
    N=dload(File_wrk);
    % MASTER
% year   doyM      Tempmedia      Tmax      Tmin      Tdevstd Tultimo Umidmedia
URmax   URmin     URdevstd      URultimo  Presmedia
% 1      2        3          4          5          6          7          8          9          10
11      12        13
% Pmax   Pmin     Pdevstd      Pultimo  RadGmedia      Rmax      Rmin      Rdevstd
Rultimo Piogsomma      VVentlVmed      Vdevstd Vraf
% 14     15        16          17          18          19          20          21          22          23
24      25        26
% Vmin   Vultimo  DVentlDprev      Draf      Dultimo numcamp Bateria
% 27     28        29          30          31          32          33
    %step 1 ordina ed elimina doppie
    [Y,I]=sort(N.doyM(:,1));
    datatemp=[N.year(I,1) N.doyM(I,1) N.Tempmedia(I,1) N.Tmax(I,1)
    N.Tmin(I,1) N.Tdevstd(I,1) N.Tultimo(I,1) N.Umidmedia(I,1)
    N.URmax(I,1) N.URmin(I,1) N.URdevstd(I,1) N.URultimo(I,1)
    N.Presmedia(I,1) N.Pmax(I,1) N.Pmin(I,1) N.Pdevstd(I,1)
    N.Pultimo(I,1) N.RadGmedia(I,1) N.Rmax(I,1) N.Rmin(I,1)
    N.Rdevstd(I,1) N.Rultimo(I,1) N.Piogsomma(I,1) N.VVentlVmed(I,1)
    N.Vdevstd(I,1) N.Vraf(I,1) N.Vmin(I,1) N.Vultimo(I,1)
    N.DVentlDprev(I,1) N.Draf(I,1) N.Dultimo(I,1) N.numcamp(I,1)
    N.Bateria(I,1)];
    [I,J]=unique(datatemp(:,2));
    data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
    datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
    datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
    datatemp(J,13) datatemp(J,14) datatemp(J,15) datatemp(J,16)
    datatemp(J,17) datatemp(J,18) datatemp(J,19) datatemp(J,20)
    datatemp(J,21) datatemp(J,22) datatemp(J,23) datatemp(J,24)
    datatemp(J,25) datatemp(J,26) datatemp(J,27) datatemp(J,28)
    datatemp(J,29) datatemp(J,30) datatemp(J,31) datatemp(J,32)
    datatemp(J,33)];
    F = data(:, :);
    F(F== -999) = NaN;
    data(:, :)=F;

% elaborazioni giornaliere
appog=data(1,2);
app2=num2str(data(1,2));
data(1,2)=1111111;
prova=num2str(data(:,2));
D=str2num(prova(:,1:3));
D(1)=appog;
doyi=D(2);
doyf=D(end);
%elaborazioni orarie
prova2=num2str(data(:,2));%
D2=str2num(prova2(:,4:7));%
%D2(1)=str2num(app2(:,4:7));
D2=D2/100;%
D3=fix((D2+0.1)*24/100) ; %
D3(1)=D3(2);
ii=1;
i2=1;%
data(1,2)=appog;

for i=doyi:doyf
    ijj = find(D==i);

```



```

if size(ijj,1)>1
    daily(ii,:) = nanmean(data(ijj,:)); % per le medie giornaliere
    daily(ii,2)=i;
    daily(ii,23)= nansum(data(ijj,23)); % cumulate giornaliere
    precipitazione mm
    daily(ii,4)= nanmax(data(ijj,4));
    daily(ii,5)= nanmin(data(ijj,5));
    daily(ii,9)= nanmax(data(ijj,9));
    daily(ii,10)= nanmin(data(ijj,10));
    daily(ii,19)= nanmax(data(ijj,19));
    daily(ii,20)= nanmin(data(ijj,20));

    if styear>2007
        daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
        sfilato km giorno rilev 5 min
        daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
        radiazione Mj /m2
    else
        if (stt=='DO' & i>=222 | stt=='BR' & i>=194 | stt=='MB' &
            i>=261 | stt=='MA' & i>=261)
            daily(ii,24)= nansum(data(ijj,24)*300/1000); % vento
            sfilato km giorno rilev 5 min
            daily(ii,18)= nansum(data(ijj,18)*300/1000000); % cumulate
            radiazione Mj /m2
        else
            daily(ii,24)= nansum(data(ijj,24)*900/1000); % vento
            sfilato km giorno rilev 15 min
            daily(ii,18)= nansum(data(ijj,18)*900/1000000); % cumulate
            radiazione Mj /m2
        end
    end
end
ii=ii+1;
else
    daily(ii,:) = NaN;
    daily(ii,2)=i;
    ii=ii+1;
end

%elaborazione oraria
if size(ijj,1)>1
    zz1=ijj(1);
    for iii=D3(ijj(1)):D3(ijj(end))
        ij2=find(D3(ijj)==iii);
        if size(ij2,1)>1
            zz=zz1+length(ij2);
            azz=data(zz1:(zz-1),:);
            zz1=zz;
            hourly(i2,:)=nanmean(azz(:,,:));
            hourly(i2,2)=i+double(iii*100/24)/100;
            hhour(i2,1)=i;
            hhour(i2,2)=iii;
            hourly(i2,23)= nansum(azz(:,23)); %
            cumulate giornaliere precipitazione mm
            hourly(i2,4)= nanmax(azz(:,4));
            hourly(i2,5)= nanmin(azz(:,5));
            hourly(i2,9)= nanmax(azz(:,9));
            hourly(i2,10)= nanmin(azz(:,10));
            hourly(i2,19)= nanmax(azz(:,19));
            hourly(i2,20)= nanmin(azz(:,20));

            if iii==24
                hourly(i2,2)=i;
            end
            i2=i2+1;
        else
            hourly(i2,:) = NaN;
            hourly(i2,2)=i+double(iii*100/24)/100;
            i2=i2+1;
        end
    end
end
hourly(i2,:) = NaN;
hourly(i2,2)=i+double(iii*100/24)/100;

```



```

    daily(xx,11), ' ', daily(xx,12), ' ', daily(xx,13), ' ',
    daily(xx,14), ' ', daily(xx,15), ' ', ...
    daily(xx,16), ' ', daily(xx,17), ' ', daily(xx,18), ' ',
    daily(xx,19), ' ', daily(xx,20), ' ', ...
    daily(xx,21), ' ', daily(xx,22), ' ', daily(xx,23), ' ',
    daily(xx,24), ' ', daily(xx,25), ' ', ...
    daily(xx,26), ' ', daily(xx,27), ' ', daily(xx,28), ' ',
    daily(xx,29), ' ', daily(xx,30), ' ', ...
    daily(xx,31), ' ', daily(xx,32), ' ', daily(xx,33));
end
fclose(fid2);

end

else
s1=1;%inizializzo indici vettori calibrazione
f1=1;
for hh=slavenumb(1):slavenumb(end)
    if (hh==74 | hh==75)
        pippo='N';
    else
        File_wrk=[File_path '\ ' hh '.dat'];
        M=dload(File_wrk);
%       File_master=[File_path_master '\M_' stt '.mat'];
%       load(File_master);
%       eval(['Master_hy=M_' stt '_HY;'])
%       eval(['Master_dy=M_' stt '_DY;'])

% SLAVE
% year DOY_hh rad Tair Tgrap Tleaf Tsoil1 Tsoil2 T_heatsoil1
T_heatsoil2 VV Bat Bagn_leaf
% 1 2 3 4 5 6 7 8 9
10 11 12 13
        %step 1 ordina ed elimina doppie
        [Y,I]=sort(M.DOY_hh(:,1));
        datatemp=[M.Year(I,1) M.DOY_hh(I,1) M.Tcjc(I,1) M.Tair(I,1)
        M.Tgrap(I,1) M.Tleaf(I,1) M.Tsoil1(I,1) M.Tsoil2(I,1)
        M.T_heatsoil1(I,1) M.T_heatsoil2(I,1) M.VV(I,1) M.Bat(I,1)
        M.Bagn_leaf(I,1)];
        [I,J]=unique(datatemp(:,2));
        data=[datatemp(J,1) datatemp(J,2) datatemp(J,3) datatemp(J,4)
        datatemp(J,5) datatemp(J,6) datatemp(J,7) datatemp(J,8)
        datatemp(J,9) datatemp(J,10) datatemp(J,11) datatemp(J,12)
        datatemp(J,13)];
        F = data(:, :);
        F(F== -999) = NaN;
        data(:, :)=F;

        wind=data(:,11);
        wind(wind==0.2)=0;
        data(:,11)=wind;

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%%%
        %applico i coefficienti di calibrazione per le slave
        data(data(:,2)>=doy_cal,4)=(data(data(:,2)>=doy_cal,4)
        *span_air(s1))+offset_air(f1);
        data(data(:,2)>=doy_cal,5)=(data(data(:,2)>=doy_cal,5)
        *span_grape(s1))+offset_grape(f1);
        data(data(:,2)>=doy_cal,6)=(data(data(:,2)>=doy_cal,6)
        *span_leaf(s1))+offset_leaf(f1);
%       data(data(:,2)>=doy_cal,7)=(data(data(:,2)>=doy_cal,7)
        *span_grape2(s1))+offset_grape2(f1);
%       data(data(:,2)>=doy_cal,8)=(data(data(:,2)>=doy_cal,8)
        *span_grape3(s1))+offset_grape3(f1);
%
%
        %sensori radiazione grappolo
        data(data(:,2)<doy_rad,3)=NaN;
        %coefficienti calibrazione radiazione grappolo

```



```

if hh==slavenu(b(5) | hh==slavenu(b(7) | hh==slavenu(b(9) |
hh==slavenu(b(11)%tolgo wind
data(data(:,2)>100,11)=NaN;
end

if hh==slavenu(1)%tair A tolgo gobba mattutina
orenull=num2str(data(:,2));%
nuoveore=str2num(orenull(:,4:5));%
nnore=find(nuoveore<50 & nuoveore>19);
data(nnore,4) = NaN;
end
if hh==slavenu(3)%tair C
data((data(:,2)>=25700 & data(:,2)<=25799),4)=NaN;
end
if hh==slavenu(3)%tair C
data((data(:,2)>=27800 & data(:,2)<=27899),4)=NaN;
end
if hh==slavenu(1)%tgrape A
data((data(:,2)>=22500 & data(:,2)<=22799),5)=NaN;
end
if hh==slavenu(1)%tgrape A
data((data(:,2)>=25700 & data(:,2)<=25799),5)=NaN;
end
if hh==slavenu(1)%tgrape A
data((data(:,2)>=26200 & data(:,2)<=26299),5)=NaN;
end
if hh==slavenu(4)%tgrape D
data((data(:,2)>=26200 & data(:,2)<=26299),5)=NaN;
end
if hh==slavenu(4)%tgrape D
data((data(:,2)>=27100 & data(:,2)<=27199),5)=NaN;
end
if hh==slavenu(4)%tgrape D
data((data(:,2)>=19300 & data(:,2)<=19599),5)=NaN;
end
if hh==slavenu(3)%tgrape C tolgo gobba mattutina
orenull=num2str(data(:,2));%
nuoveore=str2num(orenull(:,4:5));%
nnore=find(nuoveore<40 & nuoveore>19);
data(nnore,5) = NaN;
end
if hh==slavenu(6)%tgrape F tolgo gobba mattutina
orenull=num2str(data(:,2));%
nuoveore=str2num(orenull(:,4:5));%
nnore=find(nuoveore<40 & nuoveore>19);
data(nnore,5) = NaN;
end

elseif stt == 'DO'
if hh==slavenu(2) | hh==slavenu(3) | hh==slavenu(5) |
hh==slavenu(7) | hh==slavenu(8) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenu(1) | hh==slavenu(3) | hh==slavenu(6) |
hh==slavenu(8)%tolgo wind
data(data(:,2)>100,11)=NaN;
end

if hh==slavenu(5)%tgrape E
data((data(:,2)>=22300),5)=NaN;
end
if hh==slavenu(10)%tgrape L
data((data(:,2)<=15800),5)=NaN;
end
if hh==slavenu(10)%tgrape L troppo alta
data((data(:,2)>=15800),5)=data((data(:,2)>=15800),5)-83;
end
if hh==slavenu(1)%tgrape A
data((data(:,2)>=26600),5)=NaN;
end

if hh==slavenu(3)%tgrape C
data((data(:,2)>=23000 & data(:,2)<=23099),5)=NaN;

```

```

end
if hh==slavenumb(6)%tgrape F
data((data(:,2))>=23000 & data(:,2)<=23099),5)=NaN;
end
if hh==slavenumb(7)%tgrape G
data((data(:,2))>=17300 & data(:,2)<=17399),5)=NaN;
end

elseif stt == 'MB'
if hh==slavenumb(2) | hh==slavenumb(3) | hh==slavenumb(4) |
hh==slavenumb(6) | hh==slavenumb(7) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenumb(2) | hh==slavenumb(4) | hh==slavenumb(6) |
hh==slavenumb(8)%tolgo wind
data(data(:,2)>100,11)=NaN;
end
if hh==slavenumb(3)%tgrape C
data((data(:,2))>=100),5)=NaN;
end
if hh==slavenumb(5)%tgrape E
data((data(:,2))>=100),5)=NaN;
end
if hh==slavenumb(6)%tgrape F
data((data(:,2))>=100),5)=NaN;
end
if hh==slavenumb(9)%tgrape I
data((data(:,2))>=100),5)=NaN;
end

if hh==slavenumb(1)%tgrape A tolgo le ore notturne
orenull=num2str(data(:,2));%
nuoveore=str2num(orenull(:,4:5));%
nnore=find(nuoveore>72 & nuoveore<49);
data(nnore,5) = NaN;
end
if hh==slavenumb(4)%tgrape D
data((data(:,2))>=16600 & data(:,2)<=16799),5)=NaN;
data((data(:,2))>=18700 & data(:,2)<=18899),5)=NaN;
data((data(:,2))>=19500 & data(:,2)<=19699),5)=NaN;
data((data(:,2))>=20300 & data(:,2)<=20699),5)=NaN;
data((data(:,2))>=22500 & data(:,2)<=23399),5)=NaN;
data((data(:,2))>=23500 & data(:,2)<=23799),5)=NaN;
data((data(:,2))>=24000),5)=NaN;
end
if hh==slavenumb(8)%tgrape H
data((data(:,2))>=22500 & data(:,2)<=24099),5)=NaN;
end

if hh==slavenumb(1)%tair A
data((data(:,2))>=26900),4)=NaN;
end
if hh==slavenumb(4)%tair D
data((data(:,2))>=100),4)=NaN;
end
if hh==slavenumb(7)%tair G
data((data(:,2))>=100),4)=NaN;
end

%
elseif stt == 'MA'
if hh==slavenumb(3) | hh==slavenumb(5) | hh==slavenumb(6) |
hh==slavenumb(7) | hh==slavenumb(10) %tolgo pot idr
data(data(:,2)>100,7:10)=NaN;
end
if hh==slavenumb(5) | hh==slavenumb(6) | hh==
slavenumb(7)%tolgo wind
data(data(:,2)>100,11)=NaN;
end
if hh==slavenumb(9)%tgrape U
data((data(:,2))>=20000),5)=NaN;
end
if hh==slavenumb(4)%tgrape P

```

```

    data((data(:,2)>=26500),5)=NaN;
end
if hh==slavenumb(8)%tgrape T tolgo
data((data(:,2)>=100),5)=NaN;
end
if hh==slavenumb(6)%tgrape R
data((data(:,2)>=26500),5)=NaN;
end
if hh==slavenumb(6)%tgrape R
data((data(:,2)>=21700),5)=data((data(:,2)>=21700),5)-5;
end
if hh==slavenumb(8)%rad T tolgo
data((data(:,2)>=100),3)=NaN;
end

end

disp('          DESPIKE MANUALE INSERITO          ')

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

                                %inserisci giorni mancanti
hourly=data;
startday=fix(hourly(1,2)/100);
endday=fix(hourly(end,2)/100);
ndays=endday-startday+1;
ncol=13;
A=NaN*ones(ndays*24,ncol);
A(:,1)=styear;
for i=0:ndays-1
    for j=0:23
        A(i*24+j+1,2)=
            str2num([num2str(startday+i,'%3d')
                    num2str(j,'%2d')]);
    end
end

newdoy1=hourly(:,2)/100-fix(hourly(:,2)
/100);
newdoy2=int16(newdoy1*24);
appdoy=hourly(:,2);
hourly(:,2)=str2num([num2str(fix(hourly(:,2)
/100),'%3d') num2str(newdoy2,'%2d')]);
B=hourly(:,2);

for i=1:size(A,1)
    index=find(B(:,2)==A(i,2));
    if length(index)>1
        A(i,:)=B(index(1),:);
    end
end

hourly=A;

appog=hourly(1,2);
hourly(1,2)=11111;

doyi=int16(hourly(2,2)/100);
doyf=int16(hourly(end,2)/100);
prova=num2str(hourly(:,2));
D=str2num(prova(:,1:3));
ii=1;
hourly(1,2)=appog;
D(1)=D(2);
for i=doyi:doyf
    ijj = find(D==i);
    if size(ijj,1)>1
        % controllo giorni parziali
        for iii=1:13
            ina=isnan(hourly(ijj,iii));
            if sum(ina) >= 7
                hourly(ijj,iii) = NaN;
            end
        end
    end
end

```


Allegati 1.7 - 1.8 - 1.9 - Listati delle
procedure matlab di gapfilling dati slave
(2008-2009-2010)

```

clear all;
close all;
clc;
File wrk=['C:\TUSCANIA\DATI\Matrix\Tuscania_dataset_HY_318_2008.dat'];
N=dload(File_wrk);

gi = input('Giorno iniziale (doy es 9200): ');
gf = input('Giorno finale: ');
v=input('Vigneto Brolio(1)-Donna Olimpia(2)-MortelleA Cacciagrande(3)-
MortelleB Cortigliano(4): ');

%fattori di conversione

%potenziale idrico Brolio
Tasterisco30=(3.15-N.T_heatsoil1)/(3.15-0.75);
N.T_heatsoil1=-0.056*(Tasterisco30.^(-1/0.45)-1);
Tasterisco60=(3.15-N.T_heatsoil2)/(3.15-0.75);
N.T_heatsoil2=-0.056*(Tasterisco60.^(-1/0.45)-1);

%radiazione da integrale giornaliera w m-2 a MJ m-2
%N.Rad=N.Rad*3600/1000000;

%ANEMOMETRI SLAVE
N.VV=24*N.VV;

if v==1
sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss ' (mast), 'c+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'b-o ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'k+- ', 'LineWidth', 2)'])
eval(['legend(''Master'', 'A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'L', 'M', -1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);

```

```

h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

```

```
sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
```

```
eval(['figure(''Name'', sname3 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Piogsomma';
```

```
s='N.Bagn_leaf';
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
```

```
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
```

```
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
```

```
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
```

```
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
```

```
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
```

```
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
```

```
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(e),' s '(e),'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(f),' s '(f),'y+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
```

```
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])
```

```
sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
```

```
eval(['figure(''Name'', sname4 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Piogsomma';
```

```
s='N.T_heatsoill';
```

```
sss=N.Tempmedia/100;
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
```

```
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(mast), sss(mast),'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
```

```

eval(['plot(N.JDTpc(m),' s '(m),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''B'',''C'',''D'',''M'','-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'',' sname5 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T heatsoil2';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast),'r+-','LineWidth',2)'])

eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''B'',''C'',''D'',''M'','-1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'',' sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'r-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''F'',''H'',''L'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'',' sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])

```

```

eval(['plot(N.JDTpc(e), ' s '(e), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'k+-', 'LineWidth',2)'])
eval(['legend(''Master'', 'A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'L', 'M', -1)'])

```

```

sname8=['Tair Master vs Tsoil 30 cm Slave'];
eval(['figure(''Name'', sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', 'A', 'B', 'C', 'D', 'M', -1)'])

```

```

sname9=['Tair Master vs Tsoil 60 cm Slave'];
eval(['figure(''Name'', sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', 'A', 'B', 'C', 'D', 'M', -1)'])

```

```

sname10=['umidità aria Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'', sname10 )'])
grid on
hold on
ss='N.Umidmedia';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b-', 'LineWidth',2)'])

```

```

eval(['plot(N.JDTpc(c),' s '(c),' 'b-o'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'g+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'r+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'y+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'r-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'g-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'k-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'y-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),' 'k+-'',' 'LineWidth',2)'])
eval(['legend(''Master'',' 'A'',' 'B'',' 'C'',' 'D'',' 'E'',' 'F'',' 'G'',' 'H'',' 'I'',' 'L'',' 'M'','-1)'])
figure
plotyy(N.JDTpc(mast),N.Umidmedia(mast),N.JDTpc(m),N.Bagn_leaf(m))

elseif v==2 %donna olimpia

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'',' sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'r+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'm+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'c+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'b-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'k-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'm-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'g-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'r-'',' 'LineWidth',2)'])

eval(['legend(''Master'',' 'A'',' 'B'',' 'C'',' 'D'',' 'E'',' 'F'',' 'G'',' 'H'',' 'I'',' 'L'','-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'',' sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'r+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'm+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'c+-'',' 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-'',' 'LineWidth',2)'])

```



```

eval(['plot(N.JDTpc(f),' s '(f),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'r-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'','' sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'r-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'','' sname4 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T heatsoil1';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(mast),' sss(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''D'',''F'',''I'',''L'','-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'','' sname5 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T heatsoil2';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);

```

```

d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''D'',''F'',''I'',''L'','-1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'m+-','LineWidth',2)'])

eval(['legend(''Master'',''B'',''D'',''E'',''G'',''I'',''L'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'', sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'r-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname8=['Tair Master vs Tsoil 30 cm Slave'];
eval(['figure(''Name'', sname8 )'])
grid on
hold on

```

```

ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-.', 'LineWidth',2)'])
eval(['legend(''Master'', ''A'', ''D'', ''F'', ''I'', ''L'',-1)'])

sname9=['Tair Master vs Tsoil 60 cm Slave'];
eval(['figure(''Name'', sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-.', 'LineWidth',2)'])
eval(['legend(''Master'', ''A'', ''D'', ''F'', ''I'', ''L'',-1)'])

sname10=['tsoil30 vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'', sname10 )'])
grid on
hold on
ss=N.Tsoil1/100;
s='N.T_heatsoil1';
%sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(a), ss(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ss(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ss(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ss(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ss(l), 'y-.', 'LineWidth',2)'])

eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-.', 'LineWidth',2)'])
eval(['legend(''At'', ''Dt'', ''Ft'', ''It'', ''Lt'', ''Ap'', ''Dp'', ''Fp'', ''Ip'', ''Lp'',-1)'])
elseif v==3 %cacciagrande

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);

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```

n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'm+-','LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'm-','LineWidth',2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'r-','LineWidth',2)'])
eval(['legend('Master','M','N','O','P','Q','R','S','T','U','V',-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure('Name', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'm+-','LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'm-','LineWidth',2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'r-','LineWidth',2)'])
eval(['legend('Master','M','N','O','P','Q','R','S','T','U','V',-1)'])

sname3=['precipitazioni Master vs bagnatura foglie Slave'];
eval(['figure('Name', sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-','LineWidth',2)'])

```

```

eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(o),' s '(o),'m+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(q),' s '(q),'y+','LineWidth',2)'])
eval(['plot(N.JDTpc(r),' s '(r),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1),' s '(s1),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1),' s '(v1),'r-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''O'',''P'',''Q'',''R'',''S'',''T'',''U'',''V'','-1)'])

```

```

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'','' sname4 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''P'',''T'',''U'','-1)'])

```

```

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'','' sname5 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''P'',''T'',''U'','-1)'])

```

```

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'','' sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(o),' s '(o),'m+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])

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eval(['plot(N.JDTpc(v1),' s '(v1),'r-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''O'',''P'',''T'',''U'',''V'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'',' sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(o),' s '(o),'m+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(q),' s '(q),'y+','LineWidth',2)'])
eval(['plot(N.JDTpc(r),' s '(r),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1),' s '(s1),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1),' s '(v1),'r-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''O'',''P'',''Q'',''R'',''S'',''T'',''U'',''V'','-1)'])

sname8=['Tair Master vs Tsoil 30 cm Slave'];
eval(['figure(''Name'',' sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''P'',''T'',''U'','-1)'])

sname9=['Tair Master vs Tsoil 60 cm Slave'];
eval(['figure(''Name'',' sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),'g-','LineWidth',2)'])

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eval(['legend(''Master'',''M'',''N'',''P'',''T'',''U'','-1)'])

elseif v==4 %cortigliano

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'','' sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'r+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'm+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'c+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'b-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'k-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'm-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'g-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'r-.'',''LineWidth'','2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'','' sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'r+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'm+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'c+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'b-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'k-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'm-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'g-.'',''LineWidth'','2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'r-.'',''LineWidth'','2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'','' sname3 )'])

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```

grid on
hold on
ss='N.Piogssomma';
s='N.Bagn leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'm+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'c+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'y+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'b- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'k- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'm- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'g- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'r- ', 'LineWidth', 2)'])

eval(['legend(''Master'', 'A'', 'B'', 'C'', 'D'', 'E'', 'F'', 'G'', 'H'', 'I'', 'L'', -1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'', sname4 )'])
grid on
hold on
ss='N.Piogssomma';
s='N.T_heatsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'g+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'm+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y- ', 'LineWidth', 2)'])
eval(['legend(''Master'', 'A'', 'E'', 'H'', 'I'', 'L'', -1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'', sname5 )'])
grid on
hold on
ss='N.Piogssomma';
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'g+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'm+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+- ', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y- ', 'LineWidth', 2)'])
eval(['legend(''Master'', 'A'', 'E'', 'H'', 'I'', 'L'', -1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on

```



```

hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==3 & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),' r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' r-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' c+-','LineWidth',2)'])

eval(['legend('Master','A','C','E','G','I','L',-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure('Name', sname7)'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),' r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' k-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' g-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' r-','LineWidth',2)'])

eval(['legend('Master','A','B','C','D','E','F','G','H','I','L',-1)'])

sname8=['Tair Master vs Tsoil 30 cm Slave'];
eval(['figure('Name', sname8)'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),' r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' y-','LineWidth',2)'])
eval(['legend('Master','A','E','H','I','L',-1)'])

```

```

sname9=['Tair Master vs Tsoil 60 cm Slave'];
eval(['figure(''Name'', sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', 'A', 'E', 'H', 'I', 'L',-1)'])

% sname2=['Tesi 5 nei vari blocchi di Brolio (2 parametri) : ' ss sss];
% eval(['figure(''Name'', sname2 )'])
% grid on
% hold on
% mast = find(N.Doy<=gf & N.Doy>=gi & N.Vigna==1 & N.vigoria==0);
% uno = find(N.Doy<=gf & N.Doy>=gi & N.Vigna==1 & N.tesi==5 & N.vigoria==
1);
% quat=find(N.Doy<=gf & N.Doy>=gi & N.Vigna==1 & N.tesi==5 & N.vigoria==4);
% cinq=find(N.Doy<=gf & N.Doy>=gi & N.Vigna==1 & N.tesi==5 & N.vigoria==5);
% sei=find(N.Doy<=gf & N.Doy>=gi & N.Vigna==1 & N.tesi==5 & N.vigoria==6);
% eval(['plot(N.Doy(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(uno), ' ss '(uno), 'b+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(quat), ' ss '(quat), 'g+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(cinq), ' ss '(cinq), 'm+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(sei), ' ss '(sei), 'g-', 'LineWidth',2)'])
% eval(['plot(N.Doy(mast), ' sss '(mast), 'r+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(uno), ' sss '(uno), 'b+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(quat), ' sss '(quat), 'g+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(cinq), ' sss '(cinq), 'm+-', 'LineWidth',2)'])
% eval(['plot(N.Doy(sei), ' sss '(sei), 'g-', 'LineWidth',2)'])
% eval(['legend(''Master'', 'Alta', 'Medial', 'Media2', 'Bassa',-1)'])
%

end

```

```

loadde=input('carico il file dat ?? si(1) no(2) ');
if loadde==1
clear all;
close all;
clc;
choice=input('NO despiking 2008(1)-CALIBRATI(2)-DESPIKING 2009(3)-DESPIKING
2008(4): ');
if choice==1
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2008\5_DATASET_2008_FINALE\Tuscania_dataset_HY_2008_CALIBRATI.dat'];
elseif choice==2
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2009\Tuscania_dataset_HY_2009_pre_despiking.dat'];
elseif choice==3
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2009\Tuscania_dataset_HY_2009_despiking.dat'];
elseif choice==4
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2008\5_DATASET_2008_FINALE\Tuscania_dataset_HY_2008_index.dat'];
end
N=dload(File_wrk);
end
gi = input('Giorno iniziale (doy es 18200): ');
gf = input('Giorno finale (doy es 25800): ');
styear=input('Anno: ');
v=input('Vigneto Brolio(1)-Donna Olimpia(2)-MortelleA Cacciagrande(3)-
MortelleB Cortigliano(4): ');
delta=((gf-gi)/100)+1;
%fattori di conversione

%potenziale idrico
Tasterisco30=(3.15-N.T_heatsoil1)/(3.15-0.75);
N.T_heatsoil1=-0.056*(Tasterisco30.^(-1/0.45)-1);
Tasterisco60=(3.15-N.T_heatsoil2)/(3.15-0.75);
N.T_heatsoil2=-0.056*(Tasterisco60.^(-1/0.45)-1);

%radiazione da integrale giornaliera w m-2 a MJ m-2
%N.Rad=N.Rad*3600/1000000;

%ANEMOMETRI SLAVE
N.VV=24*N.VV;

if v==1
sname1='Tair Master vs Tair Slave';
eval(['figure(''Name'', sname1)'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'k+-','LineWidth',2)'])

```

```

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''M'','-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'','' sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),' c+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(b),' s '(b),' b-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(c),' s '(c),' b-o',''LineWidth'','2)'])
eval(['plot(N.JDTpc(d),' s '(d),' g+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(e),' s '(e),' r+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(f),' s '(f),' y+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(g),' s '(g),' r-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(h),' s '(h),' g-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(i),' s '(i),' k-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(l),' s '(l),' y-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(m),' s '(m),' k+-',''LineWidth'','2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''M'','-1)'])

```

```

sname22=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'','' sname22 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),' c+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(b),' s '(b),' b-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(c),' s '(c),' b-o',''LineWidth'','2)'])
eval(['plot(N.JDTpc(d),' s '(d),' g+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(e),' s '(e),' r+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(f),' s '(f),' y+-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(g),' s '(g),' r-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(h),' s '(h),' g-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(i),' s '(i),' k-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(l),' s '(l),' y-',''LineWidth'','2)'])
eval(['plot(N.JDTpc(m),' s '(m),' k+-',''LineWidth'','2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''M'','-1)'])

```

```

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'','' sname3 )'])
grid on

```

```

hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'k+-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

```

```
sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
```

```
eval(['figure(''Name'', sname4 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Piogsomma';
```

```
s='N.T_heatsoil1';
```

```
sss=N.Tempmedia/100;
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
```

```
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(mast), sss(mast), 'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(c), ' s '(c), 'm+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(d), ' s '(d), 'c+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(m), ' s '(m), 'y-','LineWidth',2)'])
```

```
eval(['legend(''Master'',''airMaster'',''A'',''B'',''C'',''D'',''M'','-1)'])
```

```
sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
```

```
eval(['figure(''Name'', sname5 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Piogsomma';
```

```
s='N.T_heatsoil2';
```

```
sss=N.Tempmedia/100;
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
```

```
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(mast), sss(mast), 'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(c), ' s '(c), 'm+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(d), ' s '(d), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', ''airMaster'', ''A'', ''B'', ''C'', ''D'', ''M'', -1)'])
```

```
sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'm-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'r-', 'LineWidth',2)'])
eval(['legend(''Master'', ''A'', ''B'', ''C'', ''D'', ''F'', ''H'', ''L'', -1)'])
```

```
sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'', sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'b-o', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'k+-', 'LineWidth',2)'])
eval(['legend(''Master'', ''A'', ''B'', ''C'', ''D'', ''E'', ''F'', ''G'', ''H'', ''I'', ''L'', ''M'', -1)'])
```

```
sname8=['Tair Master vs Tgrape2 Slave'];
eval(['figure(''Name'', sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoill';
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
```

```

l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-','LineWidth',2)'])
eval(['legend(''Master'', ''E'', ''F'', ''G'', ''H'', ''I'', ''L'', -1)'])

sname9=['Tair Master vs Tgrape3 cm Slave'];
eval(['figure(''Name'', sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-','LineWidth',2)'])
eval(['legend(''Master'', ''E'', ''F'', ''G'', ''H'', ''I'', ''L'', -1)'])

%pivot solo HV

p ora=
vertcat(N.ora(d),N.ora(e),N.ora(f),N.ora(g),N.ora(h),N.ora(i),N.ora(l),N.ora(m));
p_def=
vertcat(N.defogliazione(d),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h),N.defogliazione(i),N.defogliazione(l),N.defogliazione(m));
p_gem=
vertcat(N.gemme(d),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h),N.gemme(i),N.gemme(l),N.gemme(m));
p_dir=
vertcat(N.diradamento(d),N.diradamento(e),N.diradamento(f),N.diradamento(g),N.diradamento(h),N.diradamento(i),N.diradamento(l),N.diradamento(m));
p_grap=
vertcat(N.Tgrap(d),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h),N.Tgrap(i),N.Tgrap(l),N.Tgrap(m));
p_air=
vertcat(N.Tair(d),N.Tair(e),N.Tair(f),N.Tair(g),N.Tair(h),N.Tair(i),N.Tair(l),N.Tair(m));
p_rad=
vertcat(N.Rad(d),N.Rad(e),N.Rad(f),N.Rad(g),N.Rad(h),N.Rad(i),N.Rad(l),N.Rad(m));
p_doy=
vertcat(N.Doy(d),N.Doy(e),N.Doy(f),N.Doy(g),N.Doy(h),N.Doy(i),N.Doy(l),N.Doy(m));

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def

```

```

p_gem p_dir],p_rad,'nansum');

%gap-filling e quality check per il 2009 BROLIO
if styear==2009
    %valori dal 182 della M di brolio mancanti
    d_values_air_avg(1:9,1)=d_values_air_avg(1:9,2);
    d_values_air_max(1:9,1)=d_values_air_max(1:9,2);
    d_values_air_min(1:9,1)=d_values_air_min(1:9,2);
    d_values_gra_avg(1:9,1)=d_values_gra_avg(1:9,2);
    d_values_gra_max(1:9,1)=d_values_gra_max(1:9,2);
    d_values_gra_min(1:9,1)=d_values_gra_min(1:9,2);
    d_values_rad_sum(1:9,1)=d_values_rad_sum(1:9,2);

    % ciclo per tutte le slave gap filling aria
    for ix=1:8
        a_g=d_values_air_avg(:,ix);
        xi = find(isnan(a_g));
        x = 1:length(a_g);
        D = setdiff(x, xi);
        yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
        a_g(xi) = yi;
        d_values_air_avg(:,ix)=a_g;
        a_g=d_values_air_max(:,ix);
        xi = find(isnan(a_g));
        x = 1:length(a_g);
        D = setdiff(x, xi);
        yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
        a_g(xi) = yi;
        d_values_air_max(:,ix)=a_g;
        a_g=d_values_air_min(:,ix);
        xi = find(isnan(a_g));
        x = 1:length(a_g);
        D = setdiff(x, xi);
        yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
        a_g(xi) = yi;
        d_values_air_min(:,ix)=a_g;
        % ciclo per tutte le slave gap filling GRAPE
        g_g=d_values_gra_avg(:,ix);
        xi = find(isnan(g_g));
        x = 1:length(g_g);
        D = setdiff(x, xi);
        yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
        g_g(xi) = yi;
        d_values_gra_avg(:,ix)=g_g;
        g_g=d_values_gra_max(:,ix);
        xi = find(isnan(g_g));
        x = 1:length(g_g);
        D = setdiff(x, xi);
        yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
        g_g(xi) = yi;
        d_values_gra_max(:,ix)=g_g;
        g_g=d_values_gra_min(:,ix);
        xi = find(isnan(g_g));
        x = 1:length(g_g);
        D = setdiff(x, xi);
        yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
        g_g(xi) = yi;
        d_values_gra_min(:,ix)=g_g;
        %gap filling radiazione
        r_g=d_values_rad_sum(:,ix);
        xi = find(isnan(r_g));
        x = 1:length(r_g);
        D = setdiff(x, xi);
        yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
        r_g(xi) = yi;
        d_values_rad_sum(:,ix)=r_g;
        %%%%%%%%%%%
    end
%gap-filling e quality check per il 2008 BROLIO
% if styear==2008
%     %gap filling aria
%     a_g=d_values_air_avg(:,6);
%     xi = find(isnan(a_g));

```



```

%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,6)=a_g;
%     a_g=d_values_air_max(:,6);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,6)=a_g;
%     a_g=d_values_air_min(:,6);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,6)=a_g;
% gap filling grape G
%     g_g=d_values_gra_avg(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,6)=g_g;
%     g_g=d_values_gra_max(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,6)=g_g;
%     g_g=d_values_gra_min(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,6)=g_g;
% gap filling grape H
%     g_g=d_values_gra_avg(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,4)=g_g;
%     g_g=d_values_gra_max(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,4)=g_g;
%     g_g=d_values_gra_min(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,4)=g_g;
% gap filling grape D
%     g_g=d_values_gra_avg(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,3)=g_g;
%     g_g=d_values_gra_max(:,3);
%     xi = find(isnan(g_g));

```

```

%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,3)=g_g;
%     g_g=d_values_gra_min(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,3)=g_g;
%     %gap filling radiazione G
%     r_g=d_values_rad_sum(:,6);
%     xi = find(isnan(r_g));
%     x = 1:length(r_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%     r_g(xi) = yi;
%     d_values_rad_sum(:,6)=r_g;
%     %%%%%%%%%%%

%     %tolgo la I e tutti i suoi parametri li imposto come la M che ha le
stesse due tipologie di def e gem
%     d_values_air_avg(:,2)=d_values_air_avg(:,1);
%     d_values_air_max(:,2)=d_values_air_max(:,1);
%     d_values_air_min(:,2)=d_values_air_min(:,1);
%     d_values_gra_avg(:,2)=d_values_gra_avg(:,1);
%     d_values_gra_max(:,2)=d_values_gra_max(:,1);
%     d_values_gra_min(:,2)=d_values_gra_min(:,1);
%     d_values_rad_sum(:,2)=d_values_rad_sum(:,1);
end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h_values_air_avg(:,2)=h_values_air_avg(:,1);
    h_values_gra_avg(:,2)=h_values_gra_avg(:,1);
    h_values_rad_avg(:,2)=h_values_rad_avg(:,1);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));

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```

t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per tutte le tesi (slave)
% air_avg=nanmean(d_values_air_avg);
% air_max=nanmean(d_values_air_max);
% air_min=nanmean(d_values_air_min);
% air_avg_sum=nansum(d_values_air_avg);
% air_max_sum=nansum(d_values_air_max);
% air_min_sum=nansum(d_values_air_min);
% escursion=air_max_sum-air_min_sum;

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))
/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));

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```

ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);

ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS_grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI_air
CoolNI_air=air_min_t;
% CoolNI_grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);
%
% dMorning_5_10_air=[2.1664      2.2391  2.2652  2.3698];
% dMorning_5_10_grape=[2.1393      2.3651  2.3789  2.1366];
% dTwilight_15_20_air=[-1.7419   -1.9075  -1.9198  -1.8619];
% dTwilight_15_20_grape=[-2.1404           -2.0402  -2.1484  -1.9785];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master

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```

p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d values master air avg,d row master air avg,d column master air avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d values master air max,d row master air max,d column master air max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d values master air min,d row master air min,d column master air min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d values master rad sum,d row rad sum,d column rad sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h values ms,h row ms,h column ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');
% Master IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master IG
% Master IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==2 %donna olimpia

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])

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```

grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname22=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'', sname22 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);

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```

c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-','LineWidth',2)'])

eval(['legend(''Master'',' 'A'',' 'B'',' 'C'',' 'D'',' 'E'',' 'F'',' 'G'',' 'H'',' 'I'',' 'L',-1)'])

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'',' sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-','LineWidth',2)'])

eval(['legend(''Master'',' 'A'',' 'B'',' 'C'',' 'D'',' 'E'',' 'F'',' 'G'',' 'H'',' 'I'',' 'L',-1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'',' sname4 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T heatsoil1';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(mast),' sss(mast),' 'r+-','LineWidth',2)'])

```

```

eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''D'',''F'',''I'',''L'','-1)'])

```

```

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'','' sname5 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoil2';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''D'',''F'',''I'',''L'','-1)'])

```

```

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'','' sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'m+-','LineWidth',2)'])

```

```

eval(['legend(''Master'',''B'',''D'',''E'',''G'',''I'',''L'','-1)'])

```

```

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'','' sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k+-','LineWidth',2)'])

```



```

eval(['plot(N.JDTpc(d),' s '(d),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname8=['Tair Master vs Tgrape2 cm Slave'];
eval(['figure(''Name'',' sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'r-','LineWidth',2)'])

eval(['legend(''Master'',''B'',''C'',''E'',''G'',''H'','-1)'])

sname9=['Tair Master vs Tgrape3 cm Slave'];
eval(['figure(''Name'',' sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'r-','LineWidth',2)'])
eval(['legend(''Master'',''B'',''C'',''E'',''G'',''H'','-1)'])

%pivot solo HV
p_ora=
vertcat(N.ora(a),N.ora(b),N.ora(c),N.ora(e),N.ora(f),N.ora(g),N.ora(h),N.ora(i));
p_def=
vertcat(N.defogliazione(a),N.defogliazione(b),N.defogliazione(c),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h),N.defogliazione(i));
p_gem=
vertcat(N.gemme(a),N.gemme(b),N.gemme(c),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h),N.gemme(i));
p_dir=
vertcat(N.diradamento(a),N.diradamento(b),N.diradamento(c),N.diradamento(e),N.diradamento(f),N.diradamento(g),N.diradamento(h),N.diradamento(i));
p_grap=
vertcat(N.Tgrap(a),N.Tgrap(b),N.Tgrap(c),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h),N.Tgrap(i));
p_air=
vertcat(N.Tair(a),N.Tair(b),N.Tair(c),N.Tair(e),N.Tair(f),N.Tair(g),N.Tair(

```

```

h),N.Tair(i));
p_rad=
vertcat(N.Rad(a),N.Rad(b),N.Rad(c),N.Rad(e),N.Rad(f),N.Rad(g),N.Rad(h),N.Ra
d(i));
p_doy=
vertcat(N.Doy(a),N.Doy(b),N.Doy(c),N.Doy(e),N.Doy(f),N.Doy(g),N.Doy(h),N.Do
y(i));
%%%%%%%%%

```

```

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def
p_gem p_dir],p_rad,'nansum');

```

```

%gap-filling e quality check per il 2009 DONNA OLIMPIA
if styear==2009

```

```

%   %tolgo la RAD della B e la imposto come la G
    d_values_rad_sum(:,4)=d_values_rad_sum(:,3);

% ciclo per tutte le slave gap filling aria
for ix=1:8
    a_g=d_values_air_avg(:,ix);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_avg(:,ix)=a_g;
    a_g=d_values_air_max(:,ix);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_max(:,ix)=a_g;
    a_g=d_values_air_min(:,ix);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_min(:,ix)=a_g;
% ciclo per tutte le slave gap filling GRAPE
    g_g=d_values_gra_avg(:,ix);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_avg(:,ix)=g_g;
    g_g=d_values_gra_max(:,ix);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_max(:,ix)=g_g;
    g_g=d_values_gra_min(:,ix);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);

```

```

    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_min(:,ixe)=g_g;
    %gap filling radiazione
    r_g=d_values_rad_sum(:,ixe);
    xi = find(isnan(r_g));
    x = 1:length(r_g);
    D = setdiff(x, xi);
    yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
    r_g(xi) = yi;
    d_values_rad_sum(:,ixe)=r_g;
    %%%%%%%%%%%
end

% %gap-filling e quality check per il 2008 DONNA OLIMPIA
% if styear==2008
%     %gap filling aria E
%     a_g=d_values_air_avg(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,8)=a_g;
%     a_g=d_values_air_max(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,8)=a_g;
%     a_g=d_values_air_min(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,8)=a_g;
%     %gap filling grape E
%     g_g=d_values_gra_avg(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,8)=g_g;
%     g_g=d_values_gra_max(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,8)=g_g;
%     g_g=d_values_gra_min(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,8)=g_g;
%     %%%%%%%%%%%
%     %tolgo la B e tutti i suoi parametri li imposto come la G che ha le
%     stesse due tipologie di def e gem
%     d_values_air_avg(:,4)=d_values_air_avg(:,3);
%     d_values_air_max(:,4)=d_values_air_max(:,3);
%     d_values_air_min(:,4)=d_values_air_min(:,3);
%     d_values_gra_avg(:,4)=d_values_gra_avg(:,3);
%     d_values_gra_max(:,4)=d_values_gra_max(:,3);
%     d_values_gra_min(:,4)=d_values_gra_min(:,3);
%     d_values_rad_sum(:,4)=d_values_rad_sum(:,3);
end

```

```

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h_values_air_avg(:,4)=h_values_air_avg(:,3);
    h_values_gra_avg(:,4)=h_values_gra_avg(:,3);
    h_values_rad_avg(:,4)=h_values_rad_avg(:,3);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);

```

```

iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))
/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,
)));
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,
)));

```

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));

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);
%
% dMorning_5_10_air=[2.4988 2.6327 2.5892 2.5218];
% dMorning_5_10_grape=[2.8124 2.7129 3.1847 2.7789];
% dTwilight_15_20_air=[-1.5655 -1.5818 -1.5685 -1.5096];
% dTwilight_15_20_grape=[-1.7907 -1.712 -1.801 -1.7404];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);

% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');
% Master IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master IG
% Master IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')

```

```

hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==3 %cacciagrande

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ss '(mast), 'c-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), s '(m), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(n), s '(n), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), s '(o), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(p), s '(p), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), s '(q), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), s '(r), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), s '(s1), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(t), s '(t), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(u), s '(u), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), s '(v1), 'k-','LineWidth',2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'Q'', 'R'', 'S'', 'T'', 'U'', 'V'',-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);

```

```

v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth',2)'])
eval(['legend(''Master'', ''M'', ''N'', ''O'', ''P'', ''Q'', ''R'', ''S'', ''T'', ''
U'', ''V'',-1)'])

```

```

sname22=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'', sname22 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth',2)'])
eval(['legend(''Master'', ''M'', ''N'', ''O'', ''P'', ''Q'', ''R'', ''S'', ''T'', ''
U'', ''V'',-1)'])

```

```

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'', sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth',2)'])

```



```
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth', 2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'Q'', 'R'', 'S'', 'T'', 'U'', 'V'', -1)'])
```

```
sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'', sname4 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['bar(N.JDTpc(mast), ss(mast), 'r', 'BarWidth', 0.6)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'c+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'm-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['legend(''Master/100'', 'M'', 'N'', 'P'', 'T'', 'U'', -1)'])
```

```
sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'', sname5 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['bar(N.JDTpc(mast), ss(mast), 'r', 'BarWidth', 0.6)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'c+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'm-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['legend(''Master/100'', 'M'', 'N'', 'P'', 'T'', 'U'', -1)'])
```

```
sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'm+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'c+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'm-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'r-', 'LineWidth', 2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'T'', 'U'', 'V'', -1)'])
```

```
sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'', sname7 )'])
grid on
hold on
```

```

ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-','LineWidth',2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'Q'', 'R'', 'S'', 'T'', 'U'', 'V'',-1)'])

```

```

sname8=['Tair Master vs Tgrape2 cm Slave'];
eval(['figure(''Name'', sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-','LineWidth',2)'])
eval(['legend(''Master'', 'O'', 'Q'', 'R'', 'S'', 'V'',-1)'])

```

```

sname9=['Tair Master vs Tgrape3 cm Slave'];
eval(['figure(''Name'', sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-','LineWidth',2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'Q'', 'R'', 'S'', 'T'', 'U'', 'V'',-1)'])

```

```
%pivot solo HV
```

```
p_ora=
```

```
vertcat(N.ora(o),N.ora(p),N.ora(q),N.ora(r),N.ora(s1),N.ora(t),N.ora(u),N.ora(v1));
```

```

p_def=
vertcat(N.defogliazione(o),N.defogliazione(p),N.defogliazione(q),N.defogliazione(r),N.defogliazione(s1),N.defogliazione(t),N.defogliazione(u),N.defogliazione(v1));
p_gem=
vertcat(N.gemme(o),N.gemme(p),N.gemme(q),N.gemme(r),N.gemme(s1),N.gemme(t),N.gemme(u),N.gemme(v1));
p_dir=
vertcat(N.diradamento(o),N.diradamento(p),N.diradamento(q),N.diradamento(r),N.diradamento(s1),N.diradamento(t),N.diradamento(u),N.diradamento(v1));
p_grap=
vertcat(N.Tgrap(o),N.Tgrap(p),N.Tgrap(q),N.Tgrap(r),N.Tgrap(s1),N.Tgrap(t),N.Tgrap(u),N.Tgrap(v1));
p_air=
vertcat(N.Tair(o),N.Tair(p),N.Tair(q),N.Tair(r),N.Tair(s1),N.Tair(t),N.Tair(u),N.Tair(v1));
p_rad=
vertcat(N.Rad(o),N.Rad(p),N.Rad(q),N.Rad(r),N.Rad(s1),N.Rad(t),N.Rad(u),N.Rad(v1));
p_doy=
vertcat(N.Doy(o),N.Doy(p),N.Doy(q),N.Doy(r),N.Doy(s1),N.Doy(t),N.Doy(u),N.Doy(v1));

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def p_gem p_dir],p_rad,'nansum');

```

```

%gap-filling e quality check per il 2009 CACCIAGRANDE
if styear==2009

```

```

% %tolgo la GRAPE della T e la imposto come la O
d_values_gra_avg(:,8)=d_values_gra_avg(:,7);
d_values_gra_max(:,8)=d_values_gra_max(:,7);
d_values_gra_min(:,8)=d_values_gra_min(:,7);
%valori dal 182 della T di caccia grande mancanti
d_values_air_avg(1:9,8)=d_values_air_avg(1:9,7);
d_values_air_max(1:9,8)=d_values_air_max(1:9,7);
d_values_air_min(1:9,8)=d_values_air_min(1:9,7);
d_values_gra_avg(1:9,8)=d_values_gra_avg(1:9,7);
d_values_gra_max(1:9,8)=d_values_gra_max(1:9,7);
d_values_gra_min(1:9,8)=d_values_gra_min(1:9,7);
d_values_rad_sum(1:9,8)=d_values_rad_sum(1:9,7);

```

```

%tolgo la P e la imposto come la Q
d_values_air_avg(:,5)=d_values_air_avg(:,6);
d_values_air_max(:,5)=d_values_air_max(:,6);
d_values_air_min(:,5)=d_values_air_min(:,6);
d_values_gra_avg(:,5)=d_values_gra_avg(:,6);
d_values_gra_max(:,5)=d_values_gra_max(:,6);
d_values_gra_min(:,5)=d_values_gra_min(:,6);
d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
% ciclo per tutte le slave gap filling aria
for ix=1:8
a_g=d_values_air_avg(:,ix);
xi = find(isnan(a_g));
x = 1:length(a_g);
D = setdiff(x, xi);
yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
a_g(xi) = yi;
d_values_air_avg(:,ix)=a_g;
a_g=d_values_air_max(:,ix);

```

```

    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_max(:,ixe)=a_g;
    a_g=d_values_air_min(:,ixe);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_min(:,ixe)=a_g;
% ciclo per tutte le slave gap filling GRAPE
    g_g=d_values_gra_avg(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_avg(:,ixe)=g_g;
    g_g=d_values_gra_max(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_max(:,ixe)=g_g;
    g_g=d_values_gra_min(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_min(:,ixe)=g_g;
    %gap filling radiazione
    r_g=d_values_rad_sum(:,ixe);
    xi = find(isnan(r_g));
    x = 1:length(r_g);
    D = setdiff(x, xi);
    yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
    r_g(xi) = yi;
    d_values_rad_sum(:,ixe)=r_g;
    %%%%%%%%%%%
end

%gap-filling e quality check per il 2008 caccia grande
% if styear==2008
%     %gap filling aria U
%     a_g=d_values_air_avg(:,4);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,4)=a_g;
%     a_g=d_values_air_max(:,4);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,4)=a_g;
%     a_g=d_values_air_min(:,4);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,4)=a_g;
%     %gap filling aria T
%     a_g=d_values_air_avg(:,8);
%     xi = find(isnan(a_g));

```

```

%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,8)=a_g;
%     a_g=d_values_air_max(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,8)=a_g;
%     a_g=d_values_air_min(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,8)=a_g;
%gap filling grape R
%     g_g=d_values_gra_avg(:,2);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,2)=g_g;
%     g_g=d_values_gra_max(:,2);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,2)=g_g;
%     g_g=d_values_gra_min(:,2);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,2)=g_g;
%gap filling grape U
%     g_g=d_values_gra_avg(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,4)=g_g;
%     g_g=d_values_gra_max(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,4)=g_g;
%     g_g=d_values_gra_min(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,4)=g_g;
%gap filling grape O
%     g_g=d_values_gra_avg(:,7);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,7)=g_g;
%     g_g=d_values_gra_max(:,7);
%     xi = find(isnan(g_g));

```

```

% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values_gra_max(:,7)=g_g;
% g_g=d_values_gra_min(:,7);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values_gra_min(:,7)=g_g;
% %gap filling grape T
% g_g=d_values_gra_avg(:,8);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values_gra_avg(:,8)=g_g;
% g_g=d_values_gra_max(:,8);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values_gra_max(:,8)=g_g;
% g_g=d_values_gra_min(:,8);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values_gra_min(:,8)=g_g;
% %gap filling radiazione U
% r_g=d_values_rad_sum(:,4);
% xi = find(isnan(r_g));
% x = 1:length(r_g);
% D = setdiff(x, xi);
% yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
% r_g(xi) = yi;
% d_values_rad_sum(:,4)=r_g;
% %gap filling radiazione T
% r_g=d_values_rad_sum(:,8);
% xi = find(isnan(r_g));
% x = 1:length(r_g);
% D = setdiff(x, xi);
% yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
% r_g(xi) = yi;
% d_values_rad_sum(:,8)=r_g;
%
% %%%%%%%%%%%
% %tolgo la p e tutti i suoi parametri li imposto come la q che ha le
stesse due tipologie di def e gem
% d_values_air_avg(:,5)=d_values_air_avg(:,6);
% d_values_air_max(:,5)=d_values_air_max(:,6);
% d_values_air_min(:,5)=d_values_air_min(:,6);
% d_values_gra_avg(:,5)=d_values_gra_avg(:,6);
% d_values_gra_max(:,5)=d_values_gra_max(:,6);
% d_values_gra_min(:,5)=d_values_gra_min(:,6);
% d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
%la p la imposto come la q
h_values_air_avg(:,5)=h_values_air_avg(:,6);

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    h_values_gra_avg(:,5)=h_values_gra_avg(:,6);
    h_values_rad_avg(:,5)=h_values_rad_avg(:,6);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))
/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);

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% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,
)));
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,
)));

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);

% dMorning_5_10_air=[2.4447 2.2638 2.5846 2.4751];

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% dMorning 5 10 grape=[2.6667 2.3504 2.6263 2.9344];
% dTwilight 15 20 air=[-1.6864 -1.6683 -1.7803 -1.6947];
% dTwilight 15 20 grape=[-1.8974 -1.9125 -1.8355 -2.1824];
% d index air=vertcat(dMorning 5 10 air,dTwilight 15 20 air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');

% Master_IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master_IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master_SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master_IG
% Master_IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master_CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k+-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g+-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r+-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y+-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on

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plot(h_row air avg,h values air avg tesi(:,1),'k-','LineWidth',2)
plot(h_row air avg,h values air avg tesi(:,2),'g-','LineWidth',2)
plot(h_row air avg,h values air avg tesi(:,3),'r-','LineWidth',2)
plot(h_row air avg,h values air avg tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row rad avg,h values rad avg tesi(:,1),'k-','LineWidth',2)
plot(h_row rad avg,h values rad avg tesi(:,2),'g-','LineWidth',2)
plot(h_row rad avg,h values rad avg tesi(:,3),'r-','LineWidth',2)
plot(h_row rad avg,h values rad avg tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==4 %cortigliano

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-','LineWidth',2)'])

eval(['legend(''Master'', 'A','B','C','D','E','F','G','H','I',
'L',-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-','LineWidth',2)'])

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eval(['plot(N.JDTpc(d),' s '(d),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname22=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'', sname22 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname3=['precipitazioni Master vs bagnatura foglieare Slave'];
eval(['figure(''Name'', sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'b+-','LineWidth',2)'])

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```

eval(['plot(N.JDTpc(1),' s '(1),'b-.'','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'',' sname4 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['bar(N.JDTpc(mast),ss(mast),'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-.'','LineWidth',2)'])
eval(['legend(''Master/100'',''A'',''E'',''H'',''I'',''L'','-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'',' sname5 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['bar(N.JDTpc(mast),ss(mast),'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'m+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'c+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-.'','LineWidth',2)'])
eval(['legend(''Master/100'',''A'',''E'',''H'',''I'',''L'','-1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'',' sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==3 & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-'','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'b-.'','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'m-.'','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'r-.'','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'c+-'','LineWidth',2)'])

eval(['legend(''Master'',''A'',''C'',''E'',''G'',''I'',''L'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'',' sname7 )'])
grid on
hold on
ss='N.RadGmedia';

```

```

s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'k+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-', 'LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

sname8=['Tair Master vs Tgrape2 Slave'];
eval(['figure(''Name'','' sname8 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoill1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);

eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-', 'LineWidth',2)'])
eval(['legend(''Master'',''B'',''C'',''D'',''F'',''G'','-1)'])

sname9=['Tair Master vs Tgrape3 Slave'];
eval(['figure(''Name'','' sname9 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'y-', 'LineWidth',2)'])
eval(['legend(''Master'',''B'',''C'',''D'',''F'',''G'','-1)'])

%pivot solo HV
p_ora=
vertcat(N.ora(a),N.ora(b),N.ora(c),N.ora(d),N.ora(e),N.ora(f),N.ora(g),N.ora

```

```

a(h));
p_def=
vertcat(N.defogliazione(a),N.defogliazione(b),N.defogliazione(c),N.defogliazione(d),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h));
p_gem=
vertcat(N.gemme(a),N.gemme(b),N.gemme(c),N.gemme(d),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h));
p_dir=
vertcat(N.diradamento(a),N.diradamento(b),N.diradamento(c),N.diradamento(d),N.diradamento(e),N.diradamento(f),N.diradamento(g),N.diradamento(h));
p_grap=
vertcat(N.Tgrap(a),N.Tgrap(b),N.Tgrap(c),N.Tgrap(d),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h));
p_air=
vertcat(N.Tair(a),N.Tair(b),N.Tair(c),N.Tair(d),N.Tair(e),N.Tair(f),N.Tair(g),N.Tair(h));
p_rad=
vertcat(N.Rad(a),N.Rad(b),N.Rad(c),N.Rad(d),N.Rad(e),N.Rad(f),N.Rad(g),N.Rad(h));
p_doy=
vertcat(N.Doy(a),N.Doy(b),N.Doy(c),N.Doy(d),N.Doy(e),N.Doy(f),N.Doy(g),N.Doy(h));

```

```

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def p_gem p_dir],p_rad,'nansum');

```

```
%gap-filling e quality check per il 2009 CORTIGLIANO
```

```
if styear==2009
```

```
%
%tolgo la E
d_values_air_avg(:,3)=d_values_air_avg(:,4);
d_values_air_max(:,3)=d_values_air_max(:,4);
d_values_air_min(:,3)=d_values_air_min(:,4);
d_values_gra_avg(:,3)=d_values_gra_avg(:,4);
d_values_gra_max(:,3)=d_values_gra_max(:,4);
d_values_gra_min(:,3)=d_values_gra_min(:,4);
d_values_rad_sum(:,3)=d_values_rad_sum(:,4);
%valori dal 182 della L di cortigliano mancanti
d_values_air_avg(1:9,8)=d_values_air_avg(1:9,7);
d_values_air_max(1:9,8)=d_values_air_max(1:9,7);
d_values_air_min(1:9,8)=d_values_air_min(1:9,7);
d_values_gra_avg(1:9,8)=d_values_gra_avg(1:9,7);
d_values_gra_max(1:9,8)=d_values_gra_max(1:9,7);
d_values_gra_min(1:9,8)=d_values_gra_min(1:9,7);
d_values_rad_sum(1:9,8)=d_values_rad_sum(1:9,7);

```

```
%tolgo la D
```

```

d_values_air_avg(:,5)=d_values_air_avg(:,6);
d_values_air_max(:,5)=d_values_air_max(:,6);
d_values_air_min(:,5)=d_values_air_min(:,6);
d_values_gra_avg(:,5)=d_values_gra_avg(:,6);
d_values_gra_max(:,5)=d_values_gra_max(:,6);
d_values_gra_min(:,5)=d_values_gra_min(:,6);
d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
% ciclo per tutte le slave gap filling aria
for ix=1:8
a_g=d_values_air_avg(:,ix);
xi = find(isnan(a_g));
x = 1:length(a_g);
D = setdiff(x, xi);

```

```

    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_avg(:,ixe)=a_g;
a_g=d_values_air_max(:,ixe);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_max(:,ixe)=a_g;
a_g=d_values_air_min(:,ixe);
    xi = find(isnan(a_g));
    x = 1:length(a_g);
    D = setdiff(x, xi);
    yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
    a_g(xi) = yi;
    d_values_air_min(:,ixe)=a_g;
% ciclo per tutte le slave gap filling GRAPE
g_g=d_values_gra_avg(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_avg(:,ixe)=g_g;
g_g=d_values_gra_max(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_max(:,ixe)=g_g;
g_g=d_values_gra_min(:,ixe);
    xi = find(isnan(g_g));
    x = 1:length(g_g);
    D = setdiff(x, xi);
    yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
    g_g(xi) = yi;
    d_values_gra_min(:,ixe)=g_g;
    %gap filling radiazione
r_g=d_values_rad_sum(:,ixe);
    xi = find(isnan(r_g));
    x = 1:length(r_g);
    D = setdiff(x, xi);
    yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
    r_g(xi) = yi;
    d_values_rad_sum(:,ixe)=r_g;
    %%%%%%%%%%%
end

% %gap-filling e quality check per il 2008 cortigliano
% if styear==2008
%     %gap filling aria H
%     a_g=d_values_air_avg(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,2)=a_g;
%     a_g=d_values_air_max(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,2)=a_g;
%     a_g=d_values_air_min(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);

```

```

%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,2)=a_g;
%     %gap filling aria C
%     a_g=d_values_air_avg(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d values air avg(:,1)=a_g;
%     a_g=d_values_air_max(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d values air max(:,1)=a_g;
%     a_g=d_values_air_min(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,1)=a_g;
% %gap filling grape E
%     g_g=d_values_gra_avg(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d values gra avg(:,3)=g_g;
%     g_g=d_values_gra_max(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d values gra max(:,3)=g_g;
%     g_g=d_values_gra_min(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,3)=g_g;
% %gap filling grape D
%     g_g=d_values_gra_avg(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d values gra avg(:,5)=g_g;
%     g_g=d_values_gra_max(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d values gra max(:,5)=g_g;
%     g_g=d_values_gra_min(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,5)=g_g;
% %gap filling grape F
%     g_g=d_values_gra_avg(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);

```



```

%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra avg(:,6)=g g;
%      g_g=d_values_gra_max(:,6);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra_max(:,6)=g g;
%      g_g=d_values_gra_min(:,6);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_min(:,6)=g g;
%      %gap filling radiazione H
%      r_g=d_values_rad_sum(:,2);
%      xi = find(isnan(r_g));
%      x = 1:length(r_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%      r_g(xi) = yi;
%      d_values_rad_sum(:,2)=r_g;
%      %%%%%%%%%%%
%      %tolgo la g e tutti i suoi parametri li imposto come la b che ha le
stesse due tipologie di def e gem
%      d_values_air_avg(:,8)=d_values_air_avg(:,7);
%      d_values_air_max(:,8)=d_values_air_max(:,7);
%      d_values_air_min(:,8)=d_values_air_min(:,7);
%      d_values_gra_avg(:,8)=d_values_gra_avg(:,7);
%      d_values_gra_max(:,8)=d_values_gra_max(:,7);
%      d_values_gra_min(:,8)=d_values_gra_min(:,7);
%      d_values_rad_sum(:,8)=d_values_rad_sum(:,7);
end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h_values_air_avg(:,8)=h_values_air_avg(:,7);
    h_values_gra_avg(:,8)=h_values_gra_avg(:,7);
    h_values_rad_avg(:,8)=h_values_rad_avg(:,7);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2)));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2)));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2)));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2)));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2)));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2)));
t_values_gra_avg=

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horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))
/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))

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*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);

% dMorning 5 10 air=[2.9376 2.9804 2.9436 2.9687];
% dMorning 5 10 grape=[2.1089 2.1664 2.2037 2.8037];
% dTwilight 15 20 air=[-1.8499 -1.8052 -1.8686 -1.9616];
% dTwilight 15 20 grape=[-2.2782 -2.249 -2.4021 -2.4053];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=

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pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');

% Master_IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master_IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master_SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master_IG
% Master_IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master_CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
end

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loadde=input('carico il file dat ?? si(1) no(2) ');
if loadde==1
clear all;
close all;
clc;
choice=input('2010 no despiking(0) - NO despiking 2008(1)-CALIBRATI(2)-
DESPIKING 2009(3)-DESPIKING 2008(4)- despiking 2010(5): ');
if choice==1
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2008\5_DATASET_2008_FINALE\Tuscania_dataset_HY_2008_CALIBRATI.dat'];
elseif choice==2
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2009\Tuscania_dataset_HY_2009_pre_despiking.dat'];
elseif choice==3
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2009\Tuscania_dataset_HY_2009_despiking.dat'];
elseif choice==4
File_wrk=['C:\TUSCANIA\MATRICE Dati Tuscania\MATRICI
2008\5_DATASET_2008_FINALE\Tuscania_dataset_HY_2008_index.dat'];
elseif choice==0
File_wrk=['C:\TUSCANIA\DATI_2010\Tuscania_dataset_HY_2010.dat'];
elseif choice==5
File_wrk=['C:\TUSCANIA\DATI_2010\Tuscania_dataset_HY_2010_despiking.dat'];
end
N=dload(File_wrk);
end
gi = input('Giorno iniziale (doy es 18200): ');
gf = input('Giorno finale (doy es 25800): ');
styear=input('Anno: ');
v=input('Vigneto Brolio(1)-Donna Olimpia(2)-MortelleA Cacciagrande(3)-
MortelleB Cortigliano(4): ');
delta=(gf-gi)/100+1;
%fattori di conversione

%potenziale idrico
Tasterisco30=(3.15-N.T_heatsoil1)/(3.15-0.75);
N.T_heatsoil1=-0.056*(Tasterisco30.^(-1/0.45)-1);
Tasterisco60=(3.15-N.T_heatsoil2)/(3.15-0.75);
N.T_heatsoil2=-0.056*(Tasterisco60.^(-1/0.45)-1);

%radiazione da integrale giornaliera w m-2 a MJ m-2
%N.Rad=N.Rad*3600/1000000;

%ANEMOMETRI SLAVE
N.VV=24*N.VV;

if v==1
sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss ' (mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'r-','LineWidth',2)'])

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eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'',' sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

% sname22=['Tair Master vs Tleaf Slave'];
% eval(['figure(''Name'',' sname22 )'])
% grid on
% hold on
% ss='N.Tempmedia';
% s='N.Tleaf';
% mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
% a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
% b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
% c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
% d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
% e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
% f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
% g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
% h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
% i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
% l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
% m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
% eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
% eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
% eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
% eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])

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% eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
% eval(['plot(N.JDTpc(e),' s '(e),'r+-','LineWidth',2)'])
% eval(['plot(N.JDTpc(f),' s '(f),'y+-','LineWidth',2)'])
% eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
% eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
% eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
% eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
% eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
%
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'',' sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'b-o','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'k+-','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'',' sname4 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoil1';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-','LineWidth',2)'])

eval(['plot(N.JDTpc(mast),' sss(mast),'r+-','LineWidth',2)'])

eval(['plot(N.JDTpc(a),' s '(a),'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'y-','LineWidth',2)'])
eval(['legend(''Master'',''airMaster'',''A'',''B'',''C'',''D'',''M'','-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'',' sname5 )'])
grid on

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```

hold on
ss='N.Piogsomma';
s='N.T heatsoil2';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast),'r+-', 'LineWidth',2)'])

eval(['plot(N.JDTpc(a),' s '(a),'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'m+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'y-', 'LineWidth',2)'])
eval(['legend(''Master'', ''airMaster'', ''A'', ''B'', ''C'', ''D'', ''M'', -1)'])

% sname6=['vento Master vs vento Slave'];
% eval(['figure(''Name'', sname6 )'])
% grid on
% hold on
% ss='N.VVent1Vmed';
% s='N.VV';
% mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
% a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==
6);
% b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==
4);
% c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==
5);
% d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==
1);
% f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==
1);
% h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==
1);
% l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==
1);
% eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(a),' s '(a),'b+-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(b),' s '(b),'g+-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(c),' s '(c),'m+-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(d),' s '(d),'c+-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(f),' s '(f),'b-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(h),' s '(h),'m-', 'LineWidth',2)'])
% eval(['plot(N.JDTpc(l),' s '(l),'r-', 'LineWidth',2)'])
% eval(['legend(''Master'', ''A'', ''B'', ''C'', ''D'', ''F'', ''H'', ''L'', -1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'', sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'b-o', 'LineWidth',2)'])

```



```

eval(['plot(N.JDTpc(d),' s '(d),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'y+','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),'k+','LineWidth',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'',''M'','-1)'])

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```

% sname8=['Tair Master vs Tgrape2 Slave'];
% eval(['figure(''Name'', sname8 )'])
% grid on
% hold on
% ss='N.Tempmedia';
% s='N.Tsoil1';
% e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
% f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
% g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
% h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
% i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
% l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
% eval(['plot(N.JDTpc(mast),' ss '(mast),'c+','LineWidth',2)'])
% eval(['plot(N.JDTpc(e),' s '(e),'r+','LineWidth',2)'])
% eval(['plot(N.JDTpc(f),' s '(f),'y+','LineWidth',2)'])
% eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
% eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
% eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
% eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
% eval(['legend(''Master'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])
%

```

```

% sname9=['Tair Master vs Tgrape3 cm Slave'];
% eval(['figure(''Name'', sname9 )'])
% grid on
% hold on
% ss='N.Tempmedia';
% s='N.Tsoil2';
% e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
% f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
% g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
% h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
% i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
% l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
% eval(['plot(N.JDTpc(mast),' ss '(mast),'c+','LineWidth',2)'])
% eval(['plot(N.JDTpc(e),' s '(e),'r+','LineWidth',2)'])
% eval(['plot(N.JDTpc(f),' s '(f),'y+','LineWidth',2)'])
% eval(['plot(N.JDTpc(g),' s '(g),'r-','LineWidth',2)'])
% eval(['plot(N.JDTpc(h),' s '(h),'g-','LineWidth',2)'])
% eval(['plot(N.JDTpc(i),' s '(i),'k-','LineWidth',2)'])
% eval(['plot(N.JDTpc(l),' s '(l),'y-','LineWidth',2)'])
% eval(['legend(''Master'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

```

```
%pivot solo HV
```

```

p_ora=
vertcat(N.ora(d),N.ora(e),N.ora(f),N.ora(g),N.ora(h),N.ora(i),N.ora(l),N.ora(m));
p_def=
vertcat(N.defogliazione(d),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h),N.defogliazione(i),N.defogliazione(l),N.defogliazione(m));

```

```

zione(g),N.defogliazione(h),N.defogliazione(i),N.defogliazione(l),N.defogli
azione(m));
p_gem=
vertcat(N.gemme(d),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h),N.gemme(i),N
.gemme(l),N.gemme(m));
p_dir=
vertcat(N.diradamento(d),N.diradamento(e),N.diradamento(f),N.diradamento(g)
,N.diradamento(h),N.diradamento(i),N.diradamento(l),N.diradamento(m));
p_grap=
vertcat(N.Tgrap(d),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h),N.Tgrap(i),N
.Tgrap(l),N.Tgrap(m));
p_air=
vertcat(N.Tair(d),N.Tair(e),N.Tair(f),N.Tair(g),N.Tair(h),N.Tair(i),N.Tair(
l),N.Tair(m));
p_rad=
vertcat(N.Rad(d),N.Rad(e),N.Rad(f),N.Rad(g),N.Rad(h),N.Rad(i),N.Rad(l),N.Ra
d(m));
p_doy=
vertcat(N.Doy(d),N.Doy(e),N.Doy(f),N.Doy(g),N.Doy(h),N.Doy(i),N.Doy(l),N.Do
y(m));

```

```

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def
p_gem p_dir],p_rad,'nansum');

```

```

% %gap-filling e quality check per il 2009 BROLIO
% if styear==2009
%     %valori dal 182 della M di brolio mancanti
%     d_values_air_avg(1:9,1)=d_values_air_avg(1:9,2);
%     d_values_air_max(1:9,1)=d_values_air_max(1:9,2);
%     d_values_air_min(1:9,1)=d_values_air_min(1:9,2);
%     d_values_gra_avg(1:9,1)=d_values_gra_avg(1:9,2);
%     d_values_gra_max(1:9,1)=d_values_gra_max(1:9,2);
%     d_values_gra_min(1:9,1)=d_values_gra_min(1:9,2);
%     d_values_rad_sum(1:9,1)=d_values_rad_sum(1:9,2);
%
%     % ciclo per tutte le slave gap filling aria
%     for ix=1:8
%         a_g=d_values_air_avg(:,ix);
%         xi = find(isnan(a_g));
%         x = 1:length(a_g);
%         D = setdiff(x, xi);
%         yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%         a_g(xi) = yi;
%         d_values_air_avg(:,ix)=a_g;
%         a_g=d_values_air_max(:,ix);
%         xi = find(isnan(a_g));
%         x = 1:length(a_g);
%         D = setdiff(x, xi);
%         yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%         a_g(xi) = yi;
%         d_values_air_max(:,ix)=a_g;
%         a_g=d_values_air_min(:,ix);
%         xi = find(isnan(a_g));
%         x = 1:length(a_g);
%         D = setdiff(x, xi);
%         yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%         a_g(xi) = yi;
%         d_values_air_min(:,ix)=a_g;
%     % ciclo per tutte le slave gap filling GRAPE
%     g_g=d_values_gra_avg(:,ix);

```

```

%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,ixe)=g_g;
%     g_g=d_values_gra_max(:,ixe);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,ixe)=g_g;
%     g_g=d_values_gra_min(:,ixe);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,ixe)=g_g;
%     %gap filling radiazione
%     r_g=d_values_rad_sum(:,ixe);
%     xi = find(isnan(r_g));
%     x = 1:length(r_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%     r_g(xi) = yi;
%     d_values_rad_sum(:,ixe)=r_g;
%     %%%%%%%%%%%
%     end
%gap-filling e quality check per il 2008 BROLIO
% if styear==2008
%     %gap filling aria
%     a_g=d_values_air_avg(:,6);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,6)=a_g;
%     a_g=d_values_air_max(:,6);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,6)=a_g;
%     a_g=d_values_air_min(:,6);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,6)=a_g;
%     %gap filling grape G
%     g_g=d_values_gra_avg(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,6)=g_g;
%     g_g=d_values_gra_max(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,6)=g_g;
%     g_g=d_values_gra_min(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);

```

```

%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_min(:,6)=g_g;
%      %gap filling grape H
%      g_g=d_values_gra_avg(:,4);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra avg(:,4)=g_g;
%      g_g=d_values_gra_max(:,4);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra max(:,4)=g_g;
%      g_g=d_values_gra_min(:,4);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_min(:,4)=g_g;
%      %gap filling grape D
%      g_g=d_values_gra_avg(:,3);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra avg(:,3)=g_g;
%      g_g=d_values_gra_max(:,3);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d values gra max(:,3)=g_g;
%      g_g=d_values_gra_min(:,3);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_min(:,3)=g_g;
%      %gap filling radiazione G
%      r_g=d_values_rad_sum(:,6);
%      xi = find(isnan(r_g));
%      x = 1:length(r_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%      r_g(xi) = yi;
%      d values rad_sum(:,6)=r_g;
%      %%%%%%%%%%%

%      %tolgo la I e tutti i suoi parametri li imposto come la M che ha le
%      stesse due tipologie di def e gem
%      d_values_air_avg(:,2)=d_values_air_avg(:,1);
%      d_values_air_max(:,2)=d_values_air_max(:,1);
%      d_values_air_min(:,2)=d_values_air_min(:,1);
%      d_values_gra_avg(:,2)=d_values_gra_avg(:,1);
%      d_values_gra_max(:,2)=d_values_gra_max(:,1);
%      d_values_gra_min(:,2)=d_values_gra_min(:,1);
%      d_values_rad_sum(:,2)=d_values_rad_sum(:,1);
%end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
```

```

[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h_values_air_avg(:,2)=h_values_air_avg(:,1);
    h_values_gra_avg(:,2)=h_values_gra_avg(:,1);
    h_values_rad_avg(:,2)=h_values_rad_avg(:,1);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per tutte le tesi (slave)
% air_avg=nanmean(d_values_air_avg);
% air_max=nanmean(d_values_air_max);
% air_min=nanmean(d_values_air_min);
% air_avg_sum=nansum(d_values_air_avg);
% air_max_sum=nansum(d_values_air_max);
% air_min_sum=nansum(d_values_air_min);
% escursion=air_max_sum-air_min_sum;

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);

```

```

% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))
/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);

ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air

```

```

Night_19_02_Index_air=
nanmean(vertpcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertpcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertpcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertpcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);
%
% dMorning_5_10_air=[2.1664      2.2391  2.2652  2.3698];
% dMorning_5_10_grape=[2.1393      2.3651  2.3789  2.1366];
% dTwilight_15_20_air=[-1.7419  -1.9075  -1.9198  -1.8619];
% dTwilight_15_20_grape=[-2.1404      -2.0402  -2.1484  -1.9785];
% d_index_air=vertpcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertpcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertpcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertpcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master

p_doy m=N.Doy(mast);
p_air_avg m=N.Tempmedia(mast);
p_air_max m=N.Tmax(mast);
p_air_min m=N.Tmin(mast);
p_rad_avg m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');
% Master_IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master_IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master_SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master_IG
% Master_IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master_CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertpcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertpcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t

```

```

,gra_min_sum t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==2 %donna olimpia

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'b-','LineWidth',2)'])

eval(['legend(''Master'', 'A'', 'B'', 'C'', 'D'', 'E'', 'F'', 'G'', 'H'', 'I'',
'L'',-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);

```



```

a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

```

```

eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'b-','LineWidth',2)'])

```

```

eval(['legend(''Master'', 'A'', 'B'', 'C'', 'D'', 'E'', 'F'', 'G'', 'H'', 'I'', 'L'',-1)'])

```

```
sname22=['Tair Master vs Tleaf Slave'];
```

```
eval(['figure(''Name'', sname22 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Tempmedia';
```

```
s='N.Tleaf';
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
```

```
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
```

```
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
```

```
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
```

```
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
```

```
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
```

```
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(a), ' s '(a), 'k-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(c), ' s '(c), 'k+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(d), ' s '(d), 'b+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(e), ' s '(e), 'y+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(f), ' s '(f), 'r+-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(g), ' s '(g), 'g-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(h), ' s '(h), 'r-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(i), ' s '(i), 'y-','LineWidth',2)'])
```

```
eval(['plot(N.JDTpc(l), ' s '(l), 'b-','LineWidth',2)'])
```

```
eval(['legend(''Master'', 'A'', 'B'', 'C'', 'D'', 'E'', 'F'', 'G'', 'H'', 'I'', 'L'',-1)'])
```

```
sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
```

```
eval(['figure(''Name'', sname3 )'])
```

```
grid on
```

```
hold on
```

```
ss='N.Piogsomma';
```

```
s='N.Bagn_leaf';
```

```
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
```

```
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
```

```
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
```

```
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
```

```
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
```

```
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
```

```
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
```

```

g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'k-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'k+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'y+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'g-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'r-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'y-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'b-', 'LineWidth',2)'])

eval(['legend(''Master'', ''A'', ''B'', ''C'', ''D'', ''E'', ''F'', ''G'', ''H'', ''I'', ''L'', -1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'', sname4 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoill';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', ''airMaster'', ''A'', ''D'', ''F'', ''I'', ''L'', -1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'', sname5 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.T_heatsoil2';
sss=N.Tempmedia/100;
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(mast), sss(mast), 'r+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'b+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'g+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'm+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'c+-', 'LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'y-', 'LineWidth',2)'])
eval(['legend(''Master'', ''airMaster'', ''A'', ''D'', ''F'', ''I'', ''L'', -1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on
hold on
ss='N.VVentlVmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);

```

```

e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'r+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'b+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'c+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'b-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'m-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'r-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'m+-' , 'LineWidth',2)'])

eval(['legend(''Master'',''B'',''D'',''E'',''G'',''I'',''L'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'','' sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==5);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),'c+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),'k-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),'g+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),'k+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),'b+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),'y+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),'r+-' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),'g-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),'r-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),'y-.' , 'LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),'b-.' , 'LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

%pivot solo HV
p_ora=
vertcat(N.ora(a),N.ora(b),N.ora(c),N.ora(e),N.ora(f),N.ora(g),N.ora(h),N.ora(i));
p_def=
vertcat(N.defogliazione(a),N.defogliazione(b),N.defogliazione(c),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h),N.defogliazione(i));
p_gem=
vertcat(N.gemme(a),N.gemme(b),N.gemme(c),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h),N.gemme(i));
p_dir=
vertcat(N.diradamento(a),N.diradamento(b),N.diradamento(c),N.diradamento(e),N.diradamento(f),N.diradamento(g),N.diradamento(h),N.diradamento(i));
p_grap=
vertcat(N.Tgrap(a),N.Tgrap(b),N.Tgrap(c),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h),N.Tgrap(i));
p_air=
vertcat(N.Tair(a),N.Tair(b),N.Tair(c),N.Tair(e),N.Tair(f),N.Tair(g),N.Tair(h),N.Tair(i));
p_rad=
vertcat(N.Rad(a),N.Rad(b),N.Rad(c),N.Rad(e),N.Rad(f),N.Rad(g),N.Rad(h),N.Rad(i));
p_doy=
vertcat(N.Doy(a),N.Doy(b),N.Doy(c),N.Doy(e),N.Doy(f),N.Doy(g),N.Doy(h),N.Do

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y(i));
%%%%%%%%%%

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def
p_gem p_dir],p_rad,'nansum');

% %gap-filling e quality check per il 2009 DONNA OLIMPIA
% if styear==2009
% %   %tolgo la RAD della B e la imposto come la G
%   d_values_rad_sum(:,4)=d_values_rad_sum(:,3);
%
%   % ciclo per tutte le slave gap filling aria
%   for ix=1:8
%     a_g=d_values_air_avg(:,ix);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,ix)=a_g;
%     a_g=d_values_air_max(:,ix);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,ix)=a_g;
%     a_g=d_values_air_min(:,ix);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,ix)=a_g;
%   % ciclo per tutte le slave gap filling GRAPE
%   g_g=d_values_gra_avg(:,ix);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_avg(:,ix)=g_g;
%   g_g=d_values_gra_max(:,ix);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_max(:,ix)=g_g;
%   g_g=d_values_gra_min(:,ix);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_min(:,ix)=g_g;
%   %gap filling radiazione
%   r_g=d_values_rad_sum(:,ix);
%   xi = find(isnan(r_g));

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%       x = 1:length(r_g);
%       D = setdiff(x, xi);
%       yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%       r_g(xi) = yi;
%       d_values_rad_sum(:,ixe)=r_g;
%       %%%%%%%%%%%
%       end
%
% %gap-filling e quality check per il 2008 DONNA OLIMPIA
% if styear==2008
%     %gap filling aria E
%     a_g=d_values_air_avg(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,8)=a_g;
%     a_g=d_values_air_max(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,8)=a_g;
%     a_g=d_values_air_min(:,8);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,8)=a_g;
%     %gap filling grape E
%     g_g=d_values_gra_avg(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,8)=g_g;
%     g_g=d_values_gra_max(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,8)=g_g;
%     g_g=d_values_gra_min(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,8)=g_g;
%     %%%%%%%%%%%
%     %tolgo la B e tutti i suoi parametri li imposto come la G che ha le
%     stesse due tipologie di def e gem
%     d_values_air_avg(:,4)=d_values_air_avg(:,3);
%     d_values_air_max(:,4)=d_values_air_max(:,3);
%     d_values_air_min(:,4)=d_values_air_min(:,3);
%     d_values_gra_avg(:,4)=d_values_gra_avg(:,3);
%     d_values_gra_max(:,4)=d_values_gra_max(:,3);
%     d_values_gra_min(:,4)=d_values_gra_min(:,3);
%     d_values_rad_sum(:,4)=d_values_rad_sum(:,3);
% end
%
% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def

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p gem p dir],p rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h values air avg(:,4)=h values air avg(:,3);
    h_values_gra_avg(:,4)=h_values_gra_avg(:,3);
    h_values_rad_avg(:,4)=h_values_rad_avg(:,3);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))
/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))

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/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))
/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH_grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))
/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))
/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))
/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET_air
set_air=sum(t_values_air_max-t_values_air_min);
% SET_grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG_grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS_grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI_grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=

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vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Index_grape);
%
% dMorning 5 10 air=[2.4988 2.6327 2.5892 2.5218];
% dMorning_5_10_grape=[2.8124 2.7129 3.1847 2.7789];
% dTwilight_15_20_air=[-1.5655 -1.5818 -1.5685 -1.5096];
% dTwilight_15_20_grape=[-1.7907 -1.712 -1.801 -1.7404];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);

% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');
% Master IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master IG
% Master IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot_tgrape_alta_vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-

```



```

potato'),'A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k+','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g+','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r+','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y+','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k+','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g+','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r+','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y+','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==3 %cacciagrande

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ss '(mast), 'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(m), s '(m), 'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n), s '(n), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), s '(o), 'y+','LineWidth',2)'])
eval(['plot(N.JDTpc(p), s '(p), 'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(q), s '(q), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r), s '(r), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1), s '(s1), 'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(t), s '(t), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(u), s '(u), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1), s '(v1), 'k+','LineWidth',2)'])
eval(['legend(''Master'', 'M'', 'N'', 'O'', 'P'', 'Q'', 'R'', 'S'', 'T'', '
U'', 'V'',-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ss '(mast), 'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(m), s '(m), 'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n), s '(n), 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(o), s '(o), 'y+','LineWidth',2)'])
eval(['plot(N.JDTpc(p), s '(p), 'r+','LineWidth',2)'])

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```

eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth', 2)'])
eval(['legend(''Master'', ''M'', ''N'', ''O'', ''P'', ''Q'', ''R'', ''S'', ''T'', ''U'', ''V'', -1)'])

```

```

sname2=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'', sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth', 2)'])
eval(['legend(''Master'', ''M'', ''N'', ''O'', ''P'', ''Q'', ''R'', ''S'', ''T'', ''U'', ''V'', -1)'])

```

```

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'', sname3 )'])
grid on
hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(m), ' s '(m), 'b+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(n), ' s '(n), 'b-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(o), ' s '(o), 'y+', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(p), ' s '(p), 'r+', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(q), ' s '(q), 'r-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(r), ' s '(r), 'k-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(s1), ' s '(s1), 'g+-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(t), ' s '(t), 'y-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(u), ' s '(u), 'g-', 'LineWidth', 2)'])
eval(['plot(N.JDTpc(v1), ' s '(v1), 'k+-', 'LineWidth', 2)'])
eval(['legend(''Master'', ''M'', ''N'', ''O'', ''P'', ''Q'', ''R'', ''S'', ''T'', ''U'', ''V'', -1)'])

```

```

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];

```

```

eval(['figure(''Name'', sname4 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil1';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['bar(N.JDTpc(mast),ss(mast),'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(m),'s'(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),'s'(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),'s'(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),'s'(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),'s'(u),'g-','LineWidth',2)'])
eval(['legend(''Master/100'','M'','N'','P'','T'','U',-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'', sname5 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
eval(['bar(N.JDTpc(mast),ss(mast),'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(m),'s'(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),'s'(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),'s'(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),'s'(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),'s'(u),'g-','LineWidth',2)'])
eval(['legend(''Master/100'','M'','N'','P'','T'','U',-1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])
grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),'ss'(mast),'r+','LineWidth',2)'])
eval(['plot(N.JDTpc(m),'s'(m),'b+','LineWidth',2)'])
eval(['plot(N.JDTpc(n),'s'(n),'g+','LineWidth',2)'])
eval(['plot(N.JDTpc(o),'s'(o),'m+','LineWidth',2)'])
eval(['plot(N.JDTpc(p),'s'(p),'c+','LineWidth',2)'])
eval(['plot(N.JDTpc(t),'s'(t),'m-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),'s'(u),'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1),'s'(v1),'r-','LineWidth',2)'])
eval(['legend(''Master'','M'','N'','O'','P'','T'','U'','V',-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'', sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
m=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==6);
n=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==3);
o=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);

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```

p=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
q=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
r=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
s1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
t=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
u=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
v1=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(m),' s '(m),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(n),' s '(n),' 'b-','LineWidth',2)'])
eval(['plot(N.JDTpc(o),' s '(o),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(p),' s '(p),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(q),' s '(q),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(r),' s '(r),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(s1),' s '(s1),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(t),' s '(t),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(u),' s '(u),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(v1),' s '(v1),' 'k+-','LineWidth',2)'])
eval(['legend(''Master'',''M'',''N'',''O'',''P'',''Q'',''R'',''S'',''T'',''
U'',''V'','-1)'])

```

```

%pivot solo HV
p_ora=
vertcat(N.ora(o),N.ora(p),N.ora(q),N.ora(r),N.ora(s1),N.ora(t),N.ora(u),N.o
ra(v1));
p_def=
vertcat(N.defogliazione(o),N.defogliazione(p),N.defogliazione(q),N.defoglia
zione(r),N.defogliazione(s1),N.defogliazione(t),N.defogliazione(u),N.defogl
iazione(v1));
p_gem=
vertcat(N.gemme(o),N.gemme(p),N.gemme(q),N.gemme(r),N.gemme(s1),N.gemme(t),
N.gemme(u),N.gemme(v1));
p_dir=
vertcat(N.diradamento(o),N.diradamento(p),N.diradamento(q),N.diradamento(r)
,N.diradamento(s1),N.diradamento(t),N.diradamento(u),N.diradamento(v1));
p_grap=
vertcat(N.Tgrap(o),N.Tgrap(p),N.Tgrap(q),N.Tgrap(r),N.Tgrap(s1),N.Tgrap(t),
N.Tgrap(u),N.Tgrap(v1));
p_air=
vertcat(N.Tair(o),N.Tair(p),N.Tair(q),N.Tair(r),N.Tair(s1),N.Tair(t),N.Tair
(u),N.Tair(v1));
p_rad=
vertcat(N.Rad(o),N.Rad(p),N.Rad(q),N.Rad(r),N.Rad(s1),N.Rad(t),N.Rad(u),N.R
ad(v1));
p_doy=
vertcat(N.Doy(o),N.Doy(p),N.Doy(q),N.Doy(r),N.Doy(s1),N.Doy(t),N.Doy(u),N.D
oy(v1));

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def
p_gem p_dir],p_rad,'nansum');

```

```

% %gap-filling e quality check per il 2009 CACCIAGRANDE
% if styear==2009
% % %tolgo la GRAPE della T e la imposto come la O
% d_values_gra_avg(:,8)=d_values_gra_avg(:,7);
% d_values_gra_max(:,8)=d_values_gra_max(:,7);
% d_values_gra_min(:,8)=d_values_gra_min(:,7);
% %valori dal 182 della T di cacciagrande mancanti

```

```

% d values air avg(1:9,8)=d values air avg(1:9,7);
% d values air max(1:9,8)=d values air max(1:9,7);
% d values air min(1:9,8)=d values air min(1:9,7);
% d values gra avg(1:9,8)=d values gra avg(1:9,7);
% d values gra max(1:9,8)=d values gra max(1:9,7);
% d values gra min(1:9,8)=d values gra min(1:9,7);
% d_values_rad_sum(1:9,8)=d_values_rad_sum(1:9,7);
%
%
%tolgo la P e la imposto come la Q
% d values air avg(:,5)=d values air avg(:,6);
% d values air max(:,5)=d values air max(:,6);
% d values air min(:,5)=d values air min(:,6);
% d values gra avg(:,5)=d values gra avg(:,6);
% d values gra max(:,5)=d values gra max(:,6);
% d values gra min(:,5)=d values gra min(:,6);
% d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
% % ciclo per tutte le slave gap filling aria
% for ix=1:8
% a_g=d_values air avg(:,ix);
% xi = find(isnan(a_g));
% x = 1:length(a_g);
% D = setdiff(x, xi);
% yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
% a_g(xi) = yi;
% d_values air avg(:,ix)=a_g;
% a_g=d_values air max(:,ix);
% xi = find(isnan(a_g));
% x = 1:length(a_g);
% D = setdiff(x, xi);
% yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
% a_g(xi) = yi;
% d_values air max(:,ix)=a_g;
% a_g=d_values air min(:,ix);
% xi = find(isnan(a_g));
% x = 1:length(a_g);
% D = setdiff(x, xi);
% yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
% a_g(xi) = yi;
% d_values air min(:,ix)=a_g;
% % ciclo per tutte le slave gap filling GRAPE
% g_g=d_values gra avg(:,ix);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values gra avg(:,ix)=g_g;
% g_g=d_values gra max(:,ix);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values gra max(:,ix)=g_g;
% g_g=d_values gra min(:,ix);
% xi = find(isnan(g_g));
% x = 1:length(g_g);
% D = setdiff(x, xi);
% yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
% g_g(xi) = yi;
% d_values gra min(:,ix)=g_g;
% %gap filling radiazione
% r_g=d_values rad sum(:,ix);
% xi = find(isnan(r_g));
% x = 1:length(r_g);
% D = setdiff(x, xi);
% yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
% r_g(xi) = yi;
% d_values rad sum(:,ix)=r_g;
% %%%%%%%%%%%
% end

```

%gap-filling e quality check per il 2008 caccia grande

```

% if styear==2008
%   %gap filling aria U
%   a_g=d_values_air_avg(:,4);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_avg(:,4)=a_g;
%   a_g=d_values_air_max(:,4);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_max(:,4)=a_g;
%   a_g=d_values_air_min(:,4);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_min(:,4)=a_g;
%   %gap filling aria T
%   a_g=d_values_air_avg(:,8);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_avg(:,8)=a_g;
%   a_g=d_values_air_max(:,8);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_max(:,8)=a_g;
%   a_g=d_values_air_min(:,8);
%   xi = find(isnan(a_g));
%   x = 1:length(a_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%   a_g(xi) = yi;
%   d_values_air_min(:,8)=a_g;
%   %gap filling grape R
%   g_g=d_values_gra_avg(:,2);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_avg(:,2)=g_g;
%   g_g=d_values_gra_max(:,2);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_max(:,2)=g_g;
%   g_g=d_values_gra_min(:,2);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%   g_g(xi) = yi;
%   d_values_gra_min(:,2)=g_g;
%   %gap filling grape U
%   g_g=d_values_gra_avg(:,4);
%   xi = find(isnan(g_g));
%   x = 1:length(g_g);
%   D = setdiff(x, xi);
%   yi = interp1(D, g_g(D), xi, 'linear', 'extrap');

```

```

%     g_g(xi) = yi;
%     d_values_gra_avg(:,4)=g_g;
%     g_g=d_values_gra_max(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,4)=g_g;
%     g_g=d_values_gra_min(:,4);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,4)=g_g;
% %gap filling grape O
%     g_g=d_values_gra_avg(:,7);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,7)=g_g;
%     g_g=d_values_gra_max(:,7);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,7)=g_g;
%     g_g=d_values_gra_min(:,7);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,7)=g_g;
%     %gap filling grape T
%     g_g=d_values_gra_avg(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,8)=g_g;
%     g_g=d_values_gra_max(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,8)=g_g;
%     g_g=d_values_gra_min(:,8);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,8)=g_g;
%     %gap filling radiazione U
%     r_g=d_values_rad_sum(:,4);
%     xi = find(isnan(r_g));
%     x = 1:length(r_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%     r_g(xi) = yi;
%     d_values_rad_sum(:,4)=r_g;
%     %gap filling radiazione T
%     r_g=d_values_rad_sum(:,8);
%     xi = find(isnan(r_g));
%     x = 1:length(r_g);
%     D = setdiff(x, xi);

```

```

%      yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%      r_g(xi) = yi;
%      d_values_rad_sum(:,8)=r_g;
%
%      %%%%%%%%%%%
%      %tolgo la p e tutti i suoi parametri li imposto come la q che ha le
stesse due tipologie di def e gem
%      d_values_air_avg(:,5)=d_values_air_avg(:,6);
%      d_values_air_max(:,5)=d_values_air_max(:,6);
%      d_values_air_min(:,5)=d_values_air_min(:,6);
%      d_values_gra_avg(:,5)=d_values_gra_avg(:,6);
%      d_values_gra_max(:,5)=d_values_gra_max(:,6);
%      d_values_gra_min(:,5)=d_values_gra_min(:,6);
%      d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
%end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la p la imposto come la q
    h_values_air_avg(:,5)=h_values_air_avg(:,6);
    h_values_gra_avg(:,5)=h_values_gra_avg(:,6);
    h_values_rad_avg(:,5)=h_values_rad_avg(:,6);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);
air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);

```



```

gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);
ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape

```

```

CoolNI grap=gra min t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);

% dMorning_5_10_air=[2.4447 2.2638 2.5846 2.4751];
% dMorning_5_10_grape=[2.6667 2.3504 2.6263 2.9344];
% dTwilight_15_20_air=[-1.6864 -1.6683 -1.7803 -1.6947];
% dTwilight_15_20_grape=[-1.8974 -1.9125 -1.8355 -2.1824];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');

% Master_IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master_IH
master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master_SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master_IG
% Master_IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)

```

```

*(freddi_master+1);
% Master_CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_sum_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_air);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_index_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)

elseif v==4 %cortigliano

sname1=['Tair Master vs Tair Slave'];
eval(['figure(''Name'', sname1)'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tair';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast), ' ss '(mast), 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a), ' s '(a), 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b), ' s '(b), 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c), ' s '(c), 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d), ' s '(d), 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e), ' s '(e), 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f), ' s '(f), 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g), ' s '(g), 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h), ' s '(h), 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i), ' s '(i), 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l), ' s '(l), 'b-','LineWidth',2)'])

```

```

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''-1)'])
%
sname2=['Tair Master vs Tgrape Slave'];
eval(['figure(''Name'','' sname2 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tgrap';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'g+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'y+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'r+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'g-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'y-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'k+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'b+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-',''LineWidth'',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''-1)'])

sname22=['Tair Master vs Tleaf Slave'];
eval(['figure(''Name'','' sname22 )'])
grid on
hold on
ss='N.Tempmedia';
s='N.Tleaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'g+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'y+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'r+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'g-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'y-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'k+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'b+-',''LineWidth'',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-',''LineWidth'',2)'])
eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''
I'',''L'',''-1)'])

sname3=['precipitazioni Master vs bagnatura fogliare Slave'];
eval(['figure(''Name'','' sname3 )'])
grid on

```

```

hold on
ss='N.Piogsomma';
s='N.Bagn_leaf';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' 'y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' 'k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' 'r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' 'r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' 'y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'b-','LineWidth',2)'])

eval(['legend(''Master'',' 'A'',' 'B'',' 'C'',' 'D'',' 'E'',' 'F'',' 'G'',' 'H'',' 'I'',' 'L'','-1)'])

sname4=['precipitazioni Master vs potenziale idrico 30 cm Slave'];
eval(['figure(''Name'', sname4 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoill';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['bar(N.JDTpc(mast),ss(mast),' 'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'm+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'y-','LineWidth',2)'])
eval(['legend(''Master/100'',' 'A'',' 'E'',' 'H'',' 'I'',' 'L'','-1)'])

sname5=['precipitazioni Master vs potenziale idrico 60 cm Slave'];
eval(['figure(''Name'', sname5 )'])
grid on
hold on
ss=N.Piogsomma/100;
s='N.T_heatsoil2';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['bar(N.JDTpc(mast),ss(mast),' 'r','BarWidth',0.6)'])
eval(['plot(N.JDTpc(a),' s '(a),' 'b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' 'g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' 'm+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' 'c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' 'y-','LineWidth',2)'])
eval(['legend(''Master/100'',' 'A'',' 'E'',' 'H'',' 'I'',' 'L'','-1)'])

sname6=['vento Master vs vento Slave'];
eval(['figure(''Name'', sname6 )'])

```

```

grid on
hold on
ss='N.VVent1Vmed';
s='N.VV';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==3 & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);
eval(['plot(N.JDTpc(mast),' ss '(mast),' r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' m+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' b-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' m-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' r-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' c+-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''C'',''E'',''G'',''I'',''L'','-1)'])

sname7=['radiazione Master vs radiazione Slave'];
eval(['figure(''Name'',' sname7 )'])
grid on
hold on
ss='N.RadGmedia';
s='N.Rad';
mast = find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.vigoria==0);
a=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==6 & N.vigoria==1);
b=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==7 & N.vigoria==1);
c=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==1 & N.vigoria==1);
d=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==3 & N.vigoria==1);
e=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==1);
f=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==4 & N.vigoria==1);
g=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==8 & N.vigoria==1);
h=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==2 & N.vigoria==1);
i=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==2);
l=find(N.JDTpc<=gf & N.JDTpc>=gi & N.Vigna==v & N.tesi==5 & N.vigoria==4);

eval(['plot(N.JDTpc(mast),' ss '(mast),' c+-','LineWidth',2)'])
eval(['plot(N.JDTpc(a),' s '(a),' g+-','LineWidth',2)'])
eval(['plot(N.JDTpc(b),' s '(b),' y+-','LineWidth',2)'])
eval(['plot(N.JDTpc(c),' s '(c),' k-','LineWidth',2)'])
eval(['plot(N.JDTpc(d),' s '(d),' r+-','LineWidth',2)'])
eval(['plot(N.JDTpc(e),' s '(e),' g-','LineWidth',2)'])
eval(['plot(N.JDTpc(f),' s '(f),' r-','LineWidth',2)'])
eval(['plot(N.JDTpc(g),' s '(g),' y-','LineWidth',2)'])
eval(['plot(N.JDTpc(h),' s '(h),' k+-','LineWidth',2)'])
eval(['plot(N.JDTpc(i),' s '(i),' b+-','LineWidth',2)'])
eval(['plot(N.JDTpc(l),' s '(l),' b-','LineWidth',2)'])

eval(['legend(''Master'',''A'',''B'',''C'',''D'',''E'',''F'',''G'',''H'',''I'',''L'','-1)'])

%pivot solo HV
p_ora=
vertcat(N.ora(a),N.ora(b),N.ora(c),N.ora(d),N.ora(e),N.ora(f),N.ora(g),N.ora(h));
p_def=
vertcat(N.defogliazione(a),N.defogliazione(b),N.defogliazione(c),N.defogliazione(d),N.defogliazione(e),N.defogliazione(f),N.defogliazione(g),N.defogliazione(h));
p_gem=
vertcat(N.gemme(a),N.gemme(b),N.gemme(c),N.gemme(d),N.gemme(e),N.gemme(f),N.gemme(g),N.gemme(h));
p_dir=
vertcat(N.diradamento(a),N.diradamento(b),N.diradamento(c),N.diradamento(d),N.diradamento(e),N.diradamento(f),N.diradamento(g),N.diradamento(h));
p_grap=
vertcat(N.Tgrap(a),N.Tgrap(b),N.Tgrap(c),N.Tgrap(d),N.Tgrap(e),N.Tgrap(f),N.Tgrap(g),N.Tgrap(h));

```

```

p_air=
vertcat(N.Tair(a),N.Tair(b),N.Tair(c),N.Tair(d),N.Tair(e),N.Tair(f),N.Tair(
g),N.Tair(h));
p_rad=
vertcat(N.Rad(a),N.Rad(b),N.Rad(c),N.Rad(d),N.Rad(e),N.Rad(f),N.Rad(g),N.Ra
d(h));
p_doy=
vertcat(N.Doy(a),N.Doy(b),N.Doy(c),N.Doy(d),N.Doy(e),N.Doy(f),N.Doy(g),N.Do
y(h));

```

```

[d_values_air_avg,d_row_air_avg,d_column_air_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmean');
[d_values_air_max,d_row_air_max,d_column_air_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmax');
[d_values_air_min,d_row_air_min,d_column_air_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_air,'nanmin');
[d_values_gra_avg,d_row_gra_avg,d_column_gra_avg]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmean');
[d_values_gra_max,d_row_gra_max,d_column_gra_max]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmax');
[d_values_gra_min,d_row_gra_min,d_column_gra_min]=pivottable(p_doy,[p_def
p_gem p_dir],p_grap,'nanmin');
[d_values_rad_sum,d_row_rad_sum,d_column_rad_sum]=pivottable(p_doy,[p_def
p_gem p_dir],p_rad,'nansum');

```

```

% %gap-filling e quality check per il 2009 CORTIGLIANO
% if styear==2009
% %   %tolgo la E
%   d_values_air_avg(:,3)=d_values_air_avg(:,4);
%   d_values_air_max(:,3)=d_values_air_max(:,4);
%   d_values_air_min(:,3)=d_values_air_min(:,4);
%   d_values_gra_avg(:,3)=d_values_gra_avg(:,4);
%   d_values_gra_max(:,3)=d_values_gra_max(:,4);
%   d_values_gra_min(:,3)=d_values_gra_min(:,4);
%   d_values_rad_sum(:,3)=d_values_rad_sum(:,4);
%   %valori dal 182 della L di cortigliano mancanti
%   d_values_air_avg(1:9,8)=d_values_air_avg(1:9,7);
%   d_values_air_max(1:9,8)=d_values_air_max(1:9,7);
%   d_values_air_min(1:9,8)=d_values_air_min(1:9,7);
%   d_values_gra_avg(1:9,8)=d_values_gra_avg(1:9,7);
%   d_values_gra_max(1:9,8)=d_values_gra_max(1:9,7);
%   d_values_gra_min(1:9,8)=d_values_gra_min(1:9,7);
%   d_values_rad_sum(1:9,8)=d_values_rad_sum(1:9,7);
%
%   %tolgo la D
%   d_values_air_avg(:,5)=d_values_air_avg(:,6);
%   d_values_air_max(:,5)=d_values_air_max(:,6);
%   d_values_air_min(:,5)=d_values_air_min(:,6);
%   d_values_gra_avg(:,5)=d_values_gra_avg(:,6);
%   d_values_gra_max(:,5)=d_values_gra_max(:,6);
%   d_values_gra_min(:,5)=d_values_gra_min(:,6);
%   d_values_rad_sum(:,5)=d_values_rad_sum(:,6);
%   % ciclo per tutte le slave gap filling aria
%   for ix=1:8
%       a_g=d_values_air_avg(:,ix);
%       xi = find(isnan(a_g));
%       x = 1:length(a_g);
%       D = setdiff(x, xi);
%       yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%       a_g(xi) = yi;
%       d_values_air_avg(:,ix)=a_g;
%       a_g=d_values_air_max(:,ix);
%       xi = find(isnan(a_g));
%       x = 1:length(a_g);
%       D = setdiff(x, xi);
%       yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%       a_g(xi) = yi;
%       d_values_air_max(:,ix)=a_g;
%       a_g=d_values_air_min(:,ix);
%       xi = find(isnan(a_g));
%       x = 1:length(a_g);
%       D = setdiff(x, xi);

```

```

%      yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%      a_g(xi) = yi;
%      d_values_air_min(:,ixe)=a_g;
%      % ciclo per tutte le slave gap filling GRAPE
%      g_g=d_values_gra_avg(:,ixe);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_avg(:,ixe)=g_g;
%      g_g=d_values_gra_max(:,ixe);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_max(:,ixe)=g_g;
%      g_g=d_values_gra_min(:,ixe);
%      xi = find(isnan(g_g));
%      x = 1:length(g_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%      g_g(xi) = yi;
%      d_values_gra_min(:,ixe)=g_g;
%      %gap filling radiazione
%      r_g=d_values_rad_sum(:,ixe);
%      xi = find(isnan(r_g));
%      x = 1:length(r_g);
%      D = setdiff(x, xi);
%      yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%      r_g(xi) = yi;
%      d_values_rad_sum(:,ixe)=r_g;
%      %%%%%%%%%%%
%      end

```

```

% %gap-filling e quality check per il 2008 cortigliano
% if styear==2008
%     %gap filling aria H
%     a_g=d_values_air_avg(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,2)=a_g;
%     a_g=d_values_air_max(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_max(:,2)=a_g;
%     a_g=d_values_air_min(:,2);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_min(:,2)=a_g;
%     %gap filling aria C
%     a_g=d_values_air_avg(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values_air_avg(:,1)=a_g;
%     a_g=d_values_air_max(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);

```



```

%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d values air max(:,1)=a_g;
%     a_g=d_values_air_min(:,1);
%     xi = find(isnan(a_g));
%     x = 1:length(a_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, a_g(D), xi, 'linear', 'extrap');
%     a_g(xi) = yi;
%     d_values air min(:,1)=a_g;
% %gap filling grape E
%     g_g=d_values_gra_avg(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,3)=g_g;
%     g_g=d_values_gra_max(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,3)=g_g;
%     g_g=d_values_gra_min(:,3);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,3)=g_g;
% %gap filling grape D
%     g_g=d_values_gra_avg(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,5)=g_g;
%     g_g=d_values_gra_max(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,5)=g_g;
%     g_g=d_values_gra_min(:,5);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_min(:,5)=g_g;
% %gap filling grape F
%     g_g=d_values_gra_avg(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_avg(:,6)=g_g;
%     g_g=d_values_gra_max(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);
%     D = setdiff(x, xi);
%     yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%     g_g(xi) = yi;
%     d_values_gra_max(:,6)=g_g;
%     g_g=d_values_gra_min(:,6);
%     xi = find(isnan(g_g));
%     x = 1:length(g_g);

```

```

%       D = setdiff(x, xi);
%       yi = interp1(D, g_g(D), xi, 'linear', 'extrap');
%       g_g(xi) = yi;
%       d_values gra min(:,6)=g g;
%       %gap filling radiazione H
%       r_g=d_values_rad_sum(:,2);
%       xi = find(isnan(r_g));
%       x = 1:length(r_g);
%       D = setdiff(x, xi);
%       yi = interp1(D, r_g(D), xi, 'linear', 'extrap');
%       r_g(xi) = yi;
%       d_values_rad_sum(:,2)=r_g;
%       %%%%%%%%%%%
%       %tolgo la g e tutti i suoi parametri li imposto come la b che ha le
stesse due tipologie di def e gem
%       d_values_air_avg(:,8)=d_values_air_avg(:,7);
%       d_values_air_max(:,8)=d_values_air_max(:,7);
%       d_values_air_min(:,8)=d_values_air_min(:,7);
%       d_values_gra_avg(:,8)=d_values_gra_avg(:,7);
%       d_values_gra_max(:,8)=d_values_gra_max(:,7);
%       d_values_gra_min(:,8)=d_values_gra_min(:,7);
%       d_values_rad_sum(:,8)=d_values_rad_sum(:,7);
%end

% giorni medi (orari)
[h_values_air_avg,h_row_air_avg,h_column_air_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_air,'nanmean');
[h_values_gra_avg,h_row_gra_avg,h_column_gra_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_grap,'nanmean');
[h_values_rad_avg,h_row_rad_avg,h_column_rad_avg]=pivottable(p_ora,[p_def
p_gem p_dir],p_rad,'nanmean');
if styear==2008
    %la I la imposto come la M
    h_values_air_avg(:,8)=h_values_air_avg(:,7);
    h_values_gra_avg(:,8)=h_values_gra_avg(:,7);
    h_values_rad_avg(:,8)=h_values_rad_avg(:,7);
end
%orari solo def e gem quindi 4 serie
h_values_air_avg_tesi=
horzcat(nanmean(h_values_air_avg(:,1:2),2),nanmean(h_values_air_avg(:,3:4),
2),nanmean(h_values_air_avg(:,5:6),2),nanmean(h_values_air_avg(:,7:8),2));
h_values_gra_avg_tesi=
horzcat(nanmean(h_values_gra_avg(:,1:2),2),nanmean(h_values_gra_avg(:,3:4),
2),nanmean(h_values_gra_avg(:,5:6),2),nanmean(h_values_gra_avg(:,7:8),2));
h_values_rad_avg_tesi=
horzcat(nanmean(h_values_rad_avg(:,1:2),2),nanmean(h_values_rad_avg(:,3:4),
2),nanmean(h_values_rad_avg(:,5:6),2),nanmean(h_values_rad_avg(:,7:8),2));

%giornalieri solo def e gem quindi 4 serie
t_values_air_avg=
horzcat(mean(d_values_air_avg(:,1:2),2),mean(d_values_air_avg(:,3:4),2),mea
n(d_values_air_avg(:,5:6),2),mean(d_values_air_avg(:,7:8),2));
t_values_air_max=
horzcat(mean(d_values_air_max(:,1:2),2),mean(d_values_air_max(:,3:4),2),mea
n(d_values_air_max(:,5:6),2),mean(d_values_air_max(:,7:8),2));
t_values_air_min=
horzcat(mean(d_values_air_min(:,1:2),2),mean(d_values_air_min(:,3:4),2),mea
n(d_values_air_min(:,5:6),2),mean(d_values_air_min(:,7:8),2));
t_values_gra_avg=
horzcat(mean(d_values_gra_avg(:,1:2),2),mean(d_values_gra_avg(:,3:4),2),mea
n(d_values_gra_avg(:,5:6),2),mean(d_values_gra_avg(:,7:8),2));
t_values_gra_max=
horzcat(mean(d_values_gra_max(:,1:2),2),mean(d_values_gra_max(:,3:4),2),mea
n(d_values_gra_max(:,5:6),2),mean(d_values_gra_max(:,7:8),2));
t_values_gra_min=
horzcat(mean(d_values_gra_min(:,1:2),2),mean(d_values_gra_min(:,3:4),2),mea
n(d_values_gra_min(:,5:6),2),mean(d_values_gra_min(:,7:8),2));
t_values_rad_sum=
horzcat(mean(d_values_rad_sum(:,1:2),2),mean(d_values_rad_sum(:,3:4),2),mea
n(d_values_rad_sum(:,5:6),2),mean(d_values_rad_sum(:,7:8),2));

%indici per le tesi def e gem (4 serie nera-verde-rossa-gialla)
air_avg_t=nanmean(t_values_air_avg);

```

```

air_max_t=nanmean(t_values_air_max);
air_min_t=nanmean(t_values_air_min);
air_avg_sum_t=nansum(t_values_air_avg);
air_max_sum_t=nansum(t_values_air_max);
air_min_sum_t=nansum(t_values_air_min);
gra_avg_t=nanmean(t_values_gra_avg);
gra_max_t=nanmean(t_values_gra_max);
gra_min_t=nanmean(t_values_gra_min);
gra_avg_sum_t=nansum(t_values_gra_avg);
gra_max_sum_t=nansum(t_values_gra_max);
gra_min_sum_t=nansum(t_values_gra_min);
rad_sum_t=nansum(t_values_rad_sum);

% IW air
iw_air_nero=sum(t_values_air_avg((t_values_air_avg(:,1)>10),1)-10);
iw_air_verde=sum(t_values_air_avg((t_values_air_avg(:,2)>10),2)-10);
iw_air_rosso=sum(t_values_air_avg((t_values_air_avg(:,3)>10),3)-10);
iw_air_giallo=sum(t_values_air_avg((t_values_air_avg(:,4)>10),4)-10);
iw_air=horzcat(iw_air_nero,iw_air_verde,iw_air_rosso,iw_air_giallo);
% IW grape
iw_gra_nero=sum(t_values_gra_avg((t_values_gra_avg(:,1)>10),1)-10);
iw_gra_verde=sum(t_values_gra_avg((t_values_gra_avg(:,2)>10),2)-10);
iw_gra_rosso=sum(t_values_gra_avg((t_values_gra_avg(:,3)>10),3)-10);
iw_gra_giallo=sum(t_values_gra_avg((t_values_gra_avg(:,4)>10),4)-10);
iw_gra=horzcat(iw_gra_nero,iw_gra_verde,iw_gra_rosso,iw_gra_giallo);
% IH air
ih_air_nero=sum(((t_values_air_avg(:,1)-10)+(t_values_air_max(:,1)-10))/2);
ih_air_verde=sum(((t_values_air_avg(:,2)-10)+(t_values_air_max(:,2)-10))/2);
ih_air_rosso=sum(((t_values_air_avg(:,3)-10)+(t_values_air_max(:,3)-10))/2);
ih_air_giallo=sum(((t_values_air_avg(:,4)-10)+(t_values_air_max(:,4)-10))/2);
ih_air=horzcat(ih_air_nero,ih_air_verde,ih_air_rosso,ih_air_giallo);
% IH grape
ih_gra_nero=sum(((t_values_gra_avg(:,1)-10)+(t_values_gra_max(:,1)-10))/2);
ih_gra_verde=sum(((t_values_gra_avg(:,2)-10)+(t_values_gra_max(:,2)-10))/2);
ih_gra_rosso=sum(((t_values_gra_avg(:,3)-10)+(t_values_gra_max(:,3)-10))/2);
ih_gra_giallo=sum(((t_values_gra_avg(:,4)-10)+(t_values_gra_max(:,4)-10))/2);
ih_gra=horzcat(ih_gra_nero,ih_gra_verde,ih_gra_rosso,ih_gra_giallo);
% SET air
set_air=sum(t_values_air_max-t_values_air_min);
% SET grape
set_gra=sum(t_values_gra_max-t_values_gra_min);
% IG air
% IG grape
% IFS air
freddi_air_nero=length(find(t_values_air_min(:,1)<10));
freddi_air_verde=length(find(t_values_air_min(:,2)<10));
freddi_air_rosso=length(find(t_values_air_min(:,3)<10));
freddi_air_giallo=length(find(t_values_air_min(:,4)<10));
ifs_air_nero=sum(t_values_air_max(:,1)-t_values_air_min(:,1))
*(freddi_air_nero+1);
ifs_air_verde=sum(t_values_air_max(:,2)-t_values_air_min(:,2))
*(freddi_air_verde+1);
ifs_air_rosso=sum(t_values_air_max(:,3)-t_values_air_min(:,3))
*(freddi_air_rosso+1);
ifs_air_giallo=sum(t_values_air_max(:,4)-t_values_air_min(:,4))
*(freddi_air_giallo+1);
ifs_air=horzcat(ifs_air_nero,ifs_air_verde,ifs_air_rosso,ifs_air_giallo);

% IFS grape
freddi_gra_nero=length(find(t_values_gra_min(:,1)<10));
freddi_gra_verde=length(find(t_values_gra_min(:,2)<10));
freddi_gra_rosso=length(find(t_values_gra_min(:,3)<10));
freddi_gra_giallo=length(find(t_values_gra_min(:,4)<10));
ifs_gra_nero=sum(t_values_gra_max(:,1)-t_values_gra_min(:,1))
*(freddi_gra_nero+1);
ifs_gra_verde=sum(t_values_gra_max(:,2)-t_values_gra_min(:,2))
*(freddi_gra_verde+1);

```

```

ifs_gra_rosso=sum(t_values_gra_max(:,3)-t_values_gra_min(:,3))
*(freddi_gra_rosso+1);
ifs_gra_giallo=sum(t_values_gra_max(:,4)-t_values_gra_min(:,4))
*(freddi_gra_giallo+1);
ifs_gra=horzcat(ifs_gra_nero,ifs_gra_verde,ifs_gra_rosso,ifs_gra_giallo);
% CoolNI air
CoolNI_air=air_min_t;
% CoolNI grape
CoolNI_grap=gra_min_t;
% Morning(03-10) Index air
Morning_03_10_Index_air=nanmean(h_values_air_avg_tesi(4:11,:));
% Morning(03-10) Index grape
Morning_03_10_Index_grape=nanmean(h_values_gra_avg_tesi(4:11,:));
% Diurnal(11-18) Index air
Diurnal_11_18_Index_air=nanmean(h_values_air_avg_tesi(12:19,:));
% Diurnal(11-18) Index grape
Diurnal_11_18_Index_grape=nanmean(h_values_gra_avg_tesi(12:19,:));
% Night(19-02) Index air
Night_19_02_Index_air=
nanmean(vertcat(h_values_air_avg_tesi(20:24,:),h_values_air_avg_tesi(1:3,:))
);
% Night(19-02) Index grape
Night_19_02_Index_grape=
nanmean(vertcat(h_values_gra_avg_tesi(20:24,:),h_values_gra_avg_tesi(1:3,:))
);

MDN_index_air=
vertcat(Morning_03_10_Index_air,Diurnal_11_18_Index_air,Night_19_02_Index_a
ir);
MDN_index_gra=
vertcat(Morning_03_10_Index_grape,Diurnal_11_18_Index_grape,Night_19_02_Ind
ex_grape);

% dMorning 5 10 air=[2.9376 2.9804 2.9436 2.9687];
% dMorning 5 10 grape=[2.1089 2.1664 2.2037 2.8037];
% dTwilight 15 20 air=[-1.8499 -1.8052 -1.8686 -1.9616];
% dTwilight 15 20 grape=[-2.2782 -2.249 -2.4021 -2.4053];
% d_index_air=vertcat(dMorning_5_10_air,dTwilight_15_20_air);
% d_index_gra=vertcat(dMorning_5_10_grape,dTwilight_15_20_grape);

dMorning_rate_5_10_air=(h_values_air_avg_tesi(11,:)-
h_values_air_avg_tesi(6,:))/6;
dMorning_rate_5_10_grape=(h_values_gra_avg_tesi(11,:)-
h_values_gra_avg_tesi(6,:))/6;
dTwilight_rate_15_20_air=(h_values_air_avg_tesi(21,:)-
h_values_air_avg_tesi(16,:))/6;
dTwilight_rate_15_20_grape=(h_values_gra_avg_tesi(21,:)-
h_values_gra_avg_tesi(16,:))/6;
d_index_rate_air=vertcat(dMorning_rate_5_10_air,dTwilight_rate_15_20_air);
d_index_rate_gra=
vertcat(dMorning_rate_5_10_grape,dTwilight_rate_15_20_grape);
% per la master
p_doy_m=N.Doy(mast);
p_air_avg_m=N.Tempmedia(mast);
p_air_max_m=N.Tmax(mast);
p_air_min_m=N.Tmin(mast);
p_rad_avg_m=N.RadGmedia(mast);
p_666=N.defogliazione(mast);
[d_values_master_air_avg,d_row_master_air_avg,d_column_master_air_avg]=
pivottable(p_doy_m,p_666,p_air_avg_m,'nanmean');
[d_values_master_air_max,d_row_master_air_max,d_column_master_air_max]=
pivottable(p_doy_m,p_666,p_air_max_m,'nanmax');
[d_values_master_air_min,d_row_master_air_min,d_column_master_air_min]=
pivottable(p_doy_m,p_666,p_air_min_m,'nanmin');
[d_values_master_rad_sum,d_row_rad_sum,d_column_rad_sum]=
pivottable(p_doy_m,p_666,p_rad_avg_m,'nansum');

[h_values_ms,h_row_ms,h_column_ms]=pivottable(p_ora,[p_666
p_666],p_air_avg_m,'nanmean');

% Master_IW
master_iw=sum(d_values_master_air_avg((d_values_master_air_avg>10),1)-10);
% Master_IH

```

```

master_ih=sum(((d_values_master_air_avg-10)+(d_values_master_air_max-10))
/2);
% Master_SET
master_set=sum(d_values_master_air_max-d_values_master_air_min);
% Master_IG
% Master_IFS
freddi_master=length(find(d_values_master_air_min<10));
master_ifs=sum(d_values_master_air_max-d_values_master_air_min)
*(freddi_master+1);
% Master_CoolNI
master_cooln=nanmean(d_values_master_air_min);
%matrice indici totale
matrix_index_slave_air=
vertcat(air_avg_t,air_max_t,air_min_t,air_avg_sum_t,air_max_sum_t,air_min_s
um_t,iw_air,ih_air,set_air,ifs_air,CoolNI_air,MDN_index_air,d_index_rate_ai
r);
matrix_index_slave_gra=
vertcat(rad_sum_t,gra_avg_t,gra_max_t,gra_min_t,gra_avg_sum_t,gra_max_sum_t
,gra_min_sum_t,iw_gra,ih_gra,set_gra,ifs_gra,CoolNI_grap,MDN_index_gra,d_in
dex_rate_gra);
matrix_index_master=
horzcat(master_iw,master_ih,master_set,master_ifs,master_cooln);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
figure('name','pivot tgrape alta vigoria')
hold on
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_gra_avg,h_values_gra_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot tair alta vigoria')
hold on
plot(h_row_air_avg,h_values_air_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_air_avg,h_values_air_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
figure('name','pivot RAD alta vigoria')
hold on
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,1),'k-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,2),'g-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,3),'r-','LineWidth',2)
plot(h_row_rad_avg,h_values_rad_avg_tesi(:,4),'y-','LineWidth',2)
legend('A0-C0 non def + gemme 1(+potato)','A0-C1 non def + gemme 3(-
potato)','A1-C0 def + gemme 1(+potato)','A1-C1 def + gemme 3(-potato)',4)
end

```

Allegato 1.10 – Script R Analisi statistica GLM

Analisi statistica campione Tuscania brolio

IBIMET CNR, Istituto di Biometeorologia di Firenze -CNR

Matese.A., Genesio L., Crisci A.

Temperatura Grappolo brolio

Temperatura Grappolo Mattina

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.03	-2.77	-0.29	2.45	12.07

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.8375834	0.1249705	158.7381	< 2.2e-16	***
as.factor(def)[T.1]	0.2404122	0.0851675	2.8228	0.004773	**
as.factor(gemme)[T.3]	-0.0994415	0.0851675	-1.1676	0.243006	
as.factor(dirad)[T.1]	-0.4224085	0.0851675	-4.9597	7.215e-07	***
AnnoF[T.A2009]	0.7921477	0.1043168	7.5937	3.488e-14	***
AnnoF[T.A2010]	0.0012369	0.1043168	0.0119	0.990540	
as.factor(Mese)[T.7]	1.3129257	0.1142102	11.4957	< 2.2e-16	***
as.factor(Mese)[T.8]	2.7632262	0.1132815	24.3926	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.3679966	0.1365208	-2.6955	0.007043	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

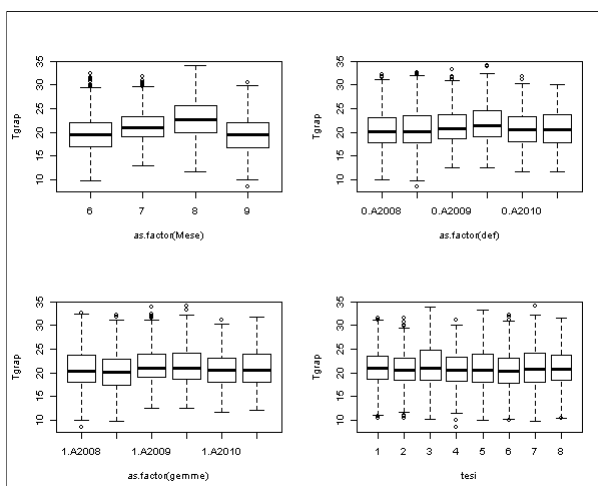
(Dispersion parameter for gaussian family taken to be 13.53504)

- Null deviance: 113429.7095 on 7463 degrees of freedom.
- Residual deviance: 100903.7396 on 7455 degrees of freedom.

AIC: 40639

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.813287579082058



Temperatura Grappolo Pieno giorno

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.86	-3.16	0.12	3.23	13.98

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	26.952550	0.088702	303.8546	< 2.2e-16	***
as.factor(def)[T.1]	0.626291	0.061013	10.2649	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.401544	0.061013	-6.5813	4.771e-11	***
as.factor(dirad)[T.1]	-0.203146	0.061013	-3.3296	0.0008713	***
AnnoF[T.A2009]	0.251382	0.074730	3.3639	0.0007700	***
AnnoF[T.A2010]	-0.646639	0.074710	-8.6554	< 2.2e-16	***
as.factor(Mese)[T.7]	3.439895	0.078981	43.5533	< 2.2e-16	***
as.factor(Mese)[T.8]	5.938208	0.081194	73.1359	< 2.2e-16	***
as.factor(Mese)[T.9]	2.401841	0.104111	23.0699	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

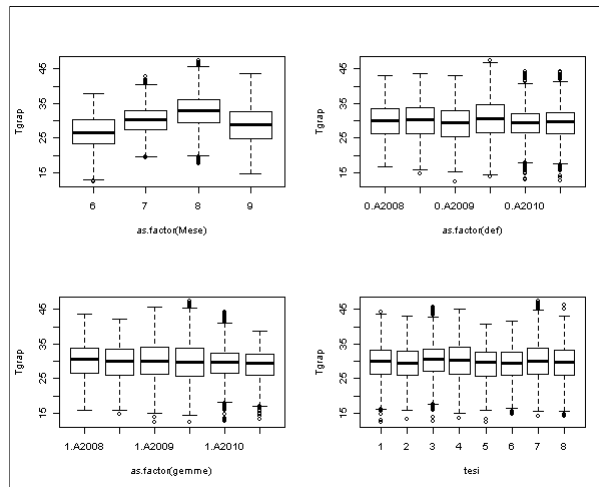
(Dispersion parameter for gaussian family taken to be 19.94551)

- Null deviance: 542433.9092 on 21431 degrees of freedom.
- Residual deviance: 427292.5778 on 21423 degrees of freedom.

AIC: 124978

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.995357037039312



Temperatura Cross Grappolo Pieno giorno

Call: glm(formula = Tgrap ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.94	-3.16	0.10	3.15	15.08

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	10.3363867	0.7468131	13.8407	< 2.2e-16	***
as.factor(gemme)[T.3]	3.5909449	0.8406753	4.2715	1.950e-05	***
as.factor(def)[T.1]	-1.9334809	0.8415030	-2.2977	0.021591	*
Rad	0.0358928	0.0016632	21.5803	< 2.2e-16	***
AnnoF[T.A2009]	0.5580496	0.0736347	7.5786	3.634e-14	***
AnnoF[T.A2010]	0.4572332	0.0813628	5.6197	1.937e-08	***
as.factor(Mese)[T.7]	3.7264631	0.0777708	47.9159	< 2.2e-16	***
as.factor(Mese)[T.8]	6.4336690	0.0814275	79.0110	< 2.2e-16	***
as.factor(Mese)[T.9]	2.7646359	0.1025138	26.9684	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.1715167	0.1200995	-1.4281	0.153272	
as.factor(def)[T.1]:Rad	0.0057663	0.0019042	3.0282	0.002463	**
as.factor(gemme)[T.3]:Rad	-0.0084361	0.0019285	-4.3745	1.223e-05	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 18.99135)

- Null deviance: 542433.9092 on 21431 degrees of freedom.
- Residual deviance: 406794.7191 on 21420 degrees of freedom.

AIC: 123931

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.998215855524105

Temperatura Grappolo Sera

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.5508	-3.0919	0.0041	3.1080	14.9776

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	24.903930	0.153132	162.6301	< 2.2e-16	***
as.factor(def)[T.1]	0.174106	0.104292	1.6694	0.09508	.
as.factor(gemme)[T.3]	-0.586818	0.104292	-5.6267	1.904e-08	***
as.factor(dirad)[T.1]	-0.016616	0.104292	-0.1593	0.87342	
AnnoF[T.A2009]	0.710616	0.127733	5.5633	2.739e-08	***
AnnoF[T.A2010]	0.165631	0.127746	1.2966	0.19482	
as.factor(Mese)[T.7]	3.876472	0.138210	28.0477	< 2.2e-16	***
as.factor(Mese)[T.8]	5.508687	0.140484	39.2123	< 2.2e-16	***
as.factor(Mese)[T.9]	3.104863	0.167435	18.5437	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 20.35887)

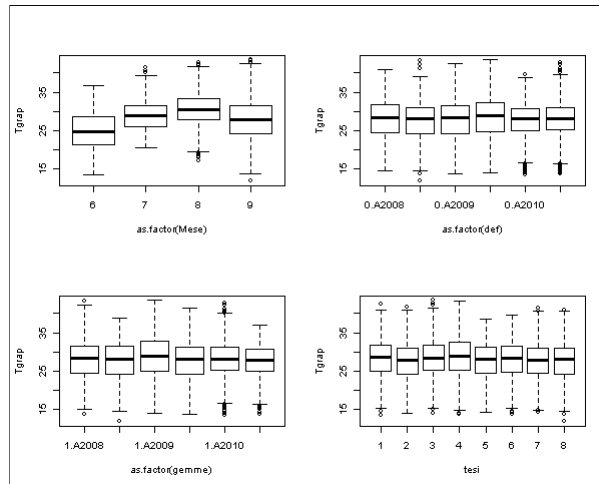
• Null deviance: 186713.6392 on 7486 degrees of freedom.

• Residual deviance: 152243.5988 on 7478 degrees of freedom.

AIC: 43820

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.989988147433551



Temperatura Grappolo Notte

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.74	-2.28	-0.25	2.07	18.23

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	18.4529688	0.0598350	308.3975	< 2.2e-16	***
as.factor(def)[T.1]	0.0216344	0.0399207	0.5419	0.5879	
as.factor(gemme)[T.3]	-0.2544214	0.0399207	-6.3732	1.884e-10	***

as.factor(dirad)[T.1]	0.0075563	0.0399207	0.1893	0.8499	
AnnoF[T.A2009]	1.1685946	0.0488977	23.8988	< 2.2e-16	***
AnnoF[T.A2010]	-0.0142868	0.0489081	-0.2921	0.7702	
as.factor(Mese)[T.7]	2.5688417	0.0552136	46.5255	< 2.2e-16	***
as.factor(Mese)[T.8]	2.8749999	0.0538479	53.3912	< 2.2e-16	***
as.factor(Mese)[T.9]	0.9205250	0.0631174	14.5843	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

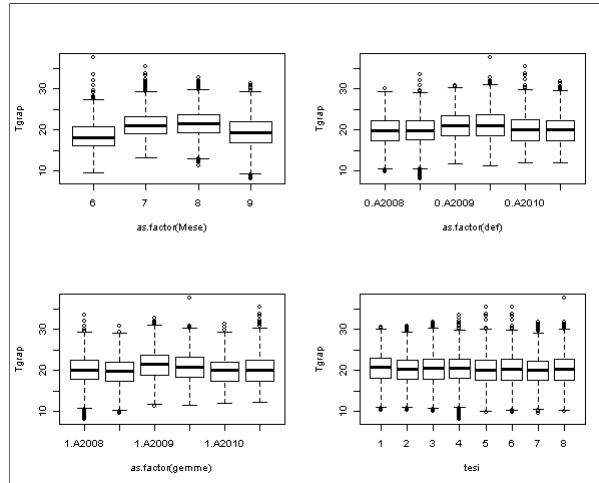
(Dispersion parameter for gaussian family taken to be 9.830097)

- Null deviance: 286260.9919 on 24672 degrees of freedom.
- Residual deviance: 242449.5103 on 24664 degrees of freedom.

AIC: 126419

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.830640242068615



Radiazione Grappolo Mattina

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-370	-33	13	41	127

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	413.4159	1.9730	209.5360	< 2.2e-16	***
as.factor(def)[T.1]	-2.2015	1.3446	-1.6373	0.1016162	
as.factor(gemme)[T.3]	-4.7672	1.3446	-3.5454	0.0003944	***
as.factor(dirad)[T.1]	1.8397	1.3446	1.3682	0.1712896	
AnnoF[T.A2009]	-12.9457	1.6469	-7.8605	4.363e-15	***
AnnoF[T.A2010]	-23.1291	1.6469	-14.0437	< 2.2e-16	***
as.factor(Mese)[T.7]	-17.6884	1.8031	-9.8099	< 2.2e-16	***
as.factor(Mese)[T.8]	2.2699	1.7885	1.2692	0.2044224	
as.factor(Mese)[T.9]	-25.1218	2.1554	-11.6555	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

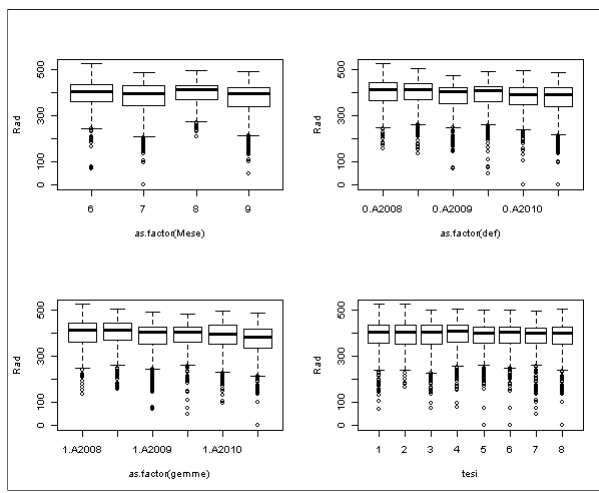
(Dispersion parameter for gaussian family taken to be 3373.663)

- Null deviance: 26781259.4948 on 7463 degrees of freedom.
- Residual deviance: 25150660.3057 on 7455 degrees of freedom.

AIC: 81829

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiazione Grappolo Pieno giorno

Call: `glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-182.6	-16.8	-1.9	17.5	98.7

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	459.70987	0.54955	836.5227	<2e-16	***
as.factor(def)[T.1]	3.90797	0.37800	10.3385	<2e-16	***
as.factor(gemme)[T.3]	-6.64234	0.37800	-17.5723	<2e-16	***
as.factor(dirad)[T.1]	0.22500	0.37800	0.5952	0.5517	
AnnoF[T.A2009]	-8.80987	0.46299	-19.0283	<2e-16	***
AnnoF[T.A2010]	-32.52705	0.46286	-70.2742	<2e-16	***
as.factor(Mese)[T.7]	-8.70228	0.48932	-17.7843	<2e-16	***
as.factor(Mese)[T.8]	-15.25704	0.50303	-30.3301	<2e-16	***
as.factor(Mese)[T.9]	-11.07925	0.64502	-17.1767	<2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 765.5775)

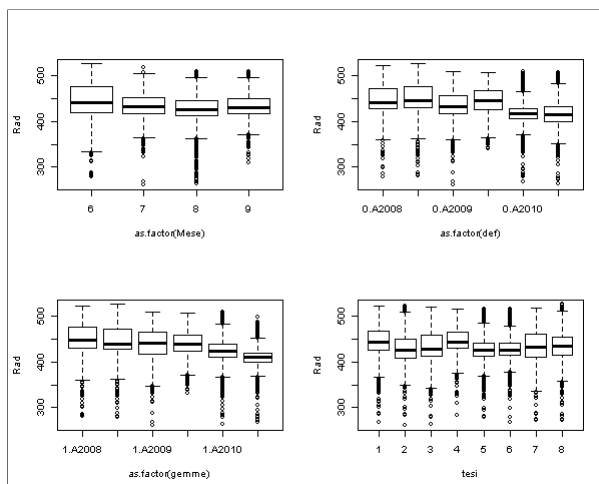
• Null deviance: 21520059.1578 on 21431 degrees of freedom.

• Residual deviance: 16400967.3276 on 21423 degrees of freedom.

AIC: 203154

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiazione Cross Grappolo Pieno giorno

Call: `glm(formula = Rad ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Tgrap + as.factor(gemme) * Tgrap + AnnoF + as.factor(Mese), family =`

gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-169.5	-16.4	-1.8	17.5	101.1

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	412.459379	1.993456	206.9067	< 2.2e-16	***
as.factor(gemme)[T.3]	21.176666	2.228280	9.5036	< 2.2e-16	***
as.factor(def)[T.1]	-2.769048	2.270925	-1.2193	0.22273	
Tgrap	1.747915	0.066906	26.1250	< 2.2e-16	***
AnnoF[T.A2009]	-9.134590	0.450716	-20.2668	< 2.2e-16	***
AnnoF[T.A2010]	-31.698072	0.450640	-70.3401	< 2.2e-16	***
as.factor(Mese)[T.7]	-13.207888	0.496244	-26.6157	< 2.2e-16	***
as.factor(Mese)[T.8]	-23.255863	0.546549	-42.5503	< 2.2e-16	***
as.factor(Mese)[T.9]	-14.579190	0.634935	-22.9617	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	3.517399	0.736811	4.7738	1.820e-06	***
as.factor(def)[T.1]:Tgrap	0.135602	0.073577	1.8430	0.06535	.
as.factor(gemme)[T.3]:Tgrap	-0.972523	0.073356	-13.2576	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 723.1125)

- Null deviance: 21520059.1578 on 21431 degrees of freedom.
- Residual deviance: 15489069.6758 on 21420 degrees of freedom.

AIC: 201934

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1

Radiation Grappolo Sera

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-263	-32	5	40	140

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	395.363687	1.889261	209.2689	< 2.2e-16	***
as.factor(def)[T.1]	3.746665	1.286701	2.9118	0.003604	**
as.factor(gemme)[T.3]	-0.098081	1.286701	-0.0762	0.939241	
as.factor(dirad)[T.1]	7.475624	1.286701	5.8099	6.507e-09	***
AnnoF[T.A2009]	-25.751243	1.575901	-16.3406	< 2.2e-16	***
AnnoF[T.A2010]	-18.238632	1.576059	-11.5723	< 2.2e-16	***
as.factor(Mese)[T.7]	-10.651164	1.705154	-6.2465	4.429e-10	***
as.factor(Mese)[T.8]	-28.389874	1.733208	-16.3800	< 2.2e-16	***
as.factor(Mese)[T.9]	-1.961249	2.065716	-0.9494	0.342434	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

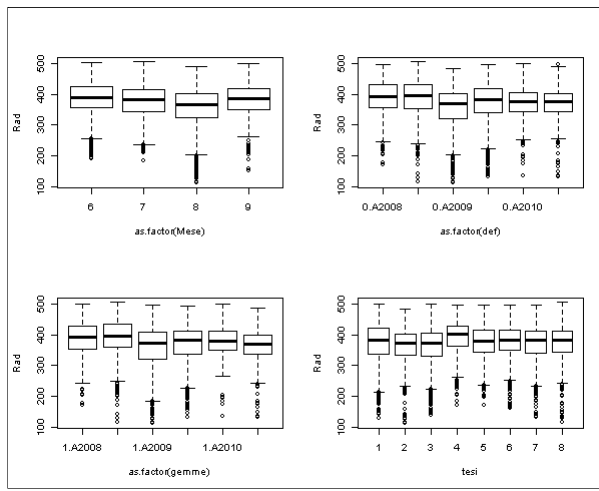
(Dispersion parameter for gaussian family taken to be 3098.87)

- Null deviance: 25150308.9245 on 7486 degrees of freedom.
- Residual deviance: 23173350.567 on 7478 degrees of freedom.

AIC: 81445

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Temperatura Aria Canopy Mattina

Call: `glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.86	-3.20	-0.36	2.94	13.93

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.279314	0.140297	144.5461	< 2.2e-16	***
as.factor(def)[T.1]	-0.037585	0.095612	-0.3931	0.69426	
as.factor(gemme)[T.3]	-0.437374	0.095612	-4.5745	4.852e-06	***
as.factor(dirad)[T.1]	-0.448567	0.095612	-4.6915	2.760e-06	***
AnnoF[T.A2009]	0.853391	0.117110	7.2871	3.492e-13	***
AnnoF[T.A2010]	-0.232588	0.117110	-1.9861	0.04706	*
as.factor(Mese)[T.7]	2.578945	0.128217	20.1140	< 2.2e-16	***
as.factor(Mese)[T.8]	4.451447	0.127174	35.0028	< 2.2e-16	***
as.factor(Mese)[T.9]	0.753340	0.153263	4.9153	9.053e-07	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 17.05841)

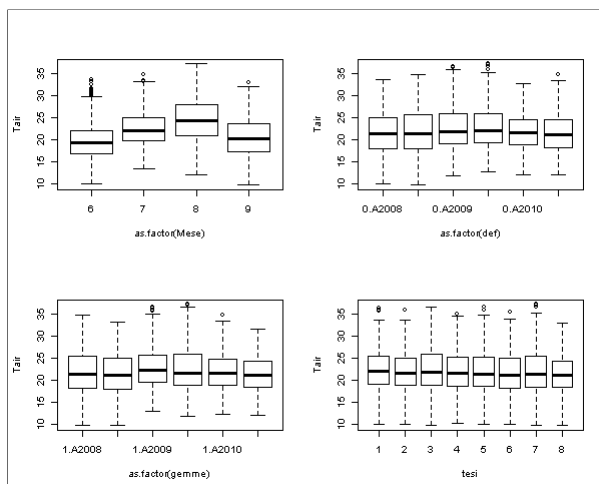
• Null deviance: 153126.5432 on 7463 degrees of freedom.

• Residual deviance: 127170.4584 on 7455 degrees of freedom.

AIC: 42366

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.969115577999602



Temperatura Aria Canopy Pieno giorno

Call: `glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset =`

((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.57	-2.94	0.24	3.25	11.92

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	27.693147	0.087133	317.8262	< 2.2e-16	***
as.factor(def)[T.1]	0.110793	0.059934	1.8486	0.0645296	.
as.factor(gemme)[T.3]	-0.280924	0.059934	-4.6873	2.786e-06	***
as.factor(dirad)[T.1]	-0.195693	0.059934	-3.2652	0.0010957	**
AnnoF[T.A2009]	0.265087	0.073409	3.6111	0.0003056	***
AnnoF[T.A2010]	-0.743082	0.073388	-10.1254	< 2.2e-16	***
as.factor(Mese)[T.7]	4.701591	0.077584	60.5999	< 2.2e-16	***
as.factor(Mese)[T.8]	5.572281	0.079758	69.8650	< 2.2e-16	***
as.factor(Mese)[T.9]	1.461912	0.102270	14.2947	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

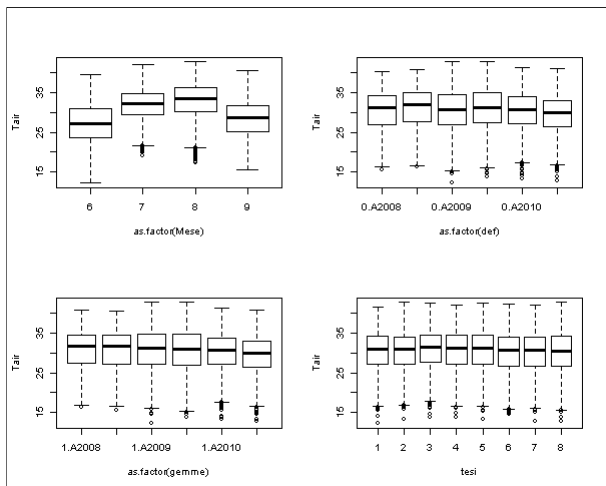
(Dispersion parameter for gaussian family taken to be 19.24608)

- Null deviance: 536233.1345 on 21431 degrees of freedom.
- Residual deviance: 412308.7033 on 21423 degrees of freedom.

AIC: 124213

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99691811844439



Temperatura Aria Crossing Canopy Pieno giorno

Call: glm(formula = Tair ~ as.factor(gemme) * as.factor(def) + as.factor(dirad) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.09	-2.94	0.21	3.13	12.35

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.3342460	0.7439200	25.9897	< 2.2e-16	***
as.factor(gemme)[T.3]	-1.4100524	0.8330089	-1.6927	0.09052	.
as.factor(def)[T.1]	-5.7995439	0.8384328	-6.9171	4.739e-12	***
as.factor(dirad)[T.1]	-0.2500962	0.0593302	-4.2153	2.504e-05	***
Rad	0.0174472	0.0016515	10.5647	< 2.2e-16	***
AnnoF[T.A2009]	0.4680698	0.0729607	6.4154	1.434e-10	***
AnnoF[T.A2010]	0.1343940	0.0806177	1.6671	0.09552	.
as.factor(Mese)[T.7]	4.9425047	0.0770583	64.1398	< 2.2e-16	***
as.factor(Mese)[T.8]	5.9915803	0.0806814	74.2622	< 2.2e-16	***
as.factor(Mese)[T.9]	1.7753891	0.1015747	17.4787	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.5784167	0.1190057	-4.8604	1.180e-06	***
as.factor(def)[T.1]:Rad	0.0139675	0.0018974	7.3615	1.884e-13	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 18.64488)

- Null deviance: 536233.1345 on 21431 degrees of freedom.
- Residual deviance: 399354.652 on 21419 degrees of freedom.

AIC: 123537

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99831609648927

Temperatura Aria Canopy Sera

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.632	-3.058	0.043	3.098	13.796

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	25.315897	0.151135	167.5053	< 2.2e-16	***
as.factor(def)[T.1]	0.093255	0.102932	0.9060	0.3649707	
as.factor(gemme)[T.3]	-0.370186	0.102932	-3.5964	0.0003247	***
as.factor(dirad)[T.1]	-0.132229	0.102932	-1.2846	0.1989647	
AnnoF[T.A2009]	0.480240	0.126067	3.8094	0.0001404	***
AnnoF[T.A2010]	-0.070734	0.126080	-0.5610	0.5747946	
as.factor(Mese)[T.7]	4.455852	0.136407	32.6659	< 2.2e-16	***
as.factor(Mese)[T.8]	4.398258	0.138651	31.7218	< 2.2e-16	***
as.factor(Mese)[T.9]	1.537906	0.165251	9.3065	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

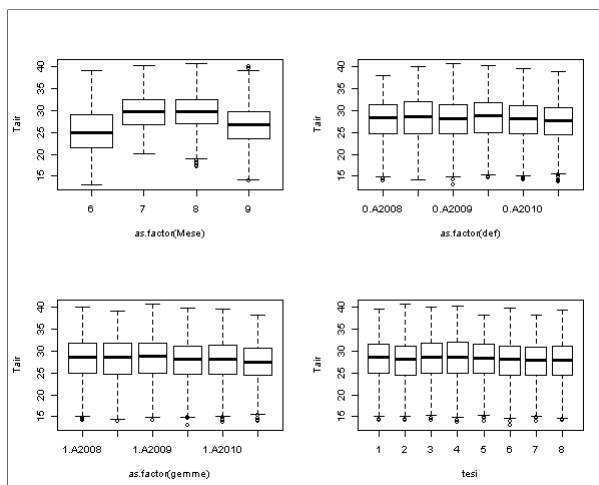
(Dispersion parameter for gaussian family taken to be 19.8312)

- Null deviance: 178303.1575 on 7486 degrees of freedom.
- Residual deviance: 148297.7247 on 7478 degrees of freedom.

AIC: 43624

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.981824322951395



Temperatura Aria Canopy Notte

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))

Deviance Residuals:

Min	1Q	Median	3Q	Max

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	18.804516	0.059224	317.5141	< 2.2e-16	***
as.factor(def)[T.1]	-0.010954	0.039513	-0.2772	0.7816028	
as.factor(gemme)[T.3]	-0.375168	0.039513	-9.4948	< 2.2e-16	***
as.factor(dirad)[T.1]	-0.173271	0.039513	-4.3851	1.164e-05	***
AnnoF[T.A2009]	0.916791	0.048399	18.9425	< 2.2e-16	***
AnnoF[T.A2010]	-0.173569	0.048409	-3.5855	0.0003371	***
as.factor(Mese)[T.7]	2.938624	0.054650	53.7717	< 2.2e-16	***
as.factor(Mese)[T.8]	2.738314	0.053298	51.3773	< 2.2e-16	***
as.factor(Mese)[T.9]	0.637432	0.062473	10.2033	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

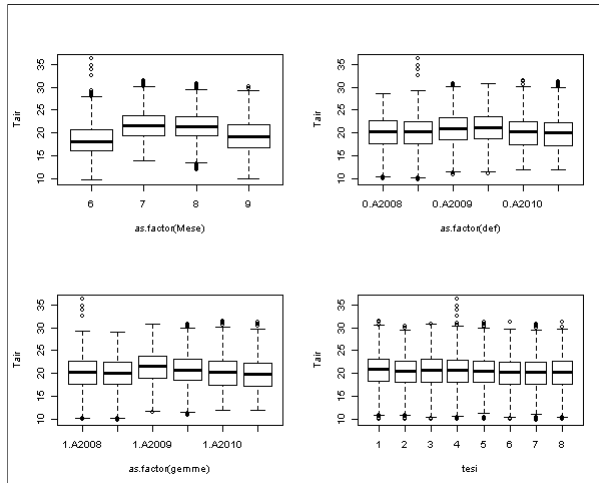
(Dispersion parameter for gaussian family taken to be 9.63042)

- Null deviance: 285159.7735 on 24672 degrees of freedom.
- Residual deviance: 237524.6602 on 24664 degrees of freedom.

AIC: 125912

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.854955386089027



Analisi indici giornalieri radiazione brolio

radiazione media : avg_rad brolio

•

Call: glm(formula = avg_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-44.790	-7.891	0.064	8.283	37.236

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	286.94970	0.75469	380.2226	< 2.2e-16	***
as.factor(def)[T.1]	2.12050	0.51305	4.1331	3.695e-05	***
as.factor(gemme)[T.3]	-2.31498	0.51305	-4.5122	6.709e-06	***
as.factor(dirad)[T.1]	1.63065	0.51305	3.1783	0.001499	**
AnnoF[T.A2009]	-10.26900	0.62836	-16.3427	< 2.2e-16	***
AnnoF[T.A2010]	-17.76073	0.62836	-28.2654	< 2.2e-16	***
as.factor(Mese)[T.7]	-11.20924	0.68231	-16.4285	< 2.2e-16	***
as.factor(Mese)[T.8]	-32.89100	0.68231	-48.2057	< 2.2e-16	***
as.factor(Mese)[T.9]	-50.07822	0.83997	-59.6190	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 167.4083)

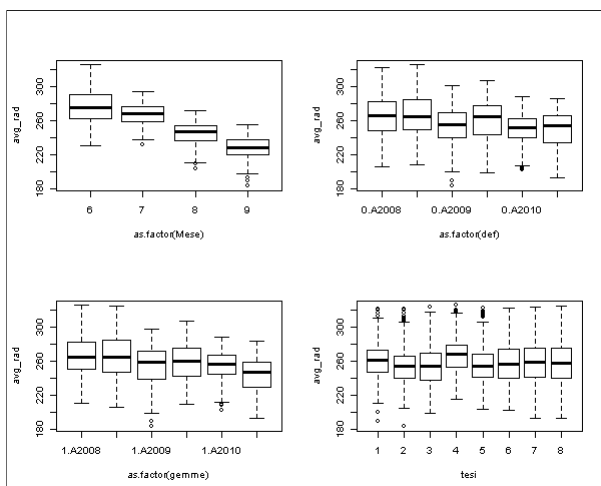
- Null deviance: 1352399.4162 on 2543 degrees of freedom.

• Residual deviance: 424380.0538 on 2535 degrees of freedom.

AIC: 20257

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



cumulata radiazione : sum_rad brolio

Call: glm(formula = sum_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1079.37	-189.29	0.54	198.85	1660.87

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6889.164	18.210	378.3230	< 2.2e-16	***
as.factor(def)[T.1]	51.911	12.379	4.1934	2.843e-05	***
as.factor(gemme)[T.3]	-56.578	12.379	-4.5704	5.101e-06	***
as.factor(dirad)[T.1]	40.155	12.379	3.2437	0.001195	**
AnnoF[T.A2009]	-247.984	15.161	-16.3562	< 2.2e-16	***
AnnoF[T.A2010]	-427.786	15.161	-28.2153	< 2.2e-16	***
as.factor(Mese)[T.7]	-270.884	16.463	-16.4539	< 2.2e-16	***
as.factor(Mese)[T.8]	-791.246	16.463	-48.0615	< 2.2e-16	***
as.factor(Mese)[T.9]	-1203.739	20.268	-59.3925	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 97465.06)

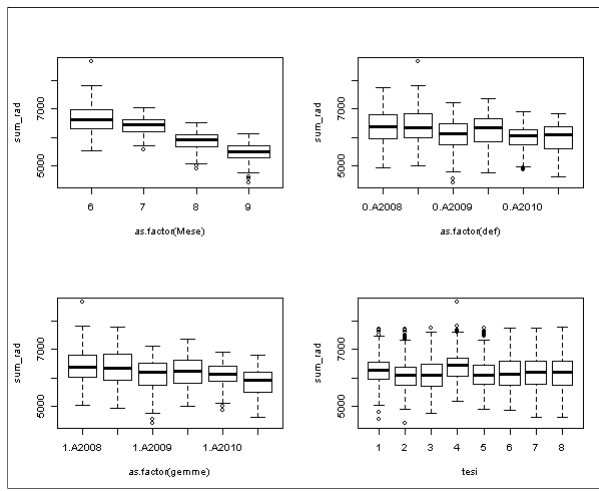
• Null deviance: 783631229.9957 on 2543 degrees of freedom.

• Residual deviance: 247073937.1971 on 2535 degrees of freedom.

AIC: 36454

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione radiazione: et_rad brolio

Call: `glm(formula = et_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-66.17	-16.33	0.58	15.36	64.04

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	486.53030	1.25502	387.6687	< 2.2e-16	***
as.factor(def)[T.1]	3.86505	0.85318	4.5302	6.167e-06	***
as.factor(gemme)[T.3]	-11.44332	0.85318	-13.4125	< 2.2e-16	***
as.factor(dirad)[T.1]	0.20202	0.85318	0.2368	0.8128463	
AnnoF[T.A2009]	-13.51530	1.04493	-12.9342	< 2.2e-16	***
AnnoF[T.A2010]	-34.22067	1.04493	-32.7493	< 2.2e-16	***
as.factor(Mese)[T.7]	-1.86048	1.13464	-1.6397	0.1011918	
as.factor(Mese)[T.8]	-5.75166	1.13464	-5.0691	4.285e-07	***
as.factor(Mese)[T.9]	-5.08686	1.39684	-3.6417	0.0002763	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

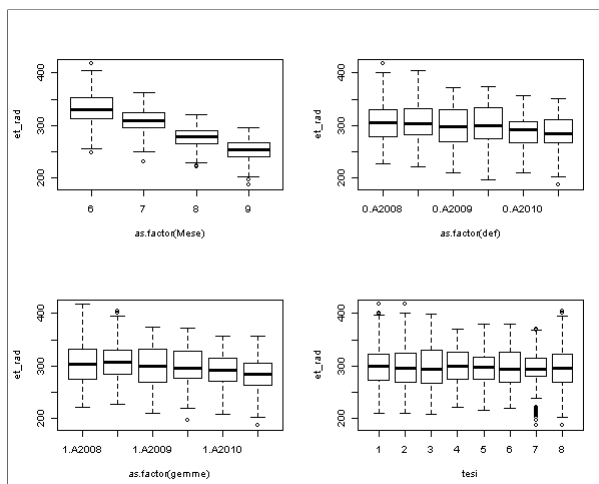
(Dispersion parameter for gaussian family taken to be 462.9559)

- Null deviance: 1784661.4866 on 2543 degrees of freedom.
- Residual deviance: 1173593.2552 on 2535 degrees of freedom.

AIC: 22845

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 3-10 radiazione: morning_03_10_index_rad brolio

Call: `glm(formula = morning_03_10_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-80.53	-15.10	-0.11	14.19	73.54

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	343.84366	1.27881	268.8783	< 2.2e-16	***
as.factor(def)[T.1]	-0.38517	0.86936	-0.4431	0.6577633	
as.factor(gemme)[T.3]	-2.86950	0.86936	-3.3007	0.0009777	***
as.factor(dirad)[T.1]	0.20739	0.86936	0.2386	0.8114681	
AnnoF[T.A2009]	-7.75788	1.06474	-7.2862	4.238e-13	***
AnnoF[T.A2010]	-21.00336	1.06474	-19.7263	< 2.2e-16	***
as.factor(Mese)[T.7]	-21.87254	1.15616	-18.9183	< 2.2e-16	***
as.factor(Mese)[T.8]	-55.12334	1.15616	-47.6782	< 2.2e-16	***
as.factor(Mese)[T.9]	-79.90826	1.42332	-56.1422	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

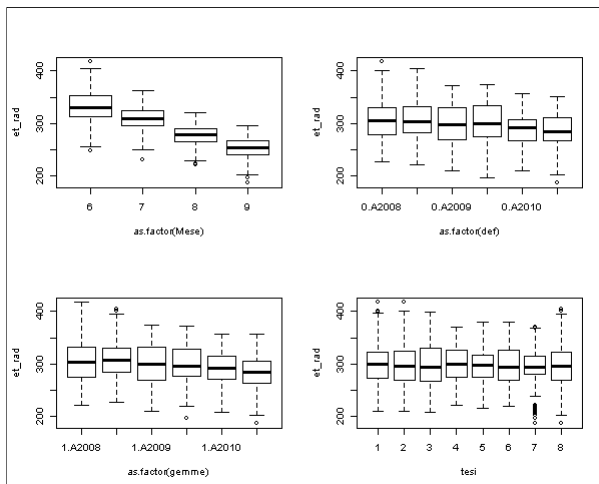
(Dispersion parameter for gaussian family taken to be 480.6754)

- Null deviance: 3419787.0276 on 2543 degrees of freedom.
- Residual deviance: 1218512.1791 on 2535 degrees of freedom.

AIC: 22940

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 11-18 radiazione: diurnal_11_18_index_rad brolio

Call: glm(formula = diurnal_11_18_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-78.39	-12.99	0.49	12.74	68.68

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	448.00633	1.20304	372.3946	< 2.2e-16	***
as.factor(def)[T.1]	4.62366	0.81785	5.6534	1.749e-08	***
as.factor(gemme)[T.3]	-5.93435	0.81785	-7.2561	5.272e-13	***
as.factor(dirad)[T.1]	1.26604	0.81785	1.5480	0.1217	
AnnoF[T.A2009]	-12.87069	1.00166	-12.8494	< 2.2e-16	***
AnnoF[T.A2010]	-31.30466	1.00166	-31.2529	< 2.2e-16	***
as.factor(Mese)[T.7]	-5.74420	1.08766	-5.2813	1.392e-07	***
as.factor(Mese)[T.8]	-12.94887	1.08766	-11.9053	< 2.2e-16	***
as.factor(Mese)[T.9]	-16.82934	1.33899	-12.5687	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

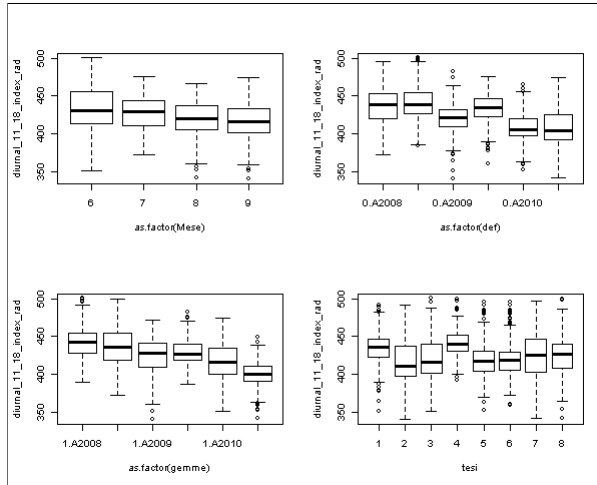
(Dispersion parameter for gaussian family taken to be 425.4054)

- Null deviance: 1629860.599 on 2543 degrees of freedom.
- Residual deviance: 1078402.6938 on 2535 degrees of freedom.

AIC: 22629

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 19-02 radiazione: night_19_02_index_rad brolio

•

Call: `glm(formula = night_19_02_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-51.3	-11.9	-0.8	12.1	202.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	69.29554	1.03108	67.2069	< 2.2e-16	***
as.factor(def)[T.1]	2.25037	0.70094	3.2105	0.001342	**
as.factor(gemme)[T.3]	1.73156	0.70094	2.4703	0.013565	*
as.factor(dirad)[T.1]	3.54589	0.70094	5.0587	4.523e-07	***
AnnoF[T.A2009]	-10.36946	0.85848	-12.0789	< 2.2e-16	***
AnnoF[T.A2010]	-1.16521	0.85848	-1.3573	0.174807	
as.factor(Mese)[T.7]	-6.24375	0.93219	-6.6980	2.594e-11	***
as.factor(Mese)[T.8]	-30.83354	0.93219	-33.0766	< 2.2e-16	***
as.factor(Mese)[T.9]	-53.72983	1.14759	-46.8195	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

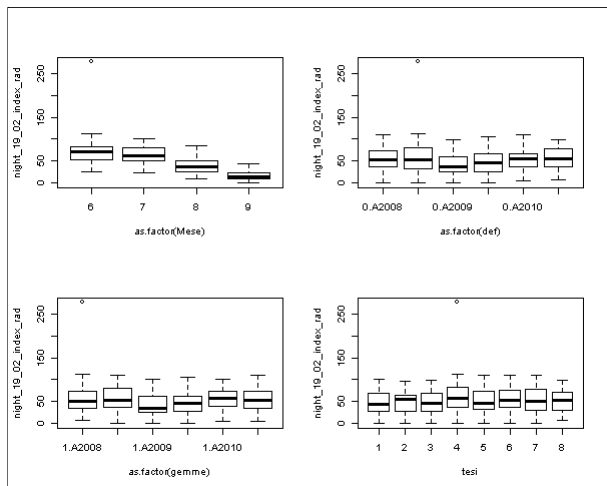
(Dispersion parameter for gaussian family taken to be 312.4816)

- Null deviance: 1769934.4933 on 2543 degrees of freedom.
- Residual deviance: 792140.9562 on 2535 degrees of freedom.

AIC: 21845

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard mattutino tra le ore 5-10 radiazione: dmorn_stderr_5_10_rad brolio

Call: glm(formula = dmorn_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-29.6	-6.8	-1.4	6.1	73.2

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	17.99572	0.63389	28.3894	< 2.2e-16	***
as.factor(def)[T.1]	-0.98005	0.43093	-2.2743	0.0230319	*
as.factor(gemme)[T.3]	-2.78616	0.43093	-6.4655	1.207e-10	***
as.factor(dirad)[T.1]	1.66727	0.43093	3.8690	0.0001120	***
AnnoF[T.A2009]	0.36946	0.52778	0.7000	0.4839707	
AnnoF[T.A2010]	2.28699	0.52778	4.3333	1.526e-05	***
as.factor(Mese)[T.7]	6.40128	0.57309	11.1697	< 2.2e-16	***
as.factor(Mese)[T.8]	23.48640	0.57309	40.9820	< 2.2e-16	***
as.factor(Mese)[T.9]	43.80192	0.70552	62.0845	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

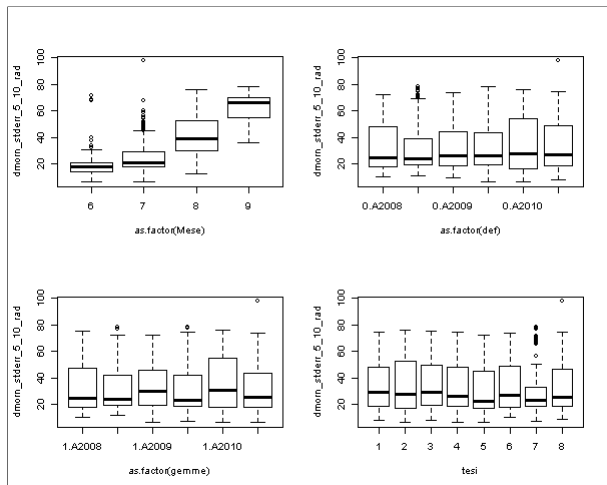
(Dispersion parameter for gaussian family taken to be 118.1047)

- Null deviance: 873088.4668 on 2543 degrees of freedom.
- Residual deviance: 299395.3648 on 2535 degrees of freedom.

AIC: 19369

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza mattutina tra le ore 5-10 radiazione: dmorn_var_5_10_rad brolio

Call: glm(formula = dmorn_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15747	-2861	-728	1915	52816

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2193.176	328.785	6.6706	3.118e-11	***
as.factor(def)[T.1]	-616.716	223.514	-2.7592	0.005836	**
as.factor(gemme)[T.3]	-1357.813	223.514	-6.0749	1.428e-09	***
as.factor(dirad)[T.1]	929.338	223.514	4.1579	3.319e-05	***
AnnoF[T.A2009]	95.963	273.747	0.3506	0.725954	
AnnoF[T.A2010]	1478.485	273.747	5.4009	7.247e-08	***
as.factor(Mese)[T.7]	1907.631	297.250	6.4176	1.647e-10	***
as.factor(Mese)[T.8]	9398.335	297.250	31.6176	< 2.2e-16	***
as.factor(Mese)[T.9]	21206.245	365.939	57.9502	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

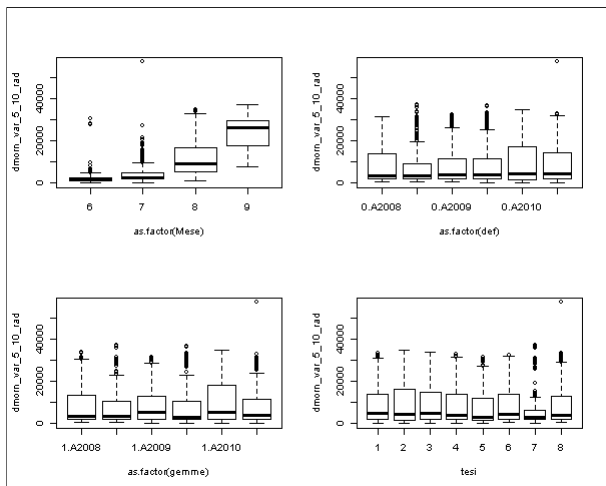
(Dispersion parameter for gaussian family taken to be 31773478)

- Null deviance: 212781084030.239 on 2543 degrees of freedom.
- Residual deviance: 80545765471.085 on 2535 degrees of freedom.

AIC: 51176

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 radiazione: dtwi_stderr_15_20_rad brolio

Call: glm(formula = dtwi_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-33.92	-9.99	0.27	9.78	31.71

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	41.33695	0.77325	53.4587	< 2.2e-16	***
as.factor(def)[T.1]	-0.99282	0.52567	-1.8887	0.05905	.
as.factor(gemme)[T.3]	-2.48416	0.52567	-4.7257	2.418e-06	***
as.factor(dirad)[T.1]	-2.46366	0.52567	-4.6867	2.923e-06	***
AnnoF[T.A2009]	4.26667	0.64381	6.6272	4.164e-11	***
AnnoF[T.A2010]	-5.31298	0.64381	-8.2524	2.465e-16	***
as.factor(Mese)[T.7]	4.51986	0.69909	6.4654	1.208e-10	***
as.factor(Mese)[T.8]	24.59178	0.69909	35.1770	< 2.2e-16	***
as.factor(Mese)[T.9]	37.03255	0.86063	43.0295	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

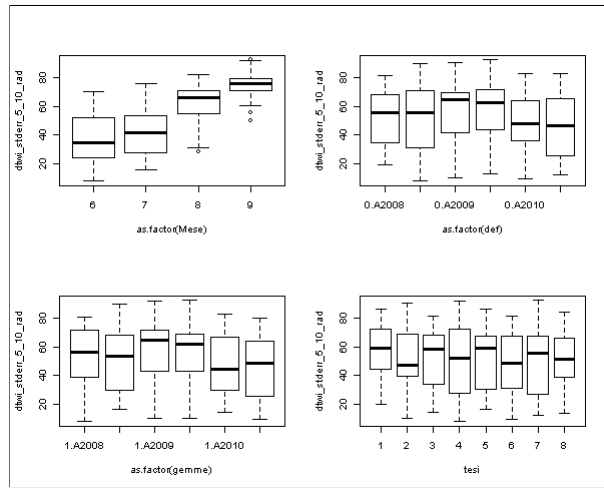
(Dispersion parameter for gaussian family taken to be 175.7445)

- Null deviance: 970562.8436 on 2543 degrees of freedom.
- Residual deviance: 445512.2769 on 2535 degrees of freedom.

AIC: 20381

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza serale tra le ore 15-20 radiazione: dtwi_var_15_20_rad brolio

Call: glm(formula = dtwi_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-19799	-5932	-717	5564	20632

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	12086.62	447.03	27.0376	< 2.2e-16	***
as.factor(def)[T.1]	-164.56	303.90	-0.5415	0.5882	
as.factor(gemme)[T.3]	-1481.66	303.90	-4.8755	1.152e-06	***
as.factor(dirad)[T.1]	-1581.28	303.90	-5.2033	2.115e-07	***
AnnoF[T.A2009]	2589.72	372.20	6.9579	4.380e-12	***
AnnoF[T.A2010]	-3302.55	372.20	-8.8731	< 2.2e-16	***
as.factor(Mese)[T.7]	2210.07	404.15	5.4684	4.986e-08	***
as.factor(Mese)[T.8]	14074.53	404.15	34.8246	< 2.2e-16	***
as.factor(Mese)[T.9]	23838.82	497.55	47.9128	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

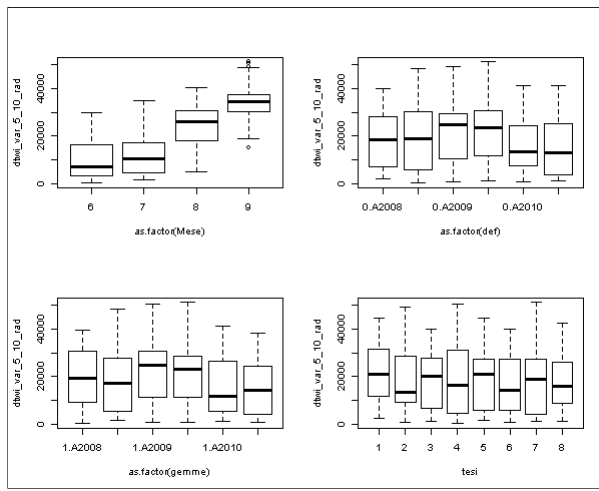
(Dispersion parameter for gaussian family taken to be 58737453)

- Null deviance: 353900132461.147 on 2543 degrees of freedom.
- Residual deviance: 148899444124.702 on 2535 degrees of freedom.

AIC: 52739

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Analisi indici giornalieri temperatura grappolo brolio

temperatura media: avg_tgrap brolio

Call: `glm(formula = avg_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.931	-1.963	0.081	2.038	6.574

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	22.66832	0.16169	140.1999	< 2.2e-16	***
as.factor(def)[T.1]	0.27962	0.10992	2.5439	0.0110205	*
as.factor(gemme)[T.3]	-0.32817	0.10992	-2.9857	0.0028567	**
as.factor(dirad)[T.1]	-0.12222	0.10992	-1.1120	0.2662591	
AnnoF[T.A2009]	0.75185	0.13462	5.5850	2.587e-08	***
AnnoF[T.A2010]	-0.20110	0.13462	-1.4938	0.1353450	
as.factor(Mese)[T.7]	2.86117	0.14618	19.5731	< 2.2e-16	***
as.factor(Mese)[T.8]	3.72093	0.14618	25.4548	< 2.2e-16	***
as.factor(Mese)[T.9]	0.65208	0.17996	3.6236	0.0002963	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

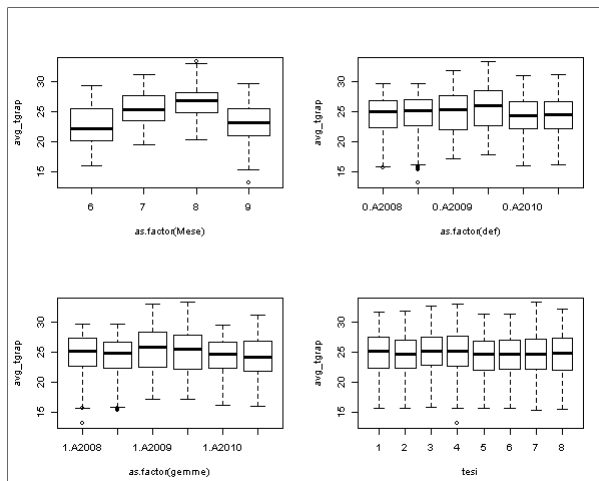
(Dispersion parameter for gaussian family taken to be 7.683953)

- Null deviance: 26226.9597 on 2543 degrees of freedom.
- Residual deviance: 19478.8212 on 2535 degrees of freedom.

AIC: 12418

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.929561111498253



cumulata temperatura grappolo: sum_tgrap brolio

Call: glm(formula = sum_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-190.4	-47.1	1.9	48.9	157.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	544.0399	3.8805	140.1999	< 2.2e-16	***
as.factor(def)[T.1]	6.7108	2.6380	2.5439	0.0110207	*
as.factor(gemme)[T.3]	-7.8762	2.6380	-2.9857	0.0028566	**
as.factor(dirad)[T.1]	-2.9334	2.6380	-1.1120	0.2662586	
AnnoF[T.A2009]	18.0442	3.2309	5.5849	2.587e-08	***
AnnoF[T.A2010]	-4.8264	3.2309	-1.4938	0.1353438	
as.factor(Mese)[T.7]	68.6680	3.5083	19.5731	< 2.2e-16	***
as.factor(Mese)[T.8]	89.3023	3.5083	25.4547	< 2.2e-16	***
as.factor(Mese)[T.9]	15.6500	4.3190	3.6235	0.0002963	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

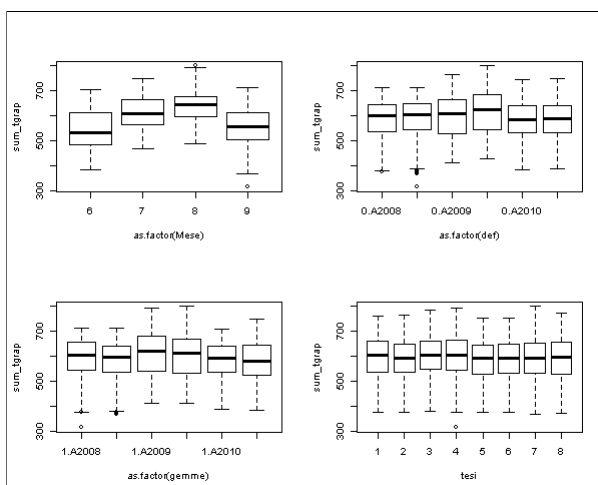
(Dispersion parameter for gaussian family taken to be 4425.96)

- Null deviance: 15106734.8635 on 2543 degrees of freedom.
- Residual deviance: 11219808.4573 on 2535 degrees of freedom.

AIC: 28588

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica grappolo: et_tgrap brolio

Call: glm(formula = et_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.15	-2.15	0.27	2.35	11.20

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.195139	0.205600	64.1788	< 2.2e-16	***
as.factor(def)[T.1]	0.884026	0.139770	6.3249	2.986e-10	***
as.factor(gemme)[T.3]	-0.598718	0.139770	-4.2836	1.908e-05	***
as.factor(dirad)[T.1]	-0.069251	0.139770	-0.4955	0.6203	
AnnoF[T.A2009]	-0.783999	0.171183	-4.5799	4.877e-06	***
AnnoF[T.A2010]	-0.885946	0.171183	-5.1754	2.452e-07	***
as.factor(Mese)[T.7]	1.785533	0.185880	9.6058	< 2.2e-16	***
as.factor(Mese)[T.8]	4.707170	0.185880	25.3237	< 2.2e-16	***
as.factor(Mese)[T.9]	2.814785	0.228833	12.3006	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

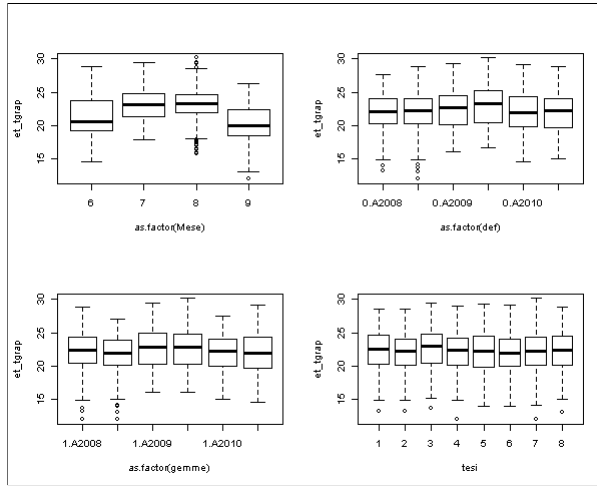
(Dispersion parameter for gaussian family taken to be 12.42471)

- Null deviance: 40908.4666 on 2543 degrees of freedom.
- Residual deviance: 31496.6315 on 2535 degrees of freedom.

AIC: 13641

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.975267195262811



media ore 3-10 temp. grappolo: morning_03_10_index_tgrap brolio

Call: glm(formula = morning_03_10_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.19	-1.79	-0.23	1.74	7.51

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.33697	0.14586	146.2810	< 2.2e-16	***
as.factor(def)[T.1]	0.21392	0.09916	2.1573	0.031074	*
as.factor(gemme)[T.3]	-0.16936	0.09916	-1.7079	0.087776	.
as.factor(dirad)[T.1]	-0.30561	0.09916	-3.0820	0.002078	**
AnnoF[T.A2009]	0.64807	0.12145	5.3363	1.033e-07	***
AnnoF[T.A2010]	-0.14345	0.12145	-1.1812	0.237643	
as.factor(Mese)[T.7]	1.76937	0.13187	13.4172	< 2.2e-16	***
as.factor(Mese)[T.8]	1.83430	0.13187	13.9096	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.81646	0.16235	-5.0292	5.271e-07	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

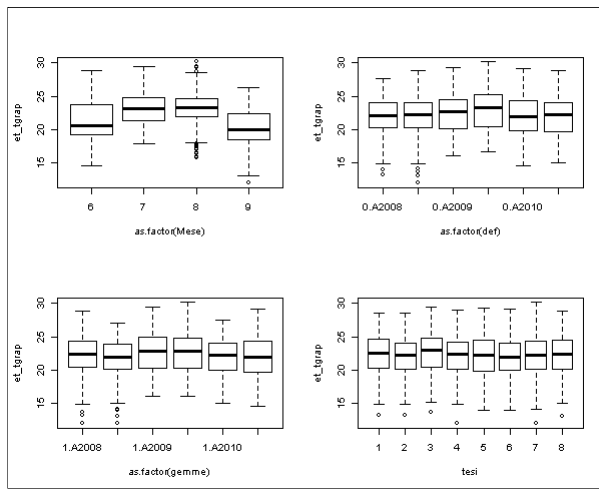
(Dispersion parameter for gaussian family taken to be 6.253611)

- Null deviance: 19093.7139 on 2543 degrees of freedom.
- Residual deviance: 15852.9034 on 2535 degrees of freedom.

AIC: 11894

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.975267195262811



media ore 11-18 temp. grappolo: diurnal_11_18_index_tgrap brolio

Call: `glm(formula = diurnal_11_18_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.5	-2.4	0.4	2.8	9.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	27.321411	0.234268	116.6246	< 2.2e-16	***
as.factor(def)[T.1]	0.565762	0.159259	3.5525	0.0003886	***
as.factor(gemme)[T.3]	-0.580627	0.159259	-3.6458	0.0002719	***
as.factor(dirad)[T.1]	-0.056782	0.159259	-0.3565	0.7214665	
AnnoF[T.A2009]	0.353638	0.195052	1.8130	0.0699431	.
AnnoF[T.A2010]	-0.612752	0.195052	-3.1415	0.0017004	**
as.factor(Mese)[T.7]	4.074072	0.211799	19.2356	< 2.2e-16	***
as.factor(Mese)[T.8]	6.181080	0.211799	29.1837	< 2.2e-16	***
as.factor(Mese)[T.9]	2.248454	0.260741	8.6233	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

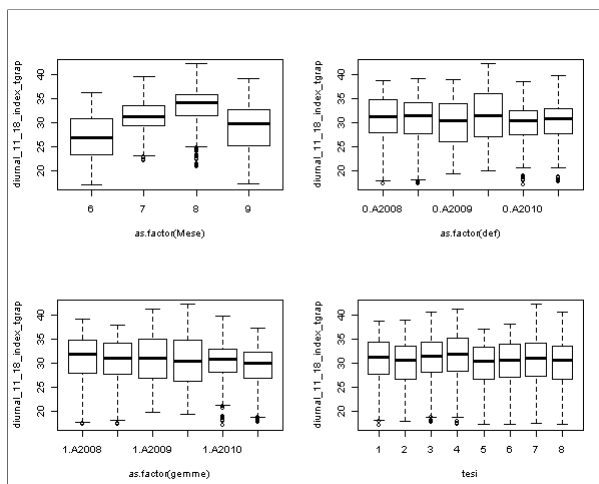
(Dispersion parameter for gaussian family taken to be 16.13123)

- Null deviance: 56316.3805 on 2543 degrees of freedom.
- Residual deviance: 40892.6717 on 2535 degrees of freedom.

AIC: 14305

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.997672076775202



media ore 19-02 temp. grappolo: night_19_02_index_tgrap brolio

Call: `glm(formula = night_19_02_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.647	-1.817	-0.066	1.745	8.993

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.3466079	0.1458046	132.6886	< 2.2e-16	***
as.factor(def)[T.1]	0.0591731	0.0991205	0.5970	0.550573	
as.factor(gemme)[T.3]	-0.2345456	0.0991205	-2.3663	0.018043	*
as.factor(dirad)[T.1]	-0.0042825	0.0991205	-0.0432	0.965541	
AnnoF[T.A2009]	1.2538175	0.1213973	10.3282	< 2.2e-16	***
AnnoF[T.A2010]	0.1529026	0.1213973	1.2595	0.207958	
as.factor(Mese)[T.7]	2.7400545	0.1318203	20.7863	< 2.2e-16	***
as.factor(Mese)[T.8]	3.1474081	0.1318203	23.8765	< 2.2e-16	***
as.factor(Mese)[T.9]	0.5242543	0.1622812	3.2305	0.001251	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

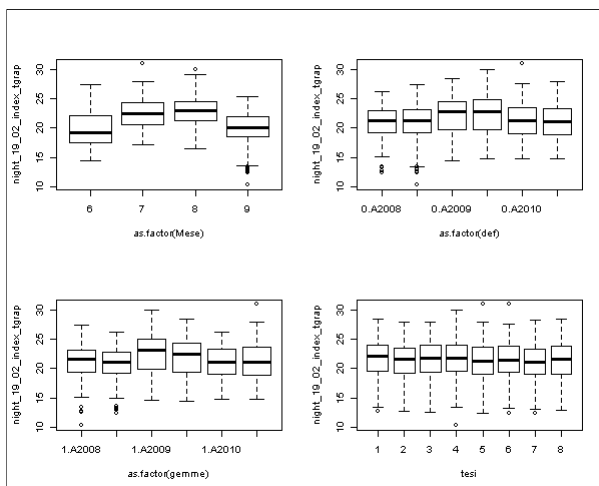
(Dispersion parameter for gaussian family taken to be 6.248619)

- Null deviance: 21520.1694 on 2543 degrees of freedom.
- Residual deviance: 15840.2488 on 2535 degrees of freedom.

AIC: 11892

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.89294788607481



errore standard mattutino tra le ore 5-10 temp. grappolo: dmorn_stderr_5_10_tgrap brolio

Call: glm(formula = dmorn_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.794	-0.335	0.048	0.333	1.333

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.585967	0.028161	56.3180	< 2.2e-16	***
as.factor(def)[T.1]	0.089343	0.019144	4.6668	3.218e-06	***
as.factor(gemme)[T.3]	0.010407	0.019144	0.5436	0.5868	
as.factor(dirad)[T.1]	-0.079580	0.019144	-4.1569	3.333e-05	***
AnnoF[T.A2009]	-0.230511	0.023447	-9.8312	< 2.2e-16	***
AnnoF[T.A2010]	-0.111683	0.023447	-4.7632	2.013e-06	***
as.factor(Mese)[T.7]	0.186933	0.025460	7.3423	2.815e-13	***
as.factor(Mese)[T.8]	0.427067	0.025460	16.7740	< 2.2e-16	***
as.factor(Mese)[T.9]	0.135081	0.031343	4.3097	1.697e-05	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

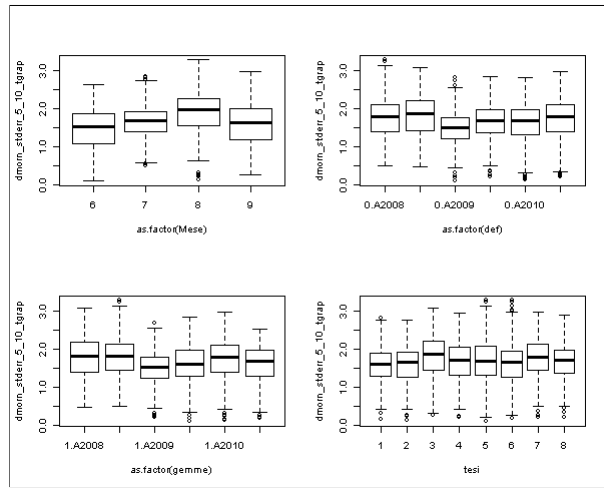
(Dispersion parameter for gaussian family taken to be 0.2330962)

- Null deviance: 690.4487 on 2543 degrees of freedom.
- Residual deviance: 590.8989 on 2535 degrees of freedom.

AIC: 3525.7

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.161451418694026



varianza mattutina tra le ore 5-10 temp. grappolo: dmorn_var_5_10_tgrap brolio

Call: glm(formula = dmorn_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-25.19	-7.34	-0.35	5.73	39.99

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	17.04874	0.56794	30.0187	< 2.2e-16	***
as.factor(def)[T.1]	1.85347	0.38609	4.8006	1.674e-06	***
as.factor(gemme)[T.3]	0.21993	0.38609	0.5696	0.569	
as.factor(dirad)[T.1]	-1.78024	0.38609	-4.6109	4.208e-06	***
AnnoF[T.A2009]	-5.08049	0.47287	-10.7440	< 2.2e-16	***
AnnoF[T.A2010]	-2.60594	0.47287	-5.5109	3.929e-08	***
as.factor(Mese)[T.7]	3.13247	0.51346	6.1007	1.218e-09	***
as.factor(Mese)[T.8]	8.93113	0.51346	17.3939	< 2.2e-16	***
as.factor(Mese)[T.9]	2.93324	0.63212	4.6403	3.655e-06	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

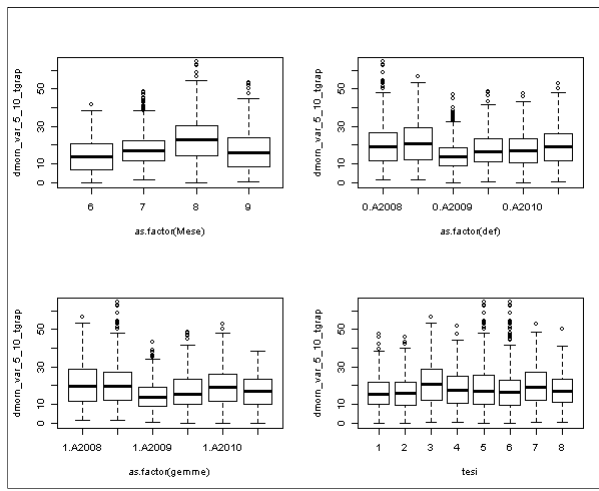
(Dispersion parameter for gaussian family taken to be 94.80704)

- Null deviance: 285702.068 on 2543 degrees of freedom.
- Residual deviance: 240335.835 on 2535 degrees of freedom.

AIC: 18810

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.999999981995403



errore standard serale tra le ore 15-20 temp. grappolo: dtwi_stderr_15_20_tgrap brolio

Call: glm(formula = dtwi_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.606	-0.307	0.019	0.304	1.583

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.2130477	0.0287312	42.2206	< 2.2e-16	***
as.factor(def)[T.1]	0.0715756	0.0195319	3.6645	0.0002529	***
as.factor(gemme)[T.3]	-0.1367303	0.0195319	-7.0003	3.257e-12	***
as.factor(dirad)[T.1]	-0.0077556	0.0195319	-0.3971	0.6913479	
AnnoF[T.A2009]	-0.1894055	0.0239216	-7.9177	3.584e-15	***
AnnoF[T.A2010]	-0.1309923	0.0239216	-5.4759	4.781e-08	***
as.factor(Mese)[T.7]	0.2741804	0.0259755	10.5553	< 2.2e-16	***
as.factor(Mese)[T.8]	0.6672967	0.0259755	25.6895	< 2.2e-16	***
as.factor(Mese)[T.9]	0.4743843	0.0319779	14.8348	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 0.2426318)

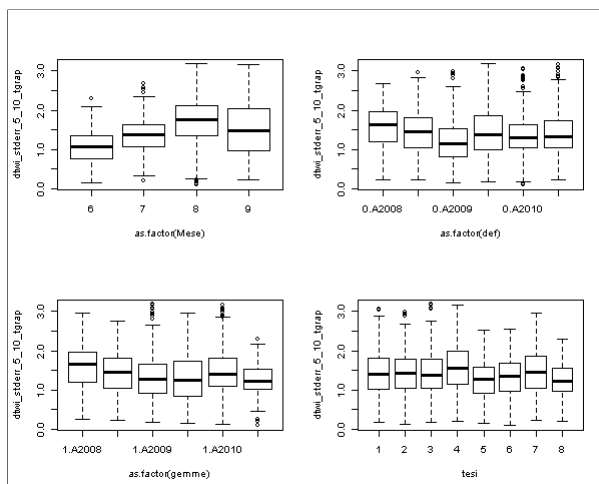
• Null deviance: 816.0797 on 2543 degrees of freedom.

• Residual deviance: 615.0716 on 2535 degrees of freedom.

AIC: 3627.7

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.276843180197595



varianza serale tra le ore 15-20 temp. grappolo: dtwi_var_15_20_tgrap brolio

Call: glm(formula = dtwi_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-21.17	-5.54	-0.92	4.42	41.18

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	10.54673	0.53130	19.8509	< 2.2e-16	***
as.factor(def)[T.1]	1.46890	0.36119	4.0669	4.910e-05	***
as.factor(gemme)[T.3]	-2.75606	0.36119	-7.6306	3.285e-14	***
as.factor(dirad)[T.1]	-0.24385	0.36119	-0.6751	0.4996	
AnnoF[T.A2009]	-3.08690	0.44236	-6.9783	3.801e-12	***
AnnoF[T.A2010]	-2.65921	0.44236	-6.0114	2.104e-09	***
as.factor(Mese)[T.7]	3.86595	0.48034	8.0484	1.276e-15	***
as.factor(Mese)[T.8]	12.12131	0.48034	25.2348	< 2.2e-16	***
as.factor(Mese)[T.9]	9.54157	0.59134	16.1356	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

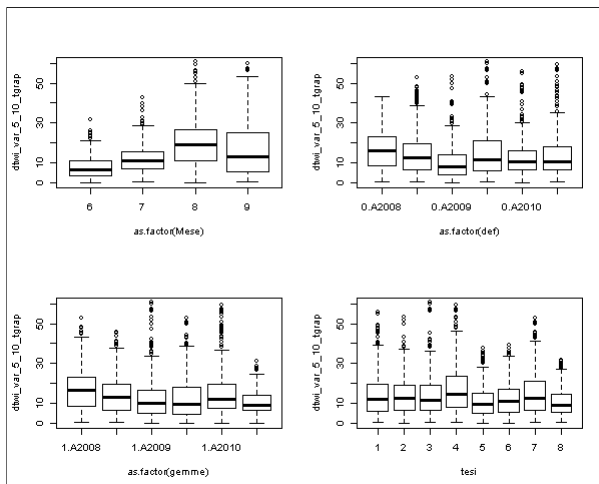
(Dispersion parameter for gaussian family taken to be 82.96948)

- Null deviance: 282148.7295 on 2543 degrees of freedom.
- Residual deviance: 210327.642 on 2535 degrees of freedom.

AIC: 18471

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.999999999999451



Analisi indici giornalieri taria canopy brolio

temperatura canopy media: avg_tair brolio

Call: glm(formula = avg_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.96	-2.14	0.11	2.18	7.21

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.202236	0.167933	138.1634	< 2.2e-16	***
as.factor(def)[T.1]	0.041579	0.114164	0.3642	0.715735	
as.factor(gemme)[T.3]	-0.349353	0.114164	-3.0601	0.002236	**
as.factor(dirad)[T.1]	-0.210035	0.114164	-1.8398	0.065919	.
AnnoF[T.A2009]	0.629410	0.139822	4.5015	7.052e-06	***
AnnoF[T.A2010]	-0.361432	0.139822	-2.5850	0.009795	**
as.factor(Mese)[T.7]	3.685664	0.151826	24.2755	< 2.2e-16	***
as.factor(Mese)[T.8]	3.560563	0.151826	23.4515	< 2.2e-16	***
as.factor(Mese)[T.9]	0.141155	0.186910	0.7552	0.450197	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

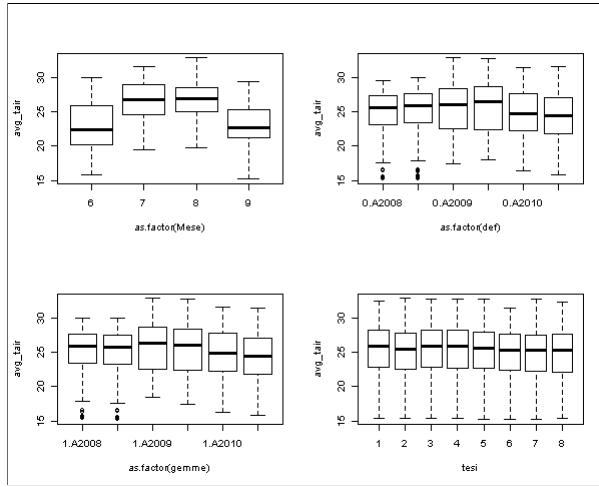
(Dispersion parameter for gaussian family taken to be 8.289241)

- Null deviance: 29451.0261 on 2543 degrees of freedom.
- Residual deviance: 21013.2258 on 2535 degrees of freedom.

AIC: 12611

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.96373835540747



cumulata temperatura canopy: sum_tair brolio

Call: glm(formula = sum_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-190.9	-51.3	2.7	52.2	173.1

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	556.85364	4.03040	138.1635	< 2.2e-16	***
as.factor(def)[T.1]	0.99789	2.73993	0.3642	0.715736	
as.factor(gemme)[T.3]	-8.38446	2.73993	-3.0601	0.002236	**
as.factor(dirad)[T.1]	-5.04084	2.73993	-1.8398	0.065919	.
AnnoF[T.A2009]	15.10588	3.35572	4.5015	7.051e-06	***
AnnoF[T.A2010]	-8.67433	3.35572	-2.5849	0.009795	**
as.factor(Mese)[T.7]	88.45591	3.64384	24.2755	< 2.2e-16	***
as.factor(Mese)[T.8]	85.45346	3.64384	23.4515	< 2.2e-16	***
as.factor(Mese)[T.9]	3.38773	4.48585	0.7552	0.450197	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

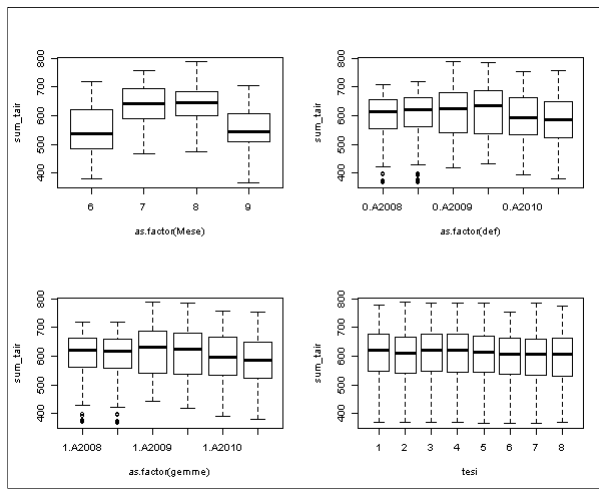
(Dispersion parameter for gaussian family taken to be 4774.601)

- Null deviance: 16963782.4544 on 2543 degrees of freedom.
- Residual deviance: 12103614.0436 on 2535 degrees of freedom.

AIC: 28781

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica canopy: et_tair brolio

Call: `glm(formula = et_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.87	-2.00	0.38	2.19	10.37

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.809492	0.185035	74.6319	< 2.2e-16	***
as.factor(def)[T.1]	-0.012819	0.125790	-0.1019	0.9188	
as.factor(gemme)[T.3]	0.127376	0.125790	1.0126	0.3113	
as.factor(dirad)[T.1]	0.014865	0.125790	0.1182	0.9059	
AnnoF[T.A2009]	-0.796851	0.154060	-5.1723	2.493e-07	***
AnnoF[T.A2010]	-0.911473	0.154060	-5.9163	3.736e-09	***
as.factor(Mese)[T.7]	2.366626	0.167288	14.1470	< 2.2e-16	***
as.factor(Mese)[T.8]	3.384639	0.167288	20.2324	< 2.2e-16	***
as.factor(Mese)[T.9]	0.944895	0.205944	4.5881	4.691e-06	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 10.06348)

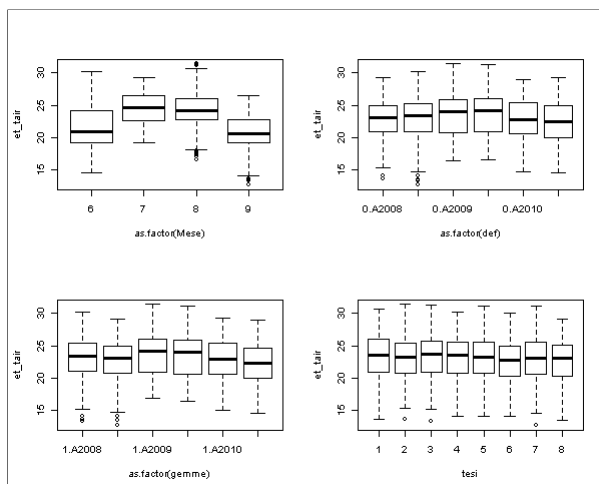
• Null deviance: 30564.2737 on 2543 degrees of freedom.

• Residual deviance: 25510.9124 on 2535 degrees of freedom.

AIC: 13104

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.862814636498442



media ore 03-10 temp. canopy: morning_03_10_index_tair brolio

Call: `glm(formula = morning_03_10_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.05	-1.81	-0.21	2.00	8.58

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.863065	0.151418	144.3883	< 2.2e-16	***
as.factor(def)[T.1]	0.025274	0.102937	0.2455	0.806067	
as.factor(gemme)[T.3]	-0.409173	0.102937	-3.9750	7.235e-05	***
as.factor(dirad)[T.1]	-0.314208	0.102937	-3.0524	0.002293	**
AnnoF[T.A2009]	0.681396	0.126071	5.4048	7.091e-08	***
AnnoF[T.A2010]	-0.309937	0.126071	-2.4584	0.014021	*
as.factor(Mese)[T.7]	2.863683	0.136896	20.9187	< 2.2e-16	***
as.factor(Mese)[T.8]	2.522972	0.136896	18.4299	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.501246	0.168529	-2.9742	0.002965	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

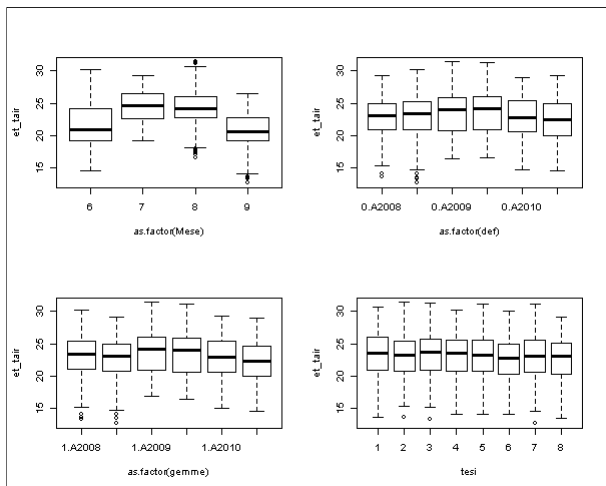
(Dispersion parameter for gaussian family taken to be 6.739058)

- Null deviance: 22859.2597 on 2543 degrees of freedom.
- Residual deviance: 17083.5123 on 2535 degrees of freedom.

AIC: 12084

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.862814636498442



media ore 11-18 temp. canopy: diurnal_11_18_index_tair brolio

Call: glm(formula = diurnal_11_18_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.48	-2.53	0.46	2.93	9.56

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	27.992537	0.237592	117.8178	< 2.2e-16	***
as.factor(def)[T.1]	0.066604	0.161519	0.4124	0.6801119	
as.factor(gemme)[T.3]	-0.276576	0.161519	-1.7123	0.0869552	.
as.factor(dirad)[T.1]	-0.119885	0.161519	-0.7422	0.4580123	
AnnoF[T.A2009]	0.229791	0.197819	1.1616	0.2455000	
AnnoF[T.A2010]	-0.748255	0.197819	-3.7825	0.0001588	***
as.factor(Mese)[T.7]	5.033925	0.214804	23.4350	< 2.2e-16	***
as.factor(Mese)[T.8]	5.323142	0.214804	24.7814	< 2.2e-16	***
as.factor(Mese)[T.9]	0.778203	0.264441	2.9428	0.0032818	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

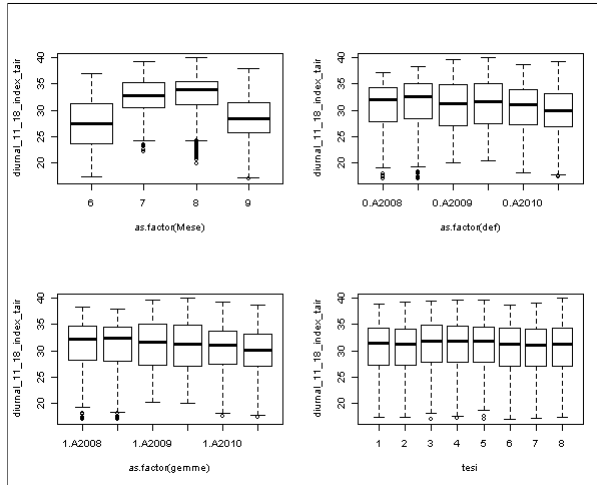
(Dispersion parameter for gaussian family taken to be 16.5922)

- Null deviance: 57650.4825 on 2543 degrees of freedom.
- Residual deviance: 42061.2236 on 2535 degrees of freedom.

AIC: 14376

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.997818741806655



media ore 19-02 temp. canopy: night_19_02_index_tair brolio

Call: `glm(formula = night_19_02_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.980	-1.805	-0.065	1.875	9.656

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.751098	0.147378	134.0169	< 2.2e-16	***
as.factor(def)[T.1]	0.032861	0.100190	0.3280	0.7429484	
as.factor(gemme)[T.3]	-0.362309	0.100190	-3.6162	0.0003048	***
as.factor(dirad)[T.1]	-0.196011	0.100190	-1.9564	0.0505285	.
AnnoF[T.A2009]	0.977051	0.122707	7.9625	2.521e-15	***
AnnoF[T.A2010]	-0.026095	0.122707	-0.2127	0.8316060	
as.factor(Mese)[T.7]	3.159381	0.133242	23.7115	< 2.2e-16	***
as.factor(Mese)[T.8]	2.835570	0.133242	21.2813	< 2.2e-16	***
as.factor(Mese)[T.9]	0.146513	0.164032	0.8932	0.3718368	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

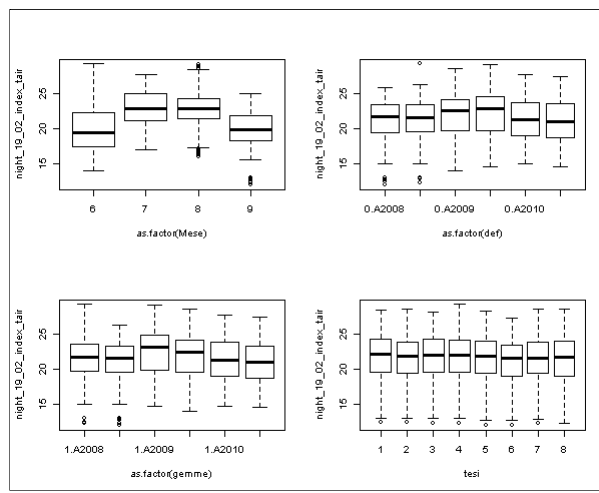
(Dispersion parameter for gaussian family taken to be 6.384178)

- Null deviance: 22257.265 on 2543 degrees of freedom.
- Residual deviance: 16183.8908 on 2535 degrees of freedom.

AIC: 11947

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.908269662331982



errore standard mattutino tra le ore 5-10 temp. canopy: dmorn_stderr_5_10_tair brolio

Call: `glm(formula = dmorn_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.900	-0.380	0.077	0.362	1.347

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.658669	0.030231	54.8670	< 2.2e-16	***
as.factor(def)[T.1]	0.071169	0.020551	3.4630	0.0005431	***
as.factor(gemme)[T.3]	0.019574	0.020551	0.9525	0.3409518	
as.factor(dirad)[T.1]	-0.040220	0.020551	-1.9570	0.0504537	.
AnnoF[T.A2009]	-0.131670	0.025170	-5.2312	1.822e-07	***
AnnoF[T.A2010]	-0.083895	0.025170	-3.3331	0.0008711	***
as.factor(Mese)[T.7]	0.353441	0.027331	12.9318	< 2.2e-16	***
as.factor(Mese)[T.8]	0.432589	0.027331	15.8277	< 2.2e-16	***
as.factor(Mese)[T.9]	0.083155	0.033647	2.4714	0.0135234	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 0.2686197)

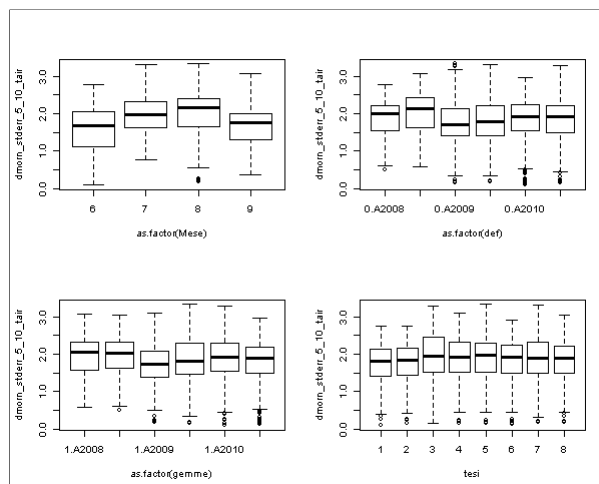
• Null deviance: 779.0888 on 2543 degrees of freedom.

• Residual deviance: 680.9509 on 2535 degrees of freedom.

AIC: 3886.6

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.143451330256536



varianza mattutina tra le ore 5-10 temp. canopy: dmorn_var_5_10_tair brolio

Call: `glm(formula = dmorn_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-28.00	-9.03	0.26	7.49	40.99

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	18.45524	0.64637	28.5522	< 2.2e-16	***
as.factor(def)[T.1]	1.83488	0.43941	4.1758	3.07e-05	***
as.factor(gemme)[T.3]	0.44409	0.43941	1.0106	0.3122856	
as.factor(dirad)[T.1]	-1.22138	0.43941	-2.7796	0.0054833	**
AnnoF[T.A2009]	-2.81874	0.53817	-5.2377	1.76e-07	***
AnnoF[T.A2010]	-1.81256	0.53817	-3.3680	0.0007684	***
as.factor(Mese)[T.7]	7.10701	0.58437	12.1617	< 2.2e-16	***
as.factor(Mese)[T.8]	9.31836	0.58437	15.9459	< 2.2e-16	***
as.factor(Mese)[T.9]	1.49286	0.71941	2.0751	0.0380771	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 122.8009)

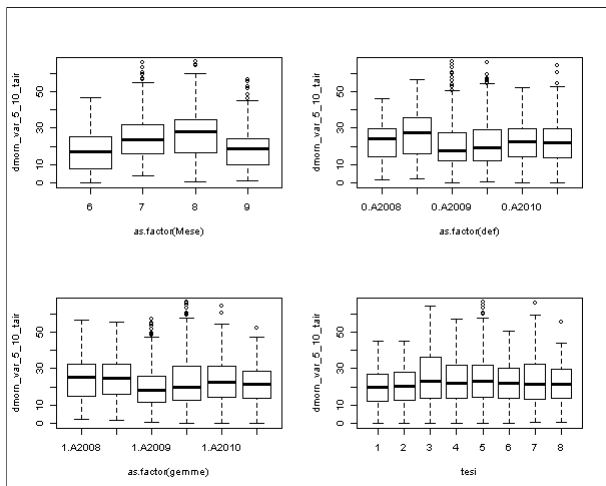
• Null deviance: 356981.1422 on 2543 degrees of freedom.

• Residual deviance: 311300.2713 on 2535 degrees of freedom.

AIC: 19469

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99999984089987



errore standard serale tra le ore 15-20 temp. canopy: dtwi_stderr_15_20_tair brolio

Call: glm(formula = dtwi_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.535	-0.279	0.031	0.297	1.105

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.2416385	0.0253058	49.0653	< 2.2e-16	***
as.factor(def)[T.1]	-0.0349016	0.0172033	-2.0288	0.04259	*
as.factor(gemme)[T.3]	0.0044362	0.0172033	0.2579	0.79653	
as.factor(dirad)[T.1]	0.0260112	0.0172033	1.5120	0.13066	
AnnoF[T.A2009]	-0.1950711	0.0210697	-9.2584	< 2.2e-16	***
AnnoF[T.A2010]	-0.1422746	0.0210697	-6.7526	1.795e-11	***
as.factor(Mese)[T.7]	0.3158714	0.0228787	13.8063	< 2.2e-16	***
as.factor(Mese)[T.8]	0.5662722	0.0228787	24.7510	< 2.2e-16	***
as.factor(Mese)[T.9]	0.2679564	0.0281655	9.5136	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

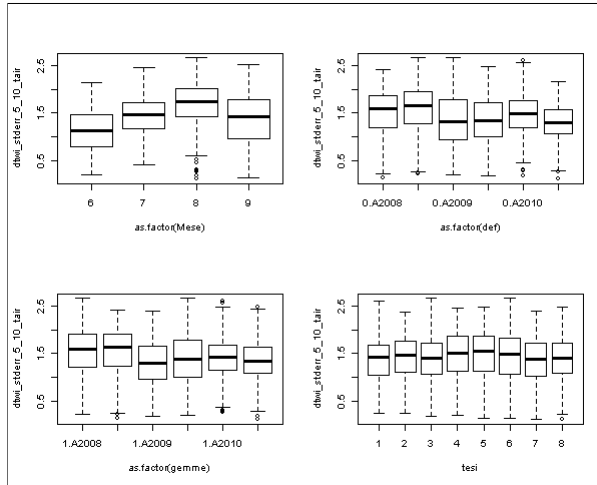
(Dispersion parameter for gaussian family taken to be 0.1882274)

- Null deviance: 611.5446 on 2543 degrees of freedom.
- Residual deviance: 477.1565 on 2535 degrees of freedom.

AIC: 2981.8

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.240805500062516



varianza serale tra le ore 15-20 temp. canopy: dtwi_var_15_20_tairbrolio

•

Call: `glm(formula = dtwi_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = brolio_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-17.88	-5.09	-0.52	4.49	24.73

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	11.015373	0.419744	26.2431	< 2e-16	***
as.factor(def)[T.1]	-0.679307	0.285349	-2.3806	0.01736	*
as.factor(gemme)[T.3]	0.084213	0.285349	0.2951	0.76792	
as.factor(dirad)[T.1]	0.386318	0.285349	1.3538	0.17591	
AnnoF[T.A2009]	-3.467278	0.349480	-9.9212	< 2e-16	***
AnnoF[T.A2010]	-3.058100	0.349480	-8.7504	< 2e-16	***
as.factor(Mese)[T.7]	4.622998	0.379486	12.1823	< 2e-16	***
as.factor(Mese)[T.8]	9.635215	0.379486	25.3902	< 2e-16	***
as.factor(Mese)[T.9]	4.832856	0.467177	10.3448	< 2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

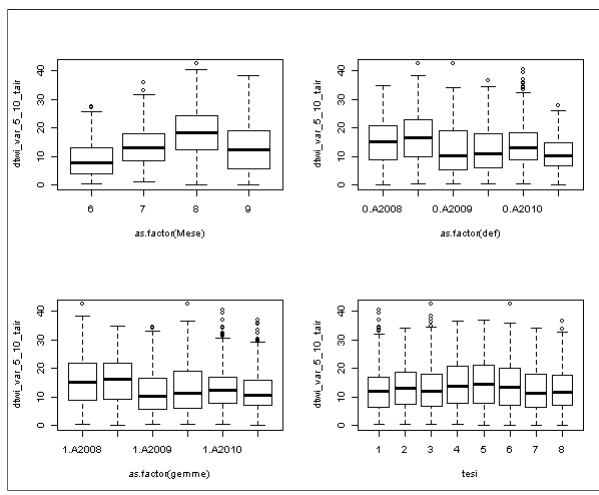
(Dispersion parameter for gaussian family taken to be 51.7858)

- Null deviance: 171205.3481 on 2543 degrees of freedom.
- Residual deviance: 131277.0024 on 2535 degrees of freedom.

AIC: 17272

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.9999984734791



Analisi statistica campione Tuscania cacciagrande

IBIMET CNR, Istituto di Biometeorologia di Firenze -CNR

Matese.A., Genesio L., Crisci A.

Temperatura Grappolo cacciagrande

Temperatura Grappolo Mattina

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.931	-2.416	-0.015	2.510	12.324

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.874023	0.118591	176.0176	< 2.2e-16	***
as.factor(def)[T.1]	0.264341	0.081537	3.2420	0.001192	**
as.factor(gemme)[T.3]	-0.356176	0.081537	-4.3683	1.270e-05	***
as.factor(dirad)[T.1]	0.053019	0.081537	0.6503	0.515550	
AnnoF[T.A2009]	-0.125984	0.101194	-1.2450	0.213181	
AnnoF[T.A2010]	-0.586676	0.101118	-5.8019	6.838e-09	***
as.factor(Mese)[T.7]	2.603224	0.109553	23.7621	< 2.2e-16	***
as.factor(Mese)[T.8]	3.349035	0.107063	31.2809	< 2.2e-16	***
as.factor(Mese)[T.9]	1.562435	0.129614	12.0545	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

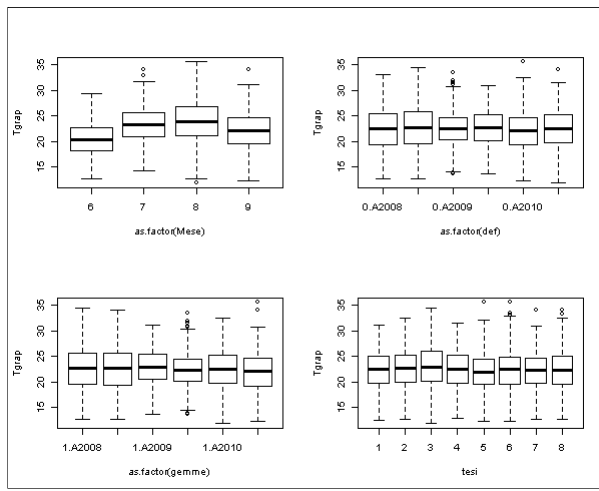
(Dispersion parameter for gaussian family taken to be 11.70153)

- Null deviance: 96656.1966 on 7121 degrees of freedom.
- Residual deviance: 83233.0184 on 7113 degrees of freedom.

AIC: 37740

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.84813381947672



Temperatura Grappolo Pieno giorno

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-16.224	-2.383	0.058	2.482	11.449

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.597296	0.073282	390.2365	< 2.2e-16	***
as.factor(def)[T.1]	0.099355	0.050812	1.9553	0.05056	.
as.factor(gemme)[T.3]	-0.624504	0.050812	-12.2904	< 2.2e-16	***
as.factor(dirad)[T.1]	-0.044047	0.050812	-0.8668	0.38604	
AnnoF[T.A2009]	-0.473297	0.063042	-7.5077	6.264e-14	***
AnnoF[T.A2010]	0.026326	0.063042	0.4176	0.67624	
as.factor(Mese)[T.7]	4.660588	0.065408	71.2538	< 2.2e-16	***
as.factor(Mese)[T.8]	6.097259	0.067313	90.5804	< 2.2e-16	***
as.factor(Mese)[T.9]	3.467416	0.085859	40.3849	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 13.09535)

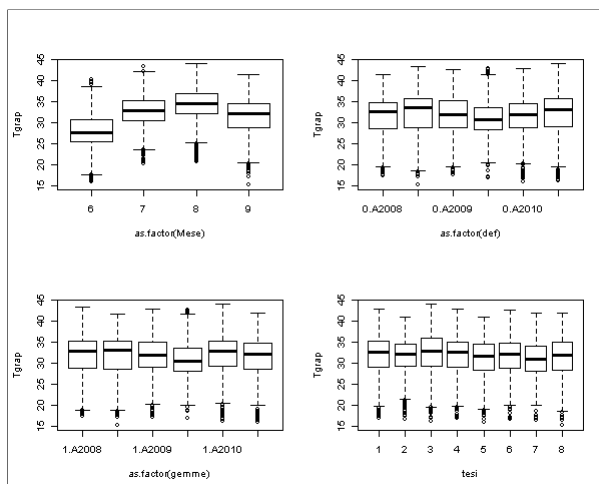
• Null deviance: 390580.4892 on 20522 degrees of freedom.

• Residual deviance: 268637.9155 on 20514 degrees of freedom.

AIC: 111043

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.997372582711401



Temperatura Cross Grappolo Pieno giorno

Call: `glm(formula = Tgrap ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.395	-2.426	0.027	2.454	11.130

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	23.1479602	0.7202795	32.1375	< 2.2e-16 ***
as.factor(gemme)[T.3]	0.8635764	0.7489163	1.1531	0.24888
as.factor(def)[T.1]	-6.3240490	0.7975713	-7.9291	2.319e-15 ***
Rad	0.0118616	0.0016405	7.2303	4.987e-13 ***
AnnoF[T.A2009]	-0.3684086	0.0638431	-5.7705	8.016e-09 ***
AnnoF[T.A2010]	0.4998461	0.0664914	7.5175	5.814e-14 ***
as.factor(Mese)[T.7]	4.7840535	0.0659248	72.5684	< 2.2e-16 ***
as.factor(Mese)[T.8]	6.1721040	0.0680038	90.7612	< 2.2e-16 ***
as.factor(Mese)[T.9]	3.4592048	0.0857726	40.3300	< 2.2e-16 ***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.0083844	0.1102951	-0.0760	0.93941
as.factor(def)[T.1]:Rad	0.0136207	0.0018042	7.5494	4.554e-14 ***
as.factor(gemme)[T.3]:Rad	-0.0029376	0.0017481	-1.6804	0.09289 .

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 12.80508)

- Null deviance: 390580.4892 on 20522 degrees of freedom.
- Residual deviance: 262645.0196 on 20511 degrees of freedom.

AIC: 110586

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.998037946017014

Temperatura Grappolo Sera

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.335	-1.976	-0.052	2.030	9.843

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	25.868600	0.107693	240.2074	< 2.2e-16 ***
as.factor(def)[T.1]	-0.175442	0.073968	-2.3719	0.01773 *
as.factor(gemme)[T.3]	-0.149808	0.073968	-2.0253	0.04287 *
as.factor(dirad)[T.1]	0.058445	0.073968	0.7901	0.42947
AnnoF[T.A2009]	0.881959	0.091784	9.6091	< 2.2e-16 ***
AnnoF[T.A2010]	0.366315	0.091784	3.9911	6.643e-05 ***
as.factor(Mese)[T.7]	4.021782	0.097337	41.3179	< 2.2e-16 ***
as.factor(Mese)[T.8]	4.492371	0.098926	45.4113	< 2.2e-16 ***
as.factor(Mese)[T.9]	2.673267	0.118659	22.5290	< 2.2e-16 ***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

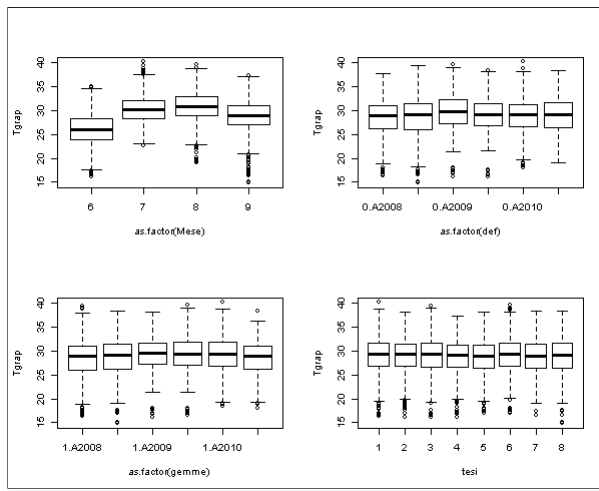
(Dispersion parameter for gaussian family taken to be 9.6721)

- Null deviance: 94337.3336 on 7152 degrees of freedom.
- Residual deviance: 69097.4848 on 7144 degrees of freedom.

AIC: 36542

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.970654914430738



Temperatura Grappolo Notte

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.85	-1.97	-0.15	1.92	13.99

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.3376537	0.0558755	346.0846	< 2.2e-16	***
as.factor(def)[T.1]	-0.2484399	0.0375415	-6.6177	3.725e-11	***
as.factor(gemme)[T.3]	-0.0025089	0.0375415	-0.0668	0.946718	
as.factor(dirad)[T.1]	0.1005869	0.0375415	2.6794	0.007381	**
AnnoF[T.A2009]	0.6020576	0.0465863	12.9235	< 2.2e-16	***
AnnoF[T.A2010]	-0.2380430	0.0465968	-5.1086	3.271e-07	***
as.factor(Mese)[T.7]	2.9297417	0.0515592	56.8229	< 2.2e-16	***
as.factor(Mese)[T.8]	3.1736694	0.0504615	62.8929	< 2.2e-16	***
as.factor(Mese)[T.9]	2.1044300	0.0590811	35.6193	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

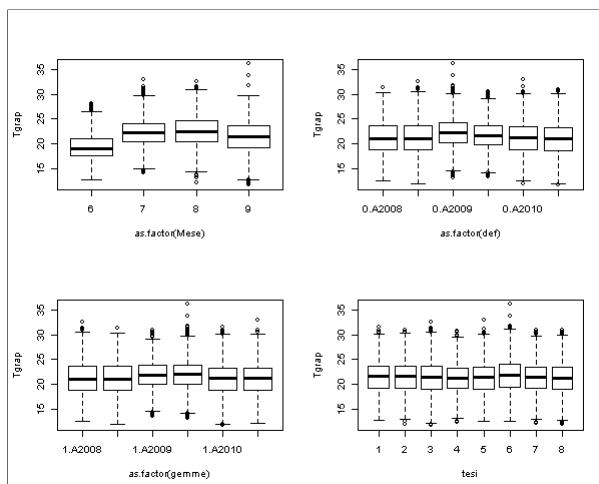
(Dispersion parameter for gaussian family taken to be 8.259946)

- Null deviance: 238008.2276 on 23713 degrees of freedom.
- Residual deviance: 195802.0292 on 23705 degrees of freedom.

AIC: 117379

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.831364659105173



Radiazione Grappolo Mattina

Call: `glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new,`

subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-393.0	-23.4	9.4	31.6	110.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	400.6941	1.6647	240.7055	< 2.2e-16	***
as.factor(def)[T.1]	19.2214	1.1356	16.9266	< 2.2e-16	***
as.factor(gemme)[T.3]	-11.8769	1.1356	-10.4589	< 2.2e-16	***
as.factor(dirad)[T.1]	-7.0477	1.1356	-6.2062	5.718e-10	***
AnnoF[T.A2009]	-7.7434	1.3913	-5.5657	2.703e-08	***
AnnoF[T.A2010]	-18.7448	1.3902	-13.4840	< 2.2e-16	***
as.factor(Mese)[T.7]	-16.2374	1.5342	-10.5839	< 2.2e-16	***
as.factor(Mese)[T.8]	-7.5929	1.4993	-5.0642	4.200e-07	***
as.factor(Mese)[T.9]	-25.2460	1.8164	-13.8990	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

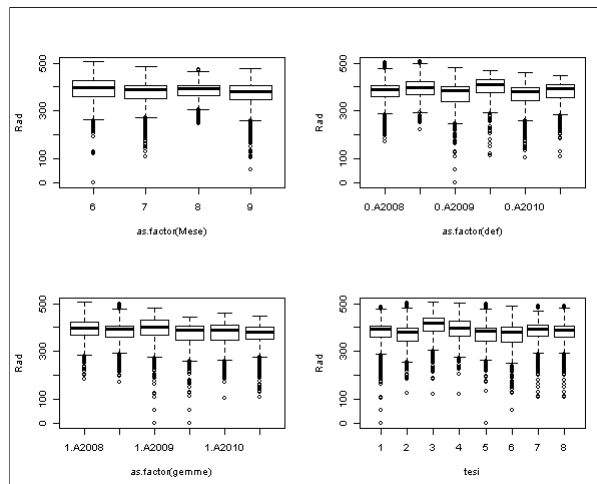
(Dispersion parameter for gaussian family taken to be 2395.946)

- Null deviance: 19832732.7728 on 7431 degrees of freedom.
- Residual deviance: 17785105.11 on 7423 degrees of freedom.

AIC: 78934

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiatione Grappolo Pieno giorno

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-285.58	-19.16	-0.71	18.42	81.98

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	449.50263	0.54555	823.9437	< 2.2e-16	***
as.factor(def)[T.1]	23.56921	0.37530	62.8008	< 2.2e-16	***
as.factor(gemme)[T.3]	-9.08343	0.37530	-24.2030	< 2.2e-16	***
as.factor(dirad)[T.1]	-4.08104	0.37530	-10.8740	< 2.2e-16	***
AnnoF[T.A2009]	-7.06513	0.45964	-15.3711	< 2.2e-16	***
AnnoF[T.A2010]	-22.84123	0.45964	-49.6940	< 2.2e-16	***
as.factor(Mese)[T.7]	-10.93457	0.48579	-22.5089	< 2.2e-16	***
as.factor(Mese)[T.8]	-8.96678	0.49993	-17.9360	< 2.2e-16	***
as.factor(Mese)[T.9]	-3.61797	0.63814	-5.6695	1.450e-08	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

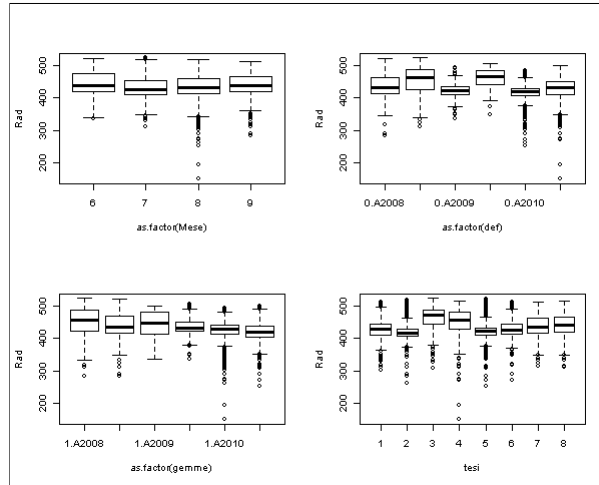
(Dispersion parameter for gaussian family taken to be 754.1152)

- Null deviance: 22055721.5185 on 21415 degrees of freedom.
- Residual deviance: 16143344.3382 on 21407 degrees of freedom.

AIC: 202680

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiazione Cross Grappolo Pieno giorno

Call: glm(formula = Rad ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Tgrap + as.factor(gemme) * Tgrap + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-294.92	-17.86	-0.55	16.44	78.70

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	449.560852	2.451819	183.3581	< 2.2e-16	***
as.factor(gemme)[T.3]	-13.718115	2.732095	-5.0211	5.181e-07	***
as.factor(def)[T.1]	-22.824101	2.767588	-8.2469	< 2.2e-16	***
Tgrap	-0.129059	0.078925	-1.6352	0.1020	
AnnoF[T.A2009]	-7.549164	0.459647	-16.4238	< 2.2e-16	***
AnnoF[T.A2010]	-24.702981	0.457672	-53.9753	< 2.2e-16	***
as.factor(Mese)[T.7]	-14.110292	0.530161	-26.6151	< 2.2e-16	***
as.factor(Mese)[T.8]	-13.757830	0.578061	-23.8000	< 2.2e-16	***
as.factor(Mese)[T.9]	-6.562568	0.647440	-10.1362	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-18.029190	0.739819	-24.3697	< 2.2e-16	***
as.factor(def)[T.1]:Tgrap	1.781103	0.084896	20.9798	< 2.2e-16	***
as.factor(gemme)[T.3]:Tgrap	0.493770	0.084687	5.8305	5.608e-09	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 689.374)

- Null deviance: 21198127.8609 on 20522 degrees of freedom.
- Residual deviance: 14139750.8806 on 20511 degrees of freedom.

AIC: 192390

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1

Radiazione Grappolo Sera

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-346.3	-25.4	7.3	34.7	123.9

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	395.54306	1.87614	210.8279	< 2.2e-16	***
as.factor(def)[T.1]	12.76763	1.27859	9.9857	< 2.2e-16	***
as.factor(gemme)[T.3]	-5.09815	1.27859	-3.9873	6.746e-05	***
as.factor(dirad)[T.1]	-4.80012	1.27859	-3.7542	0.0001752	***
AnnoF[T.A2009]	0.18013	1.56608	0.1150	0.9084323	
AnnoF[T.A2010]	-5.65149	1.56608	-3.6087	0.0003098	***
as.factor(Mese)[T.7]	-26.82115	1.69180	-15.8536	< 2.2e-16	***
as.factor(Mese)[T.8]	-49.21764	1.71943	-28.6243	< 2.2e-16	***
as.factor(Mese)[T.9]	-22.89764	2.06384	-11.0947	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

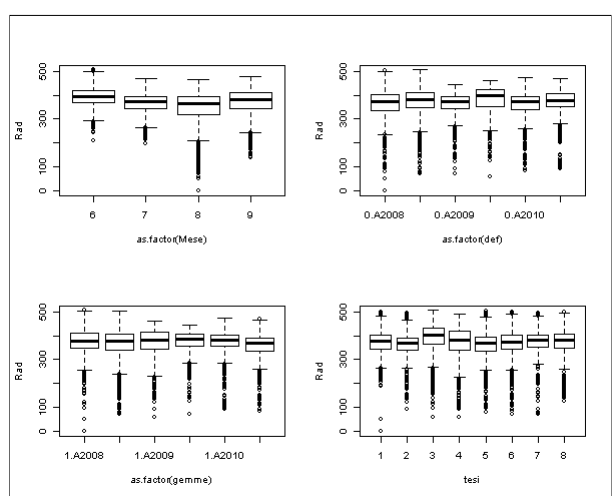
(Dispersion parameter for gaussian family taken to be 3050.533)

- Null deviance: 25705771.3525 on 7463 degrees of freedom.
- Residual deviance: 22741720.3809 on 7455 degrees of freedom.

AIC: 81077

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Temperatura Aria Canopy Mattina

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.65	-2.76	0.03	2.76	10.45

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.6591185	0.1267134	170.9300	< 2.2e-16	***
as.factor(def)[T.1]	0.2757005	0.0864393	3.1895	0.001431	**
as.factor(gemme)[T.3]	-0.4164902	0.0864393	-4.8183	1.477e-06	***
as.factor(dirad)[T.1]	-0.5197516	0.0864393	-6.0129	1.909e-09	***
AnnoF[T.A2009]	0.0058092	0.1059034	0.0549	0.956256	
AnnoF[T.A2010]	-0.6130456	0.1058176	-5.7934	7.179e-09	***
as.factor(Mese)[T.7]	3.0532034	0.1167797	26.1450	< 2.2e-16	***
as.factor(Mese)[T.8]	3.9286448	0.1141290	34.4228	< 2.2e-16	***
as.factor(Mese)[T.9]	1.6991042	0.1382621	12.2890	< 2.2e-16	***

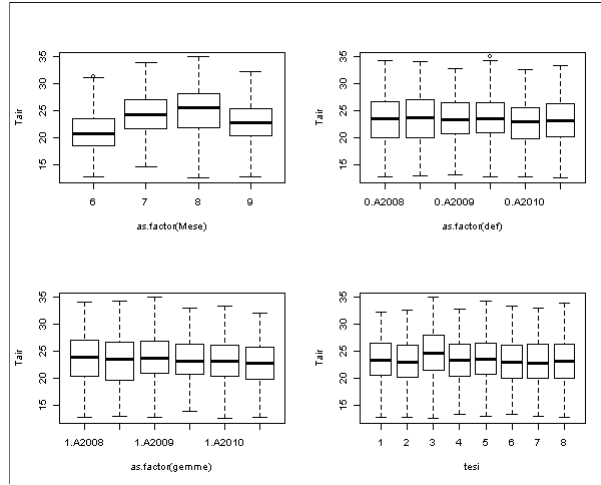
--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 13.88251)

- Null deviance: 122889.5753 on 7431 degrees of freedom.
- Residual deviance: 103049.8625 on 7423 degrees of freedom.

AIC: 40653

Number of Fisher Scoring iterations: 2



Temperatura Aria Canopy Pieno giorno

Call: `glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = caccia grande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-16.08	-2.28	0.15	2.35	10.12

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	29.058210	0.069531	417.9175	< 2.2e-16	***
as.factor(def)[T.1]	0.116789	0.047833	2.4416	0.014630	*
as.factor(gemme)[T.3]	-0.072896	0.047833	-1.5240	0.127529	
as.factor(dirad)[T.1]	0.155621	0.047833	3.2535	0.001142	**
AnnoF[T.A2009]	0.137211	0.058581	2.3422	0.019178	*
AnnoF[T.A2010]	0.018090	0.058581	0.3088	0.757477	
as.factor(Mese)[T.7]	4.626207	0.061914	74.7197	< 2.2e-16	***
as.factor(Mese)[T.8]	5.153662	0.063717	80.8840	< 2.2e-16	***
as.factor(Mese)[T.9]	2.173781	0.081332	26.7273	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 12.24969)

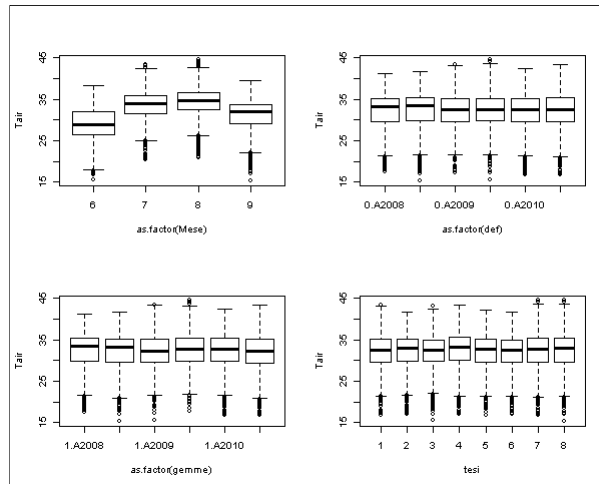
• Null deviance: 364284.8321 on 21415 degrees of freedom.

• Residual deviance: 262229.0208 on 21407 degrees of freedom.

AIC: 114445

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.991480562852282



Temperatura Aria Crossing Canopy Pieno giorno

Call: glm(formula = Tair ~ as.factor(gemme) * as.factor(def) + as.factor(dirad) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.17	-2.30	0.13	2.32	10.02

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	26.17722450	0.69707347	37.5530	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.28538618	0.71412456	-0.3996	0.6894324	
as.factor(def)[T.1]	-4.68412530	0.75703190	-6.1875	6.224e-10	***
as.factor(dirad)[T.1]	0.20701027	0.04819064	4.2957	1.749e-05	***
Rad	0.00636759	0.00158413	4.0196	5.849e-05	***
AnnoF[T.A2009]	0.16587634	0.05949495	2.7881	0.0053069	**
AnnoF[T.A2010]	0.29653735	0.06181337	4.7973	1.619e-06	***
as.factor(Mese)[T.7]	4.72568565	0.06271773	75.3485	< 2.2e-16	***
as.factor(Mese)[T.8]	5.21508700	0.06457452	80.7608	< 2.2e-16	***
as.factor(Mese)[T.9]	2.16957486	0.08156704	26.5987	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	0.35353998	0.10552093	3.3504	0.0008083	***
as.factor(def)[T.1]:Rad	0.00993240	0.00171165	5.8028	6.613e-09	***
as.factor(gemme)[T.3]:Rad	0.00044522	0.00166662	0.2671	0.7893639	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 12.12163)

• Null deviance: 364284.8321 on 21415 degrees of freedom.

• Residual deviance: 259439.2113 on 21403 degrees of freedom.

AIC: 114224

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.9925211272728

Temperatura Aria Canopy Sera

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.162	-2.036	-0.018	2.064	9.096

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	26.576274	0.107678	246.8122	< 2.2e-16	***
as.factor(def)[T.1]	0.045941	0.073383	0.6261	0.531300	
as.factor(gemme)[T.3]	-0.075251	0.073383	-1.0255	0.305179	
as.factor(dirad)[T.1]	-0.037426	0.073383	-0.5100	0.610056	
AnnoF[T.A2009]	0.586675	0.089882	6.5271	7.143e-11	***
AnnoF[T.A2010]	0.256472	0.089882	2.8534	0.004337	**
as.factor(Mese)[T.7]	4.057275	0.097098	41.7852	< 2.2e-16	***
as.factor(Mese)[T.8]	3.542636	0.098684	35.8987	< 2.2e-16	***
as.factor(Mese)[T.9]	1.682611	0.118451	14.2051	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

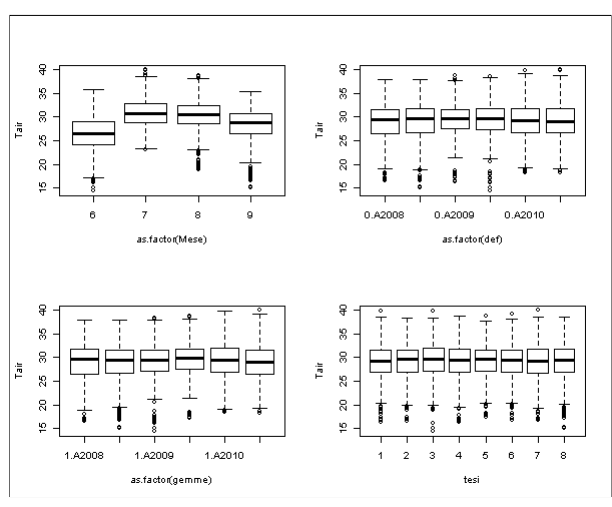
(Dispersion parameter for gaussian family taken to be 10.04845)

- Null deviance: 96535.347 on 7463 degrees of freedom.
- Residual deviance: 74911.1884 on 7455 degrees of freedom.

AIC: 38415

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.944820773482649



Temperatura Aria Canopy Notte

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.90	-2.04	-0.14	1.98	10.03

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.699259	0.056137	350.9163	< 2.2e-16	***
as.factor(def)[T.1]	0.140019	0.037429	3.7409	0.0001838	***
as.factor(gemme)[T.3]	-0.151894	0.037429	-4.0582	4.961e-05	***
as.factor(dirad)[T.1]	-0.315556	0.037429	-8.4307	< 2.2e-16	***
AnnoF[T.A2009]	0.283307	0.045845	6.1797	6.524e-10	***
AnnoF[T.A2010]	-0.284281	0.045856	-6.1994	5.757e-10	***
as.factor(Mese)[T.7]	2.993573	0.051685	57.9191	< 2.2e-16	***
as.factor(Mese)[T.8]	3.040543	0.050585	60.1076	< 2.2e-16	***
as.factor(Mese)[T.9]	2.027538	0.059263	34.2124	< 2.2e-16	***

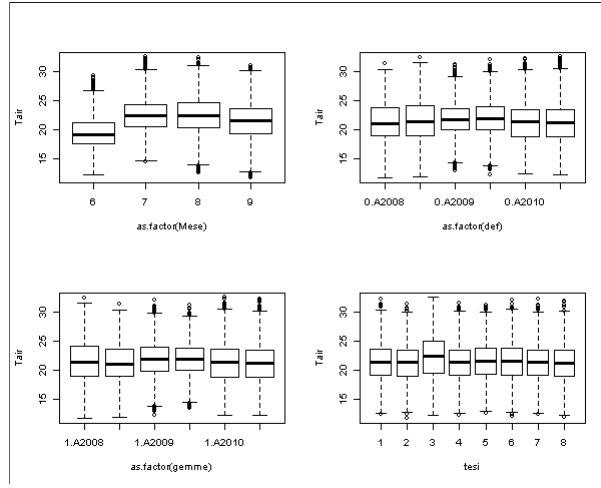
--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 8.666216)

- Null deviance: 255803.6109 on 24743 degrees of freedom.
- Residual deviance: 214358.8623 on 24735 degrees of freedom.

AIC: 123664

Number of Fisher Scoring iterations: 2



Analisi indici giornalieri radiazione cacciagrande

radiazione media : avg_rad cacciagrande

Call: `glm(formula = avg_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-55.99	-8.35	-0.30	7.42	47.09

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	280.83863	0.76476	367.2266	< 2.2e-16	***
as.factor(def)[T.1]	13.32235	0.51989	25.6251	< 2.2e-16	***
as.factor(gemme)[T.3]	-5.22162	0.51989	-10.0436	< 2.2e-16	***
as.factor(dirad)[T.1]	-2.85043	0.51989	-5.4827	4.603e-08	***
AnnoF[T.A2009]	-3.74817	0.63674	-5.8865	4.465e-09	***
AnnoF[T.A2010]	-11.08848	0.63674	-17.4145	< 2.2e-16	***
as.factor(Mese)[T.7]	-15.08576	0.69141	-21.8189	< 2.2e-16	***
as.factor(Mese)[T.8]	-34.44106	0.69141	-49.8130	< 2.2e-16	***
as.factor(Mese)[T.9]	-50.87074	0.85118	-59.7652	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

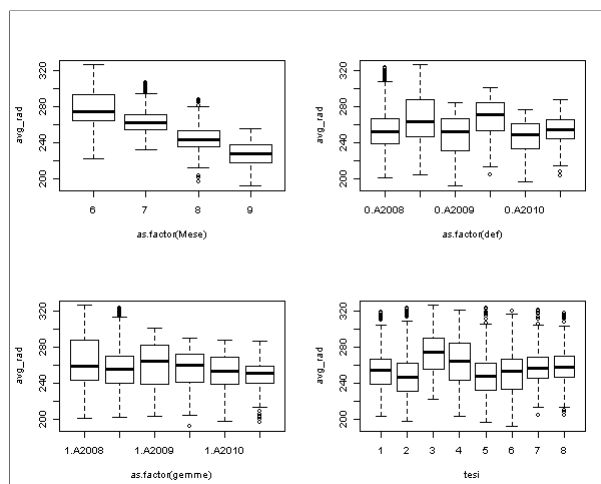
(Dispersion parameter for gaussian family taken to be 171.9043)

- Null deviance: 1412522.2942 on 2543 degrees of freedom.
- Residual deviance: 435777.4255 on 2535 degrees of freedom.

AIC: 20324

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



cumulata radiazione : sum_rad cacciagrande

Call: `glm(formula = sum_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1343.8	-200.4	-7.1	178.2	1130.2

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6740.127	18.354	367.2267	< 2.2e-16	***
as.factor(def)[T.1]	319.736	12.477	25.6251	< 2.2e-16	***
as.factor(gemme)[T.3]	-125.319	12.477	-10.0436	< 2.2e-16	***
as.factor(dirad)[T.1]	-68.410	12.477	-5.4827	4.603e-08	***
AnnoF[T.A2009]	-89.956	15.282	-5.8865	4.465e-09	***
AnnoF[T.A2010]	-266.124	15.282	-17.4145	< 2.2e-16	***
as.factor(Mese)[T.7]	-362.058	16.594	-21.8189	< 2.2e-16	***
as.factor(Mese)[T.8]	-826.586	16.594	-49.8130	< 2.2e-16	***
as.factor(Mese)[T.9]	-1220.898	20.428	-59.7652	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

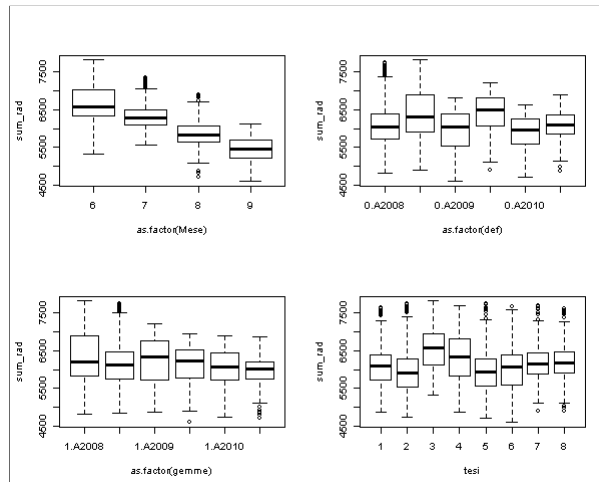
(Dispersion parameter for gaussian family taken to be 99016.87)

- Null deviance: 813612807.0459 on 2543 degrees of freedom.
- Residual deviance: 251007766.5946 on 2535 degrees of freedom.

AIC: 36494

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione radiazione: et_rad cacciagrande

Call: `glm(formula = et_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-70.34	-14.88	0.69	16.79	60.84

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	465.91234	1.34075	347.5020	< 2.2e-16	***
as.factor(def)[T.1]	27.45146	0.91146	30.1180	< 2.2e-16	***
as.factor(gemme)[T.3]	-4.35610	0.91146	-4.7792	1.860e-06	***
as.factor(dirad)[T.1]	-5.85828	0.91146	-6.4273	1.546e-10	***
AnnoF[T.A2009]	-13.64718	1.11631	-12.2253	< 2.2e-16	***
AnnoF[T.A2010]	-23.88002	1.11631	-21.3919	< 2.2e-16	***
as.factor(Mese)[T.7]	-1.89107	1.21215	-1.5601	0.1189	
as.factor(Mese)[T.8]	-0.94939	1.21215	-0.7832	0.4336	

as.factor(Mese)[T.9] 1.87834 1.49226 1.2587 0.2082

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

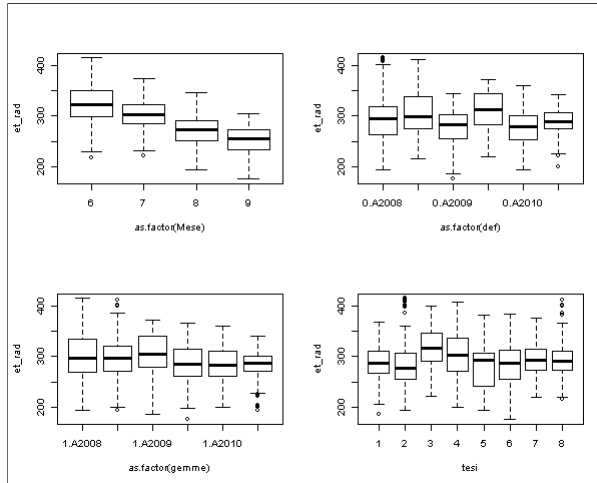
(Dispersion parameter for gaussian family taken to be 528.3664)

- Null deviance: 2099808.4933 on 2543 degrees of freedom.
- Residual deviance: 1339408.9075 on 2535 degrees of freedom.

AIC: 23181

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 3-10 radiazione: morning_03_10_index_rad cacciagrande

Call: glm(formula = morning_03_10_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-109.0	-18.6	-2.0	20.0	87.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	331.5933	1.4851	223.2785	< 2.2e-16	***
as.factor(def)[T.1]	17.8600	1.0096	17.6901	< 2.2e-16	***
as.factor(gemme)[T.3]	-9.4747	1.0096	-9.3845	< 2.2e-16	***
as.factor(dirad)[T.1]	-4.5522	1.0096	-4.5089	6.814e-06	***
AnnoF[T.A2009]	-5.0065	1.2365	-4.0489	5.300e-05	***
AnnoF[T.A2010]	-16.3520	1.2365	-13.2243	< 2.2e-16	***
as.factor(Mese)[T.7]	-21.8521	1.3427	-16.2751	< 2.2e-16	***
as.factor(Mese)[T.8]	-53.5649	1.3427	-39.8943	< 2.2e-16	***
as.factor(Mese)[T.9]	-71.4809	1.6529	-43.2449	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

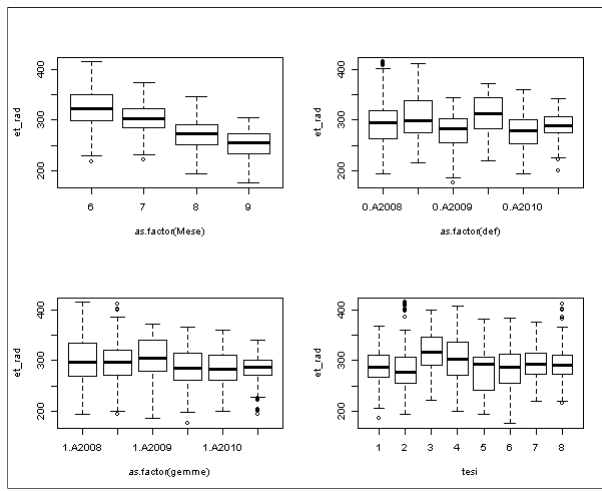
(Dispersion parameter for gaussian family taken to be 648.2745)

- Null deviance: 3731957.1113 on 2543 degrees of freedom.
- Residual deviance: 1643375.8485 on 2535 degrees of freedom.

AIC: 23701

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 11-18 radiazione: diurnal_11_18_index_rad cacciagrande

Call: glm(formula = diurnal_11_18_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-90.60	-12.70	-0.37	13.14	71.65

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	441.54186	1.22966	359.0775	< 2.2e-16	***
as.factor(def)[T.1]	22.77445	0.83594	27.2441	< 2.2e-16	***
as.factor(gemme)[T.3]	-7.91336	0.83594	-9.4664	< 2.2e-16	***
as.factor(dirad)[T.1]	-4.47820	0.83594	-5.3571	9.219e-08	***
AnnoF[T.A2009]	-6.46371	1.02381	-6.3134	3.213e-10	***
AnnoF[T.A2010]	-20.09968	1.02381	-19.6321	< 2.2e-16	***
as.factor(Mese)[T.7]	-9.85811	1.11172	-8.8675	< 2.2e-16	***
as.factor(Mese)[T.8]	-13.01248	1.11172	-11.7048	< 2.2e-16	***
as.factor(Mese)[T.9]	-20.93452	1.36861	-15.2962	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

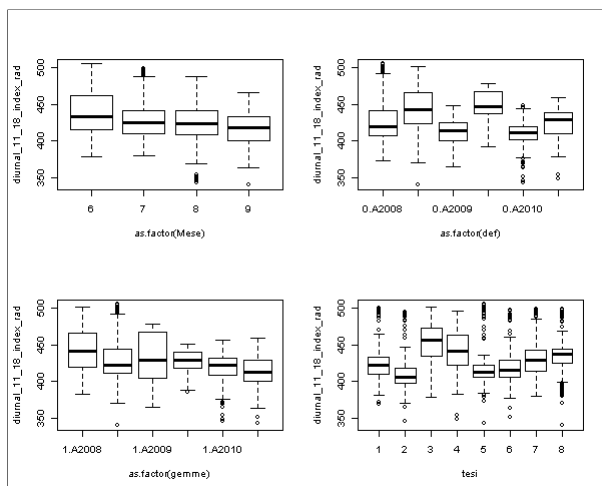
(Dispersion parameter for gaussian family taken to be 444.4354)

- Null deviance: 1806768.2267 on 2543 degrees of freedom.
- Residual deviance: 1126643.8451 on 2535 degrees of freedom.

AIC: 22741

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 19-02 radiazione: night_19_02_index_rad cacciagrande

Call: glm(formula = night_19_02_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-48.1	-11.3	-2.1	12.2	40.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	69.38077	0.90962	76.2746	< 2.2e-16	***
as.factor(def)[T.1]	-0.66738	0.61837	-1.0793	0.280576	
as.factor(gemme)[T.3]	1.72317	0.61837	2.7866	0.005366	**
as.factor(dirad)[T.1]	0.47908	0.61837	0.7747	0.438561	
AnnoF[T.A2009]	0.22571	0.75735	0.2980	0.765705	
AnnoF[T.A2010]	3.18619	0.75735	4.2070	2.677e-05	***
as.factor(Mese)[T.7]	-13.54705	0.82237	-16.4731	< 2.2e-16	***
as.factor(Mese)[T.8]	-36.74581	0.82237	-44.6825	< 2.2e-16	***
as.factor(Mese)[T.9]	-60.19679	1.01241	-59.4590	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

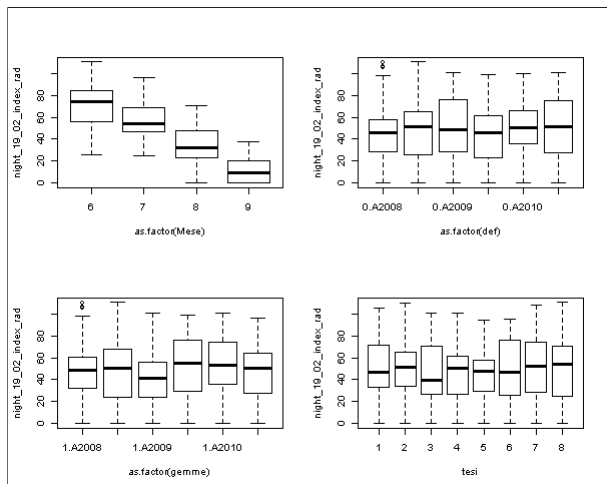
(Dispersion parameter for gaussian family taken to be 243.1977)

- Null deviance: 1697394.0208 on 2543 degrees of freedom.
- Residual deviance: 616506.0433 on 2535 degrees of freedom.

AIC: 21207

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard mattutino tra le ore 5-10 radiazione: dmorn_stderr_5_10_rad cacciagrande

Call: glm(formula = dmorn_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-28.6	-7.1	-1.2	6.6	50.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	16.67815	0.65362	25.5166	< 2.2e-16	***
as.factor(def)[T.1]	-1.70358	0.44434	-3.8340	0.0001292	***
as.factor(gemme)[T.3]	0.55713	0.44434	1.2538	0.2100169	
as.factor(dirad)[T.1]	-0.19970	0.44434	-0.4494	0.6531666	
AnnoF[T.A2009]	1.72344	0.54420	3.1669	0.0015591	**
AnnoF[T.A2010]	0.52606	0.54420	0.9667	0.3338093	
as.factor(Mese)[T.7]	2.98239	0.59093	5.0470	4.807e-07	***
as.factor(Mese)[T.8]	22.79944	0.59093	38.5824	< 2.2e-16	***
as.factor(Mese)[T.9]	40.81302	0.72748	56.1019	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

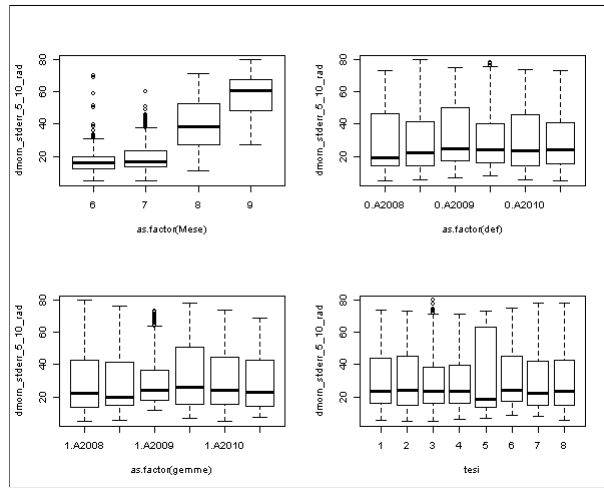
(Dispersion parameter for gaussian family taken to be 125.5711)

- Null deviance: 863676.3401 on 2543 degrees of freedom.
- Residual deviance: 318322.8264 on 2535 degrees of freedom.

AIC: 19525

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza mattutina tra le ore 5-10 radiazione: dmorn_var_5_10_rad cacciagrande

Call: `glm(formula = dmorn_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15464	-2304	-551	1802	25893

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2429.815	318.002	7.6409	3.038e-14	***
as.factor(def)[T.1]	-1142.746	216.183	-5.2860	1.357e-07	***
as.factor(gemme)[T.3]	517.942	216.183	2.3958	0.016654	*
as.factor(dirad)[T.1]	-464.111	216.183	-2.1468	0.031901	*
AnnoF[T.A2009]	320.648	264.769	1.2110	0.225990	
AnnoF[T.A2010]	-46.436	264.769	-0.1754	0.860792	
as.factor(Mese)[T.7]	826.557	287.502	2.8750	0.004074	**
as.factor(Mese)[T.8]	8848.857	287.502	30.7784	< 2.2e-16	***
as.factor(Mese)[T.9]	18768.693	353.937	53.0283	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

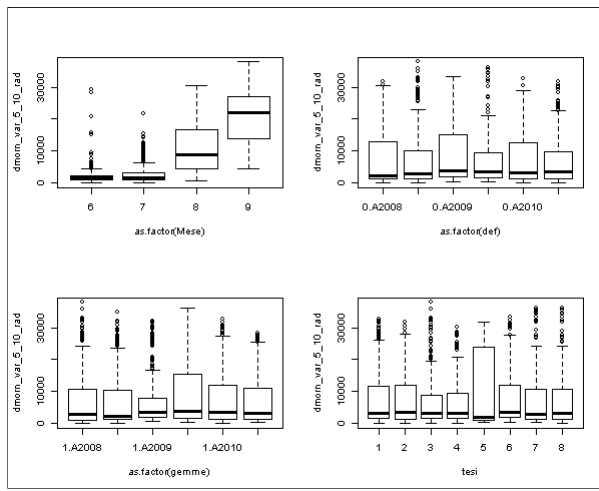
(Dispersion parameter for gaussian family taken to be 29723550)

- Null deviance: 185583124897.419 on 2543 degrees of freedom.
- Residual deviance: 75349199046.1903 on 2535 degrees of freedom.

AIC: 51006

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 radiazione: dtwi_stderr_15_20_rad cacciagrande

Call: glm(formula = dtwi_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-37.4	-9.7	1.1	9.3	34.0

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	41.37530	0.78515	52.6972	< 2.2e-16	***
as.factor(def)[T.1]	4.40555	0.53376	8.2538	2.437e-16	***
as.factor(gemme)[T.3]	-4.45139	0.53376	-8.3397	< 2.2e-16	***
as.factor(dirad)[T.1]	-1.74397	0.53376	-3.2673	0.0011001	**
AnnoF[T.A2009]	-2.16099	0.65372	-3.3057	0.0009606	***
AnnoF[T.A2010]	-6.48806	0.65372	-9.9248	< 2.2e-16	***
as.factor(Mese)[T.7]	8.78256	0.70985	12.3725	< 2.2e-16	***
as.factor(Mese)[T.8]	26.78769	0.70985	37.7373	< 2.2e-16	***
as.factor(Mese)[T.9]	39.14195	0.87388	44.7911	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

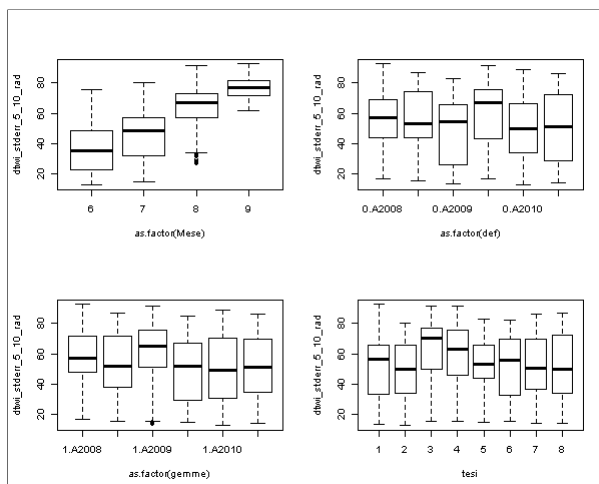
(Dispersion parameter for gaussian family taken to be 181.1959)

- Null deviance: 1001027.3626 on 2543 degrees of freedom.
- Residual deviance: 459331.7274 on 2535 degrees of freedom.

AIC: 20458

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza serale tra le ore 15-20 radiazione: dtwi_var_15_20_rad cacciagrande

Call: glm(formula = dtwi_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-21551	-5769	-164	5550	22542

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	11799.83	467.72	25.2285	< 2.2e-16	***
as.factor(def)[T.1]	3290.99	317.96	10.3502	< 2.2e-16	***
as.factor(gemme)[T.3]	-3139.04	317.96	-9.8723	< 2.2e-16	***
as.factor(dirad)[T.1]	-1035.81	317.96	-3.2576	0.001138	**
AnnoF[T.A2009]	-684.68	389.42	-1.7582	0.078838	.
AnnoF[T.A2010]	-3398.48	389.42	-8.7269	< 2.2e-16	***
as.factor(Mese)[T.7]	4513.07	422.86	10.6727	< 2.2e-16	***
as.factor(Mese)[T.8]	15801.23	422.86	37.3676	< 2.2e-16	***
as.factor(Mese)[T.9]	25592.67	520.57	49.1625	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

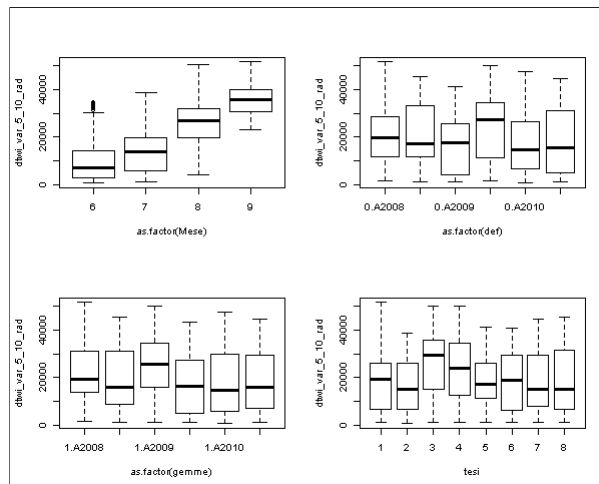
(Dispersion parameter for gaussian family taken to be 64300053)

- Null deviance: 386395623378.894 on 2543 degrees of freedom.
- Residual deviance: 163000633840.193 on 2535 degrees of freedom.

AIC: 52969

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Analisi indici giornalieri temperatura grappolo cacciagrande

temperatura media: avg_tgrap cacciagrande

Call: glm(formula = avg_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.947	-1.314	0.083	1.327	5.590

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.882070	0.119376	200.0580	< 2.2e-16	***
as.factor(def)[T.1]	-0.055107	0.081797	-0.6737	0.5005634	
as.factor(gemme)[T.3]	-0.281736	0.081797	-3.4443	0.0005822	***
as.factor(dirad)[T.1]	0.038914	0.081797	0.4757	0.6343079	
AnnoF[T.A2009]	0.191056	0.101490	1.8825	0.0598869	.
AnnoF[T.A2010]	-0.093189	0.101490	-0.9182	0.3586001	
as.factor(Mese)[T.7]	3.556645	0.108158	32.8838	< 2.2e-16	***
as.factor(Mese)[T.8]	3.784285	0.108158	34.9885	< 2.2e-16	***
as.factor(Mese)[T.9]	1.689462	0.133151	12.6883	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

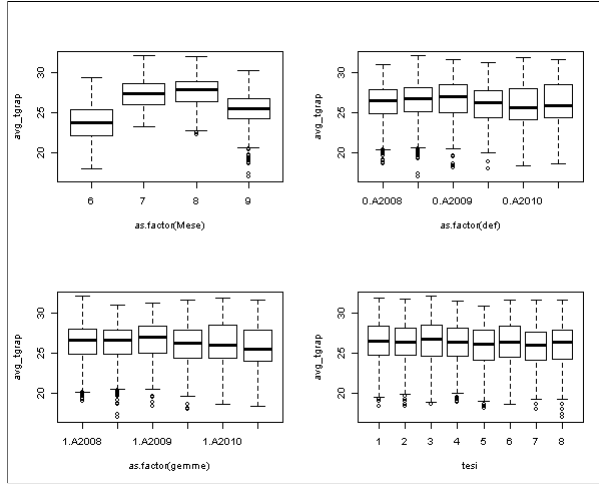
(Dispersion parameter for gaussian family taken to be 4.031367)

- Null deviance: 16211.5715 on 2437 degrees of freedom.
- Residual deviance: 9792.1904 on 2429 degrees of freedom.

AIC: 10329

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.929344345293706



cumulata temperatura grappolo: sum_tgrap cacciagrande

Call: glm(formula = sum_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-191	-32	2	32	134

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	573.16974	2.86502	200.0581	< 2.2e-16	***
as.factor(def)[T.1]	-1.32253	1.96313	-0.6737	0.5005759	
as.factor(gemme)[T.3]	-6.76168	1.96313	-3.4443	0.0005822	***
as.factor(dirad)[T.1]	0.93391	1.96313	0.4757	0.6343130	
AnnoF[T.A2009]	4.58532	2.43577	1.8825	0.0598880	.
AnnoF[T.A2010]	-2.23658	2.43577	-0.9182	0.3585918	
as.factor(Mese)[T.7]	85.35942	2.59579	32.8838	< 2.2e-16	***
as.factor(Mese)[T.8]	90.82278	2.59579	34.9885	< 2.2e-16	***
as.factor(Mese)[T.9]	40.54704	3.19562	12.6883	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

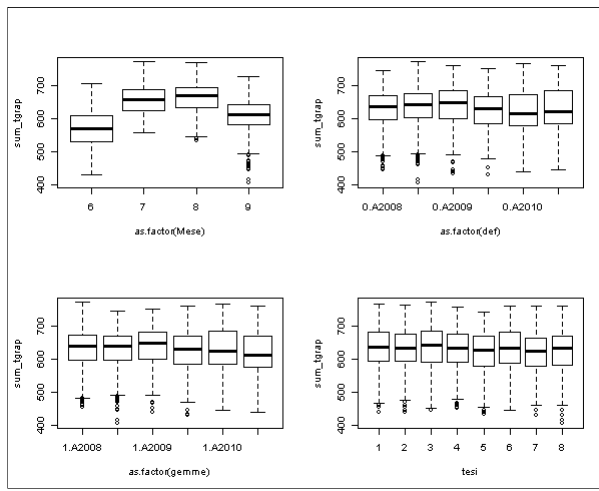
(Dispersion parameter for gaussian family taken to be 2322.067)

- Null deviance: 9337859.1174 on 2437 degrees of freedom.
- Residual deviance: 5640300.3421 on 2429 degrees of freedom.

AIC: 25825

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica grappolo: et_tgrap cacciagrande

Call: `glm(formula = et_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.299	-2.429	-0.091	2.643	10.852

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.11312	0.20913	62.7029	< 2.2e-16	***
as.factor(def)[T.1]	0.49771	0.14330	3.4732	0.0005232	***
as.factor(gemme)[T.3]	-0.48697	0.14330	-3.3983	0.0006890	***
as.factor(dirad)[T.1]	-0.32292	0.14330	-2.2535	0.0243186	*
AnnoF[T.A2009]	-0.78046	0.17780	-4.3896	1.184e-05	***
AnnoF[T.A2010]	0.72705	0.17780	4.0892	4.470e-05	***
as.factor(Mese)[T.7]	2.53430	0.18948	13.3751	< 2.2e-16	***
as.factor(Mese)[T.8]	3.79803	0.18948	20.0446	< 2.2e-16	***
as.factor(Mese)[T.9]	2.02614	0.23326	8.6861	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 12.37247)

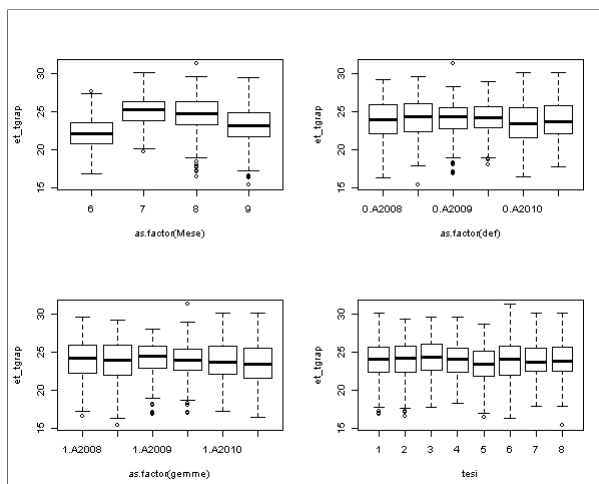
• Null deviance: 36517.6021 on 2437 degrees of freedom.

• Residual deviance: 30052.7217 on 2429 degrees of freedom.

AIC: 13062

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.929470118224397



media ore 3-10 temp. grappolo: morning_03_10_index_tgrap cacciagrande

Call: `glm(formula = morning_03_10_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.7334	-1.3665	-0.0039	1.4578	6.6751

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	22.425327	0.127057	176.4982	< 2.2e-16	***
as.factor(def)[T.1]	0.256371	0.087060	2.9447	0.003263	**
as.factor(gemme)[T.3]	-0.240453	0.087060	-2.7619	0.005790	**
as.factor(dirad)[T.1]	0.043942	0.087060	0.5047	0.613796	
AnnoF[T.A2009]	-0.019245	0.108021	-0.1782	0.858610	
AnnoF[T.A2010]	-0.455111	0.108021	-4.2132	2.61e-05	***
as.factor(Mese)[T.7]	2.810714	0.115117	24.4161	< 2.2e-16	***
as.factor(Mese)[T.8]	2.408460	0.115117	20.9218	< 2.2e-16	***
as.factor(Mese)[T.9]	1.038598	0.141718	7.3286	3.15e-13	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

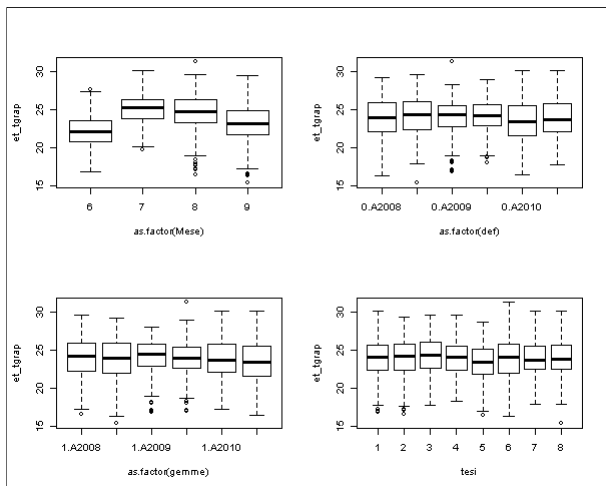
(Dispersion parameter for gaussian family taken to be 4.566856)

- Null deviance: 14597.0517 on 2437 degrees of freedom.
- Residual deviance: 11092.8934 on 2429 degrees of freedom.

AIC: 10633

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.929470118224397



media ore 11-18 temp. grappolo: diurnal_11_18_index_tgrap cacciagrande

Call: glm(formula = diurnal_11_18_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.85	-1.94	0.12	2.08	8.13

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.845723	0.181063	159.3132	< 2.2e-16	***
as.factor(def)[T.1]	-0.164280	0.124066	-1.3241	0.18558	
as.factor(gemme)[T.3]	-0.593122	0.124066	-4.7807	1.851e-06	***
as.factor(dirad)[T.1]	-0.004723	0.124066	-0.0381	0.96964	
AnnoF[T.A2009]	-0.051667	0.153935	-0.3356	0.73717	
AnnoF[T.A2010]	0.293515	0.153935	1.9067	0.05667	.
as.factor(Mese)[T.7]	4.776130	0.164048	29.1142	< 2.2e-16	***
as.factor(Mese)[T.8]	5.620599	0.164048	34.2619	< 2.2e-16	***
as.factor(Mese)[T.9]	2.465951	0.201956	12.2103	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

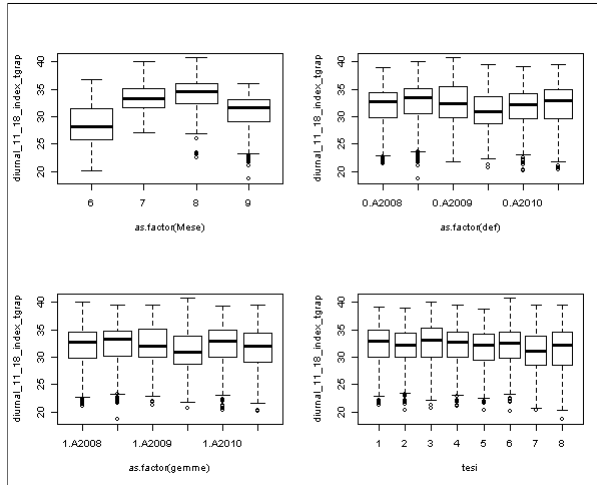
(Dispersion parameter for gaussian family taken to be 9.274266)

- Null deviance: 35681.1273 on 2437 degrees of freedom.
- Residual deviance: 22527.193 on 2429 degrees of freedom.

AIC: 12360

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.995462938565478



media ore 19-02 temp. grappolo: night_19_02_index_tgrap cacciagrande

•

Call: `glm(formula = night_19_02_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.8302	-1.3097	0.0034	1.2286	5.9746

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.375171	0.113612	179.3404	< 2.2e-16	***
as.factor(def)[T.1]	-0.257410	0.077848	-3.3066	0.0009582	***
as.factor(gemme)[T.3]	-0.011632	0.077848	-0.1494	0.8812392	
as.factor(dirad)[T.1]	0.077521	0.077848	0.9958	0.3194413	
AnnoF[T.A2009]	0.644071	0.096590	6.6681	3.197e-11	***
AnnoF[T.A2010]	-0.117980	0.096590	-1.2215	0.2220313	
as.factor(Mese)[T.7]	3.083083	0.102935	29.9516	< 2.2e-16	***
as.factor(Mese)[T.8]	3.323787	0.102935	32.2900	< 2.2e-16	***
as.factor(Mese)[T.9]	1.563833	0.126722	12.3407	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

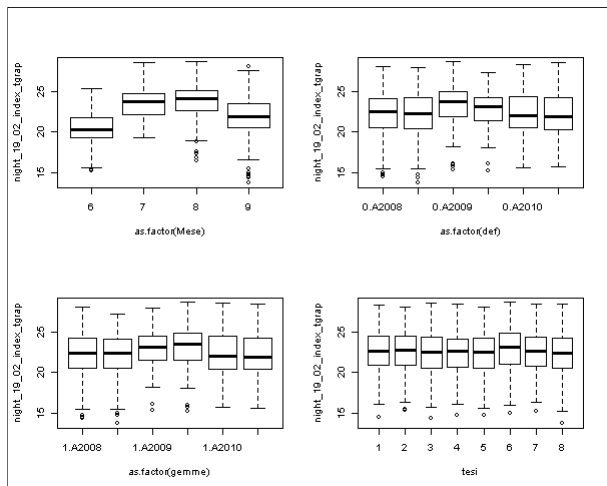
(Dispersion parameter for gaussian family taken to be 3.65146)

- Null deviance: 13984.8394 on 2437 degrees of freedom.
- Residual deviance: 8869.3964 on 2429 degrees of freedom.

AIC: 10087

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.880164849192521



errore standard mattutino tra le ore 5-10 temp. grappolo: dmorn_stderr_5_10_tgrap cacciagrande

```
Call: glm(formula = dmorn_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.098	-0.410	-0.018	0.426	2.156

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.616871	0.035765	45.2088	< 2.2e-16	***
as.factor(def)[T.1]	0.182442	0.024506	7.4447	1.341e-13	***
as.factor(gemme)[T.3]	-0.035131	0.024506	-1.4335	0.1518	
as.factor(dirad)[T.1]	-0.039766	0.024506	-1.6227	0.1048	
AnnoF[T.A2009]	-0.163214	0.030406	-5.3678	8.726e-08	***
AnnoF[T.A2010]	-0.038545	0.030406	-1.2677	0.2050	
as.factor(Mese)[T.7]	0.339548	0.032404	10.4787	< 2.2e-16	***
as.factor(Mese)[T.8]	0.543577	0.032404	16.7751	< 2.2e-16	***
as.factor(Mese)[T.9]	0.221830	0.039892	5.5608	2.979e-08	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

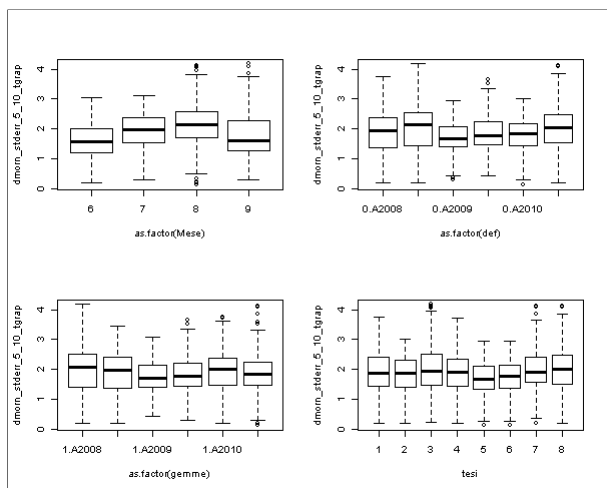
(Dispersion parameter for gaussian family taken to be 0.3618479)

- Null deviance: 1016.6669 on 2437 degrees of freedom.
- Residual deviance: 878.9286 on 2429 degrees of freedom.

AIC: 4451.4

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.161092712191829



varianza mattutina tra le ore 5-10 temp. grappolo: dmorn_var_5_10_tgrap cacciagrande

```
Call: glm(formula = dmorn_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =
```

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-34.3	-10.1	-2.4	8.5	75.2

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	18.14155	0.85835	21.1354	< 2.2e-16	***
as.factor(def)[T.1]	5.07319	0.58815	8.6257	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.95655	0.58815	-1.6264	0.10400	
as.factor(dirad)[T.1]	-1.34102	0.58815	-2.2801	0.02269	*
AnnoF[T.A2009]	-4.98469	0.72975	-6.8307	1.064e-11	***
AnnoF[T.A2010]	-1.69570	0.72975	-2.3237	0.02022	*
as.factor(Mese)[T.7]	7.02284	0.77769	9.0304	< 2.2e-16	***
as.factor(Mese)[T.8]	13.03139	0.77769	16.7566	< 2.2e-16	***
as.factor(Mese)[T.9]	6.31372	0.95740	6.5947	5.211e-11	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

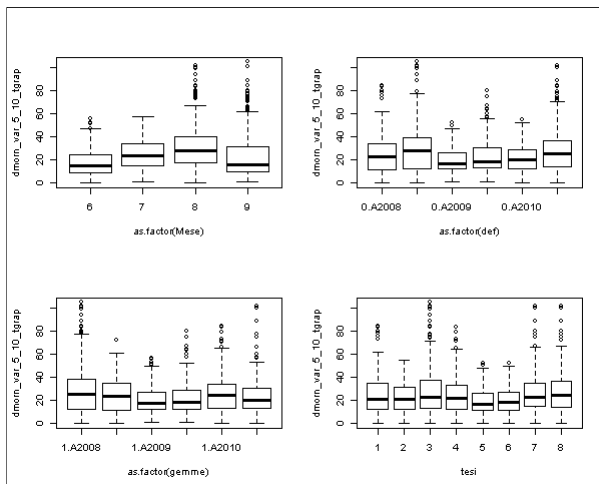
(Dispersion parameter for gaussian family taken to be 208.4239)

- Null deviance: 591174.5803 on 2437 degrees of freedom.
- Residual deviance: 506261.6217 on 2429 degrees of freedom.

AIC: 19948

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 temp. grappolo: dtwi_stderr_15_20_tgrap cacciagrande

Call: glm(formula = dtwi_stderr_15_20_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.2891	-0.2440	-0.0048	0.2668	1.2823

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.1275460	0.0236163	47.7443	< 2.2e-16	***
as.factor(def)[T.1]	0.0276103	0.0161821	1.7062	0.088094	.
as.factor(gemme)[T.3]	-0.0880611	0.0161821	-5.4419	5.800e-08	***
as.factor(dirad)[T.1]	0.0041656	0.0161821	0.2574	0.796874	
AnnoF[T.A2009]	-0.1289375	0.0200780	-6.4218	1.614e-10	***
AnnoF[T.A2010]	0.0550185	0.0200780	2.7402	0.006184	**
as.factor(Mese)[T.7]	0.2033194	0.0213971	9.5022	< 2.2e-16	***
as.factor(Mese)[T.8]	0.4098589	0.0213971	19.1549	< 2.2e-16	***
as.factor(Mese)[T.9]	0.3552857	0.0263415	13.4877	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

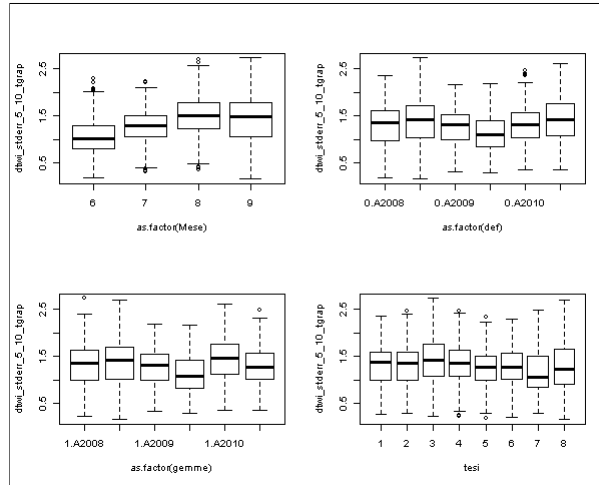
(Dispersion parameter for gaussian family taken to be 0.1577778)

- Null deviance: 468.0557 on 2437 degrees of freedom.
- Residual deviance: 383.2422 on 2429 degrees of freedom.

AIC: 2427.8

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.195729858433260



varianza serale tra le ore 15-20 temp. grappolo: dtwi_var_15_20_tgrap cacciagrande

Call: `glm(formula = dtwi_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.47	-4.29	-0.88	3.76	29.50

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	8.583742	0.381801	22.4823	< 2.2e-16	***
as.factor(def)[T.1]	0.855555	0.261613	3.2703	0.001089	**
as.factor(gemme)[T.3]	-1.236144	0.261613	-4.7251	2.431e-06	***
as.factor(dirad)[T.1]	0.062948	0.261613	0.2406	0.809872	
AnnoF[T.A2009]	-2.473198	0.324597	-7.6193	3.631e-14	***
AnnoF[T.A2010]	0.554709	0.324597	1.7089	0.087595	.
as.factor(Mese)[T.7]	2.738237	0.345922	7.9158	3.703e-15	***
as.factor(Mese)[T.8]	6.538210	0.345922	18.9008	< 2.2e-16	***
as.factor(Mese)[T.9]	6.255260	0.425858	14.6886	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

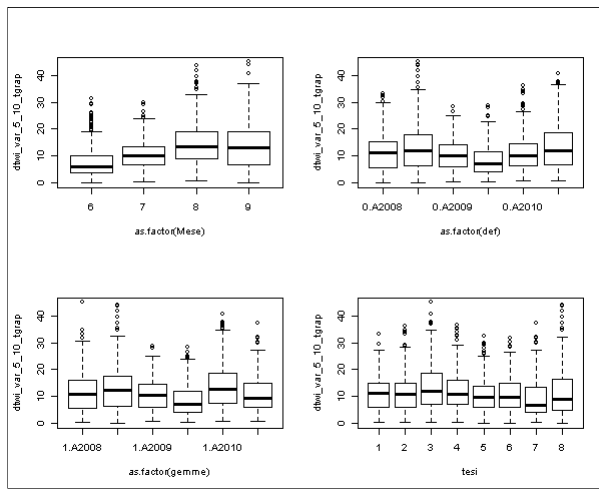
(Dispersion parameter for gaussian family taken to be 41.23764)

- Null deviance: 123755.9285 on 2437 degrees of freedom.
- Residual deviance: 100166.2299 on 2429 degrees of freedom.

AIC: 15998

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.999937217880801



Analisi indici giornalieri tarja canopy cacciagrande

temperatura canopy media: avg_tair cacciagrande

Call: `glm(formula = avg_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.975	-1.424	0.040	1.449	5.576

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	24.403547	0.122876	198.6023	< 2.2e-16	***
as.factor(def)[T.1]	0.136887	0.083534	1.6387	0.10140	.
as.factor(gemme)[T.3]	-0.147023	0.083534	-1.7600	0.07852	.
as.factor(dirad)[T.1]	-0.141141	0.083534	-1.6896	0.09122	.
AnnoF[T.A2009]	0.250330	0.102307	2.4468	0.01448	*
AnnoF[T.A2010]	-0.136624	0.102307	-1.3354	0.18186	
as.factor(Mese)[T.7]	3.605546	0.111091	32.4557	< 2.2e-16	***
as.factor(Mese)[T.8]	3.321202	0.111091	29.8962	< 2.2e-16	***
as.factor(Mese)[T.9]	1.099432	0.136762	8.0390	1.375e-15	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

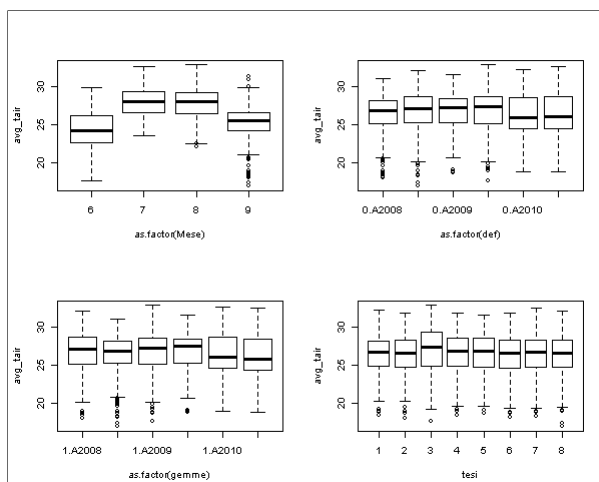
(Dispersion parameter for gaussian family taken to be 4.437912)

- Null deviance: 17562.6518 on 2543 degrees of freedom.
- Residual deviance: 11250.1067 on 2535 degrees of freedom.

AIC: 11022

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.917290593885808



cumulata temperatura canopy: sum_tair cacciagrande

Call: glm(formula = sum_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-191.40	-34.18	0.95	34.77	133.82

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	585.6852	2.9490	198.6024	< 2.2e-16	***
as.factor(def)[T.1]	3.2853	2.0048	1.6387	0.10140	
as.factor(gemme)[T.3]	-3.5286	2.0048	-1.7600	0.07852	.
as.factor(dirad)[T.1]	-3.3874	2.0048	-1.6896	0.09122	.
AnnoF[T.A2009]	6.0079	2.4554	2.4468	0.01448	*
AnnoF[T.A2010]	-3.2789	2.4554	-1.3354	0.18187	
as.factor(Mese)[T.7]	86.5331	2.6662	32.4557	< 2.2e-16	***
as.factor(Mese)[T.8]	79.7088	2.6662	29.8962	< 2.2e-16	***
as.factor(Mese)[T.9]	26.3864	3.2823	8.0390	1.375e-15	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

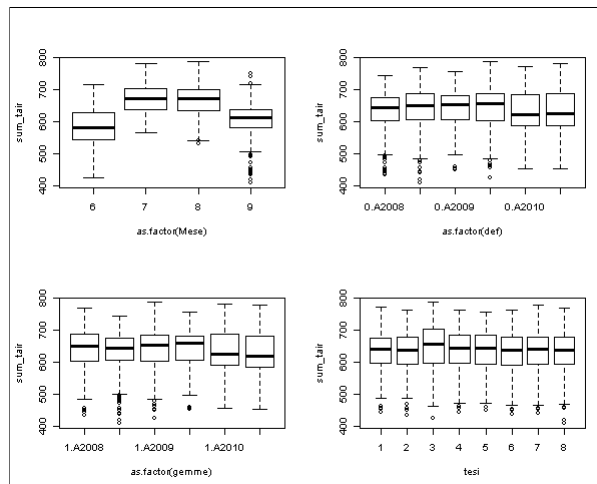
(Dispersion parameter for gaussian family taken to be 2556.235)

- Null deviance: 10116076.1445 on 2543 degrees of freedom.
- Residual deviance: 6480054.5913 on 2535 degrees of freedom.

AIC: 27192

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica canopy: et_tair cacciagrande

Call: glm(formula = et_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.03	-2.41	-0.24	2.71	10.00

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.2734977	0.2016425	65.8269	< 2.2e-16	***
as.factor(def)[T.1]	0.0018182	0.1370800	0.0133	0.9894182	
as.factor(gemme)[T.3]	0.1548289	0.1370800	1.1295	0.2588029	
as.factor(dirad)[T.1]	0.5191645	0.1370800	3.7873	0.0001558	***
AnnoF[T.A2009]	0.0534633	0.1678880	0.3184	0.7501727	
AnnoF[T.A2010]	0.6400165	0.1678880	3.8122	0.0001410	***
as.factor(Mese)[T.7]	2.1773072	0.1823026	11.9434	< 2.2e-16	***
as.factor(Mese)[T.8]	2.5447665	0.1823026	13.9590	< 2.2e-16	***
as.factor(Mese)[T.9]	0.3590601	0.2244289	1.5999	0.1097490	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

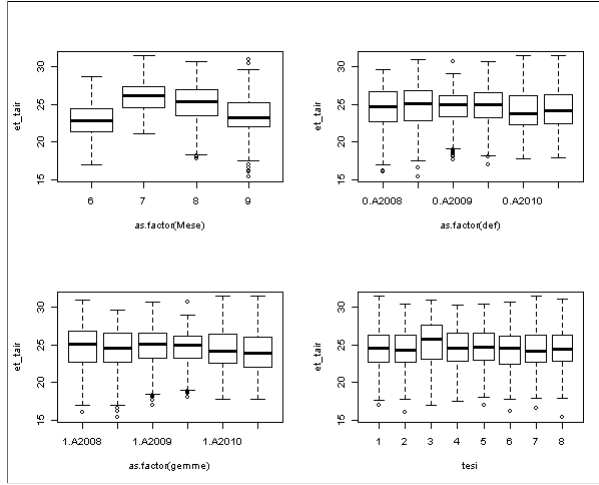
(Dispersion parameter for gaussian family taken to be 11.95103)

- Null deviance: 33872.5834 on 2543 degrees of freedom.
- Residual deviance: 30295.8659 on 2535 degrees of freedom.

AIC: 13542

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.75486531297534



media ore 03-10 temp. canopy: morning_03_10_index_tair cacciagrande

Call: glm(formula = morning_03_10_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.998	-1.494	-0.034	1.589	6.782

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.240751	0.129044	180.0999	< 2.2e-16	***
as.factor(def)[T.1]	0.211827	0.087726	2.4146	0.015821	*
as.factor(gemme)[T.3]	-0.268020	0.087726	-3.0552	0.002273	**
as.factor(dirad)[T.1]	-0.275668	0.087726	-3.1424	0.001695	**
AnnoF[T.A2009]	0.058540	0.107442	0.5448	0.585907	
AnnoF[T.A2010]	-0.427389	0.107442	-3.9779	7.149e-05	***
as.factor(Mese)[T.7]	3.032190	0.116667	25.9902	< 2.2e-16	***
as.factor(Mese)[T.8]	2.307272	0.116667	19.7766	< 2.2e-16	***
as.factor(Mese)[T.9]	0.614081	0.143626	4.2755	1.977e-05	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

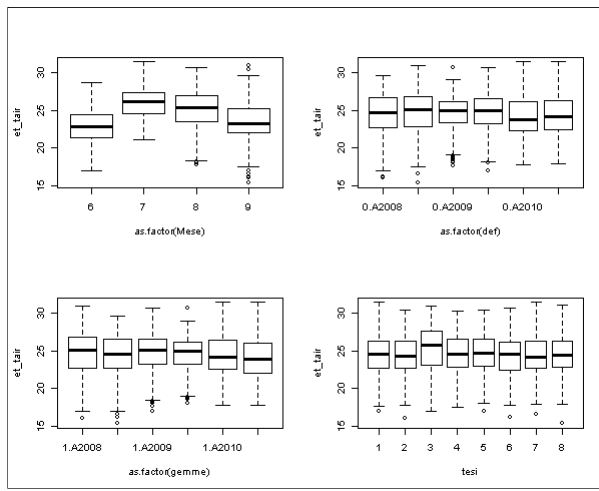
(Dispersion parameter for gaussian family taken to be 4.894573)

- Null deviance: 16673.6137 on 2543 degrees of freedom.
- Residual deviance: 12407.7437 on 2535 degrees of freedom.

AIC: 11271

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.75486531297534



media ore 11-18 temp. canopy: diurnal_11_18_index_tair cacciagrande

Call: glm(formula = diurnal_11_18_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.94	-1.90	0.17	1.98	7.21

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	29.175537	0.177539	164.3336	< 2.2e-16	***
as.factor(def)[T.1]	0.085551	0.120694	0.7088	0.47850	
as.factor(gemme)[T.3]	-0.028016	0.120694	-0.2321	0.81646	
as.factor(dirad)[T.1]	0.133381	0.120694	1.1051	0.26921	
AnnoF[T.A2009]	0.403221	0.147819	2.7278	0.00642	**
AnnoF[T.A2010]	0.189152	0.147819	1.2796	0.20080	
as.factor(Mese)[T.7]	4.638598	0.160510	28.8990	< 2.2e-16	***
as.factor(Mese)[T.8]	4.610192	0.160510	28.7221	< 2.2e-16	***
as.factor(Mese)[T.9]	1.301496	0.197601	6.5865	5.457e-11	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

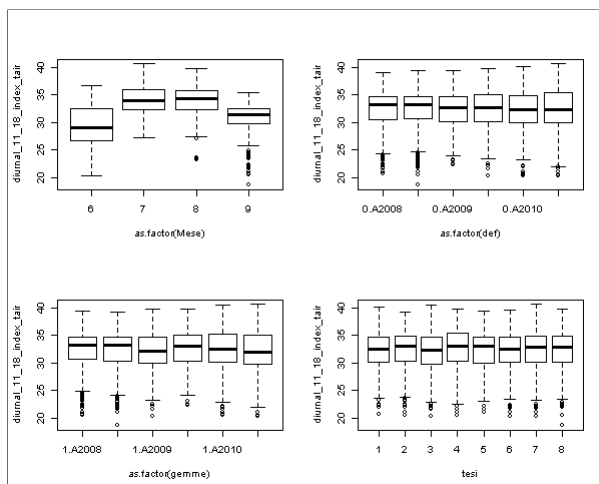
(Dispersion parameter for gaussian family taken to be 9.264596)

- Null deviance: 34769.0992 on 2543 degrees of freedom.
- Residual deviance: 23485.7511 on 2535 degrees of freedom.

AIC: 12894

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.988149379469373



media ore 19-02 temp. canopy: night_19_02_index_tair cacciagrande

Call: glm(formula = night_19_02_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.967	-1.395	-0.029	1.353	7.692

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.794362	0.118548	175.4083	< 2.2e-16	***
as.factor(def)[T.1]	0.113280	0.080591	1.4056	0.1599611	.
as.factor(gemme)[T.3]	-0.145036	0.080591	-1.7997	0.0720338	.
as.factor(dirad)[T.1]	-0.281137	0.080591	-3.4884	0.0004941	***
AnnoF[T.A2009]	0.289228	0.098704	2.9303	0.0034170	**
AnnoF[T.A2010]	-0.171627	0.098704	-1.7388	0.0821895	.
as.factor(Mese)[T.7]	3.145841	0.107178	29.3515	< 2.2e-16	***
as.factor(Mese)[T.8]	3.046134	0.107178	28.4212	< 2.2e-16	***
as.factor(Mese)[T.9]	1.382719	0.131945	10.4795	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

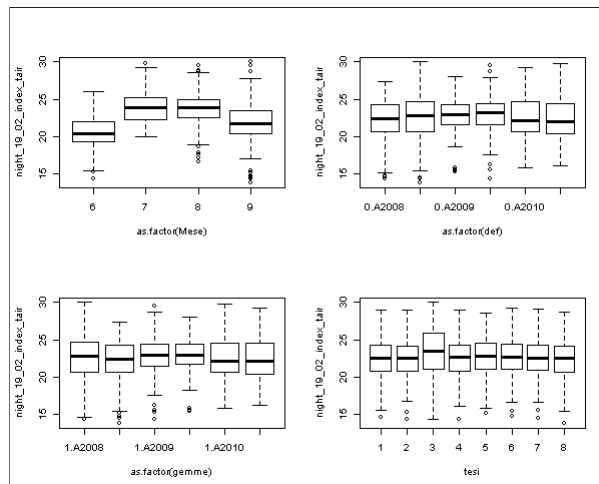
(Dispersion parameter for gaussian family taken to be 4.130784)

- Null deviance: 15347.4828 on 2543 degrees of freedom.
- Residual deviance: 10471.5366 on 2535 degrees of freedom.

AIC: 10839

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.854951217793666



errore standard mattutino tra le ore 5-10 temp. canopy: dmorn_stderr_5_10_tair cacciagrande

Call: glm(formula = dmorn_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.0903	-0.3863	-0.0012	0.4451	1.7493

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.7397029	0.0342991	50.7215	< 2.2e-16	***
as.factor(def)[T.1]	-0.0018404	0.0233171	-0.0789	0.937095	.
as.factor(gemme)[T.3]	0.0115824	0.0233171	0.4967	0.619420	.
as.factor(dirad)[T.1]	0.1123758	0.0233171	4.8195	1.525e-06	***
AnnoF[T.A2009]	-0.0794618	0.0285575	-2.7825	0.005434	**
AnnoF[T.A2010]	0.0224024	0.0285575	0.7845	0.432841	.
as.factor(Mese)[T.7]	0.2997714	0.0310094	9.6671	< 2.2e-16	***
as.factor(Mese)[T.8]	0.3628525	0.0310094	11.7014	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.1113775	0.0381751	-2.9175	0.003559	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

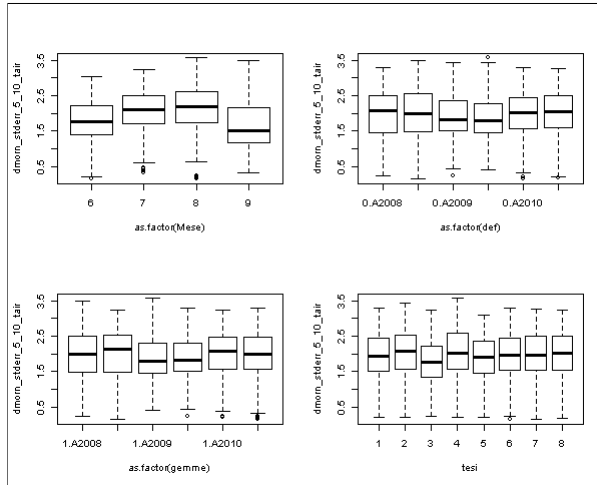
(Dispersion parameter for gaussian family taken to be 0.3457859)

- Null deviance: 978.1992 on 2543 degrees of freedom.
- Residual deviance: 876.5671 on 2535 degrees of freedom.

AIC: 4529

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.122682038598181



varianza mattutina tra le ore 5-10 temp. canopy: dmorn_var_5_10_tair cacciagrande

Call: `glm(formula = dmorn_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-32.0	-10.2	-2.1	9.7	51.6

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.325155	0.793244	25.6228	< 2.2e-16	***
as.factor(def)[T.1]	0.087223	0.539261	0.1617	0.8715197	
as.factor(gemme)[T.3]	0.140971	0.539261	0.2614	0.7937944	
as.factor(dirad)[T.1]	2.780830	0.539261	5.1567	2.707e-07	***
AnnoF[T.A2009]	-2.296008	0.660457	-3.4764	0.0005168	***
AnnoF[T.A2010]	0.096102	0.660457	0.1455	0.8843216	
as.factor(Mese)[T.7]	6.661389	0.717163	9.2885	< 2.2e-16	***
as.factor(Mese)[T.8]	8.753226	0.717163	12.2054	< 2.2e-16	***
as.factor(Mese)[T.9]	-1.789288	0.882884	-2.0266	0.0428038	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

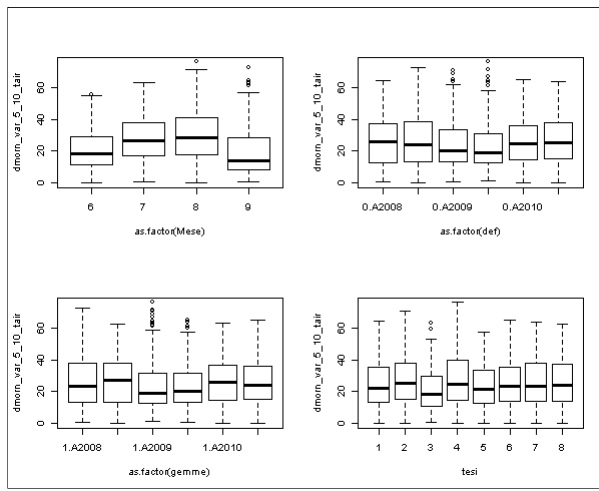
(Dispersion parameter for gaussian family taken to be 184.9505)

- Null deviance: 522010.8923 on 2543 degrees of freedom.
- Residual deviance: 468849.4747 on 2535 degrees of freedom.

AIC: 20510

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99999999159298



errore standard serale tra le ore 15-20 temp. canopy: dtwi_stderr_15_20_tair cacciagrande

Call: glm(formula = dtwi_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.2582	-0.2196	-0.0091	0.2283	1.2946

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.0618217	0.0224482	47.3009	< 2.2e-16	***
as.factor(def)[T.1]	0.0052689	0.0152607	0.3453	0.729925	
as.factor(gemme)[T.3]	0.0104856	0.0152607	0.6871	0.492083	
as.factor(dirad)[T.1]	0.0673691	0.0152607	4.4146	1.055e-05	***
AnnoF[T.A2009]	0.0576103	0.0186904	3.0823	0.002076	**
AnnoF[T.A2010]	0.0915713	0.0186904	4.8994	1.022e-06	***
as.factor(Mese)[T.7]	0.1742623	0.0202952	8.5864	< 2.2e-16	***
as.factor(Mese)[T.8]	0.3428951	0.0202952	16.8954	< 2.2e-16	***
as.factor(Mese)[T.9]	0.2724248	0.0249850	10.9035	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 0.1481171)

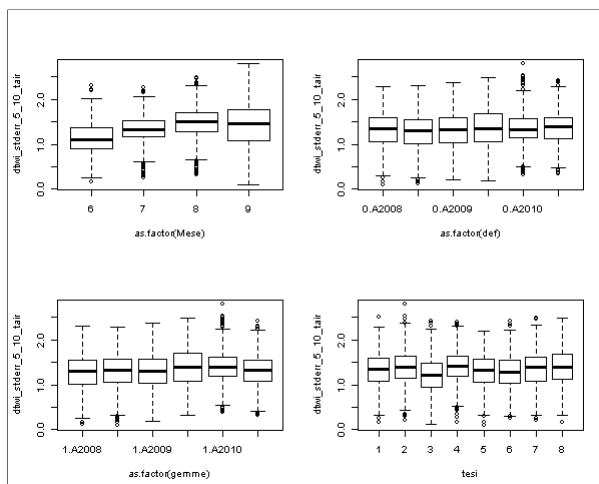
• Null deviance: 427.2334 on 2543 degrees of freedom.

• Residual deviance: 375.4769 on 2535 degrees of freedom.

AIC: 2372.1

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.130270660818011



varianza serale tra le ore 15-20 temp. canopy: dtwi_var_15_20_taircacciagrande

Call: glm(formula = dtwi_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

cacciagrande_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.65	-4.11	-0.96	3.11	31.89

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	7.36898	0.36629	20.1179	< 2.2e-16	***
as.factor(def)[T.1]	0.19925	0.24901	0.8002	0.423678	
as.factor(gemme)[T.3]	0.10589	0.24901	0.4252	0.670696	
as.factor(dirad)[T.1]	1.05163	0.24901	4.2232	2.493e-05	***
AnnoF[T.A2009]	0.93643	0.30497	3.0705	0.002160	**
AnnoF[T.A2010]	1.23072	0.30497	4.0355	5.610e-05	***
as.factor(Mese)[T.7]	2.47532	0.33116	7.4747	1.059e-13	***
as.factor(Mese)[T.8]	5.52153	0.33116	16.6734	< 2.2e-16	***
as.factor(Mese)[T.9]	5.08494	0.40768	12.4728	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

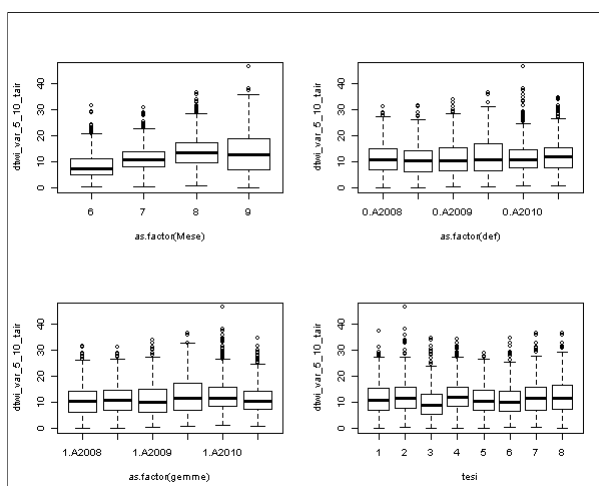
(Dispersion parameter for gaussian family taken to be 39.43595)

- Null deviance: 114161.8853 on 2543 degrees of freedom.
- Residual deviance: 99970.1229 on 2535 degrees of freedom.

AIC: 16579

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.996221858156876



Analisi statistica campione Tuscania cortigliano

IBIMET CNR, Istituto di Biometeorologia di Firenze -CNR

Matese.A., Genesio L., Crisci A.

Temperatura Grappolo cortigliano

Temperatura Grappolo Mattina

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-10.97	-2.53	-0.15	2.43	14.22

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.958008	0.120799	165.2170	< 2.2e-16	***
as.factor(def)[T.1]	0.734746	0.081437	9.0222	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.108216	0.081437	-1.3288	0.1839	
as.factor(dirad)[T.1]	0.011733	0.081437	0.1441	0.8854	

AnnoF[T.A2009]	0.420359	0.099963	4.2052	2.64e-05	***
AnnoF[T.A2010]	0.011198	0.099881	0.1121	0.9107	
as.factor(Mese)[T.7]	2.178058	0.110409	19.7273	< 2.2e-16	***
as.factor(Mese)[T.8]	2.972163	0.107903	27.5447	< 2.2e-16	***
as.factor(Mese)[T.9]	1.704053	0.130436	13.0643	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

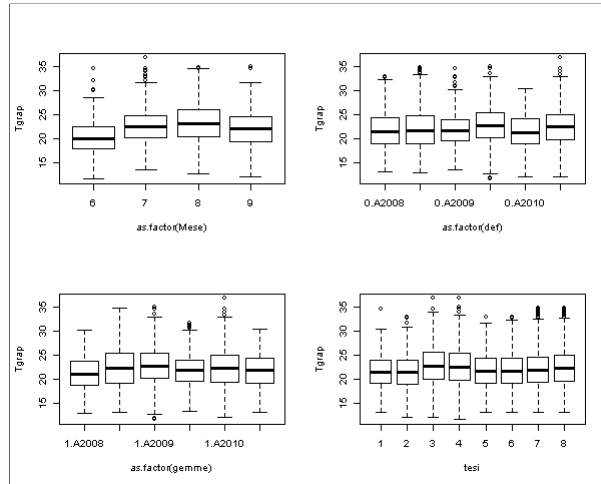
(Dispersion parameter for gaussian family taken to be 12.23269)

- Null deviance: 101207.2125 on 7377 degrees of freedom.
- Residual deviance: 90142.6624 on 7369 degrees of freedom.

AIC: 39424

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.776796594241002



Temperatura Grappolo Pieno giorno

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-18.38	-2.15	0.13	2.38	13.55

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.377217	0.072237	392.8373	< 2.2e-16	***
as.factor(def)[T.1]	0.567788	0.049111	11.5612	< 2.2e-16	***
as.factor(gemme)[T.3]	0.077555	0.049111	1.5792	0.1143104	
as.factor(dirad)[T.1]	0.190957	0.049111	3.8882	0.0001013	***
AnnoF[T.A2009]	-0.521578	0.060266	-8.6546	< 2.2e-16	***
AnnoF[T.A2010]	-1.310622	0.060266	-21.7472	< 2.2e-16	***
as.factor(Mese)[T.7]	4.109339	0.063801	64.4084	< 2.2e-16	***
as.factor(Mese)[T.8]	5.994868	0.065611	91.3703	< 2.2e-16	***
as.factor(Mese)[T.9]	3.009559	0.083541	36.0248	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

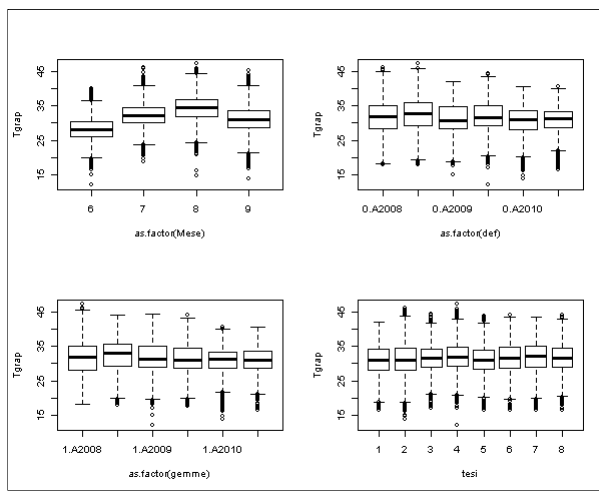
(Dispersion parameter for gaussian family taken to be 12.8151)

- Null deviance: 393424.7035 on 21252 degrees of freedom.
- Residual deviance: 272243.9702 on 21244 degrees of freedom.

AIC: 114533

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.996660120353337



Temperatura Cross Grappolo Pieno giorno

Call: `glm(formula = Tgrap ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-18.40	-2.19	0.13	2.40	13.65

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	26.8982171	0.6996193	38.4469	< 2.2e-16	***
as.factor(gemme)[T.3]	-1.6365283	0.8081883	-2.0249	0.04289	*
as.factor(def)[T.1]	-1.5203195	0.8170577	-1.8607	0.06280	.
Rad	0.0033288	0.0016282	2.0444	0.04092	*
AnnoF[T.A2009]	-0.3779735	0.0621174	-6.0848	1.187e-09	***
AnnoF[T.A2010]	-1.1672683	0.0621427	-18.7837	< 2.2e-16	***
as.factor(Mese)[T.7]	4.1832350	0.0667438	62.6760	< 2.2e-16	***
as.factor(Mese)[T.8]	6.0937758	0.0690238	88.2852	< 2.2e-16	***
as.factor(Mese)[T.9]	3.1436623	0.0859428	36.5785	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.1764448	0.1014137	-1.7399	0.08190	.
as.factor(def)[T.1]:Rad	0.0047866	0.0019074	2.5096	0.01210	*
as.factor(gemme)[T.3]:Rad	0.0041071	0.0019107	2.1496	0.03160	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 12.79047)

• Null deviance: 379681.4159 on 20636 degrees of freedom.

• Residual deviance: 263803.5286 on 20625 degrees of freedom.

AIC: 111177

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99635740035453

Temperatura Grappolo Sera

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.6	-2.2	-0.1	2.1	14.5

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	27.469500	0.119157	230.5318	< 2.2e-16	***
as.factor(def)[T.1]	0.232230	0.080252	2.8938	0.003818	**
as.factor(gemme)[T.3]	-0.402502	0.080252	-5.0155	5.413e-07	***
as.factor(dirad)[T.1]	0.316140	0.080252	3.9393	8.246e-05	***
AnnoF[T.A2009]	-0.664094	0.098481	-6.7434	1.664e-11	***

AnnoF[T.A2010]	-2.133926	0.098481	-21.6685	< 2.2e-16	***
as.factor(Mese)[T.7]	3.620381	0.106596	33.9634	< 2.2e-16	***
as.factor(Mese)[T.8]	4.745639	0.108275	43.8296	< 2.2e-16	***
as.factor(Mese)[T.9]	2.704828	0.129705	20.8536	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

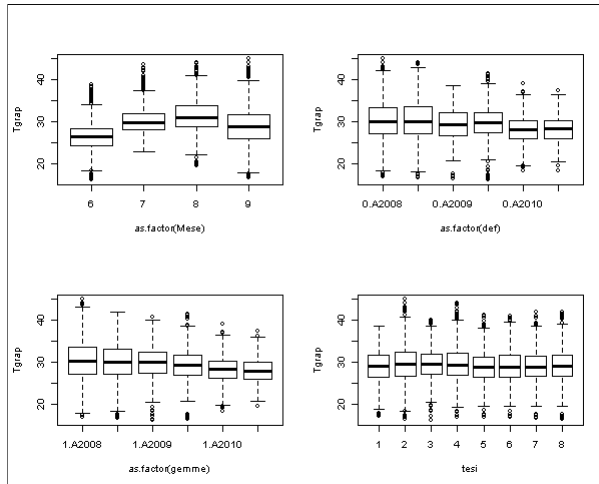
(Dispersion parameter for gaussian family taken to be 11.93077)

- Null deviance: 119936.4703 on 7409 degrees of freedom.
- Residual deviance: 88299.6468 on 7401 degrees of freedom.

AIC: 39410

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.986010992594123



Temperatura Grappolo Notte

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.91	-2.04	-0.16	1.95	16.37

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.797263	0.057382	345.0062	< 2.2e-16	***
as.factor(def)[T.1]	0.088954	0.037809	2.3527	0.01865	*
as.factor(gemme)[T.3]	-0.315386	0.037809	-8.3415	< 2.2e-16	***
as.factor(dirad)[T.1]	-0.277114	0.037809	-7.3293	2.386e-13	***
AnnoF[T.A2009]	0.469846	0.046389	10.1285	< 2.2e-16	***
AnnoF[T.A2010]	-0.561323	0.046400	-12.0975	< 2.2e-16	***
as.factor(Mese)[T.7]	2.898640	0.052436	55.2793	< 2.2e-16	***
as.factor(Mese)[T.8]	3.324763	0.051322	64.7823	< 2.2e-16	***
as.factor(Mese)[T.9]	2.169600	0.060008	36.1550	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

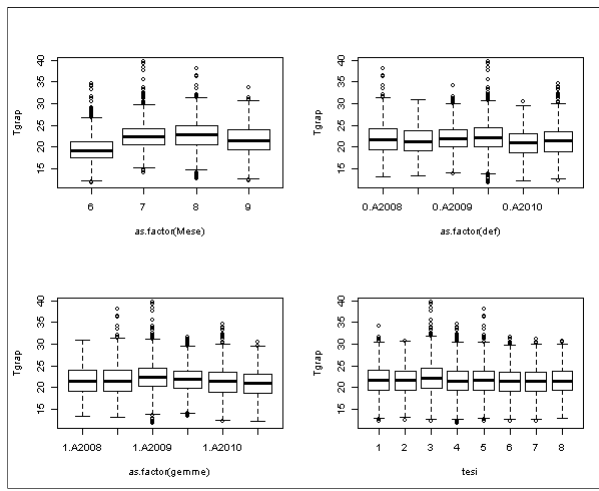
(Dispersion parameter for gaussian family taken to be 8.785526)

- Null deviance: 263202.0908 on 24582 degrees of freedom.
- Residual deviance: 215895.5089 on 24574 degrees of freedom.

AIC: 123196

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.854050217094768



Radiazione Grappolo Mattina

Call: `glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-395	-24	8	33	109

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	391.1087	1.7341	225.5365	< 2.2e-16	***
as.factor(def)[T.1]	12.3884	1.1813	10.4874	< 2.2e-16	***
as.factor(gemme)[T.3]	-5.7125	1.1813	-4.8359	1.353e-06	***
as.factor(dirad)[T.1]	3.0504	1.1814	2.5820	0.009844	**
AnnoF[T.A2009]	-9.7347	1.4623	-6.6570	3.000e-11	***
AnnoF[T.A2010]	-11.0909	1.4612	-7.5901	3.600e-14	***
as.factor(Mese)[T.7]	-16.8331	1.6087	-10.4639	< 2.2e-16	***
as.factor(Mese)[T.8]	-5.0539	1.5635	-3.2324	0.001233	**
as.factor(Mese)[T.9]	-29.2098	1.8761	-15.5690	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

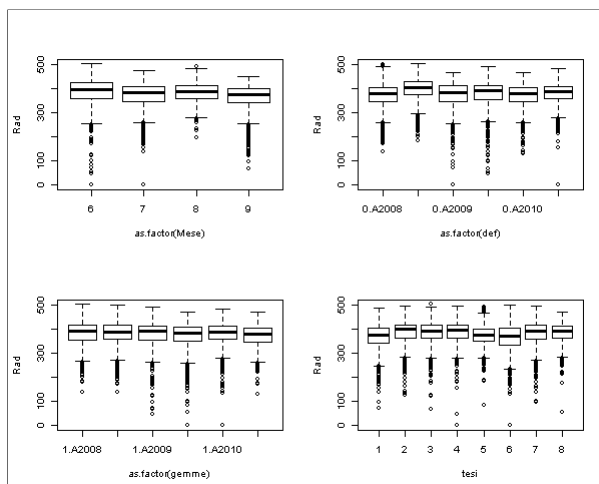
(Dispersion parameter for gaussian family taken to be 2491.609)

- Null deviance: 19104412.8532 on 7175 degrees of freedom.
- Residual deviance: 17857362.3256 on 7167 degrees of freedom.

AIC: 76497

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiazione Grappolo Pieno giorno

Call: `glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))`

= ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-168.5	-16.0	-3.4	12.9	84.3

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	436.71026	0.50997	856.3410	< 2.2e-16	***
as.factor(def)[T.1]	9.62021	0.35079	27.4246	< 2.2e-16	***
as.factor(gemme)[T.3]	-1.22795	0.35079	-3.5006	0.0004653	***
as.factor(dirad)[T.1]	1.54254	0.35084	4.3967	1.104e-05	***
AnnoF[T.A2009]	-5.01505	0.43443	-11.5441	< 2.2e-16	***
AnnoF[T.A2010]	-5.08906	0.43443	-11.7144	< 2.2e-16	***
as.factor(Mese)[T.7]	-14.75822	0.45720	-32.2795	< 2.2e-16	***
as.factor(Mese)[T.8]	-18.39609	0.46747	-39.3523	< 2.2e-16	***
as.factor(Mese)[T.9]	-17.80518	0.59038	-30.1589	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

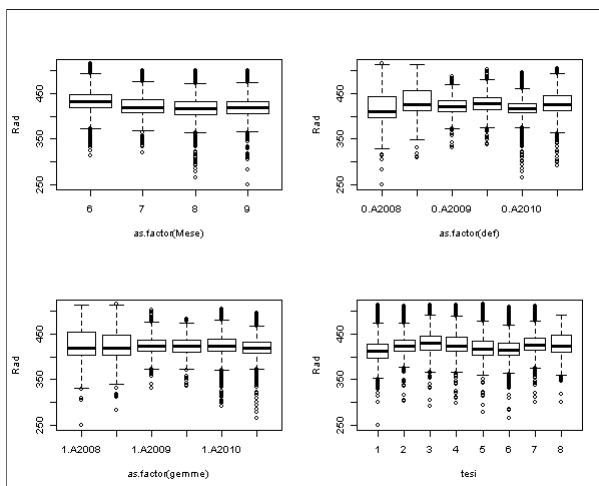
(Dispersion parameter for gaussian family taken to be 631.512)

- Null deviance: 14785677.7202 on 20636 degrees of freedom.
- Residual deviance: 13026829.2848 on 20628 degrees of freedom.

AIC: 191646

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiation Cross Grappolo Pieno giorno

Call: glm(formula = Rad ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Tgrap + as.factor(gemme) * Tgrap + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-168.7	-16.1	-3.2	13.0	84.9

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	436.867759	2.215594	197.1786	< 2.2e-16	***
as.factor(gemme)[T.3]	-7.744856	2.602818	-2.9756	0.002928	**
as.factor(def)[T.1]	-5.635727	2.625485	-2.1465	0.031841	*
Tgrap	0.049710	0.071942	0.6910	0.489586	
AnnoF[T.A2009]	-4.740604	0.434617	-10.9075	< 2.2e-16	***
AnnoF[T.A2010]	-4.478745	0.437874	-10.2284	< 2.2e-16	***
as.factor(Mese)[T.7]	-16.291285	0.497913	-32.7192	< 2.2e-16	***
as.factor(Mese)[T.8]	-20.642706	0.549995	-37.5325	< 2.2e-16	***
as.factor(Mese)[T.9]	-18.919534	0.607517	-31.1424	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.233920	0.701694	-0.3334	0.738862	
as.factor(def)[T.1]:Tgrap	0.480208	0.081843	5.8674	4.494e-09	***
as.factor(gemme)[T.3]:Tgrap	0.209181	0.081821	2.5566	0.010578	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 629.2172)

- Null deviance: 14785677.7202 on 20636 degrees of freedom.
- Residual deviance: 12977604.8949 on 20625 degrees of freedom.

AIC: 191574

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1

Radiation Grappolo Sera

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-257.1	-29.1	7.6	37.1	112.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	400.9112	1.8220	220.0441	< 2.2e-16	***
as.factor(def)[T.1]	10.9953	1.2406	8.8627	< 2.2e-16	***
as.factor(gemme)[T.3]	3.3058	1.2406	2.6646	0.007724	**
as.factor(dirad)[T.1]	3.4495	1.2408	2.7801	0.005449	**
AnnoF[T.A2009]	-14.1294	1.5360	-9.1986	< 2.2e-16	***
AnnoF[T.A2010]	-2.7665	1.5360	-1.8011	0.071736	.
as.factor(Mese)[T.7]	-24.8194	1.6541	-15.0051	< 2.2e-16	***
as.factor(Mese)[T.8]	-51.8322	1.6711	-31.0168	< 2.2e-16	***
as.factor(Mese)[T.9]	-31.5933	1.9868	-15.9017	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

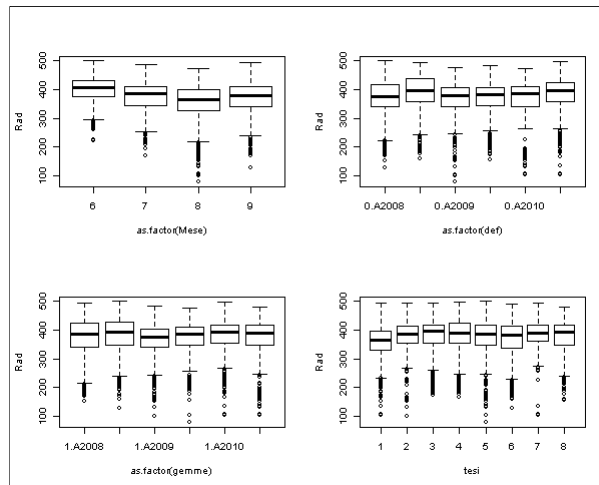
(Dispersion parameter for gaussian family taken to be 2756.577)

- Null deviance: 22958851.6565 on 7199 degrees of freedom.
- Residual deviance: 19822545.0936 on 7191 degrees of freedom.

AIC: 77480

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Temperatura Aria Canopy Mattina

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.57	-3.03	-0.19	2.92	22.94

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.493853	0.144350	148.9006	< 2.2e-16	***
as.factor(def)[T.1]	0.131192	0.097315	1.3481	0.17766	
as.factor(gemme)[T.3]	-0.128000	0.097315	-1.3153	0.18844	
as.factor(dirad)[T.1]	-0.136153	0.097315	-1.3991	0.16183	
AnnoF[T.A2009]	0.301557	0.119452	2.5245	0.01161	*
AnnoF[T.A2010]	-0.653319	0.119355	-5.4738	4.551e-08	***
as.factor(Mese)[T.7]	2.821590	0.131934	21.3863	< 2.2e-16	***
as.factor(Mese)[T.8]	3.801887	0.128941	29.4855	< 2.2e-16	***
as.factor(Mese)[T.9]	1.405965	0.155866	9.0203	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

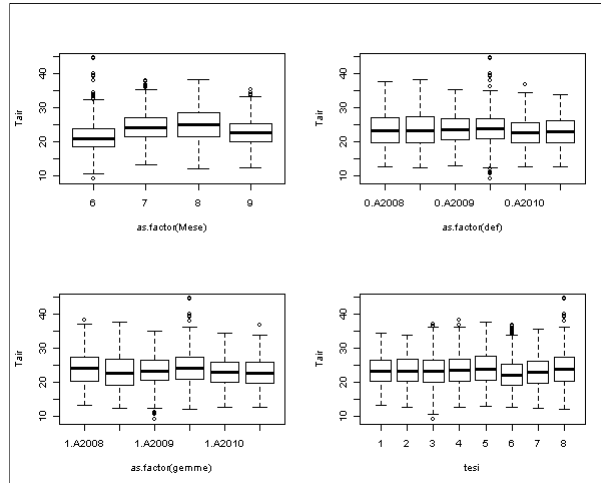
(Dispersion parameter for gaussian family taken to be 17.46759)

- Null deviance: 146814.4331 on 7377 degrees of freedom.
- Residual deviance: 128718.6742 on 7369 degrees of freedom.

AIC: 42052

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.913936040349835



Temperatura Aria Canopy Pieno giorno

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.06	-2.01	0.25	2.26	11.02

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	30.0881289	0.0669976	449.0928	< 2.2e-16	***
as.factor(def)[T.1]	0.4730956	0.0455495	10.3864	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.0059771	0.0455495	-0.1312	0.8956	
as.factor(dirad)[T.1]	-0.2600321	0.0455496	-5.7088	1.153e-08	***
AnnoF[T.A2009]	-0.5005442	0.0558953	-8.9550	< 2.2e-16	***
AnnoF[T.A2010]	-0.7120571	0.0558953	-12.7391	< 2.2e-16	***
as.factor(Mese)[T.7]	4.1598420	0.0591740	70.2984	< 2.2e-16	***
as.factor(Mese)[T.8]	4.5082684	0.0608522	74.0855	< 2.2e-16	***
as.factor(Mese)[T.9]	1.6004850	0.0774825	20.6561	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

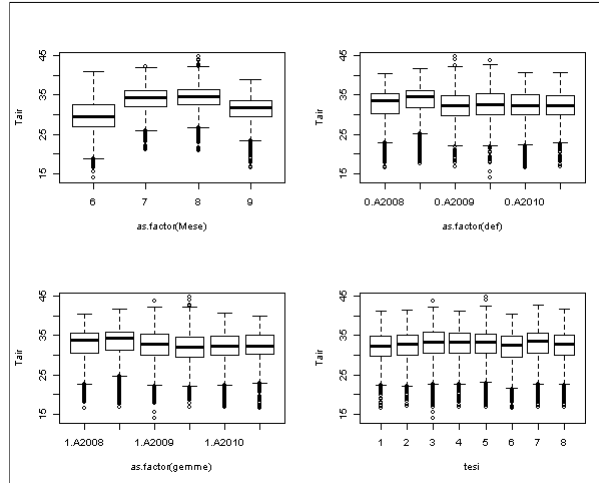
(Dispersion parameter for gaussian family taken to be 11.02366)

- Null deviance: 318077.6919 on 21252 degrees of freedom.
- Residual deviance: 234186.653 on 21244 degrees of freedom.

AIC: 111333

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.980692707510384



Temperatura Aria Crossing Canopy Pieno giorno

Call: `glm(formula = Tair ~ as.factor(gemme) * as.factor(def) + as.factor(dirad) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.22	-1.95	0.25	2.24	10.40

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	31.6205819	0.6475219	48.8332	< 2.2e-16 ***
as.factor(gemme)[T.3]	3.0535945	0.7485964	4.0791	4.538e-05 ***
as.factor(def)[T.1]	2.2189962	0.7564686	2.9334	0.0033568 **
as.factor(dirad)[T.1]	-0.3037098	0.0463737	-6.5492	5.921e-11 ***
Rad	-0.0035349	0.0015106	-2.3401	0.0192888 *
AnnoF[T.A2009]	-0.4898840	0.0574580	-8.5260	< 2.2e-16 ***
AnnoF[T.A2010]	-0.7071108	0.0574805	-12.3017	< 2.2e-16 ***
as.factor(Mese)[T.7]	4.0144915	0.0616828	65.0828	< 2.2e-16 ***
as.factor(Mese)[T.8]	4.3019917	0.0638041	67.4250	< 2.2e-16 ***
as.factor(Mese)[T.9]	1.4588154	0.0794480	18.3619	< 2.2e-16 ***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-0.4123210	0.0937523	-4.3980	1.098e-05 ***
as.factor(def)[T.1]:Rad	-0.0035152	0.0017660	-1.9905	0.0465524 *
as.factor(gemme)[T.3]:Rad	-0.0068183	0.0017698	-3.8526	0.0001172 ***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 10.92348)

• Null deviance: 305905.0342 on 20636 degrees of freedom.

• Residual deviance: 225285.7966 on 20624 degrees of freedom.

AIC: 107922

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.97989036885585

Temperatura Aria Canopy Sera

Call: `glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.634	-2.208	-0.054	2.207	21.907

Coefficients:

Estimate	Std. Error	t value	Pr(> t)
----------	------------	---------	----------

(Intercept)	27.520728	0.114805	239.7176	< 2.2e-16	***
as.factor(def)[T.1]	0.278942	0.077321	3.6076	0.0003111	***
as.factor(gemme)[T.3]	-0.104992	0.077321	-1.3579	0.1745445	
as.factor(dirad)[T.1]	-0.378874	0.077321	-4.9000	9.786e-07	***
AnnoF[T.A2009]	0.640852	0.094883	6.7541	1.546e-11	***
AnnoF[T.A2010]	-0.159426	0.094883	-1.6802	0.0929557	.
as.factor(Mese)[T.7]	3.157524	0.102703	30.7443	< 2.2e-16	***
as.factor(Mese)[T.8]	2.704740	0.104320	25.9274	< 2.2e-16	***
as.factor(Mese)[T.9]	0.922075	0.124968	7.3785	1.775e-13	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

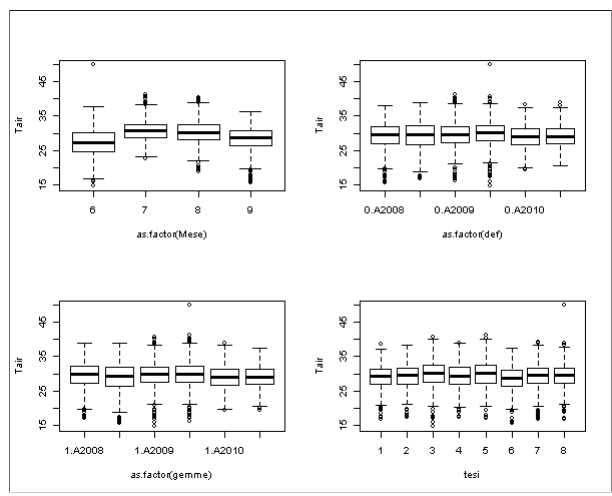
(Dispersion parameter for gaussian family taken to be 11.07513)

- Null deviance: 96333.786 on 7409 degrees of freedom.
- Residual deviance: 81967.0299 on 7401 degrees of freedom.

AIC: 38859

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.856130252553648



Temperatura Aria Canopy Notte

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.50	-2.10	-0.15	1.96	29.46

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.775703	0.060165	328.6930	< 2.2e-16	***
as.factor(def)[T.1]	-0.022881	0.039642	-0.5772	0.563825	
as.factor(gemme)[T.3]	-0.048466	0.039642	-1.2226	0.221497	
as.factor(dirad)[T.1]	-0.332463	0.039643	-8.3865	< 2.2e-16	***
AnnoF[T.A2009]	0.846661	0.048638	17.4075	< 2.2e-16	***
AnnoF[T.A2010]	-0.133619	0.048650	-2.7466	0.006027	**
as.factor(Mese)[T.7]	2.665704	0.054979	48.4861	< 2.2e-16	***
as.factor(Mese)[T.8]	2.610793	0.053811	48.5182	< 2.2e-16	***
as.factor(Mese)[T.9]	1.654574	0.062918	26.2973	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

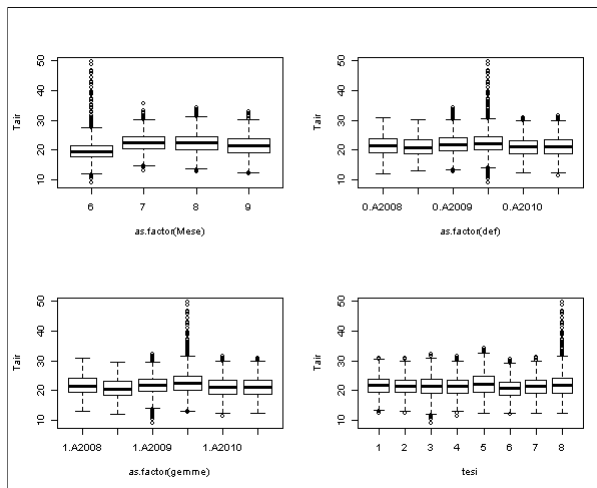
(Dispersion parameter for gaussian family taken to be 9.658155)

- Null deviance: 272080.8483 on 24582 degrees of freedom.
- Residual deviance: 237339.5068 on 24574 degrees of freedom.

AIC: 125524

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.756654929334545



Analisi indici giornalieri radiazione cortigliano

radiazione media : avg_rad cortigliano

Call: `glm(formula = avg_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-47.0802	-7.1309	-0.0061	7.2054	45.1565

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	276.53872	0.73225	377.6545	< 2.2e-16	***
as.factor(def)[T.1]	6.92675	0.49726	13.9298	< 2.2e-16	***
as.factor(gemme)[T.3]	-0.99943	0.49726	-2.0099	0.04455	*
as.factor(dirad)[T.1]	0.95901	0.49733	1.9283	0.05393	.
AnnoF[T.A2009]	-5.78063	0.61566	-9.3893	< 2.2e-16	***
AnnoF[T.A2010]	-3.88595	0.61566	-6.3118	3.263e-10	***
as.factor(Mese)[T.7]	-16.06743	0.66640	-24.1107	< 2.2e-16	***
as.factor(Mese)[T.8]	-37.46211	0.66258	-56.5399	< 2.2e-16	***
as.factor(Mese)[T.9]	-56.29396	0.80796	-69.6739	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 151.0928)

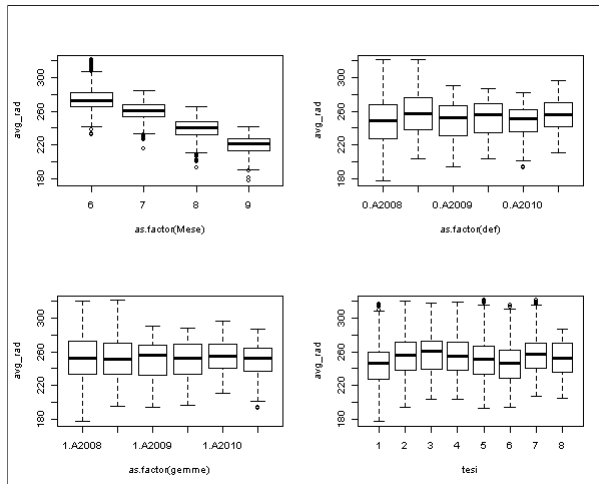
• Null deviance: 1328137.068 on 2455 degrees of freedom.

• Residual deviance: 369724.0114 on 2447 degrees of freedom.

AIC: 19305

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



cumulata radiazione : sum_rad cortigliano

Call: glm(formula = sum_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1129.92	-171.14	-0.15	172.93	1083.76

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6636.929	17.574	377.6545	< 2.2e-16	***
as.factor(def)[T.1]	166.242	11.934	13.9298	< 2.2e-16	***
as.factor(gemme)[T.3]	-23.986	11.934	-2.0099	0.04455	*
as.factor(dirad)[T.1]	23.016	11.936	1.9283	0.05393	.
AnnoF[T.A2009]	-138.735	14.776	-9.3893	< 2.2e-16	***
AnnoF[T.A2010]	-93.263	14.776	-6.3118	3.263e-10	***
as.factor(Mese)[T.7]	-385.618	15.994	-24.1107	< 2.2e-16	***
as.factor(Mese)[T.8]	-899.091	15.902	-56.5399	< 2.2e-16	***
as.factor(Mese)[T.9]	-1351.055	19.391	-69.6739	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

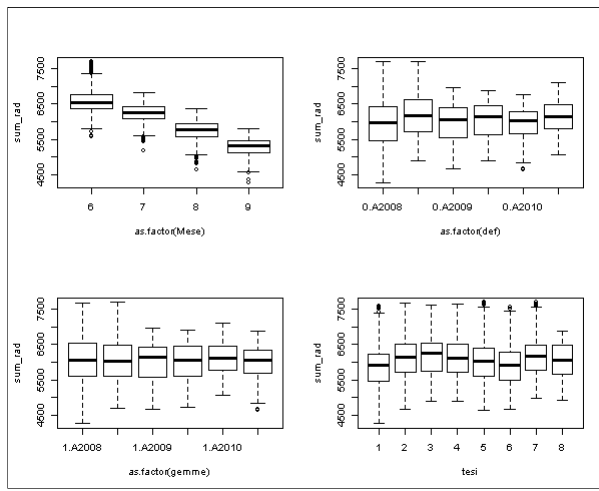
(Dispersion parameter for gaussian family taken to be 87029.44)

- Null deviance: 765006963.5982 on 2455 degrees of freedom.
- Residual deviance: 212961040.5077 on 2447 degrees of freedom.

AIC: 34915

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione radiazione: et_rad cortigliano

Call: `glm(formula = et_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-69.66	-13.42	-0.82	12.68	53.83

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	465.95946	1.20961	385.2138	< 2.2e-16	***
as.factor(def)[T.1]	13.50940	0.82143	16.4462	< 2.2e-16	***
as.factor(gemme)[T.3]	-2.42002	0.82143	-2.9461	0.003248	**
as.factor(dirad)[T.1]	2.07841	0.82154	2.5299	0.011472	*
AnnoF[T.A2009]	-19.43373	1.01701	-19.1087	< 2.2e-16	***
AnnoF[T.A2010]	-8.09608	1.01701	-7.9607	2.594e-15	***
as.factor(Mese)[T.7]	-10.12535	1.10083	-9.1979	< 2.2e-16	***
as.factor(Mese)[T.8]	-11.10130	1.09452	-10.1427	< 2.2e-16	***
as.factor(Mese)[T.9]	-15.29566	1.33468	-11.4602	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 412.3004)

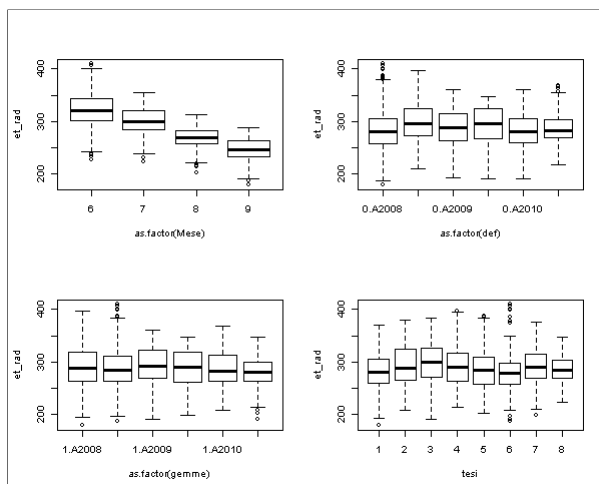
• Null deviance: 1336070.0148 on 2455 degrees of freedom.

• Residual deviance: 1008898.9997 on 2447 degrees of freedom.

AIC: 21770

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 3-10 radiazione: morning_03_10_index_rad cortigliano

Call: `glm(formula = morning_03_10_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-93.31	-14.81	-0.44	15.20	88.39

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	325.98473	1.33816	243.6066	< 2.2e-16	***
as.factor(def)[T.1]	8.93190	0.90872	9.8291	< 2.2e-16	***
as.factor(gemme)[T.3]	-4.28385	0.90872	-4.7141	2.564e-06	***
as.factor(dirad)[T.1]	-0.20058	0.90884	-0.2207	0.8253474	
AnnoF[T.A2009]	-3.73454	1.12509	-3.3193	0.0009156	***
AnnoF[T.A2010]	-9.19839	1.12509	-8.1757	4.668e-16	***
as.factor(Mese)[T.7]	-22.55175	1.21782	-18.5181	< 2.2e-16	***
as.factor(Mese)[T.8]	-55.20841	1.21083	-45.5954	< 2.2e-16	***
as.factor(Mese)[T.9]	-77.33362	1.47652	-52.3757	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

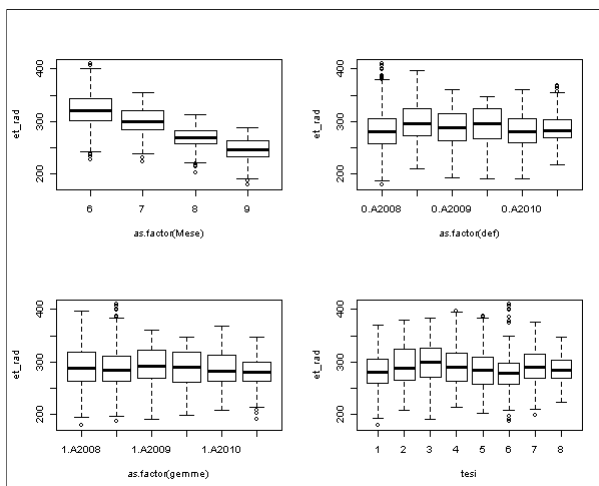
(Dispersion parameter for gaussian family taken to be 504.5888)

- Null deviance: 3162031.9233 on 2455 degrees of freedom.
- Residual deviance: 1234728.7013 on 2447 degrees of freedom.

AIC: 22266

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 11-18 radiazione: diurnal_11_18_index_rad cortigliano

Call: glm(formula = diurnal_11_18_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-69.05	-10.40	-0.77	10.28	69.48

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	431.67992	1.05832	407.8928	< 2.2e-16	***
as.factor(def)[T.1]	11.41458	0.71869	15.8826	< 2.2e-16	***
as.factor(gemme)[T.3]	0.65294	0.71869	0.9085	0.3637	
as.factor(dirad)[T.1]	2.99451	0.71878	4.1661	3.206e-05	***
AnnoF[T.A2009]	-9.60881	0.88981	-10.7988	< 2.2e-16	***
AnnoF[T.A2010]	-4.04714	0.88981	-4.5483	5.671e-06	***
as.factor(Mese)[T.7]	-14.90913	0.96314	-15.4797	< 2.2e-16	***
as.factor(Mese)[T.8]	-22.45284	0.95762	-23.4466	< 2.2e-16	***
as.factor(Mese)[T.9]	-31.79182	1.16774	-27.2251	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

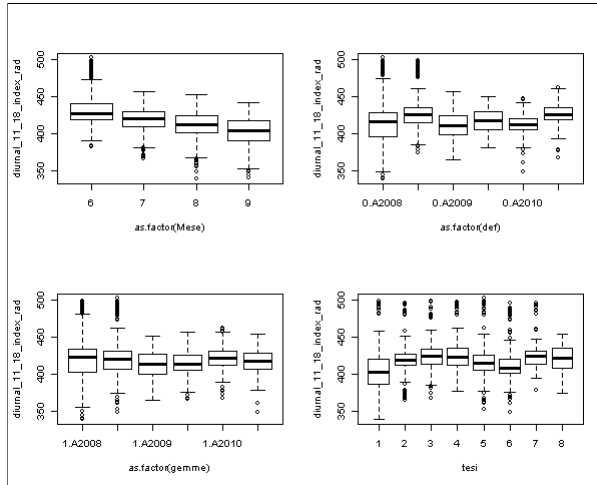
(Dispersion parameter for gaussian family taken to be 315.6115)

- Null deviance: 1165540.435 on 2455 degrees of freedom.
- Residual deviance: 772301.3354 on 2447 degrees of freedom.

AIC: 21114

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 19-02 radiazione: night_19_02_index_rad cortigliano

•

Call: `glm(formula = night_19_02_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-45.50	-11.11	-0.83	10.62	40.13

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	71.951496	0.877534	81.9928	< 2.2e-16	***
as.factor(def)[T.1]	0.433784	0.595919	0.7279	0.46673	
as.factor(gemme)[T.3]	0.632607	0.595919	1.0616	0.28854	
as.factor(dirad)[T.1]	0.083103	0.595997	0.1394	0.88912	
AnnoF[T.A2009]	-3.998525	0.737809	-5.4195	6.563e-08	***
AnnoF[T.A2010]	1.587679	0.737809	2.1519	0.03150	*
as.factor(Mese)[T.7]	-10.741434	0.798617	-13.4500	< 2.2e-16	***
as.factor(Mese)[T.8]	-34.725087	0.794035	-43.7324	< 2.2e-16	***
as.factor(Mese)[T.9]	-59.756431	0.968265	-61.7150	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

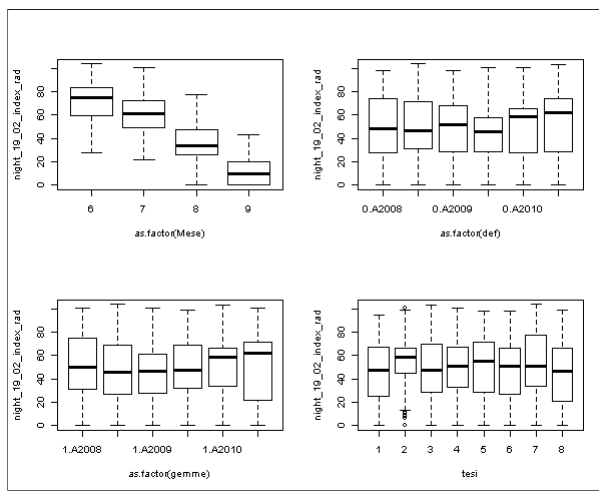
(Dispersion parameter for gaussian family taken to be 216.9948)

- Null deviance: 1579507.3709 on 2455 degrees of freedom.
- Residual deviance: 530986.3946 on 2447 degrees of freedom.

AIC: 20194

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard mattutino tra le ore 5-10 radiazione: dmorn_stderr_5_10_rad cortigliano

Call: `glm(formula = dmorn_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-31.4	-6.5	-1.1	6.8	53.6

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	16.09430	0.64252	25.0486	< 2.2e-16	***
as.factor(def)[T.1]	-0.98544	0.43633	-2.2585	0.0240	*
as.factor(gemme)[T.3]	-0.56155	0.43633	-1.2870	0.1982	
as.factor(dirad)[T.1]	0.96252	0.43638	2.2057	0.0275	*
AnnoF[T.A2009]	2.17115	0.54022	4.0190	6.020e-05	***
AnnoF[T.A2010]	2.29558	0.54022	4.2494	2.224e-05	***
as.factor(Mese)[T.7]	3.44062	0.58474	5.8840	4.552e-09	***
as.factor(Mese)[T.8]	22.96334	0.58138	39.4977	< 2.2e-16	***
as.factor(Mese)[T.9]	39.96945	0.70895	56.3780	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 116.3315)

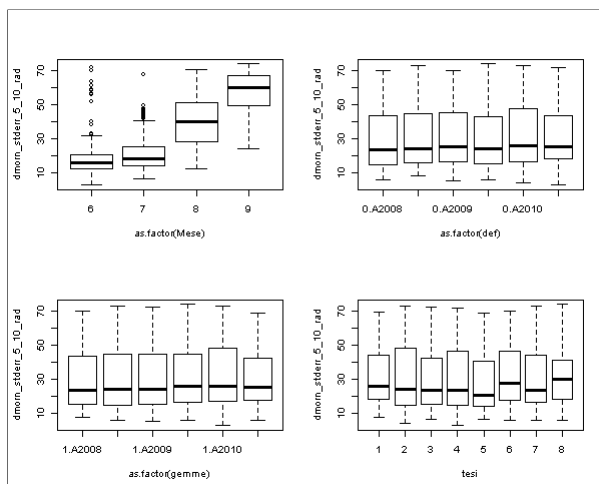
• Null deviance: 794376.5915 on 2455 degrees of freedom.

• Residual deviance: 284663.2722 on 2447 degrees of freedom.

AIC: 18663

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza mattutina tra le ore 5-10 radiazione: dmorn_var_5_10_rad cortigliano

Call: `glm(formula = dmorn_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-16285	-2069	-698	1942	28563

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1836.52	311.72	5.8916	4.35e-09	***
as.factor(def)[T.1]	-545.12	211.68	-2.5752	0.0100764	*
as.factor(gemme)[T.3]	-407.41	211.68	-1.9246	0.0543926	.
as.factor(dirad)[T.1]	402.97	211.71	1.9034	0.0571028	.
AnnoF[T.A2009]	925.91	262.08	3.5329	0.0004186	***
AnnoF[T.A2010]	847.00	262.08	3.2318	0.0012465	**
as.factor(Mese)[T.7]	833.33	283.68	2.9376	0.0033390	**
as.factor(Mese)[T.8]	8779.00	282.06	31.1251	< 2.2e-16	***
as.factor(Mese)[T.9]	18330.51	343.94	53.2949	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

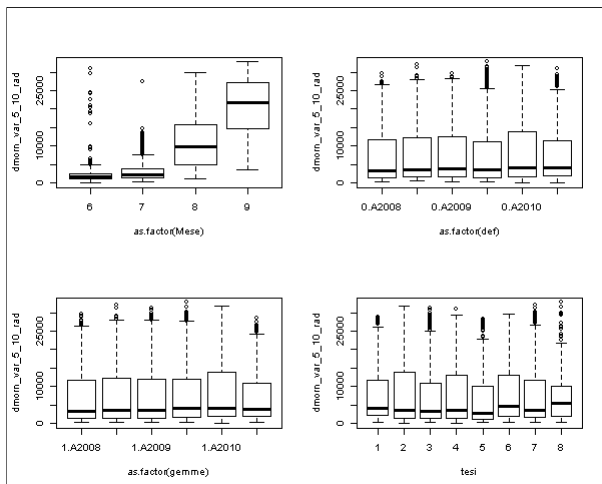
(Dispersion parameter for gaussian family taken to be 27380335)

- Null deviance: 169583420913.985 on 2455 degrees of freedom.
- Residual deviance: 66999679064.906 on 2447 degrees of freedom.

AIC: 49041

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 radiazione: dtwi_stderr_15_20_rad cortigliano

Call: glm(formula = dtwi_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-37.04	-8.90	0.42	8.80	36.82

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	38.05593	0.70332	54.1090	< 2.2e-16	***
as.factor(def)[T.1]	3.05277	0.47761	6.3917	1.957e-10	***
as.factor(gemme)[T.3]	-0.36846	0.47761	-0.7715	0.4405	
as.factor(dirad)[T.1]	0.68191	0.47768	1.4276	0.1535	
AnnoF[T.A2009]	-0.40869	0.59133	-0.6911	0.4895	
AnnoF[T.A2010]	-3.23421	0.59133	-5.4694	4.975e-08	***
as.factor(Mese)[T.7]	6.37468	0.64007	9.9594	< 2.2e-16	***
as.factor(Mese)[T.8]	25.12184	0.63640	39.4751	< 2.2e-16	***
as.factor(Mese)[T.9]	37.41707	0.77604	48.2155	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

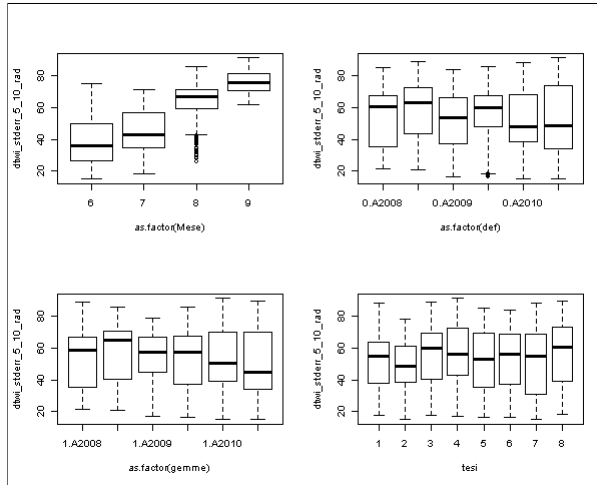
(Dispersion parameter for gaussian family taken to be 139.3883)

- Null deviance: 811036.8017 on 2455 degrees of freedom.
- Residual deviance: 341083.2743 on 2447 degrees of freedom.

AIC: 19107

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza serale tra le ore 15-20 radiazione: dtwi_var_15_20_rad cortigliano

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Call: `glm(formula = dtwi_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-21308	-5495	-228	5324	23981

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	9436.28	416.04	22.6813	< 2.2e-16	***
as.factor(def)[T.1]	2273.28	282.52	8.0463	1.317e-15	***
as.factor(gemme)[T.3]	188.78	282.52	0.6682	0.5041	
as.factor(dirad)[T.1]	318.50	282.56	1.1272	0.2598	
AnnoF[T.A2009]	-542.53	349.79	-1.5510	0.1210	
AnnoF[T.A2010]	-1861.73	349.79	-5.3224	1.117e-07	***
as.factor(Mese)[T.7]	3162.77	378.62	8.3533	< 2.2e-16	***
as.factor(Mese)[T.8]	15031.40	376.45	39.9292	< 2.2e-16	***
as.factor(Mese)[T.9]	24854.37	459.05	54.1427	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

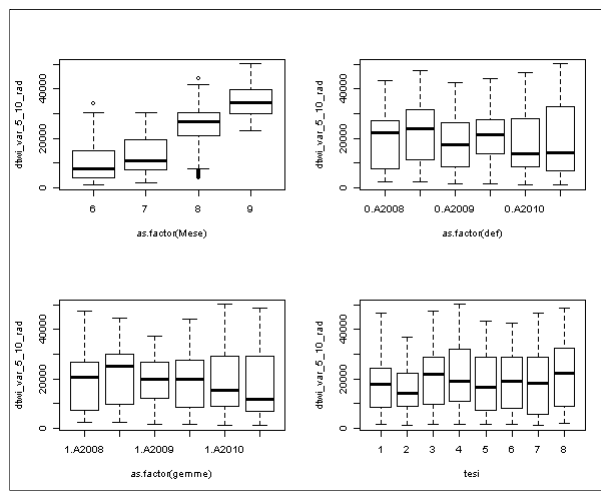
(Dispersion parameter for gaussian family taken to be 48773820)

- Null deviance: 319814726149.115 on 2455 degrees of freedom.
- Residual deviance: 119349536732.279 on 2447 degrees of freedom.

AIC: 50459

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Analisi indici giornalieri temperatura grappolo cortigliano

temperatura media: avg_tgrap cortigliano

Call: `glm(formula = avg_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.964	-1.154	0.052	1.256	6.283

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	24.0687220	0.1158534	207.7515	< 2.2e-16	***
as.factor(def)[T.1]	0.3559555	0.0778570	4.5719	5.067e-06	***
as.factor(gemme)[T.3]	-0.1600430	0.0778570	-2.0556	0.03992	*
as.factor(dirad)[T.1]	-0.0021214	0.0778570	-0.0272	0.97826	
AnnoF[T.A2009]	-0.0086430	0.0955407	-0.0905	0.92793	
AnnoF[T.A2010]	-0.9327639	0.0955407	-9.7630	< 2.2e-16	***
as.factor(Mese)[T.7]	3.2567641	0.1038905	31.3481	< 2.2e-16	***
as.factor(Mese)[T.8]	3.7968266	0.1038905	36.5464	< 2.2e-16	***
as.factor(Mese)[T.9]	1.6277012	0.1275983	12.7564	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

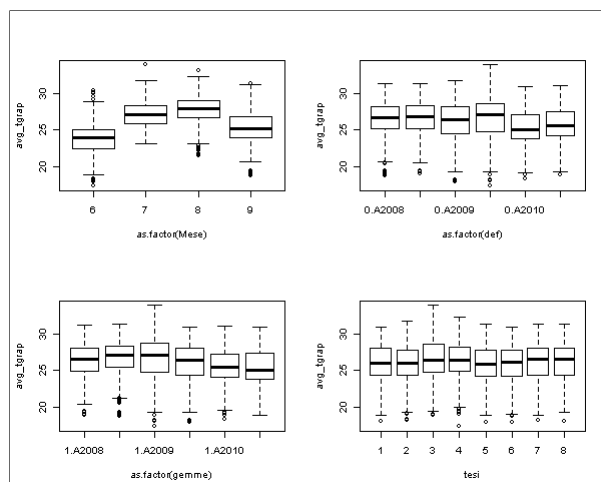
(Dispersion parameter for gaussian family taken to be 3.827949)

- Null deviance: 16339.1495 on 2525 degrees of freedom.
- Residual deviance: 9634.9474 on 2517 degrees of freedom.

AIC: 10570

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.931081126848165



cumulata temperatura grappolo: sum_tgrap cortigliano

Call: glm(formula = sum_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-167.1	-27.7	1.2	30.1	150.8

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	577.64930	2.78048	207.7515	< 2.2e-16	***
as.factor(def)[T.1]	8.54295	1.86857	4.5719	5.066e-06	***
as.factor(gemme)[T.3]	-3.84099	1.86857	-2.0556	0.03993	*
as.factor(dirad)[T.1]	-0.05095	1.86857	-0.0273	0.97825	
AnnoF[T.A2009]	-0.20741	2.29298	-0.0905	0.92793	
AnnoF[T.A2010]	-22.38627	2.29298	-9.7630	< 2.2e-16	***
as.factor(Mese)[T.7]	78.16240	2.49337	31.3481	< 2.2e-16	***
as.factor(Mese)[T.8]	91.12382	2.49337	36.5464	< 2.2e-16	***
as.factor(Mese)[T.9]	39.06484	3.06236	12.7565	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

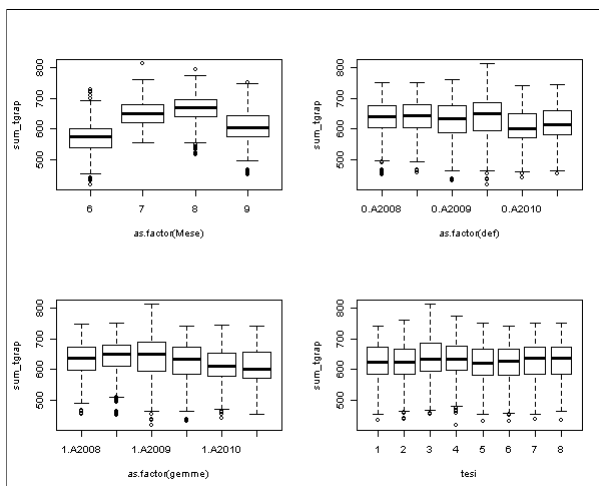
(Dispersion parameter for gaussian family taken to be 2204.897)

- Null deviance: 9411345.4881 on 2525 degrees of freedom.
- Residual deviance: 5549725.39 on 2517 degrees of freedom.

AIC: 26626

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica grappolo: et_tgrap cortigliano

Call: glm(formula = et_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.40	-2.51	-0.32	2.70	11.26

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	14.30777	0.21717	65.8841	< 2.2e-16	***
as.factor(def)[T.1]	0.24286	0.14594	1.6641	0.09622	.
as.factor(gemme)[T.3]	-0.18870	0.14594	-1.2930	0.19614	
as.factor(dirad)[T.1]	0.75343	0.14594	5.1625	2.627e-07	***
AnnoF[T.A2009]	-2.52838	0.17909	-14.1179	< 2.2e-16	***
AnnoF[T.A2010]	-2.10921	0.17909	-11.7774	< 2.2e-16	***
as.factor(Mese)[T.7]	2.07230	0.19474	10.6413	< 2.2e-16	***
as.factor(Mese)[T.8]	4.16934	0.19474	21.4096	< 2.2e-16	***
as.factor(Mese)[T.9]	1.54463	0.23918	6.4580	1.269e-10	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

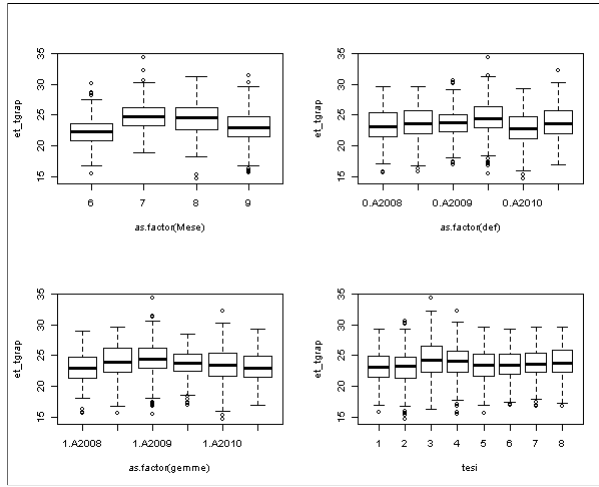
(Dispersion parameter for gaussian family taken to be 13.45027)

- Null deviance: 43741.9279 on 2525 degrees of freedom.
- Residual deviance: 33854.3383 on 2517 degrees of freedom.

AIC: 13745

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.980046052946158



media ore 3-10 temp. grappolo: morning_03_10_index_tgrap cortigliano

Call: glm(formula = morning_03_10_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.013	-1.573	-0.035	1.558	8.815

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.899473	0.136100	160.9073	< 2.2e-16	***
as.factor(def)[T.1]	0.684169	0.091463	7.4803	1.019e-13	***
as.factor(gemme)[T.3]	-0.041729	0.091463	-0.4562	0.64826	
as.factor(dirad)[T.1]	-0.077916	0.091463	-0.8519	0.39436	
AnnoF[T.A2009]	0.485671	0.112237	4.3272	1.569e-05	***
AnnoF[T.A2010]	-0.232808	0.112237	-2.0742	0.03816	*
as.factor(Mese)[T.7]	2.443075	0.122046	20.0176	< 2.2e-16	***
as.factor(Mese)[T.8]	2.109082	0.122046	17.2810	< 2.2e-16	***
as.factor(Mese)[T.9]	0.916378	0.149897	6.1134	1.127e-09	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

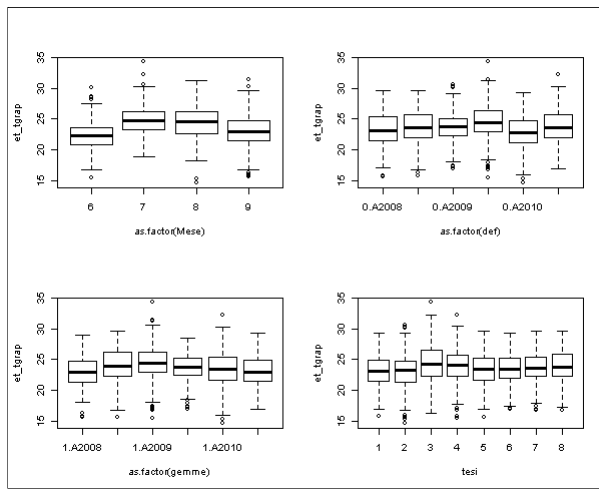
(Dispersion parameter for gaussian family taken to be 5.282801)

- Null deviance: 16402.1619 on 2525 degrees of freedom.
- Residual deviance: 13296.8099 on 2517 degrees of freedom.

AIC: 11384

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.980046052946158



media ore 11-18 temp. grappolo: diurnal_11_18_index_tgrap cortigliano

Call: `glm(formula = diurnal_11_18_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.17	-1.79	0.25	1.98	6.71

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	29.36462	0.17022	172.5105	< 2.2e-16	***
as.factor(def)[T.1]	0.35730	0.11439	3.1235	0.001808	**
as.factor(gemme)[T.3]	-0.15246	0.11439	-1.3328	0.182714	
as.factor(dirad)[T.1]	0.33765	0.11439	2.9517	0.003190	**
AnnoF[T.A2009]	-0.94086	0.14037	-6.7025	2.519e-11	***
AnnoF[T.A2010]	-1.89734	0.14037	-13.5163	< 2.2e-16	***
as.factor(Mese)[T.7]	4.28864	0.15264	28.0960	< 2.2e-16	***
as.factor(Mese)[T.8]	5.76003	0.15264	37.7354	< 2.2e-16	***
as.factor(Mese)[T.9]	2.33899	0.18748	12.4762	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

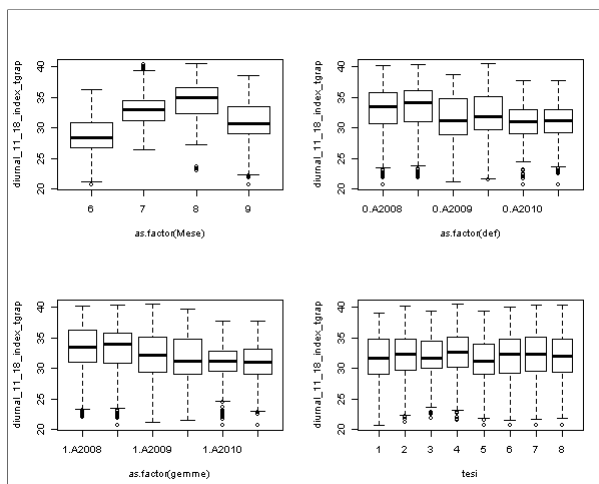
(Dispersion parameter for gaussian family taken to be 8.263536)

- Null deviance: 35535.1321 on 2525 degrees of freedom.
- Residual deviance: 20799.319 on 2517 degrees of freedom.

AIC: 12514

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.997073417535997



media ore 19-02 temp. grappolo: night_19_02_index_tgrap cortigliano

Call: `glm(formula = night_19_02_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.328	-1.212	0.039	1.166	6.399

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.942069	0.112987	185.3502	< 2.2e-16	***
as.factor(def)[T.1]	0.026399	0.075930	0.3477	0.7281184	
as.factor(gemme)[T.3]	-0.285930	0.075930	-3.7657	0.0001699	***
as.factor(dirad)[T.1]	-0.266104	0.075930	-3.5046	0.0004653	***
AnnoF[T.A2009]	0.429266	0.093176	4.6070	4.288e-06	***
AnnoF[T.A2010]	-0.668135	0.093176	-7.1707	9.769e-13	***
as.factor(Mese)[T.7]	3.038583	0.101320	29.9901	< 2.2e-16	***
as.factor(Mese)[T.8]	3.521370	0.101320	34.7551	< 2.2e-16	***
as.factor(Mese)[T.9]	1.627738	0.124441	13.0804	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

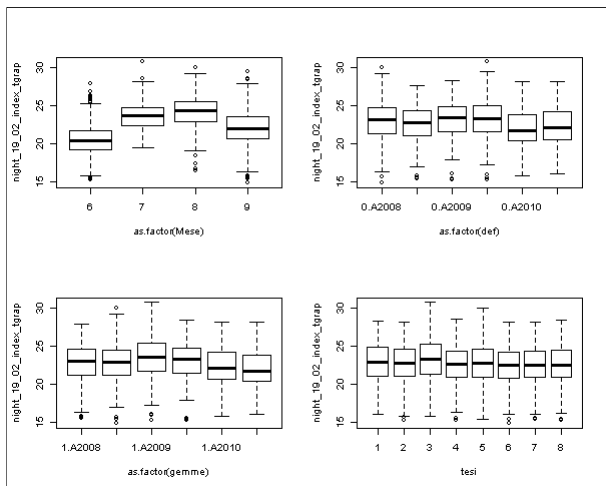
(Dispersion parameter for gaussian family taken to be 3.640839)

- Null deviance: 15003.9971 on 2525 degrees of freedom.
- Residual deviance: 9163.9906 on 2517 degrees of freedom.

AIC: 10444

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.903310892616456



errore standard mattutino tra le ore 5-10 temp. grappolo: dmorn_stderr_5_10_tgrap cortigliano

Call: glm(formula = dmorn_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.9322	-0.3965	-0.0083	0.4086	1.7802

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.5584899	0.0347237	44.8825	< 2.2e-16	***
as.factor(def)[T.1]	0.0967527	0.0233354	4.1462	3.493e-05	***
as.factor(gemme)[T.3]	0.1338559	0.0233354	5.7362	1.084e-08	***
as.factor(dirad)[T.1]	0.0179924	0.0233354	0.7710	0.4408	
AnnoF[T.A2009]	0.0044696	0.0286356	0.1561	0.8760	
AnnoF[T.A2010]	-0.1188234	0.0286356	-4.1495	3.443e-05	***
as.factor(Mese)[T.7]	0.2533081	0.0311382	8.1350	6.401e-16	***
as.factor(Mese)[T.8]	0.3901844	0.0311382	12.5307	< 2.2e-16	***
as.factor(Mese)[T.9]	0.0200703	0.0382439	0.5248	0.5998	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

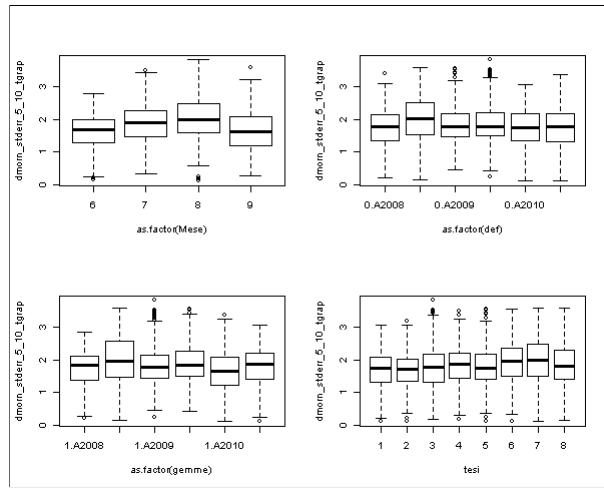
(Dispersion parameter for gaussian family taken to be 0.3438756)

- Null deviance: 959.076 on 2525 degrees of freedom.
- Residual deviance: 865.5349 on 2517 degrees of freedom.

AIC: 4483

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.115073310878342



varianza mattutina tra le ore 5-10 temp. grappolo: dmorn_var_5_10_tgrap cortigliano

Call: glm(formula = dmorn_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-28.3	-9.8	-2.2	7.8	59.9

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	15.650285	0.795004	19.6858	< 2.2e-16	***
as.factor(def)[T.1]	2.588503	0.534266	4.8450	1.343e-06	***
as.factor(gemme)[T.3]	3.213257	0.534266	6.0143	2.069e-09	***
as.factor(dirad)[T.1]	0.283853	0.534266	0.5313	0.5953	
AnnoF[T.A2009]	0.044663	0.655615	0.0681	0.9457	
AnnoF[T.A2010]	-2.825810	0.655615	-4.3102	1.694e-05	***
as.factor(Mese)[T.7]	5.880968	0.712912	8.2492	2.538e-16	***
as.factor(Mese)[T.8]	9.710814	0.712912	13.6213	< 2.2e-16	***
as.factor(Mese)[T.9]	1.184408	0.875599	1.3527	0.1763	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

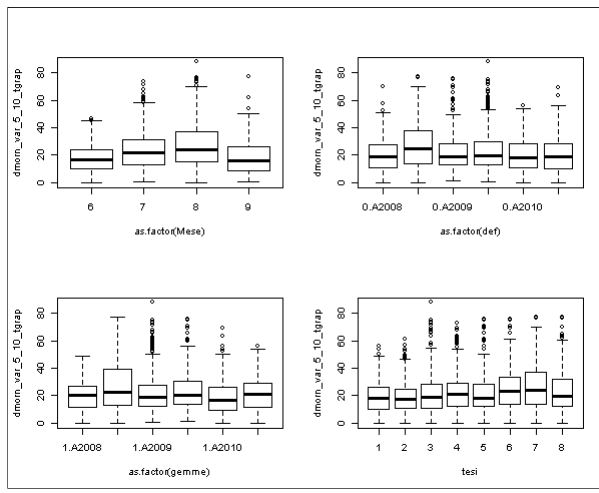
(Dispersion parameter for gaussian family taken to be 180.2547)

- Null deviance: 508319.8317 on 2525 degrees of freedom.
- Residual deviance: 453701.1136 on 2517 degrees of freedom.

AIC: 20300

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.99999999593164



errore standard serale tra le ore 15-20 temp. grappolo: dtwi_stderr_15_20_tgrap cortigliano

Call: glm(formula = dtwi_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.711	-0.268	0.015	0.280	1.732

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.3428631	0.0267304	50.2373	< 2.2e-16	***
as.factor(def)[T.1]	0.0792523	0.0179636	4.4118	1.068e-05	***
as.factor(gemme)[T.3]	-0.0019986	0.0179636	-0.1113	0.9114	
as.factor(dirad)[T.1]	0.1672089	0.0179636	9.3082	< 2.2e-16	***
AnnoF[T.A2009]	-0.4799711	0.0220437	-21.7736	< 2.2e-16	***
AnnoF[T.A2010]	-0.3153575	0.0220437	-14.3060	< 2.2e-16	***
as.factor(Mese)[T.7]	0.1893912	0.0239702	7.9011	4.095e-15	***
as.factor(Mese)[T.8]	0.5205803	0.0239702	21.7178	< 2.2e-16	***
as.factor(Mese)[T.9]	0.3892207	0.0294403	13.2207	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

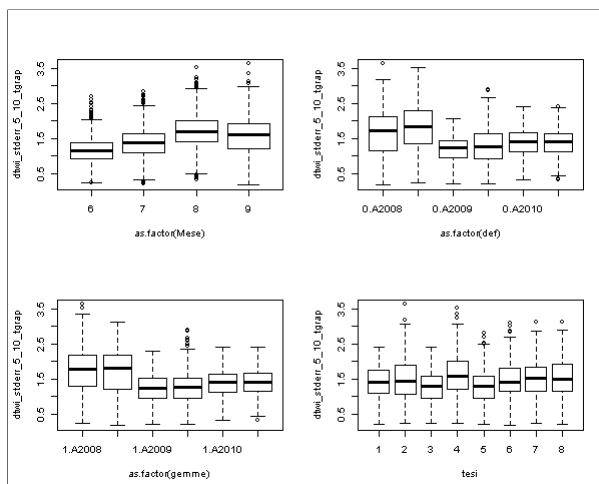
(Dispersion parameter for gaussian family taken to be 0.2037791)

- Null deviance: 742.7095 on 2525 degrees of freedom.
- Residual deviance: 512.912 on 2517 degrees of freedom.

AIC: 3161.3

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.341354202871621



varianza serale tra le ore 15-20 temp. grappolo: dtwi_var_15_20_tgrap cortigliano

Call: glm(formula = dtwi_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-24.67	-5.27	-0.66	4.25	55.01

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.136739	0.511356	25.6900	< 2.2e-16	***
as.factor(def)[T.1]	1.544231	0.343647	4.4937	7.317e-06	***
as.factor(gemme)[T.3]	-0.083497	0.343647	-0.2430	0.808	
as.factor(dirad)[T.1]	3.614587	0.343647	10.5183	< 2.2e-16	***
AnnoF[T.A2009]	-10.017343	0.421699	-23.7547	< 2.2e-16	***
AnnoF[T.A2010]	-7.548996	0.421699	-17.9014	< 2.2e-16	***
as.factor(Mese)[T.7]	3.150687	0.458554	6.8709	8.011e-12	***
as.factor(Mese)[T.8]	9.665918	0.458554	21.0791	< 2.2e-16	***
as.factor(Mese)[T.9]	7.358975	0.563196	13.0665	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

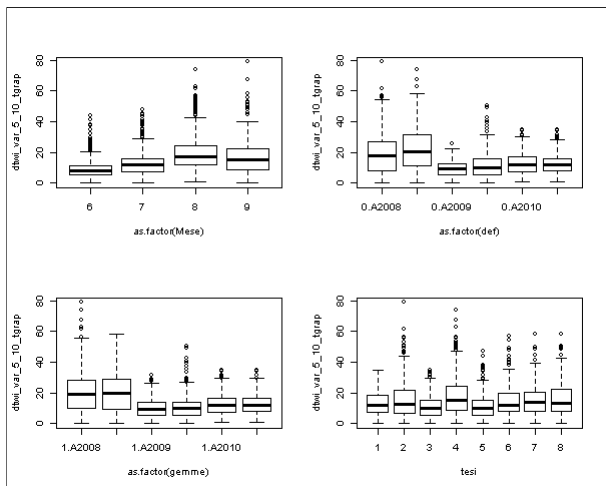
(Dispersion parameter for gaussian family taken to be 74.57537)

- Null deviance: 281675.4626 on 2525 degrees of freedom.
- Residual deviance: 187706.1964 on 2517 degrees of freedom.

AIC: 18071

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Analisi indici giornalieri tarra canopy cortigliano

temperatura canopy media: avg_tair cortigliano

Call: glm(formula = avg_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.9363	-1.3759	-0.0065	1.4523	6.5402

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	24.910589	0.125566	198.3858	< 2.2e-16	***
as.factor(def)[T.1]	0.208852	0.084384	2.4750	0.0133886	*
as.factor(gemme)[T.3]	-0.047960	0.084384	-0.5684	0.5698432	
as.factor(dirad)[T.1]	-0.286381	0.084384	-3.3938	0.0007001	***
AnnoF[T.A2009]	0.287660	0.103551	2.7780	0.0055106	**
AnnoF[T.A2010]	-0.397390	0.103551	-3.8376	0.0001273	***
as.factor(Mese)[T.7]	3.146463	0.112600	27.9436	< 2.2e-16	***
as.factor(Mese)[T.8]	2.784295	0.112600	24.7272	< 2.2e-16	***
as.factor(Mese)[T.9]	0.598979	0.138296	4.3311	1.541e-05	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

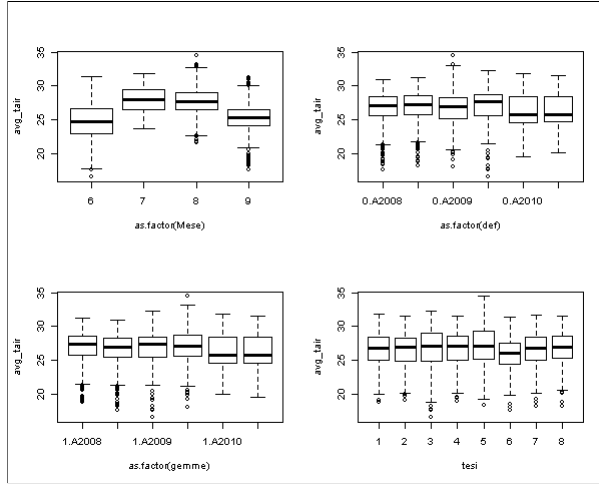
(Dispersion parameter for gaussian family taken to be 4.496713)

- Null deviance: 16383.4877 on 2525 degrees of freedom.
- Residual deviance: 11318.2269 on 2517 degrees of freedom.

AIC: 10977

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.866694796943022



cumulata temperatura canopy: sum_tair cortigliano

Call: glm(formula = sum_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-190.47	-33.02	-0.16	34.85	156.97

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	597.8541	3.0136	198.3858	< 2.2e-16	***
as.factor(def)[T.1]	5.0125	2.0252	2.4750	0.0133881	*
as.factor(gemme)[T.3]	-1.1510	2.0252	-0.5683	0.5698498	
as.factor(dirad)[T.1]	-6.8732	2.0252	-3.3938	0.0007000	***
AnnoF[T.A2009]	6.9039	2.4852	2.7780	0.0055105	**
AnnoF[T.A2010]	-9.5374	2.4852	-3.8377	0.0001273	***
as.factor(Mese)[T.7]	75.5151	2.7024	27.9436	< 2.2e-16	***
as.factor(Mese)[T.8]	66.8231	2.7024	24.7272	< 2.2e-16	***
as.factor(Mese)[T.9]	14.3755	3.3191	4.3312	1.541e-05	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

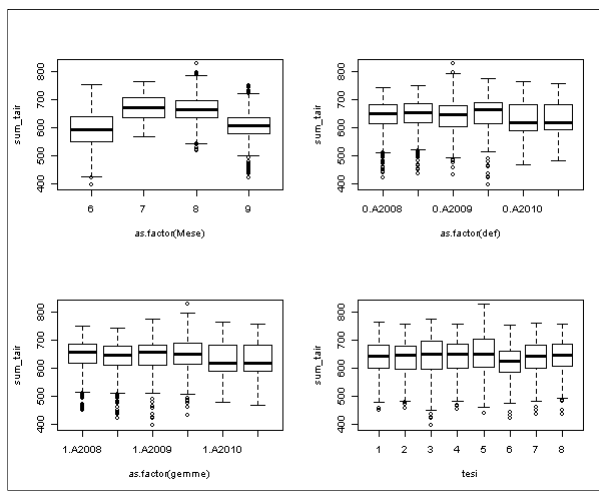
(Dispersion parameter for gaussian family taken to be 2590.107)

- Null deviance: 9436890.6316 on 2525 degrees of freedom.
- Residual deviance: 6519299.4914 on 2517 degrees of freedom.

AIC: 27032

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica canopy: et_tair cortigliano

Call: `glm(formula = et_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.229	-2.328	-0.017	2.496	21.241

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	14.56015	0.20947	69.5080	< 2.2e-16	***
as.factor(def)[T.1]	0.65617	0.14077	4.6612	3.308e-06	***
as.factor(gemme)[T.3]	0.19170	0.14077	1.3618	0.1733903	
as.factor(dirad)[T.1]	0.26223	0.14077	1.8628	0.0626063	.
AnnoF[T.A2009]	-0.88327	0.17275	-5.1131	3.407e-07	***
AnnoF[T.A2010]	-0.42634	0.17275	-2.4680	0.0136537	*
as.factor(Mese)[T.7]	1.24342	0.18784	6.6194	4.392e-11	***
as.factor(Mese)[T.8]	1.30384	0.18784	6.9411	4.932e-12	***
as.factor(Mese)[T.9]	-0.82781	0.23071	-3.5881	0.0003394	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

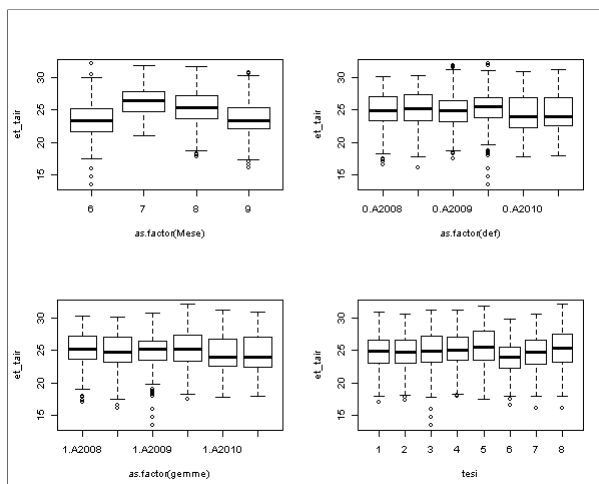
(Dispersion parameter for gaussian family taken to be 12.51442)

- Null deviance: 33826.5338 on 2525 degrees of freedom.
- Residual deviance: 31498.7892 on 2517 degrees of freedom.

AIC: 13562

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.60208481425923



media ore 03-10 temp. canopy: morning_03_10_index_tair cortigliano

Call: `glm(formula = morning_03_10_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.40	-1.58	-0.11	1.70	8.43

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.776486	0.138124	172.1387	< 2.2e-16	***
as.factor(def)[T.1]	0.180118	0.092823	1.9404	0.052438	.
as.factor(gemme)[T.3]	-0.030528	0.092823	-0.3289	0.742272	
as.factor(dirad)[T.1]	-0.288184	0.092823	-3.1047	0.001926	**
AnnoF[T.A2009]	0.082040	0.113907	0.7202	0.471443	
AnnoF[T.A2010]	-0.585952	0.113907	-5.1441	2.894e-07	***
as.factor(Mese)[T.7]	2.750385	0.123861	22.2053	< 2.2e-16	***
as.factor(Mese)[T.8]	1.907194	0.123861	15.3978	< 2.2e-16	***
as.factor(Mese)[T.9]	0.164312	0.152127	1.0801	0.280201	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

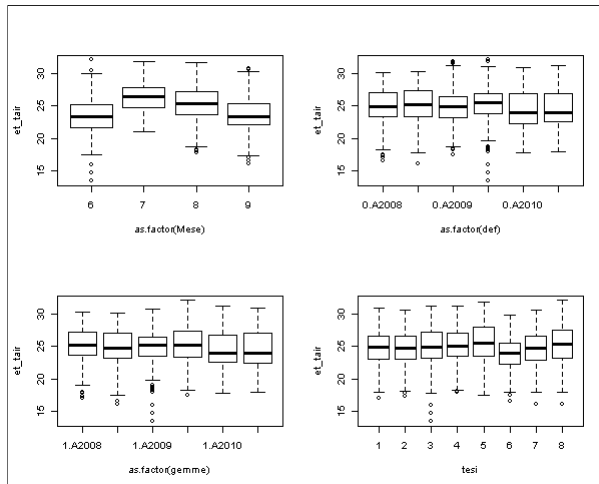
(Dispersion parameter for gaussian family taken to be 5.441099)

- Null deviance: 17427.286 on 2525 degrees of freedom.
- Residual deviance: 13695.2466 on 2517 degrees of freedom.

AIC: 11458

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.60208481425923



media ore 11-18 temp. canopy: diurnal_11_18_index_tair cortigliano

Call: glm(formula = diurnal_11_18_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-10.70	-1.85	0.22	2.00	7.62

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	29.984145	0.174132	172.1922	< 2.2e-16	***
as.factor(def)[T.1]	0.450602	0.117022	3.8506	0.0001208	***
as.factor(gemme)[T.3]	-0.065047	0.117022	-0.5558	0.5783628	
as.factor(dirad)[T.1]	-0.194329	0.117022	-1.6606	0.0969145	.
AnnoF[T.A2009]	-0.032607	0.143601	-0.2271	0.8203921	
AnnoF[T.A2010]	-0.519002	0.143601	-3.6142	0.0003072	***
as.factor(Mese)[T.7]	3.934178	0.156151	25.1947	< 2.2e-16	***
as.factor(Mese)[T.8]	3.832687	0.156151	24.5447	< 2.2e-16	***
as.factor(Mese)[T.9]	0.657876	0.191785	3.4303	0.0006127	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

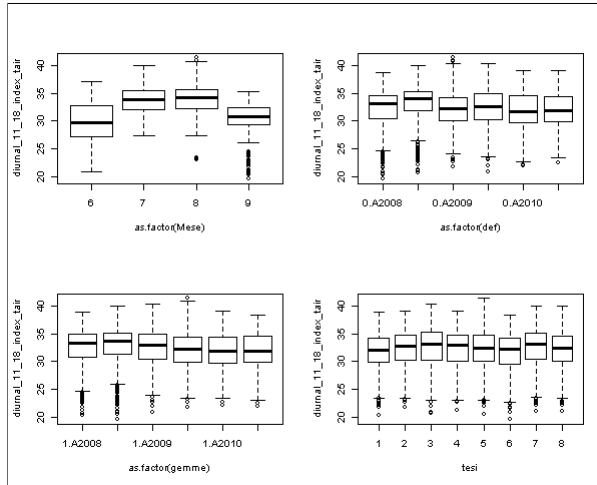
(Dispersion parameter for gaussian family taken to be 8.647778)

- Null deviance: 30352.3564 on 2525 degrees of freedom.
- Residual deviance: 21766.4583 on 2517 degrees of freedom.

AIC: 12629

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.966599504746068



media ore 19-02 temp. canopy: night_19_02_index_tair cortigliano

•

Call: `glm(formula = night_19_02_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.606	-1.433	-0.082	1.342	13.227

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.9711301	0.1258968	166.5739	< 2.2e-16	***
as.factor(def)[T.1]	-0.0041586	0.0846064	-0.0492	0.9608	
as.factor(gemme)[T.3]	-0.0483040	0.0846064	-0.5709	0.5681	
as.factor(dirad)[T.1]	-0.3766320	0.0846064	-4.4516	8.896e-06	***
AnnoF[T.A2009]	0.8135486	0.1038231	7.8359	6.810e-15	***
AnnoF[T.A2010]	-0.0872222	0.1038231	-0.8401	0.4009	
as.factor(Mese)[T.7]	2.7548259	0.1128968	24.4013	< 2.2e-16	***
as.factor(Mese)[T.8]	2.6130059	0.1128968	23.1451	< 2.2e-16	***
as.factor(Mese)[T.9]	0.9747596	0.1386599	7.0299	2.652e-12	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

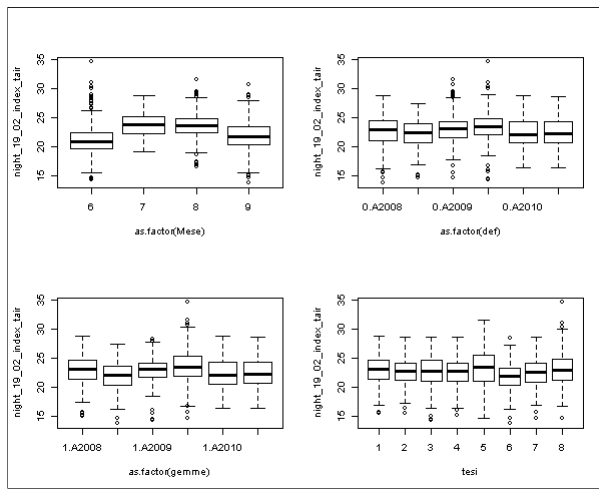
(Dispersion parameter for gaussian family taken to be 4.52041)

- Null deviance: 15465.9843 on 2525 degrees of freedom.
- Residual deviance: 11377.8724 on 2517 degrees of freedom.

AIC: 10990

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.803548805794079



errore standard mattutino tra le ore 5-10 temp. canopy: dmorn_stderr_5_10_tair cortigliano

Call: `glm(formula = dmorn_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.3498	-0.4639	0.0021	0.5044	2.0110

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2.156731	0.040877	52.7616	< 2.2e-16	***
as.factor(def)[T.1]	0.092356	0.027471	3.3620	0.0007854	***
as.factor(gemme)[T.3]	0.076043	0.027471	2.7682	0.0056788	**
as.factor(dirad)[T.1]	-0.041727	0.027471	-1.5190	0.1288970	
AnnoF[T.A2009]	-0.361318	0.033710	-10.7185	< 2.2e-16	***
AnnoF[T.A2010]	-0.188624	0.033710	-5.5955	2.438e-08	***
as.factor(Mese)[T.7]	0.294319	0.036656	8.0292	1.491e-15	***
as.factor(Mese)[T.8]	0.383265	0.036656	10.4557	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.122115	0.045021	-2.7124	0.0067251	**

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

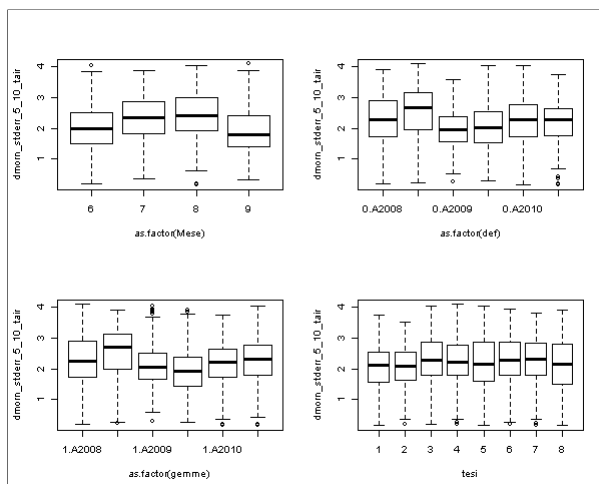
(Dispersion parameter for gaussian family taken to be 0.4765459)

- Null deviance: 1360.8494 on 2525 degrees of freedom.
- Residual deviance: 1199.4661 on 2517 degrees of freedom.

AIC: 5307.2

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.148591894194682



varianza mattutina tra le ore 5-10 temp. canopy: dmorn_var_5_10_tair cortigliano

Call: `glm(formula = dmorn_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-41.3	-13.5	-2.6	12.3	70.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	31.38319	1.08807	28.8429	< 2.2e-16	***
as.factor(def)[T.1]	2.71610	0.73122	3.7145	0.0002080	***
as.factor(gemme)[T.3]	2.66760	0.73122	3.6482	0.0002695	***
as.factor(dirad)[T.1]	-1.41456	0.73122	-1.9345	0.0531600	.
AnnoF[T.A2009]	-10.61508	0.89730	-11.8300	< 2.2e-16	***
AnnoF[T.A2010]	-6.16620	0.89730	-6.8720	7.954e-12	***
as.factor(Mese)[T.7]	7.58416	0.97572	7.7729	1.109e-14	***
as.factor(Mese)[T.8]	10.58837	0.97572	10.8519	< 2.2e-16	***
as.factor(Mese)[T.9]	-2.43985	1.19838	-2.0360	0.0418584	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

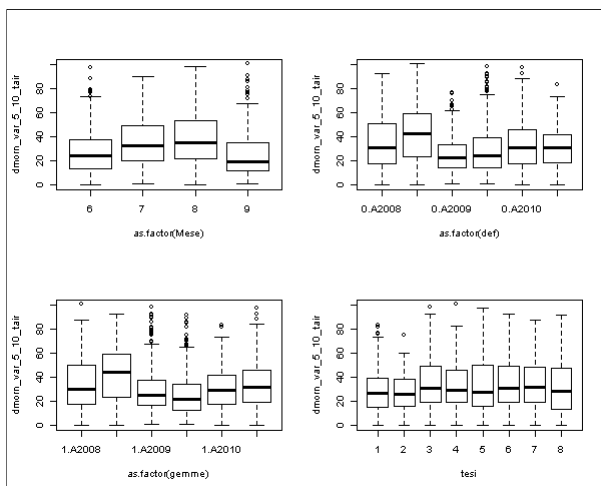
(Dispersion parameter for gaussian family taken to be 337.6473)

- Null deviance: 973981.6638 on 2525 degrees of freedom.
- Residual deviance: 849858.3522 on 2517 degrees of freedom.

AIC: 21886

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 temp. canopy: dtwi_stderr_15_20_tair cortigliano

Call: glm(formula = dtwi_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.347	-0.245	0.017	0.264	4.051

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.2340739	0.0254798	48.4334	< 2.2e-16	***
as.factor(def)[T.1]	0.0846034	0.0171232	4.9409	8.288e-07	***
as.factor(gemme)[T.3]	0.0013797	0.0171232	0.0806	0.9357875	
as.factor(dirad)[T.1]	0.0699700	0.0171232	4.0863	4.521e-05	***
AnnoF[T.A2009]	-0.0676371	0.0210124	-3.2189	0.0013031	**
AnnoF[T.A2010]	-0.0230402	0.0210124	-1.0965	0.2729630	
as.factor(Mese)[T.7]	0.0864248	0.0228488	3.7825	0.0001589	***
as.factor(Mese)[T.8]	0.2415172	0.0228488	10.5702	< 2.2e-16	***
as.factor(Mese)[T.9]	0.1753006	0.0280629	6.2467	4.906e-10	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

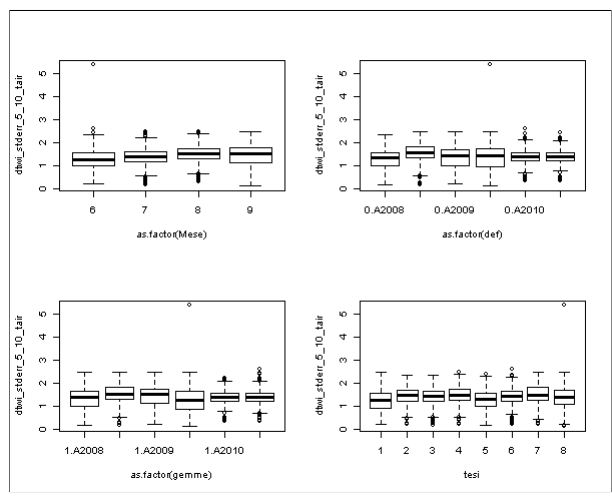
(Dispersion parameter for gaussian family taken to be 0.1851576)

- Null deviance: 498.4616 on 2525 degrees of freedom.
- Residual deviance: 466.0417 on 2517 degrees of freedom.

AIC: 2919.2

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.0712101520640617



varianza serale tra le ore 15-20 temp. canopy: dtwi_var_15_20_taircortigliano

Call: glm(formula = dtwi_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = cortigliano_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.84	-4.79	-0.69	3.63	160.97

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	10.40082	0.45394	22.9125	< 2.2e-16	***
as.factor(def)[T.1]	1.57076	0.30506	5.1490	2.821e-07	***
as.factor(gemme)[T.3]	0.19344	0.30506	0.6341	0.5260792	
as.factor(dirad)[T.1]	1.08229	0.30506	3.5478	0.0003956	***
AnnoF[T.A2009]	-0.96377	0.37435	-2.5745	0.0100944	*
AnnoF[T.A2010]	-1.06082	0.37435	-2.8338	0.0046367	**
as.factor(Mese)[T.7]	1.06490	0.40706	2.6160	0.0089486	**
as.factor(Mese)[T.8]	3.74218	0.40706	9.1931	< 2.2e-16	***
as.factor(Mese)[T.9]	3.06724	0.49996	6.1350	9.859e-10	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

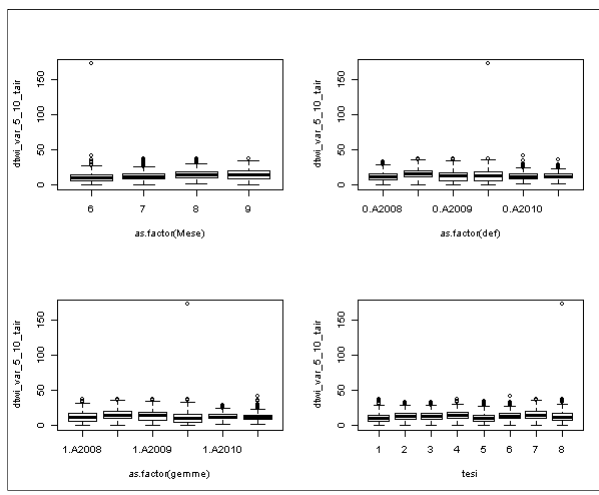
(Dispersion parameter for gaussian family taken to be 58.76775)

- Null deviance: 156836.1211 on 2525 degrees of freedom.
- Residual deviance: 147918.4238 on 2517 degrees of freedom.

AIC: 17469

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.970705723925557



Analisi statistica campione Toscana donna olimpia

IBIMET CNR, Istituto di Biometeorologia di Firenze -CNR

Matese.A., Genesio L., Crisci A.

Temperatura Grappolo donna

Temperatura Grappolo Mattina

•
Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.169	-2.759	0.014	2.803	12.235

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	21.026513	0.130760	160.8029	< 2.2e-16	***
as.factor(def)[T.1]	0.594926	0.089103	6.6768	2.617e-11	***
as.factor(gemme)[T.3]	-0.311818	0.089103	-3.4995	0.0004688	***
as.factor(dirad)[T.1]	-0.030383	0.089103	-0.3410	0.7331222	
AnnoF[T.A2009]	-0.601297	0.109109	-5.5110	3.687e-08	***
AnnoF[T.A2010]	-0.513167	0.109109	-4.7033	2.606e-06	***
as.factor(Mese)[T.7]	1.893705	0.120116	15.7656	< 2.2e-16	***
as.factor(Mese)[T.8]	3.194905	0.117962	27.0841	< 2.2e-16	***
as.factor(Mese)[T.9]	1.038117	0.142906	7.2643	4.128e-13	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

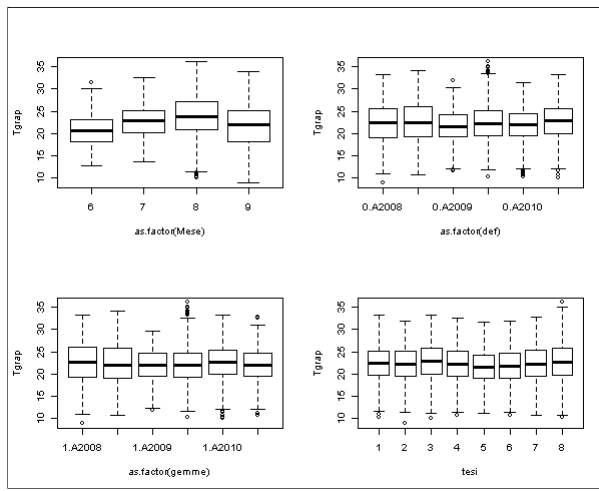
(Dispersion parameter for gaussian family taken to be 14.83076)

- Null deviance: 123460.3505 on 7471 degrees of freedom.
- Residual deviance: 110681.9801 on 7463 degrees of freedom.

AIC: 41365

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.81916452510368



Temperatura Grappolo Pieno giorno

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-18.24	-1.86	0.17	2.18	12.11

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.987795	0.064284	450.9310	< 2.2e-16	***
as.factor(def)[T.1]	0.350235	0.044414	7.8858	3.273e-15	***
as.factor(gemme)[T.3]	-0.680913	0.044414	-15.3312	< 2.2e-16	***
as.factor(dirad)[T.1]	-0.321498	0.044414	-7.2387	4.680e-13	***
AnnoF[T.A2009]	-0.558459	0.054405	-10.2649	< 2.2e-16	***
AnnoF[T.A2010]	-0.660580	0.054389	-12.1456	< 2.2e-16	***
as.factor(Mese)[T.7]	3.886892	0.057160	67.9998	< 2.2e-16	***
as.factor(Mese)[T.8]	5.012977	0.059112	84.8050	< 2.2e-16	***
as.factor(Mese)[T.9]	2.978880	0.075649	39.3777	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

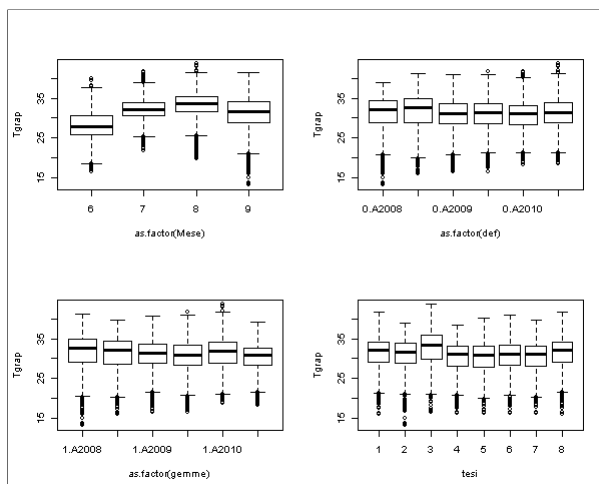
(Dispersion parameter for gaussian family taken to be 10.74214)

- Null deviance: 325954.5239 on 21782 degrees of freedom.
- Residual deviance: 233899.296 on 21774 degrees of freedom.

AIC: 113545

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.985389776960965



Temperatura Cross Grappolo Pieno giorno

Call: `glm(formula = Tgrap ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Rad + as.factor(gemme) * Rad + AnnoF + as.factor(Mese), family = gaussian(identity),`

data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-16.33	-1.95	0.11	2.17	11.69

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	17.9537205	0.5715371	31.4130	< 2.2e-16 ***
as.factor(gemme)[T.3]	-5.6122583	0.5964435	-9.4095	< 2.2e-16 ***
as.factor(def)[T.1]	4.6595999	0.6003531	7.7614	8.766e-15 ***
Rad	0.0244415	0.0012952	18.8707	< 2.2e-16 ***
AnnoF[T.A2009]	-0.3707577	0.0544325	-6.8113	9.922e-12 ***
AnnoF[T.A2010]	-0.3854465	0.0542385	-7.1065	1.227e-12 ***
as.factor(Mese)[T.7]	4.3161945	0.0570352	75.6760	< 2.2e-16 ***
as.factor(Mese)[T.8]	5.4209652	0.0587758	92.2312	< 2.2e-16 ***
as.factor(Mese)[T.9]	3.1366446	0.0738936	42.4481	< 2.2e-16 ***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	0.0124371	0.0875265	0.1421	0.887
as.factor(def)[T.1]:Rad	-0.0105456	0.0013793	-7.6455	2.168e-14 ***
as.factor(gemme)[T.3]:Rad	0.0116854	0.0013885	8.4158	< 2.2e-16 ***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 10.16377)

- Null deviance: 325954.5239 on 21782 degrees of freedom.
- Residual deviance: 221275.4993 on 21771 degrees of freedom.

AIC: 112343

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.991816014323555

Temperatura Grappolo Sera

Call: glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-14.473	-1.707	0.006	1.775	10.107

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	25.606647	0.098227	260.6884	< 2.2e-16 ***
as.factor(def)[T.1]	0.318370	0.066190	4.8100	1.541e-06 ***
as.factor(gemme)[T.3]	-0.406735	0.066190	-6.1450	8.429e-10 ***
as.factor(dirad)[T.1]	0.088493	0.066190	1.3370	0.1813
AnnoF[T.A2009]	0.666159	0.081024	8.2217	2.364e-16 ***
AnnoF[T.A2010]	0.476028	0.081101	5.8696	4.566e-09 ***
as.factor(Mese)[T.7]	3.243552	0.089531	36.2282	< 2.2e-16 ***
as.factor(Mese)[T.8]	3.319921	0.088973	37.3137	< 2.2e-16 ***
as.factor(Mese)[T.9]	2.104747	0.106561	19.7516	< 2.2e-16 ***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

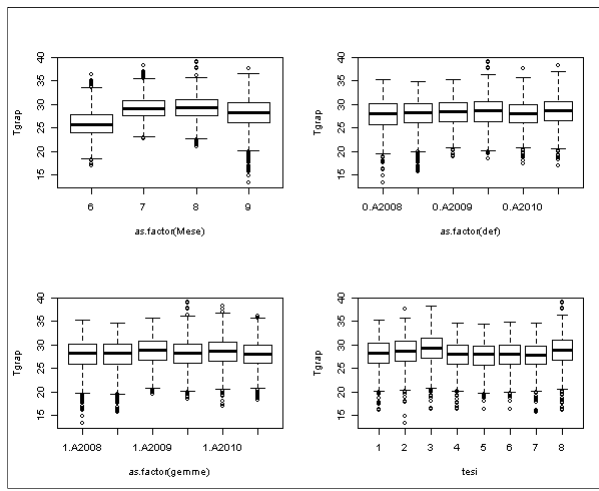
(Dispersion parameter for gaussian family taken to be 7.799427)

- Null deviance: 70369.1947 on 7120 degrees of freedom.
- Residual deviance: 55469.5278 on 7112 degrees of freedom.

AIC: 34846

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.876648713773008



Temperatura Grappolo Notte

Call: `glm(formula = Tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Notte"))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.11	-2.24	-0.13	2.22	21.54

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	19.108181	0.059954	318.7164	< 2.2e-16	***
as.factor(def)[T.1]	0.079249	0.039933	1.9846	0.04721	*
as.factor(gemme)[T.3]	-0.162655	0.039933	-4.0732	4.651e-05	***
as.factor(dirad)[T.1]	0.037212	0.039933	0.9319	0.35142	
AnnoF[T.A2009]	0.598508	0.048919	12.2346	< 2.2e-16	***
AnnoF[T.A2010]	0.050070	0.048919	1.0235	0.30606	
as.factor(Mese)[T.7]	2.169602	0.055152	39.3385	< 2.2e-16	***
as.factor(Mese)[T.8]	2.351346	0.054047	43.5053	< 2.2e-16	***
as.factor(Mese)[T.9]	0.992623	0.063229	15.6989	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

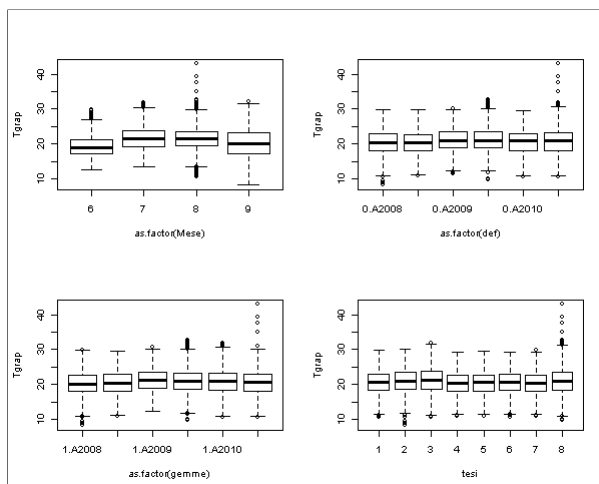
(Dispersion parameter for gaussian family taken to be 9.83878)

- Null deviance: 268107.5989 on 24679 degrees of freedom.
- Residual deviance: 242732.5513 on 24671 degrees of freedom.

AIC: 126476

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.642348692002927



Radiatione Grappolo Mattina

Call: `glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset =`

((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-364	-23	11	33	125

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	394.75881	1.85958	212.2835	< 2.2e-16	***
as.factor(def)[T.1]	5.09358	1.26717	4.0196	5.886e-05	***
as.factor(gemme)[T.3]	-7.58132	1.26717	-5.9829	2.294e-09	***
as.factor(dirad)[T.1]	3.39850	1.26717	2.6820	0.007335	**
AnnoF[T.A2009]	-11.79006	1.55168	-7.5983	3.366e-14	***
AnnoF[T.A2010]	-0.38156	1.55168	-0.2459	0.805767	
as.factor(Mese)[T.7]	-25.06310	1.70822	-14.6720	< 2.2e-16	***
as.factor(Mese)[T.8]	-9.21480	1.67759	-5.4929	4.085e-08	***
as.factor(Mese)[T.9]	-24.12180	2.03232	-11.8691	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

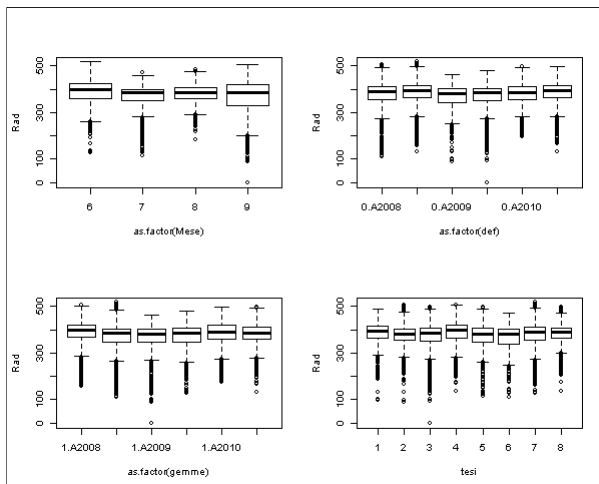
(Dispersion parameter for gaussian family taken to be 2999.493)

- Null deviance: 23606926.7771 on 7471 degrees of freedom.
- Residual deviance: 22385213.1013 on 7463 degrees of freedom.

AIC: 81038

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiatione Grappolo Pieno giorno

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-198.1	-20.2	-1.7	19.4	89.0

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	446.21779	0.59936	744.493	< 2.2e-16	***
as.factor(def)[T.1]	9.61202	0.41409	23.212	< 2.2e-16	***
as.factor(gemme)[T.3]	-4.91513	0.41409	-11.870	< 2.2e-16	***
as.factor(dirad)[T.1]	4.76884	0.41409	11.516	< 2.2e-16	***
AnnoF[T.A2009]	-12.17322	0.50725	-23.999	< 2.2e-16	***
AnnoF[T.A2010]	-14.33696	0.50709	-28.273	< 2.2e-16	***
as.factor(Mese)[T.7]	-17.08924	0.53294	-32.066	< 2.2e-16	***
as.factor(Mese)[T.8]	-15.54265	0.55113	-28.201	< 2.2e-16	***
as.factor(Mese)[T.9]	-5.35685	0.70532	-7.595	3.203e-14	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

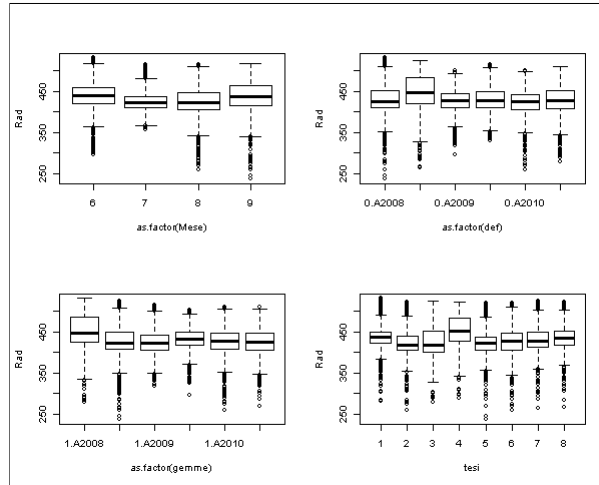
(Dispersion parameter for gaussian family taken to be 933.7987)

- Null deviance: 23184757.7917 on 21782 degrees of freedom.
- Residual deviance: 20332532.9836 on 21774 degrees of freedom.

AIC: 210808

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Radiazione Cross Grappolo Pieno giorno

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Call: glm(formula = Rad ~ as.factor(gemme) * as.factor(def) + as.factor(def) * Tgrap + as.factor(gemme) * Tgrap + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Full_day"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-176.7	-19.8	-1.3	18.8	94.2

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	402.705914	3.037802	132.5649	< 2.2e-16	***
as.factor(gemme)[T.3]	-25.278496	3.319162	-7.6159	2.725e-14	***
as.factor(def)[T.1]	2.735210	3.374409	0.8106	0.41762	
Tgrap	1.623561	0.098145	16.5425	< 2.2e-16	***
AnnoF[T.A2009]	-11.028191	0.496725	-22.2018	< 2.2e-16	***
AnnoF[T.A2010]	-12.876637	0.497352	-25.8904	< 2.2e-16	***
as.factor(Mese)[T.7]	-25.200158	0.573137	-43.9688	< 2.2e-16	***
as.factor(Mese)[T.8]	-25.923747	0.621026	-41.7434	< 2.2e-16	***
as.factor(Mese)[T.9]	-11.397654	0.713920	-15.9649	< 2.2e-16	***
as.factor(gemme)[T.3]:as.factor(def)[T.1]	-1.010894	0.813032	-1.2434	0.21375	
as.factor(def)[T.1]:Tgrap	0.212239	0.105460	2.0125	0.04418	*
as.factor(gemme)[T.3]:Tgrap	0.714032	0.105349	6.7778	1.251e-11	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 891.0879)

- Null deviance: 23184757.7917 on 21782 degrees of freedom.
- Residual deviance: 19399874.2973 on 21771 degrees of freedom.

AIC: 209791

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1

Radiazione Grappolo Sera

•

Call: glm(formula = Rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Evening"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-302.6	-20.5	9.6	27.4	140.9

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	388.33810	1.59271	243.8219	< 2.2e-16	***
as.factor(def)[T.1]	9.85150	1.07324	9.1792	< 2.2e-16	***
as.factor(gemme)[T.3]	-7.17768	1.07324	-6.6879	2.437e-11	***
as.factor(dirad)[T.1]	2.90559	1.07324	2.7073	0.0068	**
AnnoF[T.A2009]	-6.79975	1.31377	-5.1757	2.332e-07	***
AnnoF[T.A2010]	0.55466	1.31502	0.4218	0.6732	
as.factor(Mese)[T.7]	-13.84207	1.45171	-9.5350	< 2.2e-16	***
as.factor(Mese)[T.8]	-39.86784	1.44267	-27.6348	< 2.2e-16	***
as.factor(Mese)[T.9]	-25.57781	1.72784	-14.8033	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

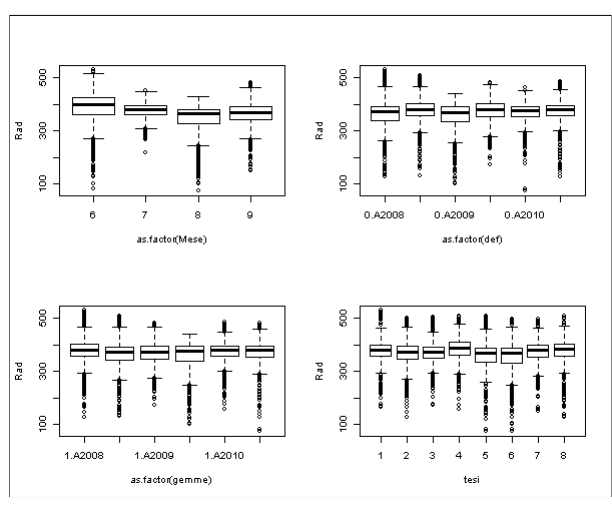
(Dispersion parameter for gaussian family taken to be 2050.573)

- Null deviance: 16628955.5237 on 7120 degrees of freedom.
- Residual deviance: 14583678.2562 on 7112 degrees of freedom.

AIC: 74523

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Temperatura Aria Canopy Mattina

Call: glm(formula = Tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new, subset = ((jd > 153) & (jd < 260) & Code == "Morning"))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.470	-2.735	0.083	2.846	10.898

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.9805214	0.1270538	165.1310	< 2.2e-16	***
as.factor(def)[T.1]	0.1415977	0.0865780	1.6355	0.1020	
as.factor(gemme)[T.3]	-0.0095119	0.0865780	-0.1099	0.9125	
as.factor(dirad)[T.1]	0.0931147	0.0865780	1.0755	0.2822	
AnnoF[T.A2009]	-0.5125185	0.1060164	-4.8343	1.363e-06	***
AnnoF[T.A2010]	-0.7752821	0.1060164	-7.3129	2.887e-13	***
as.factor(Mese)[T.7]	2.3161735	0.1167122	19.8452	< 2.2e-16	***
as.factor(Mese)[T.8]	3.2515477	0.1146194	28.3682	< 2.2e-16	***
as.factor(Mese)[T.9]	0.9553649	0.1388561	6.8803	6.459e-12	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

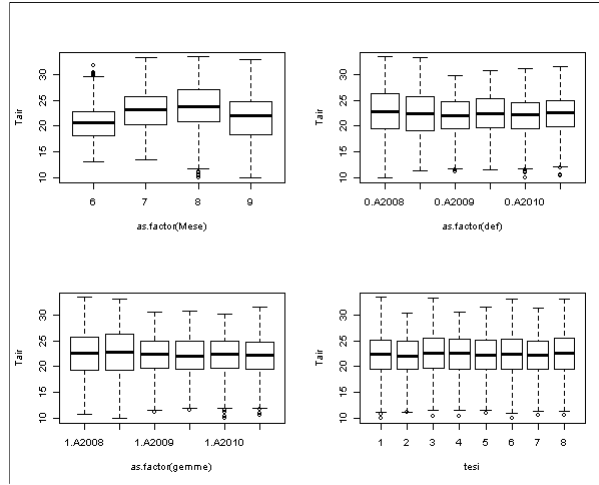
(Dispersion parameter for gaussian family taken to be 14.00206)

- Null deviance: 117966.257 on 7471 degrees of freedom.
- Residual deviance: 104497.359 on 7463 degrees of freedom.

AIC: 40936

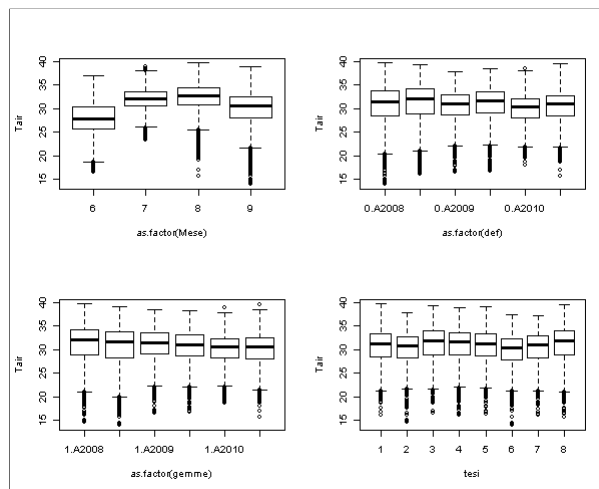
Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.835127595998237



Temperatura Aria Canopy Pieno giorno

Pseudo Rsquared: 0.835127595998237

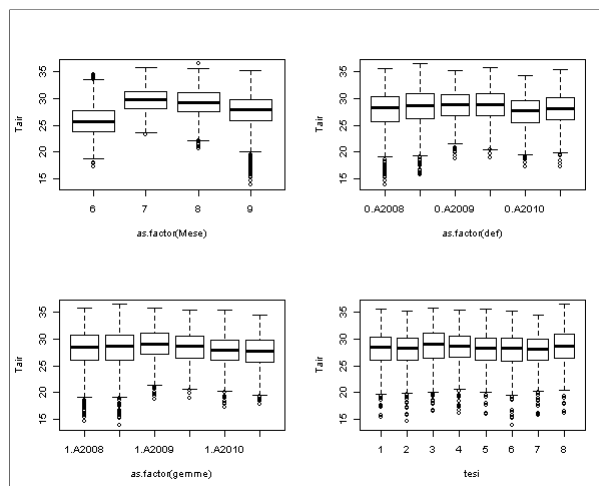


Temperatura Aria Crossing Canopy Pieno giorno

Pseudo Rsquared: 0.835127595998237

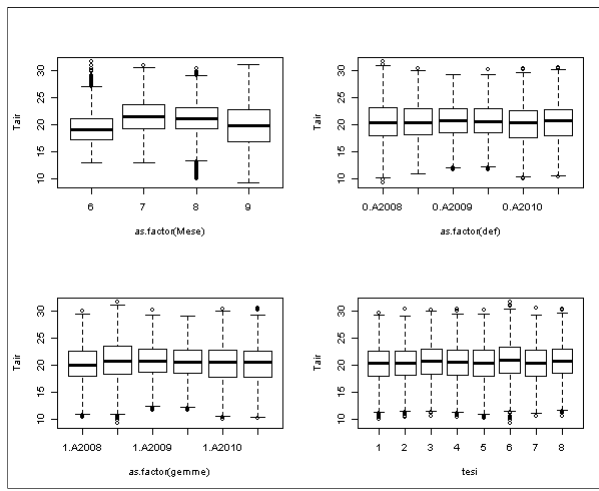
Temperatura Aria Canopy Sera

Pseudo Rsquared: 0.835127595998237



Temperatura Aria Canopy Notte

Pseudo Rsquared: 0.835127595998237



Analisi indici giornalieri radiazione donna

radiazione media : avg_rad donna

Call: `glm(formula = avg_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-59.30	-9.16	0.42	8.76	57.74

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	278.63229	0.87563	318.2080	< 2e-16	***
as.factor(def)[T.1]	8.70314	0.59527	14.6205	< 2e-16	***
as.factor(gemme)[T.3]	-1.18245	0.59527	-1.9864	0.04709	*
as.factor(dirad)[T.1]	5.76451	0.59527	9.6839	< 2e-16	***
AnnoF[T.A2009]	-7.59936	0.72905	-10.4236	< 2e-16	***
AnnoF[T.A2010]	-7.61921	0.72905	-10.4509	< 2e-16	***
as.factor(Mese)[T.7]	-17.82173	0.79165	-22.5122	< 2e-16	***
as.factor(Mese)[T.8]	-37.99721	0.79165	-47.9977	< 2e-16	***
as.factor(Mese)[T.9]	-53.16444	0.97458	-54.5512	< 2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

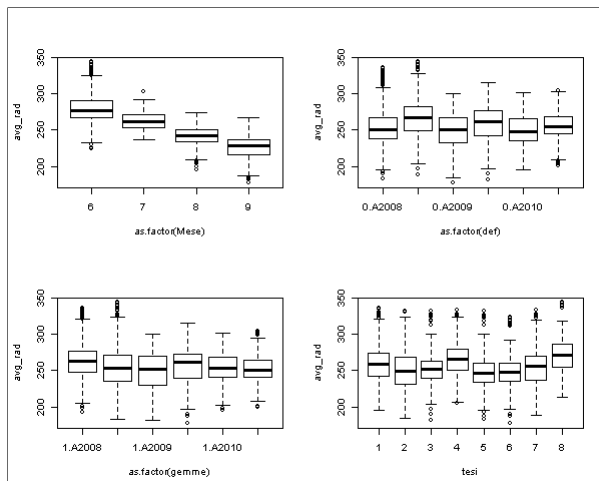
(Dispersion parameter for gaussian family taken to be 225.3627)

- Null deviance: 1555430.1555 on 2543 degrees of freedom.
- Residual deviance: 571294.3359 on 2535 degrees of freedom.

AIC: 21013

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



cumulata radiazione : sum_rad donna

Call: glm(formula = sum_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1423	-220	10	210	1386

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6687.175	21.015	318.2080	< 2e-16	***
as.factor(def)[T.1]	208.875	14.286	14.6205	< 2e-16	***
as.factor(gemme)[T.3]	-28.379	14.286	-1.9864	0.04709	*
as.factor(dirad)[T.1]	138.348	14.286	9.6839	< 2e-16	***
AnnoF[T.A2009]	-182.385	17.497	-10.4236	< 2e-16	***
AnnoF[T.A2010]	-182.861	17.497	-10.4509	< 2e-16	***
as.factor(Mese)[T.7]	-427.721	19.000	-22.5122	< 2e-16	***
as.factor(Mese)[T.8]	-911.933	19.000	-47.9977	< 2e-16	***
as.factor(Mese)[T.9]	-1275.946	23.390	-54.5512	< 2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 129808.9)

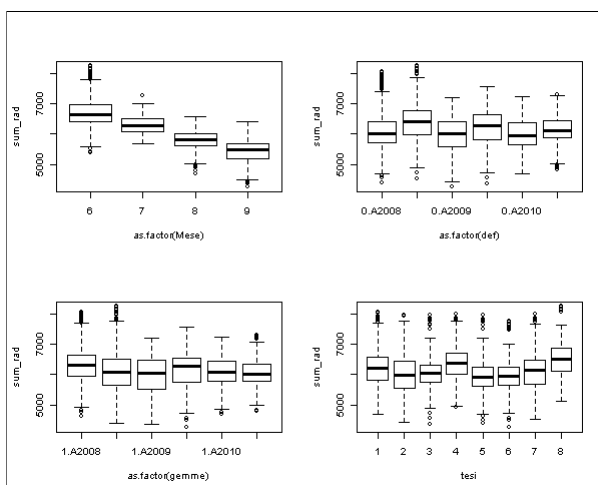
• Null deviance: 895927795.7837 on 2543 degrees of freedom.

• Residual deviance: 329065573.781 on 2535 degrees of freedom.

AIC: 37183

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione radiazione: et_rad donna

Call: glm(formula = et_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-106.288	-19.072	-0.035	21.680	69.075

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	472.5587	1.6729	282.4707	< 2.2e-16	***
as.factor(def)[T.1]	6.5162	1.1373	5.7295	1.126e-08	***
as.factor(gemme)[T.3]	-11.7551	1.1373	-10.3359	< 2.2e-16	***
as.factor(dirad)[T.1]	-1.5700	1.1373	-1.3805	0.16757	
AnnoF[T.A2009]	-22.1487	1.3929	-15.9012	< 2.2e-16	***
AnnoF[T.A2010]	-11.9192	1.3929	-8.5571	< 2.2e-16	***
as.factor(Mese)[T.7]	-10.6313	1.5125	-7.0290	2.664e-12	***
as.factor(Mese)[T.8]	-3.3086	1.5125	-2.1875	0.02879	*
as.factor(Mese)[T.9]	2.1386	1.8620	1.1486	0.25084	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

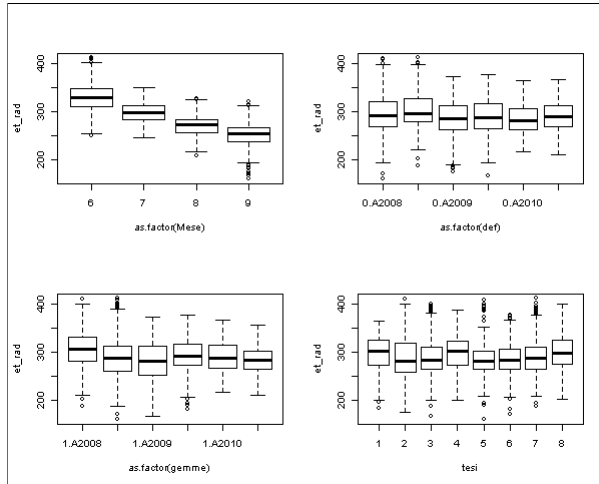
(Dispersion parameter for gaussian family taken to be 822.6329)

- Null deviance: 2467878.041 on 2543 degrees of freedom.
- Residual deviance: 2085374.4168 on 2535 degrees of freedom.

AIC: 24307

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 3-10 radiazione: morning_03_10_index_rad donna

Call: glm(formula = morning_03_10_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-92.03	-15.28	0.74	15.96	79.43

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	332.63407	1.37502	241.9113	< 2.2e-16	***
as.factor(def)[T.1]	6.05050	0.93477	6.4727	1.151e-10	***
as.factor(gemme)[T.3]	-3.47409	0.93477	-3.7165	0.0002063	***
as.factor(dirad)[T.1]	4.45966	0.93477	4.7709	1.938e-06	***
AnnoF[T.A2009]	-11.18713	1.14485	-9.7717	< 2.2e-16	***
AnnoF[T.A2010]	-8.98296	1.14485	-7.8464	6.259e-15	***
as.factor(Mese)[T.7]	-31.13859	1.24314	-25.0483	< 2.2e-16	***
as.factor(Mese)[T.8]	-59.55601	1.24314	-47.9076	< 2.2e-16	***
as.factor(Mese)[T.9]	-76.39411	1.53041	-49.9175	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

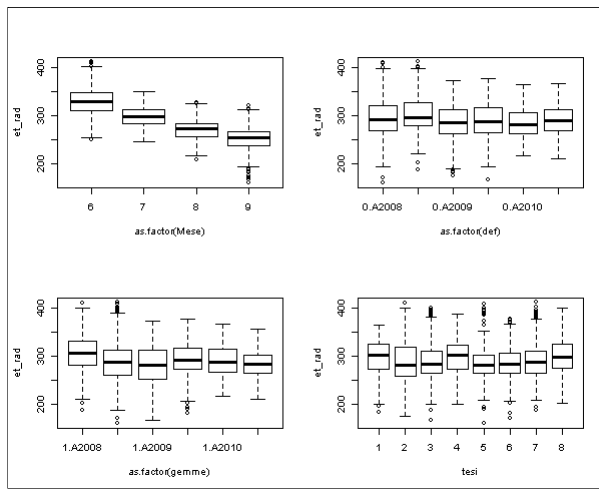
(Dispersion parameter for gaussian family taken to be 555.7283)

- Null deviance: 3427248.7841 on 2543 degrees of freedom.
- Residual deviance: 1408771.1828 on 2535 degrees of freedom.

AIC: 23309

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 11-18 radiazione: diurnal_11_18_index_rad donna

Call: glm(formula = diurnal_11_18_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-106.75	-12.57	-0.32	12.81	89.74

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	435.67313	1.28940	337.8882	< 2.2e-16	***
as.factor(def)[T.1]	10.43144	0.87656	11.9005	< 2.2e-16	***
as.factor(gemme)[T.3]	-4.02870	0.87656	-4.5961	4.517e-06	***
as.factor(dirad)[T.1]	5.35221	0.87656	6.1059	1.179e-09	***
AnnoF[T.A2009]	-8.22079	1.07356	-7.6575	2.677e-14	***
AnnoF[T.A2010]	-9.69158	1.07356	-9.0275	< 2.2e-16	***
as.factor(Mese)[T.7]	-18.67175	1.16573	-16.0172	< 2.2e-16	***
as.factor(Mese)[T.8]	-24.16817	1.16573	-20.7322	< 2.2e-16	***
as.factor(Mese)[T.9]	-26.73022	1.43511	-18.6259	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

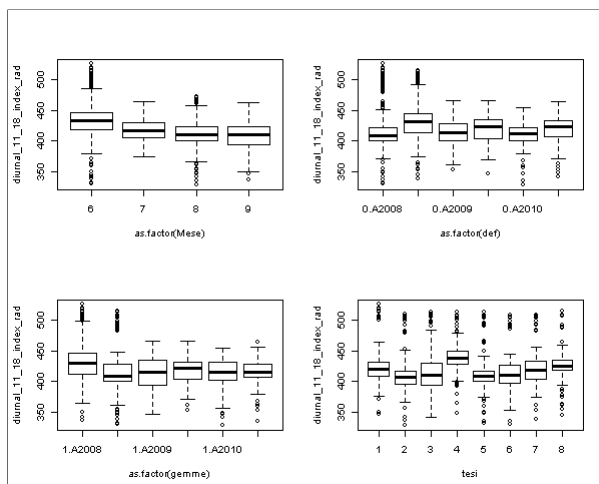
(Dispersion parameter for gaussian family taken to be 488.6713)

- Null deviance: 1657122.9504 on 2543 degrees of freedom.
- Residual deviance: 1238781.8536 on 2535 degrees of freedom.

AIC: 22982

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



media ore 19-02 radiazione: night_19_02_index_rad donna

Call: glm(formula = night_19_02_index_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-63.21	-12.99	0.36	11.58	75.76

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	67.58967	1.07238	63.0275	< 2.2e-16	***
as.factor(def)[T.1]	9.62747	0.72903	13.2060	< 2.2e-16	***
as.factor(gemme)[T.3]	3.95544	0.72903	5.4257	6.321e-08	***
as.factor(dirad)[T.1]	7.48165	0.72903	10.2625	< 2.2e-16	***
AnnoF[T.A2009]	-3.39018	0.89287	-3.7969	0.0001499	***
AnnoF[T.A2010]	-4.18311	0.89287	-4.6850	2.947e-06	***
as.factor(Mese)[T.7]	-3.65485	0.96953	-3.7697	0.0001672	***
as.factor(Mese)[T.8]	-30.26746	0.96953	-31.2187	< 2.2e-16	***
as.factor(Mese)[T.9]	-56.36898	1.19357	-47.2273	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

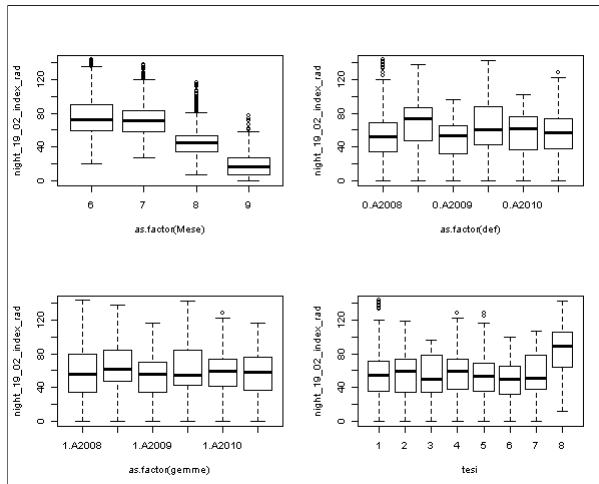
(Dispersion parameter for gaussian family taken to be 338.0199)

- Null deviance: 1990283.1496 on 2543 degrees of freedom.
- Residual deviance: 856880.5397 on 2535 degrees of freedom.

AIC: 22045

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard mattutino tra le ore 5-10 radiazione: dmorn_stderr_5_10_rad donna

Call: glm(formula = dmorn_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-33.1	-7.8	-0.6	6.9	41.1

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	18.80994	0.66348	28.3506	< 2.2e-16	***
as.factor(def)[T.1]	0.48240	0.45104	1.0695	0.284939	
as.factor(gemme)[T.3]	-1.10910	0.45104	-2.4590	0.014000	*
as.factor(dirad)[T.1]	-0.96213	0.45104	-2.1331	0.033010	*
AnnoF[T.A2009]	-1.81811	0.55241	-3.2912	0.001011	**
AnnoF[T.A2010]	-1.41587	0.55241	-2.5631	0.010432	*
as.factor(Mese)[T.7]	6.66034	0.59984	11.1035	< 2.2e-16	***
as.factor(Mese)[T.8]	29.60060	0.59984	49.3474	< 2.2e-16	***
as.factor(Mese)[T.9]	47.74272	0.73845	64.6524	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

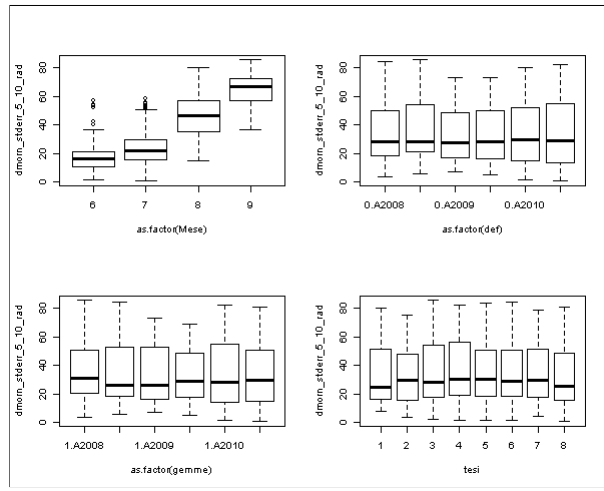
(Dispersion parameter for gaussian family taken to be 129.3875)

- Null deviance: 1069786.3657 on 2543 degrees of freedom.
- Residual deviance: 327997.3816 on 2535 degrees of freedom.

AIC: 19602

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza mattutina tra le ore 5-10 radiazione: dmorn_var_5_10_rad donna

Call: glm(formula = dmorn_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-17961	-2814	-581	2300	23310

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2775.59	343.45	8.0815	9.800e-16	***
as.factor(def)[T.1]	455.15	233.48	1.9494	0.0513577	.
as.factor(gemme)[T.3]	-779.69	233.48	-3.3394	0.0008518	***
as.factor(dirad)[T.1]	-433.40	233.48	-1.8562	0.0635371	.
AnnoF[T.A2009]	-981.35	285.96	-3.4318	0.0006092	***
AnnoF[T.A2010]	162.39	285.96	0.5679	0.5701729	
as.factor(Mese)[T.7]	1891.69	310.51	6.0922	1.283e-09	***
as.factor(Mese)[T.8]	12207.72	310.51	39.3151	< 2.2e-16	***
as.factor(Mese)[T.9]	23561.92	382.26	61.6381	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

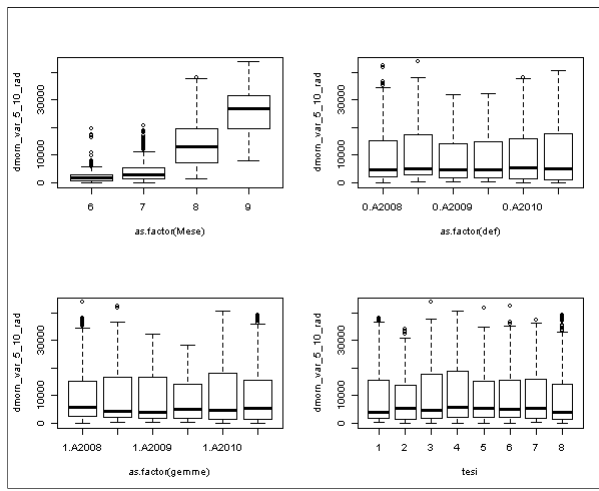
(Dispersion parameter for gaussian family taken to be 34671349)

- Null deviance: 261089488499.476 on 2543 degrees of freedom.
- Residual deviance: 87891869651.8161 on 2535 degrees of freedom.

AIC: 51398

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 radiazione: dtwi_stderr_15_20_rad donna

Call: glm(formula = dtwi_stderr_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-34.72	-9.53	-0.25	9.56	33.06

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	38.56773	0.72852	52.9396	< 2.2e-16	***
as.factor(def)[T.1]	-1.25382	0.49526	-2.5316	0.0114133	*
as.factor(gemme)[T.3]	0.27170	0.49526	0.5486	0.5833347	
as.factor(dirad)[T.1]	-0.61250	0.49526	-1.2367	0.2163075	
AnnoF[T.A2009]	2.15998	0.60657	3.5610	0.0003763	***
AnnoF[T.A2010]	-1.29482	0.60657	-2.1347	0.0328842	*
as.factor(Mese)[T.7]	-4.43448	0.65865	-6.7327	2.053e-11	***
as.factor(Mese)[T.8]	17.83623	0.65865	27.0800	< 2.2e-16	***
as.factor(Mese)[T.9]	32.39355	0.81085	39.9501	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

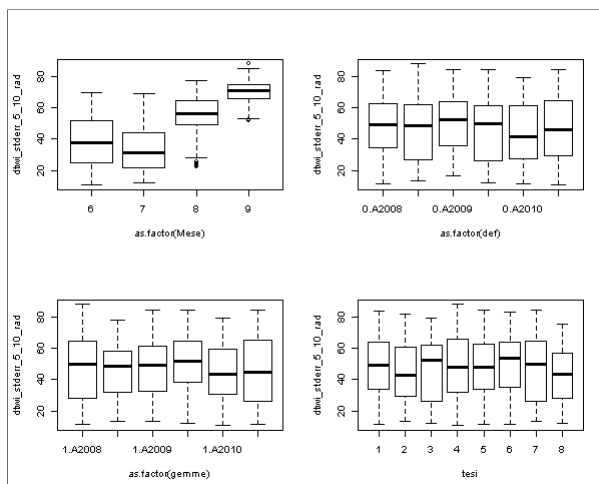
(Dispersion parameter for gaussian family taken to be 156.0016)

- Null deviance: 847049.0487 on 2543 degrees of freedom.
- Residual deviance: 395463.9474 on 2535 degrees of freedom.

AIC: 20077

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



varianza serale tra le ore 15-20 radiazione: dtwi_var_15_20_rad donna

Call: glm(formula = dtwi_var_5_10_rad ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-17032	-5172	-1102	4815	19681

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	10280.990	384.385	26.7466	< 2.2e-16	***
as.factor(def)[T.1]	-377.136	261.312	-1.4432	0.1490757	
as.factor(gemme)[T.3]	97.707	261.312	0.3739	0.7085024	
as.factor(dirad)[T.1]	-408.589	261.312	-1.5636	0.1180341	
AnnoF[T.A2009]	1146.414	320.040	3.5821	0.0003473	***
AnnoF[T.A2010]	-533.163	320.040	-1.6659	0.0958516	.
as.factor(Mese)[T.7]	-2257.677	347.518	-6.4966	9.856e-11	***
as.factor(Mese)[T.8]	9263.169	347.518	26.6552	< 2.2e-16	***
as.factor(Mese)[T.9]	19898.307	427.822	46.5107	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

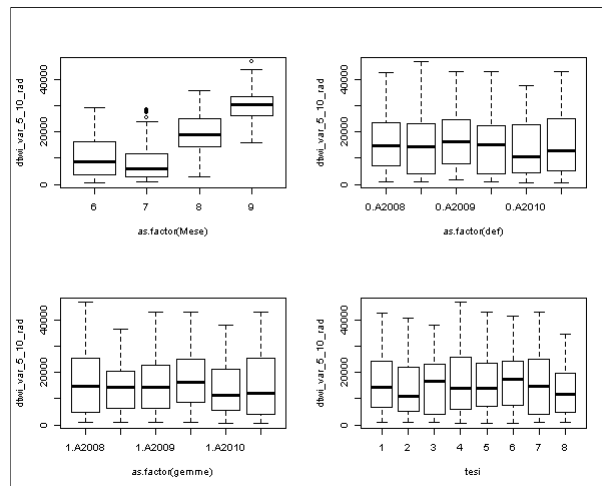
(Dispersion parameter for gaussian family taken to be 43428443)

- Null deviance: 261513112236.451 on 2543 degrees of freedom.
- Residual deviance: 110091104107.977 on 2535 degrees of freedom.

AIC: 51971

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



Analisi indici giornalieri temperatura grappolo donna

temperatura media: avg_tgrap donna

Call: glm(formula = avg_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-9.45	-1.34	-0.16	1.36	6.57

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.958749	0.119296	200.8348	< 2.2e-16	***
as.factor(def)[T.1]	0.266977	0.081099	3.2920	0.001009	**
as.factor(gemme)[T.3]	-0.394331	0.081099	-4.8623	1.231e-06	***
as.factor(dirad)[T.1]	-0.093111	0.081099	-1.1481	0.251030	
AnnoF[T.A2009]	0.059571	0.099326	0.5998	0.548724	
AnnoF[T.A2010]	-0.208540	0.099326	-2.0996	0.035867	*
as.factor(Mese)[T.7]	2.743011	0.107854	25.4326	< 2.2e-16	***
as.factor(Mese)[T.8]	2.925041	0.107854	27.1204	< 2.2e-16	***
as.factor(Mese)[T.9]	0.841721	0.132777	6.3394	2.722e-10	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

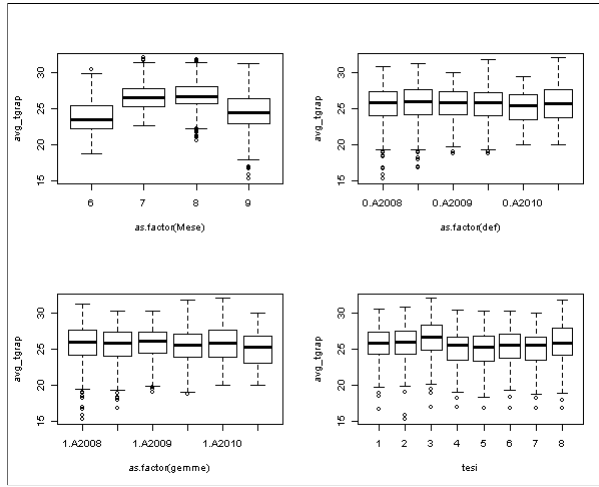
(Dispersion parameter for gaussian family taken to be 4.183037)

- Null deviance: 14974.9198 on 2543 degrees of freedom.
- Residual deviance: 10603.9996 on 2535 degrees of freedom.

AIC: 10871

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.822883692460685



cumulata temperatura grappolo: sum_tgrap donna

Call: glm(formula = sum_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-226.9	-32.1	-3.8	32.6	157.6

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	575.0100	2.8631	200.8348	< 2.2e-16	***
as.factor(def)[T.1]	6.4074	1.9464	3.2919	0.001009	**
as.factor(gemme)[T.3]	-9.4639	1.9464	-4.8623	1.231e-06	***
as.factor(dirad)[T.1]	-2.2347	1.9464	-1.1481	0.251028	
AnnoF[T.A2009]	1.4297	2.3838	0.5998	0.548714	
AnnoF[T.A2010]	-5.0050	2.3838	-2.0996	0.035866	*
as.factor(Mese)[T.7]	65.8323	2.5885	25.4326	< 2.2e-16	***
as.factor(Mese)[T.8]	70.2010	2.5885	27.1204	< 2.2e-16	***
as.factor(Mese)[T.9]	20.2013	3.1866	6.3394	2.722e-10	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

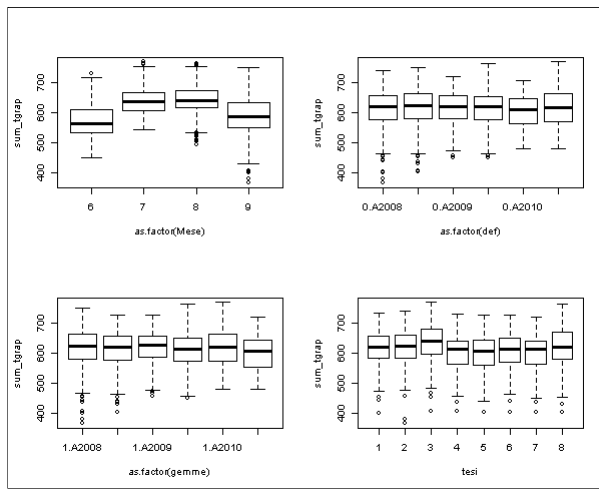
(Dispersion parameter for gaussian family taken to be 2409.429)

- Null deviance: 8625552.896 on 2543 degrees of freedom.
- Residual deviance: 6107902.6363 on 2535 degrees of freedom.

AIC: 27041

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica grappolo: et_tgrap donna

Call: `glm(formula = et_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.998	-2.082	-0.013	2.208	12.652

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	14.01107	0.19292	72.6262	< 2.2e-16	***
as.factor(def)[T.1]	0.15273	0.13115	1.1646	0.2443063	
as.factor(gemme)[T.3]	-0.50277	0.13115	-3.8335	0.0001294	***
as.factor(dirad)[T.1]	-0.35981	0.13115	-2.7435	0.0061223	**
AnnoF[T.A2009]	-1.04933	0.16063	-6.5327	7.777e-11	***
AnnoF[T.A2010]	-0.75932	0.16063	-4.7273	2.400e-06	***
as.factor(Mese)[T.7]	2.05972	0.17442	11.8092	< 2.2e-16	***
as.factor(Mese)[T.8]	3.09477	0.17442	17.7435	< 2.2e-16	***
as.factor(Mese)[T.9]	3.01286	0.21472	14.0315	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 10.93950)

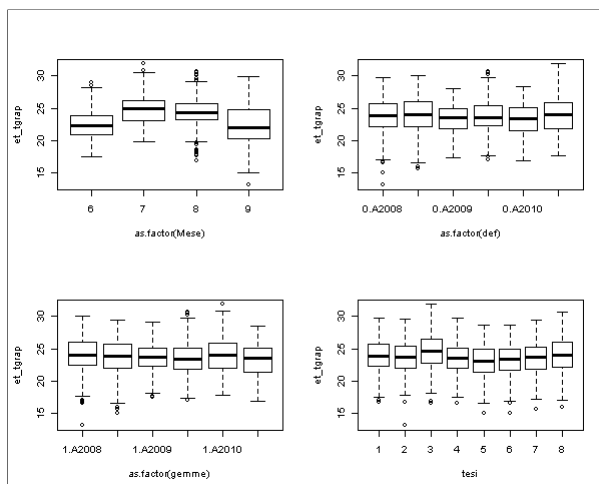
• Null deviance: 32524.0983 on 2543 degrees of freedom.

• Residual deviance: 27731.6407 on 2535 degrees of freedom.

AIC: 13317

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.847995222690575



media ore 3-10 temp. grappolo: morning_03_10_index_tgrap donna

Call: `glm(formula = morning_03_10_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-6.99	-1.59	-0.12	1.46	7.30

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	22.841391	0.130019	175.6772	< 2.2e-16	***
as.factor(def)[T.1]	0.426493	0.088389	4.8252	1.482e-06	***
as.factor(gemme)[T.3]	-0.447051	0.088389	-5.0578	4.545e-07	***
as.factor(dirad)[T.1]	-0.141149	0.088389	-1.5969	0.110411	
AnnoF[T.A2009]	-0.352077	0.108254	-3.2523	0.001160	**
AnnoF[T.A2010]	-0.423098	0.108254	-3.9084	9.535e-05	***
as.factor(Mese)[T.7]	2.156945	0.117549	18.3494	< 2.2e-16	***
as.factor(Mese)[T.8]	1.939254	0.117549	16.4975	< 2.2e-16	***
as.factor(Mese)[T.9]	0.083782	0.144712	0.5790	0.562669	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

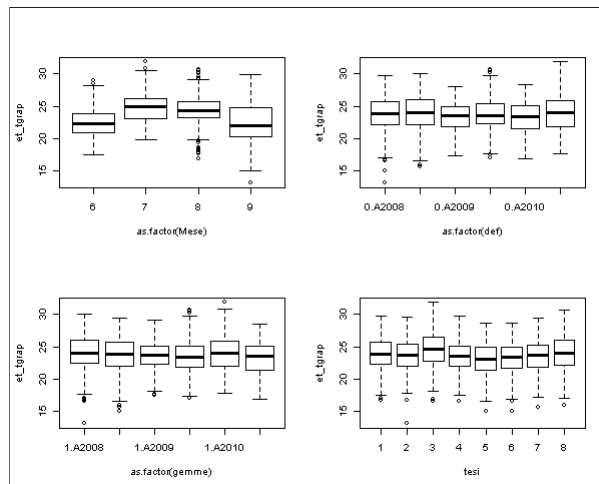
(Dispersion parameter for gaussian family taken to be 4.968846)

- Null deviance: 15476.9955 on 2543 degrees of freedom.
- Residual deviance: 12596.0235 on 2535 degrees of freedom.

AIC: 11309

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.847995222690575



media ore 11-18 temp. grappolo: diurnal_11_18_index_tgrap donna

Call: glm(formula = diurnal_11_18_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.64	-1.46	0.22	1.77	9.39

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.811916	0.163942	175.7442	< 2.2e-16	***
as.factor(def)[T.1]	0.258663	0.111451	2.3209	0.020373	*
as.factor(gemme)[T.3]	-0.553980	0.111451	-4.9706	7.119e-07	***
as.factor(dirad)[T.1]	-0.188919	0.111451	-1.6951	0.090181	.
AnnoF[T.A2009]	-0.074471	0.136499	-0.5456	0.585404	
AnnoF[T.A2010]	-0.374809	0.136499	-2.7459	0.006078	**
as.factor(Mese)[T.7]	3.794799	0.148218	25.6028	< 2.2e-16	***
as.factor(Mese)[T.8]	4.344729	0.148218	29.3130	< 2.2e-16	***
as.factor(Mese)[T.9]	1.975353	0.182468	10.8257	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

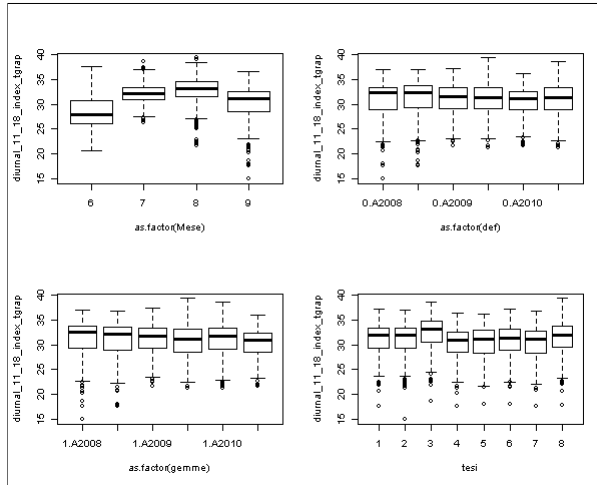
(Dispersion parameter for gaussian family taken to be 7.899937)

- Null deviance: 28516.7189 on 2543 degrees of freedom.
- Residual deviance: 20026.3398 on 2535 degrees of freedom.

AIC: 12489

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.96448431956953



media ore 19-02 temp. grappolo: night_19_02_index_tgrap donna

Call: `glm(formula = night_19_02_index_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.618	-1.406	-0.069	1.286	8.792

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.222943	0.121347	166.6533	< 2.2e-16	***
as.factor(def)[T.1]	0.115769	0.082494	1.4034	0.1606322	
as.factor(gemme)[T.3]	-0.181958	0.082494	-2.2057	0.0274933	*
as.factor(dirad)[T.1]	0.050734	0.082494	0.6150	0.5386095	
AnnoF[T.A2009]	0.605265	0.101034	5.9907	2.386e-09	***
AnnoF[T.A2010]	0.172280	0.101034	1.7052	0.0882857	.
as.factor(Mese)[T.7]	2.277288	0.109709	20.7576	< 2.2e-16	***
as.factor(Mese)[T.8]	2.491138	0.109709	22.7068	< 2.2e-16	***
as.factor(Mese)[T.9]	0.466029	0.135060	3.4505	0.0005686	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

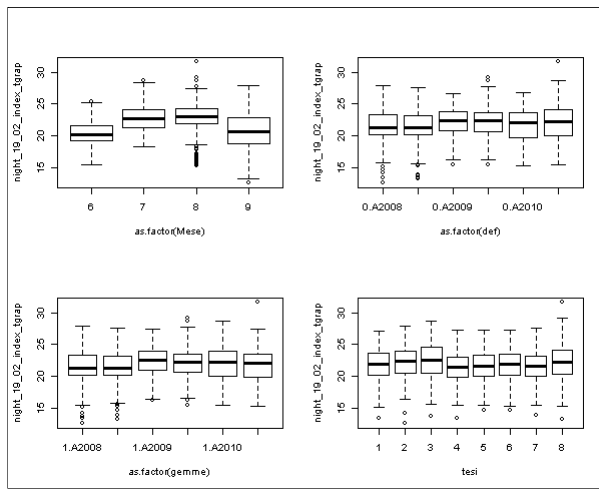
(Dispersion parameter for gaussian family taken to be 4.32815)

- Null deviance: 14295.2972 on 2543 degrees of freedom.
- Residual deviance: 10971.86 on 2535 degrees of freedom.

AIC: 10958

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.731856905387191



errore standard mattutino tra le ore 5-10 temp. grappolo: dmorn_stderr_5_10_tgrap donna

Call: glm(formula = dmorn_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-2.08	-0.39	0.03	0.44	2.85

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.878287	0.038863	48.3306	< 2.2e-16	***
as.factor(def)[T.1]	0.084896	0.026420	3.2133	0.0013285	**
as.factor(gemme)[T.3]	-0.135730	0.026420	-5.1374	2.998e-07	***
as.factor(dirad)[T.1]	-0.092860	0.026420	-3.5148	0.0004478	***
AnnoF[T.A2009]	-0.229451	0.032358	-7.0911	1.719e-12	***
AnnoF[T.A2010]	-0.165578	0.032358	-5.1171	3.335e-07	***
as.factor(Mese)[T.7]	0.443965	0.035136	12.6357	< 2.2e-16	***
as.factor(Mese)[T.8]	0.705027	0.035136	20.0657	< 2.2e-16	***
as.factor(Mese)[T.9]	0.445855	0.043255	10.3076	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

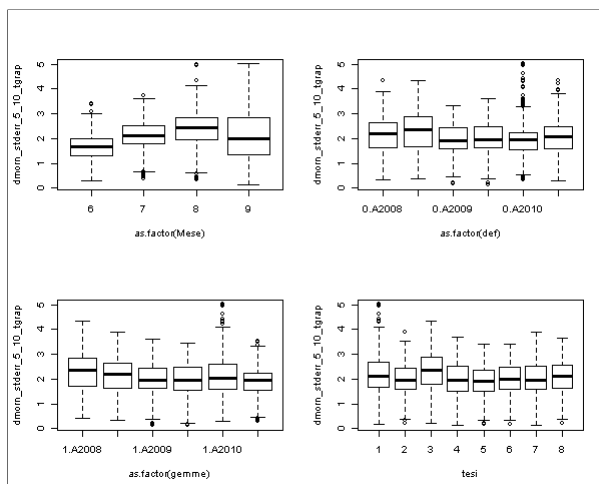
(Dispersion parameter for gaussian family taken to be 0.4439364)

- Null deviance: 1353.7239 on 2543 degrees of freedom.
- Residual deviance: 1125.3788 on 2535 degrees of freedom.

AIC: 5164.6

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.208043531647799



varianza mattutina tra le ore 5-10 temp. grappolo: dmorn_var_5_10_tgrap donna

Call: glm(formula = dmorn_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-39.9	-11.2	-1.4	9.5	116.3

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	24.75352	1.00691	24.5837	< 2.2e-16	***
as.factor(def)[T.1]	2.38083	0.68451	3.4781	0.0005134	***
as.factor(gemme)[T.3]	-4.24428	0.68451	-6.2004	6.554e-10	***
as.factor(dirad)[T.1]	-3.02452	0.68451	-4.4185	1.036e-05	***
AnnoF[T.A2009]	-6.74432	0.83835	-8.0447	1.314e-15	***
AnnoF[T.A2010]	-4.34107	0.83835	-5.1781	2.418e-07	***
as.factor(Mese)[T.7]	10.29948	0.91033	11.3140	< 2.2e-16	***
as.factor(Mese)[T.8]	18.42373	0.91033	20.2384	< 2.2e-16	***
as.factor(Mese)[T.9]	14.02104	1.12069	12.5110	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 298.0037)

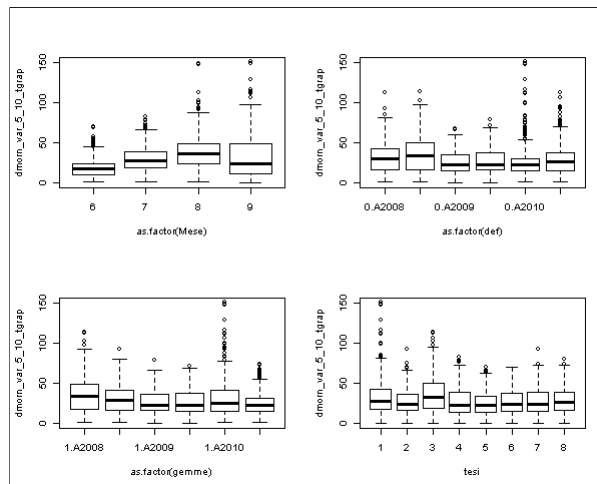
• Null deviance: 924041.4298 on 2543 degrees of freedom.

• Residual deviance: 755439.3629 on 2535 degrees of freedom.

AIC: 21724

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



errore standard serale tra le ore 15-20 temp. grappolo: dtwi_stderr_15_20_tgrap donna

Call: glm(formula = dtwi_stderr_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.3547	-0.1884	-0.0017	0.2016	1.5859

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.0849701	0.0198093	54.7708	< 2.2e-16	***
as.factor(def)[T.1]	0.0062848	0.0134667	0.4667	0.6407590	
as.factor(gemme)[T.3]	-0.0658818	0.0134667	-4.8922	1.060e-06	***
as.factor(dirad)[T.1]	-0.0348439	0.0134667	-2.5874	0.0097251	**
AnnoF[T.A2009]	-0.1450086	0.0164933	-8.7920	< 2.2e-16	***
AnnoF[T.A2010]	-0.0616959	0.0164933	-3.7407	0.0001876	***
as.factor(Mese)[T.7]	0.1430833	0.0179094	7.9893	2.040e-15	***
as.factor(Mese)[T.8]	0.2942594	0.0179094	16.4305	< 2.2e-16	***
as.factor(Mese)[T.9]	0.4229819	0.0220478	19.1847	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

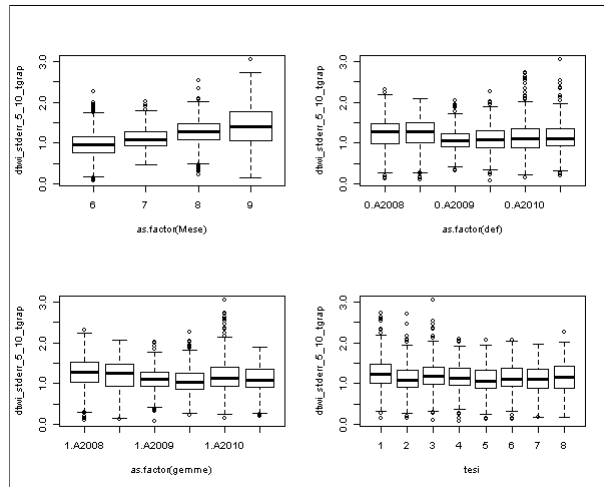
(Dispersion parameter for gaussian family taken to be 0.1153399)

- Null deviance: 359.3048 on 2543 degrees of freedom.
- Residual deviance: 292.3866 on 2535 degrees of freedom.

AIC: 1735.8

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.197101733199746



varianza serale tra le ore 15-20 temp. grappolo: dtwi_var_15_20_tgrap donna

Call: glm(formula = dtwi_var_5_10_tgrap ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-15.00	-2.98	-0.62	2.38	41.14

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	8.232221	0.293105	28.0862	< 2.2e-16	***
as.factor(def)[T.1]	0.013347	0.199258	0.0670	0.9465985	
as.factor(gemme)[T.3]	-1.035627	0.199258	-5.1974	2.182e-07	***
as.factor(dirad)[T.1]	-0.579479	0.199258	-2.9082	0.0036670	**
AnnoF[T.A2009]	-2.441045	0.244040	-10.0026	< 2.2e-16	***
AnnoF[T.A2010]	-0.903156	0.244040	-3.7008	0.0002195	***
as.factor(Mese)[T.7]	1.473721	0.264993	5.5614	2.957e-08	***
as.factor(Mese)[T.8]	3.929161	0.264993	14.8274	< 2.2e-16	***
as.factor(Mese)[T.9]	6.905158	0.326227	21.1667	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

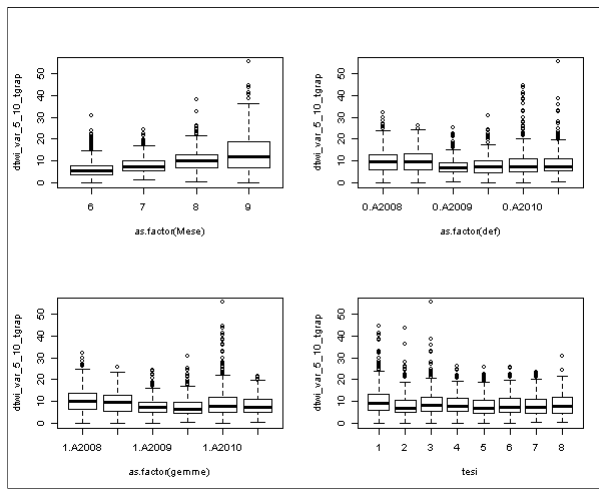
(Dispersion parameter for gaussian family taken to be 25.25158)

- Null deviance: 81122.8453 on 2543 degrees of freedom.
- Residual deviance: 64012.7576 on 2535 degrees of freedom.

AIC: 15445

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.998800275566707



Analisi indici giornalieri tarja canopy donna

temperatura canopy media: avg_tair donna

Call: `glm(formula = avg_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-8.344	-1.340	-0.042	1.319	6.599

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	23.870252	0.115987	205.8004	< 2.2e-16	***
as.factor(def)[T.1]	0.273861	0.078850	3.4732	0.000523	***
as.factor(gemme)[T.3]	-0.061554	0.078850	-0.7807	0.435081	
as.factor(dirad)[T.1]	0.010450	0.078850	0.1325	0.894572	
AnnoF[T.A2009]	-0.020999	0.096571	-0.2174	0.827875	
AnnoF[T.A2010]	-0.640962	0.096571	-6.6372	3.896e-11	***
as.factor(Mese)[T.7]	2.802603	0.104863	26.7264	< 2.2e-16	***
as.factor(Mese)[T.8]	2.464267	0.104863	23.4999	< 2.2e-16	***
as.factor(Mese)[T.9]	0.327314	0.129094	2.5355	0.011290	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

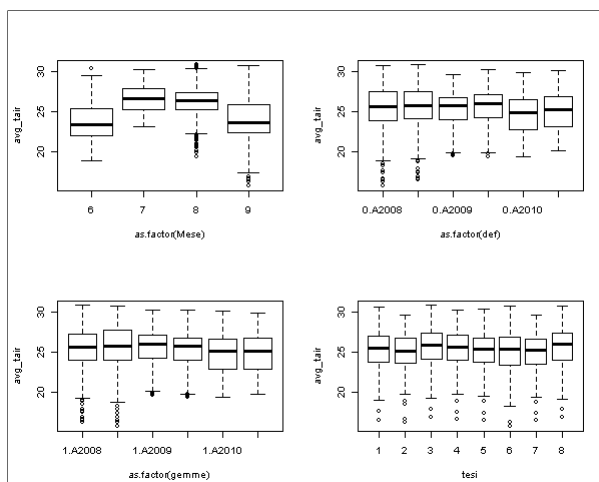
(Dispersion parameter for gaussian family taken to be 3.954238)

- Null deviance: 14295.2271 on 2543 degrees of freedom.
- Residual deviance: 10023.994 on 2535 degrees of freedom.

AIC: 10728

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.816390586291943



cumulata temperatura canopy: sum_tair donna

Call: glm(formula = sum_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-200.2	-32.2	-1.0	31.6	158.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	572.88609	2.78370	205.8004	< 2.2e-16	***
as.factor(def)[T.1]	6.57265	1.89241	3.4732	0.000523	***
as.factor(gemme)[T.3]	-1.47735	1.89241	-0.7807	0.435067	
as.factor(dirad)[T.1]	0.25077	1.89241	0.1325	0.894588	
AnnoF[T.A2009]	-0.50399	2.31771	-0.2175	0.827874	
AnnoF[T.A2010]	-15.38310	2.31771	-6.6372	3.896e-11	***
as.factor(Mese)[T.7]	67.26249	2.51671	26.7264	< 2.2e-16	***
as.factor(Mese)[T.8]	59.14242	2.51671	23.4999	< 2.2e-16	***
as.factor(Mese)[T.9]	7.85554	3.09827	2.5355	0.011290	*

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

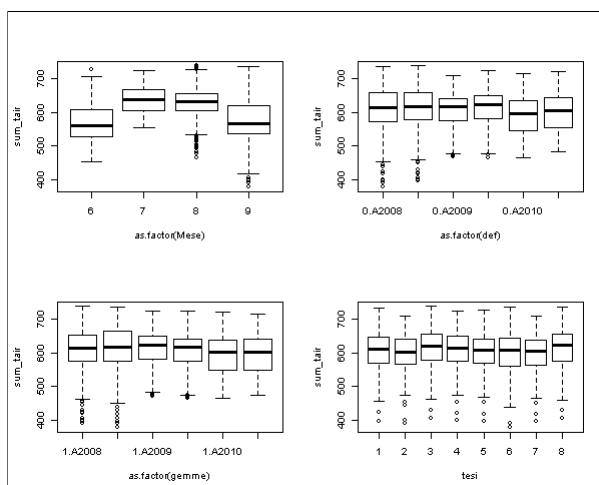
(Dispersion parameter for gaussian family taken to be 2277.642)

- Null deviance: 8234053.8983 on 2543 degrees of freedom.
- Residual deviance: 5773821.8782 on 2535 degrees of freedom.

AIC: 26898

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 1



escursione termica canopy: et_tair donna

Call: glm(formula = et_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-11.52	-1.94	0.11	2.03	9.08

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	13.19497	0.17623	74.8736	< 2.2e-16	***
as.factor(def)[T.1]	0.48091	0.11980	4.0141	6.139e-05	***
as.factor(gemme)[T.3]	-0.34170	0.11980	-2.8521	0.0043778	**
as.factor(dirad)[T.1]	-0.42347	0.11980	-3.5347	0.0004155	***
AnnoF[T.A2009]	-0.26501	0.14673	-1.8061	0.0710232	.
AnnoF[T.A2010]	-0.67649	0.14673	-4.6104	4.218e-06	***
as.factor(Mese)[T.7]	2.01079	0.15933	12.6205	< 2.2e-16	***
as.factor(Mese)[T.8]	2.45829	0.15933	15.4292	< 2.2e-16	***
as.factor(Mese)[T.9]	2.10827	0.19614	10.7485	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

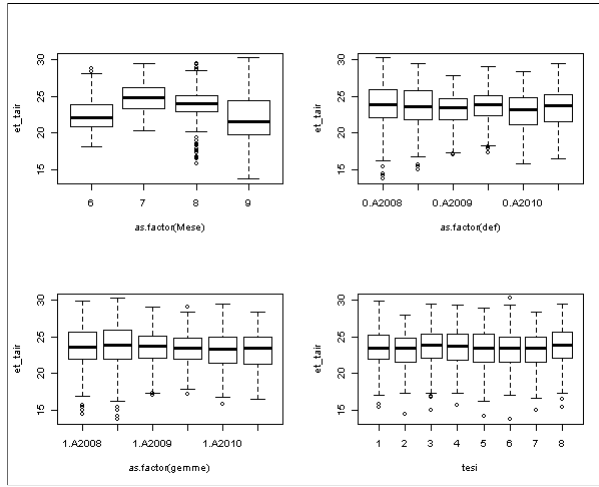
(Dispersion parameter for gaussian family taken to be 9.128522)

- Null deviance: 26221.5817 on 2543 degrees of freedom.
- Residual deviance: 23140.8042 on 2535 degrees of freedom.

AIC: 12856

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.702123450575838



media ore 03-10 temp. canopy: morning_03_10_index_tair donna

Call: glm(formula = morning_03_10_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-7.803	-1.469	-0.088	1.409	7.851

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	22.667601	0.126665	178.9569	< 2.2e-16	***
as.factor(def)[T.1]	0.216201	0.086109	2.5108	0.012108	*
as.factor(gemme)[T.3]	-0.061334	0.086109	-0.7123	0.476357	
as.factor(dirad)[T.1]	0.033924	0.086109	0.3940	0.693636	
AnnoF[T.A2009]	-0.309148	0.105462	-2.9314	0.003405	**
AnnoF[T.A2010]	-0.661480	0.105462	-6.2722	4.171e-10	***
as.factor(Mese)[T.7]	2.283779	0.114516	19.9428	< 2.2e-16	***
as.factor(Mese)[T.8]	1.631566	0.114516	14.2474	< 2.2e-16	***
as.factor(Mese)[T.9]	-0.213806	0.140979	-1.5166	0.129496	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

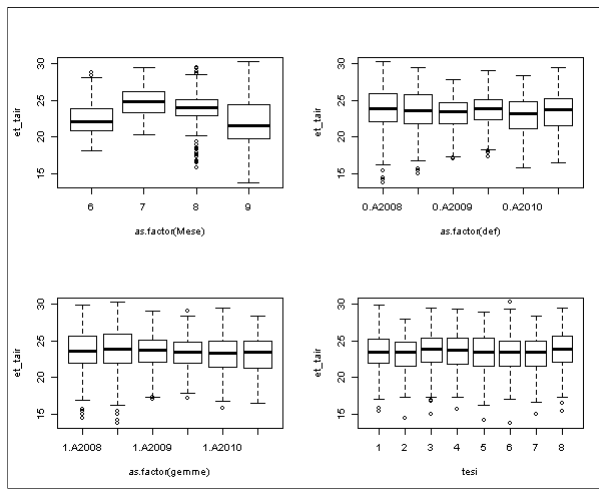
(Dispersion parameter for gaussian family taken to be 4.7158)

- Null deviance: 14889.0051 on 2543 degrees of freedom.
- Residual deviance: 11954.5526 on 2535 degrees of freedom.

AIC: 11176

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.702123450575838



media ore 11-18 temp. canopy: diurnal_11_18_index_tair donna

Call: `glm(formula = diurnal_11_18_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))`

Deviance Residuals:

Min	1Q	Median	3Q	Max
-13.68	-1.39	0.26	1.58	7.17

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	28.48287	0.15691	181.5221	< 2.2e-16	***
as.factor(def)[T.1]	0.50123	0.10667	4.6989	2.756e-06	***
as.factor(gemme)[T.3]	-0.26189	0.10667	-2.4551	0.01415	*
as.factor(dirad)[T.1]	-0.19728	0.10667	-1.8495	0.06451	.
AnnoF[T.A2009]	0.22535	0.13064	1.7249	0.08466	.
AnnoF[T.A2010]	-0.82314	0.13064	-6.3006	3.485e-10	***
as.factor(Mese)[T.7]	3.88183	0.14186	27.3635	< 2.2e-16	***
as.factor(Mese)[T.8]	3.78781	0.14186	26.7008	< 2.2e-16	***
as.factor(Mese)[T.9]	1.18108	0.17464	6.7629	1.674e-11	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

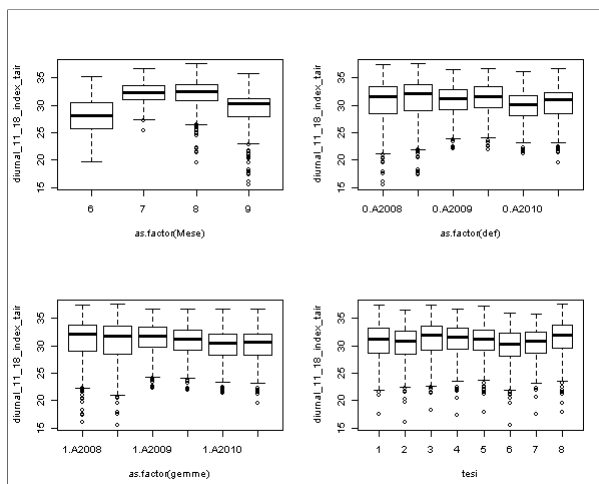
(Dispersion parameter for gaussian family taken to be 7.236851)

- Null deviance: 26700.4131 on 2543 degrees of freedom.
- Residual deviance: 18345.4167 on 2535 degrees of freedom.

AIC: 12266

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.962555953352373



media ore 19-02 temp. canopy: night_19_02_index_tair donna

Call: `glm(formula = night_19_02_index_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =`

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-8.107	-1.399	0.047	1.264	8.109

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.460295	0.121236	168.7638	< 2.2e-16	***
as.factor(def)[T.1]	0.104147	0.082418	1.2636	0.20648	
as.factor(gemme)[T.3]	0.138560	0.082418	1.6812	0.09285	.
as.factor(dirad)[T.1]	0.194706	0.082418	2.3624	0.01823	*
AnnoF[T.A2009]	0.020798	0.100942	0.2060	0.83677	
AnnoF[T.A2010]	-0.438267	0.100942	-4.3418	1.469e-05	***
as.factor(Mese)[T.7]	2.242205	0.109608	20.4565	< 2.2e-16	***
as.factor(Mese)[T.8]	1.973420	0.109608	18.0043	< 2.2e-16	***
as.factor(Mese)[T.9]	0.014662	0.134936	0.1087	0.91348	

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

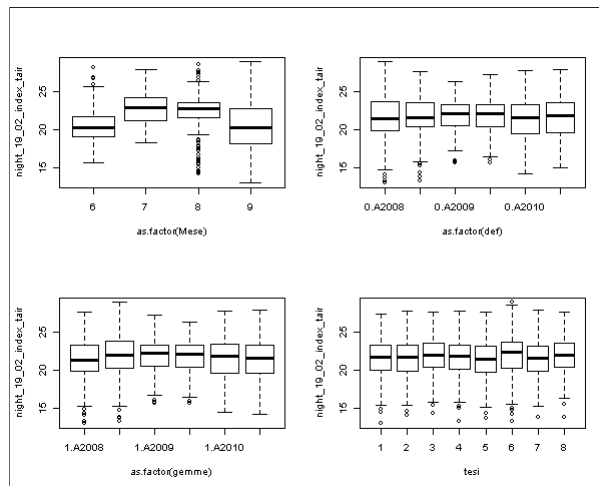
(Dispersion parameter for gaussian family taken to be 4.320225)

- Null deviance: 13867.0899 on 2543 degrees of freedom.
- Residual deviance: 10951.7697 on 2535 degrees of freedom.

AIC: 10953

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.685021587778185



errore standard mattutino tra le ore 5-10 temp. canopy: dmorn_stderr_5_10_tair donna

Call: glm(formula = dmorn_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.861	-0.348	0.042	0.410	1.942

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	1.753548	0.035835	48.9342	< 2.2e-16	***
as.factor(def)[T.1]	0.076941	0.024361	3.1584	0.0016053	**
as.factor(gemme)[T.3]	-0.062029	0.024361	-2.5462	0.0109484	*
as.factor(dirad)[T.1]	-0.058735	0.024361	-2.4110	0.0159795	*
AnnoF[T.A2009]	-0.080773	0.029836	-2.7072	0.0068306	**
AnnoF[T.A2010]	-0.099154	0.029836	-3.3233	0.0009023	***
as.factor(Mese)[T.7]	0.384887	0.032398	11.8800	< 2.2e-16	***
as.factor(Mese)[T.8]	0.536522	0.032398	16.5604	< 2.2e-16	***
as.factor(Mese)[T.9]	0.288079	0.039884	7.2229	6.7e-13	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

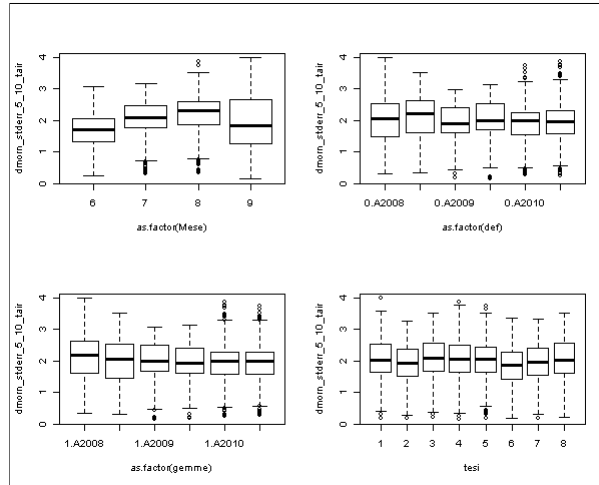
(Dispersion parameter for gaussian family taken to be 0.3774427)

- Null deviance: 1079.3491 on 2543 degrees of freedom.
- Residual deviance: 956.8171 on 2535 degrees of freedom.

AIC: 4751.8

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.136002768382631



varianza mattutina tra le ore 5-10 temp. canopy: dmorn_var_5_10_tair donna

Call: glm(formula = dmorn_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-32.1	-9.8	-1.0	8.6	65.3

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	20.91597	0.83271	25.1180	< 2.2e-16	***
as.factor(def)[T.1]	1.93936	0.56609	3.4259	0.0006225	***
as.factor(gemme)[T.3]	-1.49636	0.56609	-2.6433	0.0082600	**
as.factor(dirad)[T.1]	-1.46194	0.56609	-2.5825	0.0098634	**
AnnoF[T.A2009]	-2.67488	0.69332	-3.8581	0.0001171	***
AnnoF[T.A2010]	-2.92817	0.69332	-4.2234	2.491e-05	***
as.factor(Mese)[T.7]	8.60289	0.75284	11.4272	< 2.2e-16	***
as.factor(Mese)[T.8]	12.94242	0.75284	17.1914	< 2.2e-16	***
as.factor(Mese)[T.9]	8.99789	0.92681	9.7085	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

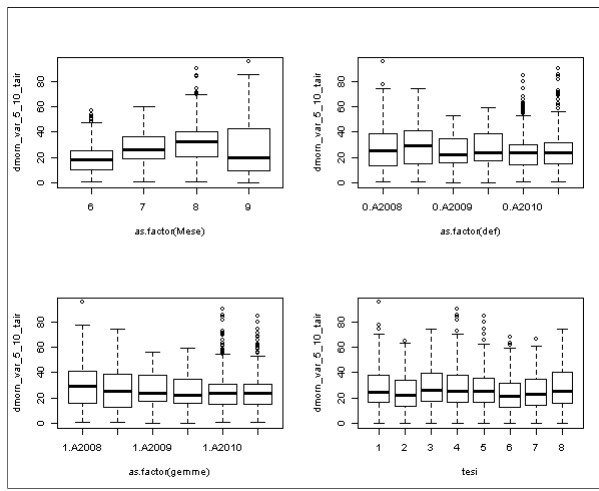
(Dispersion parameter for gaussian family taken to be 203.8108)

- Null deviance: 589185.85 on 2543 degrees of freedom.
- Residual deviance: 516660.2814 on 2535 degrees of freedom.

AIC: 20757

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.999999999999584



errore standard serale tra le ore 15-20 temp. canopy: dtwi_stderr_15_20_tair donna

Call: glm(formula = dtwi_stderr_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data = donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.2308	-0.1914	-0.0032	0.2016	1.0857

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	0.972435	0.019277	50.4443	< 2.2e-16	***
as.factor(def)[T.1]	0.083845	0.013105	6.3979	1.870e-10	***
as.factor(gemme)[T.3]	-0.066961	0.013105	-5.1095	3.470e-07	***
as.factor(dirad)[T.1]	-0.063659	0.013105	-4.8576	1.261e-06	***
AnnoF[T.A2009]	0.027840	0.016050	1.7346	0.08294	.
AnnoF[T.A2010]	-0.066308	0.016050	-4.1313	3.725e-05	***
as.factor(Mese)[T.7]	0.171509	0.017428	9.8407	< 2.2e-16	***
as.factor(Mese)[T.8]	0.306910	0.017428	17.6097	< 2.2e-16	***
as.factor(Mese)[T.9]	0.368312	0.021456	17.1661	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 0.1092291)

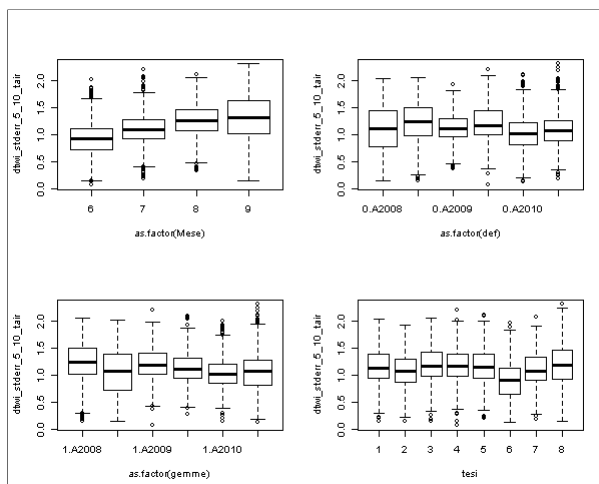
• Null deviance: 337.7482 on 2543 degrees of freedom.

• Residual deviance: 276.8958 on 2535 degrees of freedom.

AIC: 1597.3

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.190112581910978



varianza serale tra le ore 15-20 temp. canopy: dtwi_var_15_20_tairdonna

Call: glm(formula = dtwi_var_5_10_tair ~ as.factor(def) + as.factor(gemme) + as.factor(dirad) + AnnoF + as.factor(Mese), family = gaussian(identity), data =

donna_new_indici, subset = ((jd > 153) & (jd < 260)))

Deviance Residuals:

Min	1Q	Median	3Q	Max
-12.91	-2.94	-0.59	2.38	21.60

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6.56624	0.26722	24.5726	< 2.2e-16	***
as.factor(def)[T.1]	1.12816	0.18166	6.2103	6.161e-10	***
as.factor(gemme)[T.3]	-0.76437	0.18166	-4.2077	2.670e-05	***
as.factor(dirad)[T.1]	-0.71778	0.18166	-3.9513	7.986e-05	***
AnnoF[T.A2009]	-0.12790	0.22249	-0.5748	0.5654	
AnnoF[T.A2010]	-1.23873	0.22249	-5.5677	2.853e-08	***
as.factor(Mese)[T.7]	1.95668	0.24159	8.0992	8.500e-16	***
as.factor(Mese)[T.8]	3.96088	0.24159	16.3952	< 2.2e-16	***
as.factor(Mese)[T.9]	5.44208	0.29741	18.2980	< 2.2e-16	***

--- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

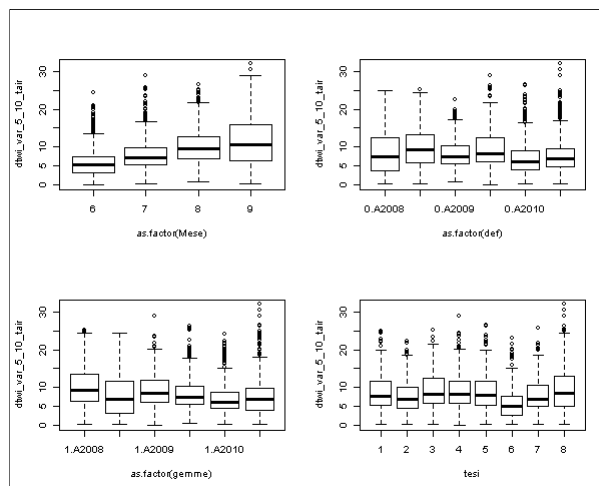
(Dispersion parameter for gaussian family taken to be 20.98797)

- Null deviance: 64773.9222 on 2543 degrees of freedom.
- Residual deviance: 53204.4966 on 2535 degrees of freedom.

AIC: 14974

Number of Fisher Scoring iterations: 2

Pseudo Rsquared: 0.989408782676975



Allegato 2.1 – Anova Multiscala Sangiovese

ANOVA PER VIGNETI SANGIOVESE (BROLIO E CORTIGLIANO) NEGLI ANNI 2008-2009-2010

▼ General Linear Model

IBIMET 1 2009-2010

Data for the following results were selected according to
SELECT (VARIETA\$ = 'Sangiovese') AND (ANNO <> 2008)

Effects coding used for categorical variables in model.
The categorical values encountered during processing are

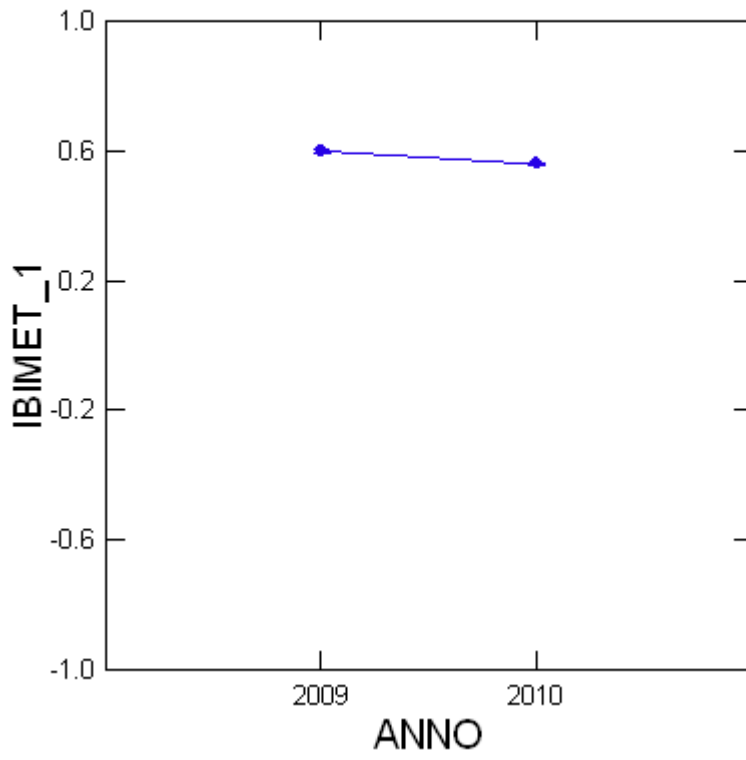
Variables	Levels			
ANNO (2 levels)	2,009.000	2,010.000		
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		
DIRAD\$ (2 levels)	D0	D1		

Dependent Variable	IBIMET_1
N	128
Multiple R	0.989
Squared Multiple R	0.978

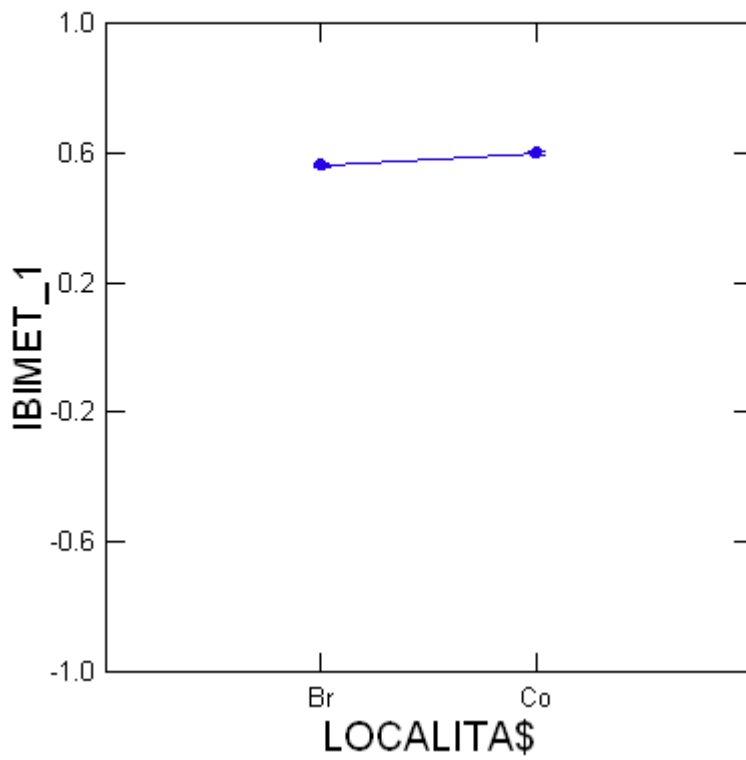
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.030	1	0.030	81.565	0.000
LOCALITA\$	0.025	1	0.025	68.392	0.000
N_GEMME\$	0.002	1	0.002	5.288	0.024
DEFOGL\$	0.035	1	0.035	95.441	0.000
DIRAD\$	0.000	1	0.000	0.890	0.348
LOCALITA\$*ANNO	0.683	1	0.683	1,836.044	0.000
N_GEMME\$*ANNO	0.000	1	0.000	0.005	0.943
DEFOGL\$*ANNO	0.008	1	0.008	20.333	0.000
DIRAD\$*ANNO	0.000	1	0.000	0.403	0.527
N_GEMME\$*LOCALITA\$	0.002	1	0.002	4.764	0.031
DEFOGL\$*LOCALITA\$	0.001	1	0.001	1.457	0.230
DIRAD\$*LOCALITA\$	0.000	1	0.000	1.159	0.284
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.129	0.720
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.925	0.338
DIRAD\$*DEFOGL\$	0.003	1	0.003	9.057	0.003
BLOCCO(LOCALITA\$)	0.299	6	0.050	133.988	0.000
BLOCCO*ANNO(LOCALITA\$)	0.011	6	0.002	4.771	0.000
Error	0.037	100	0.000		

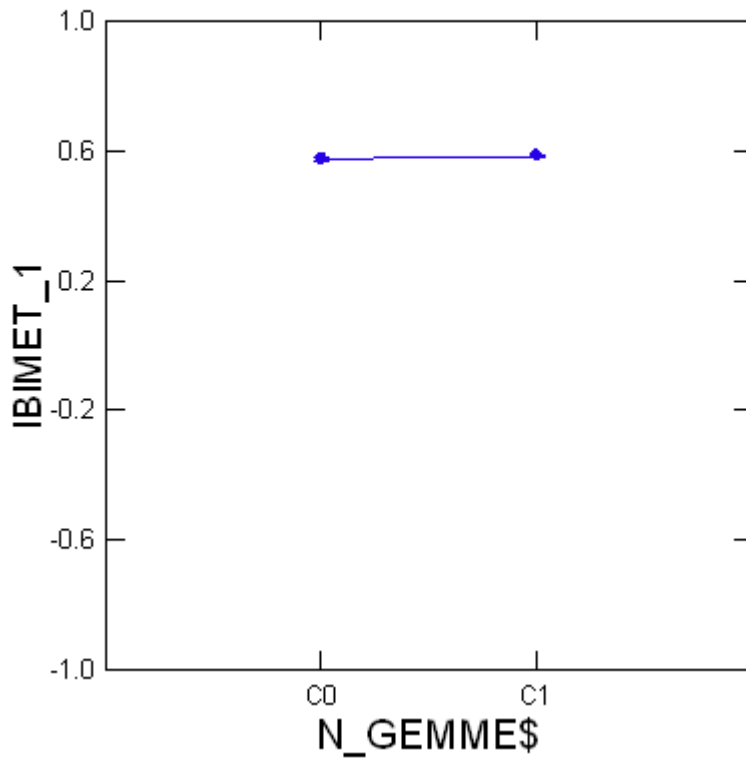
Least Squares Means



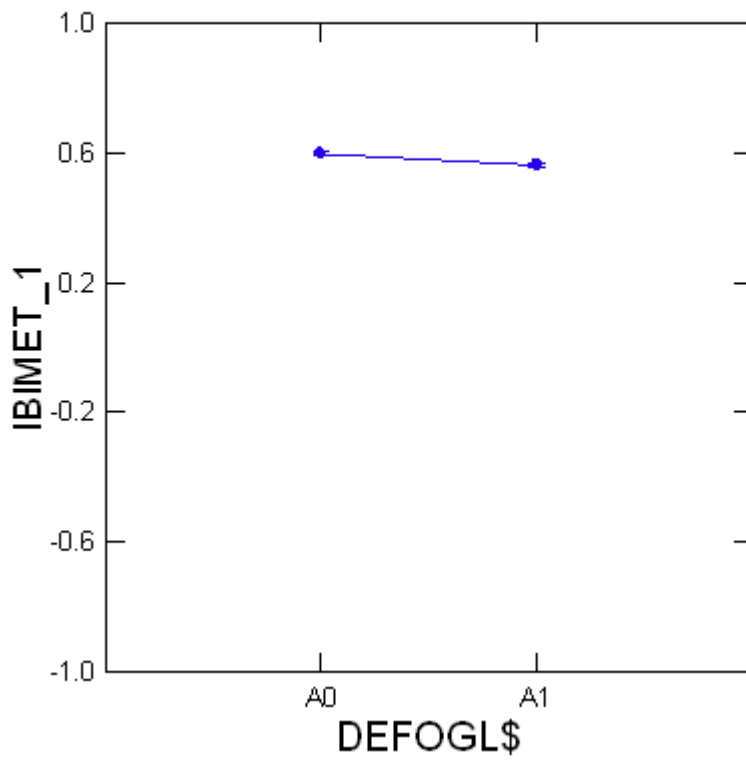
Least Squares Means



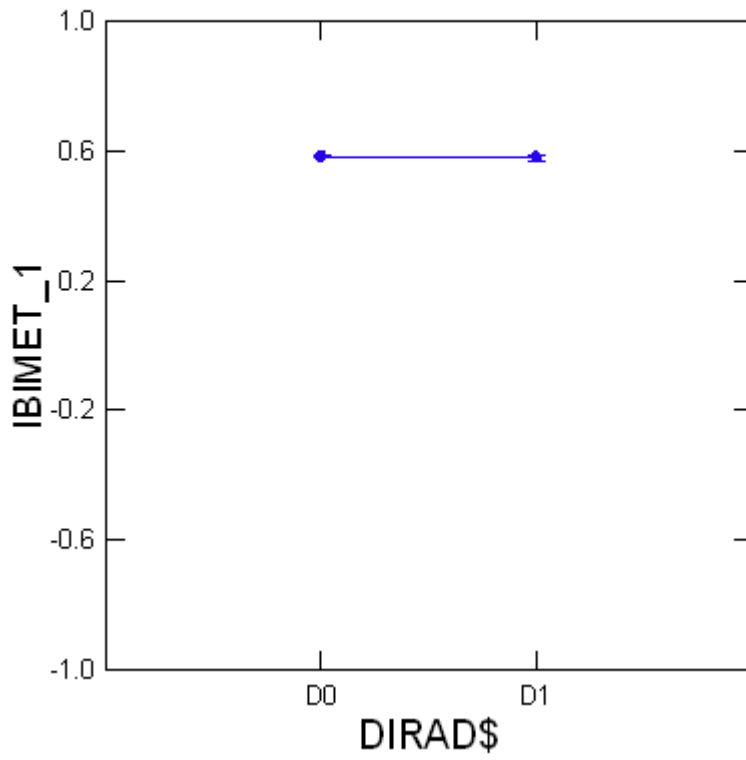
Least Squares Means



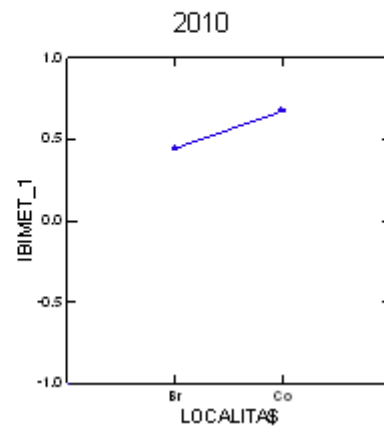
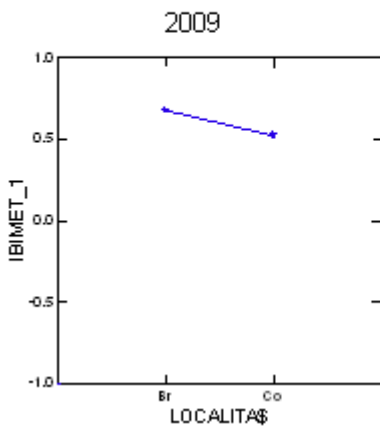
Least Squares Means



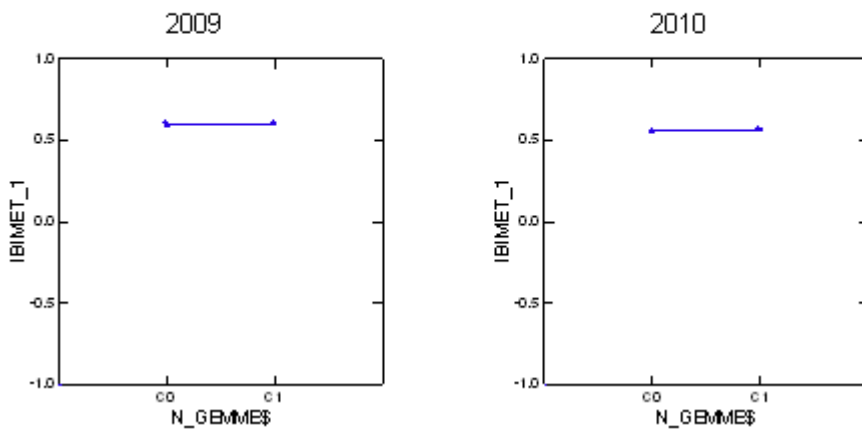
Least Squares Means



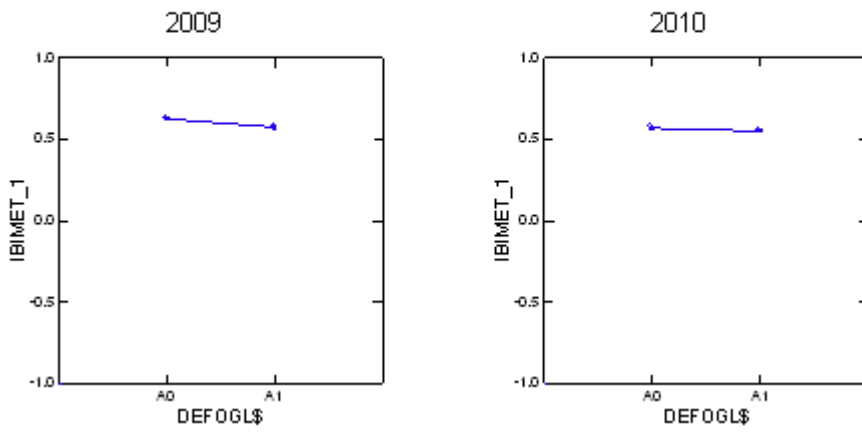
Least Squares Means



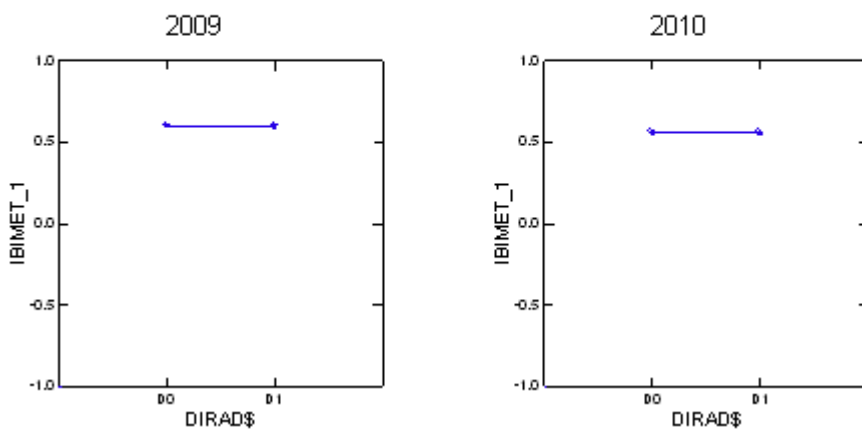
Least Squares Means



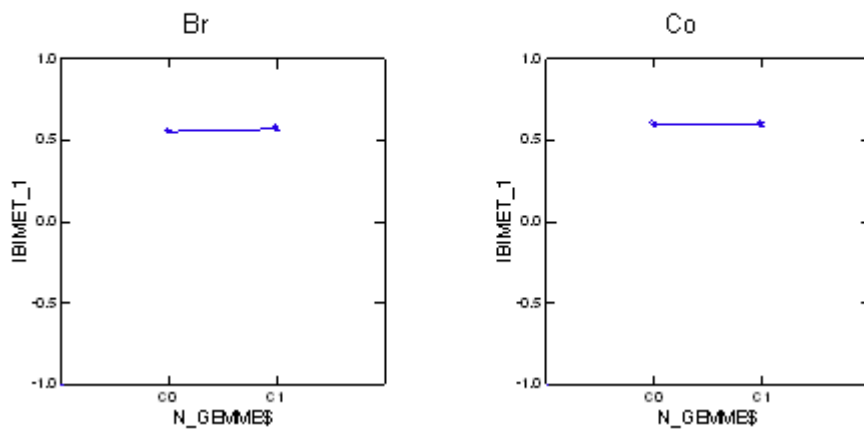
Least Squares Means



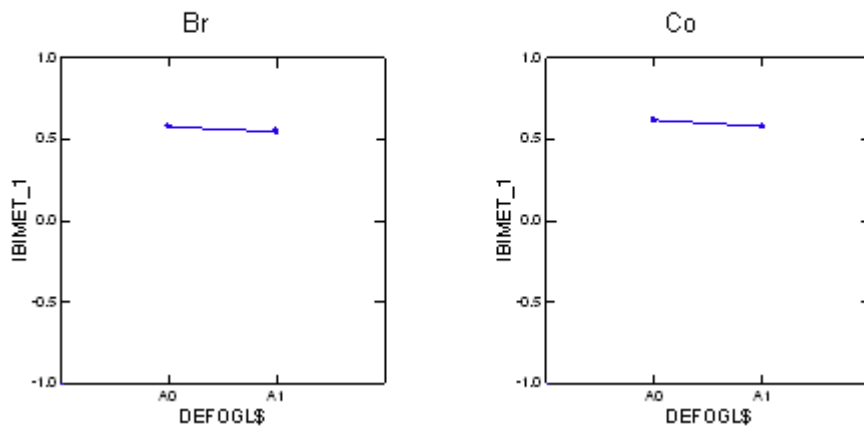
Least Squares Means



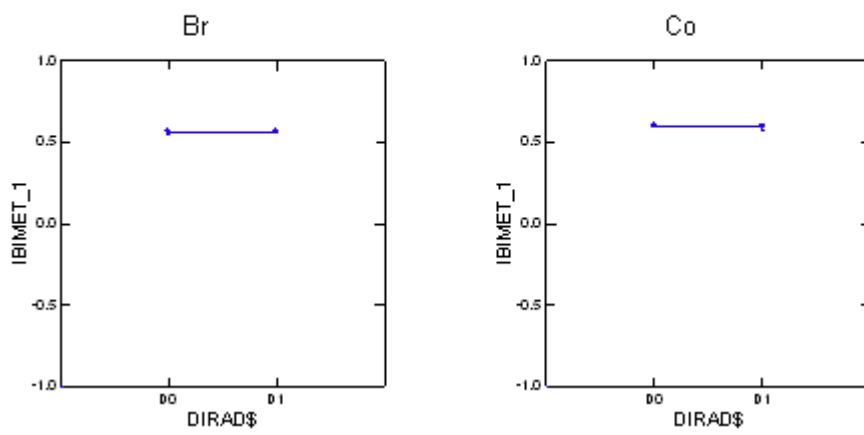
Least Squares Means



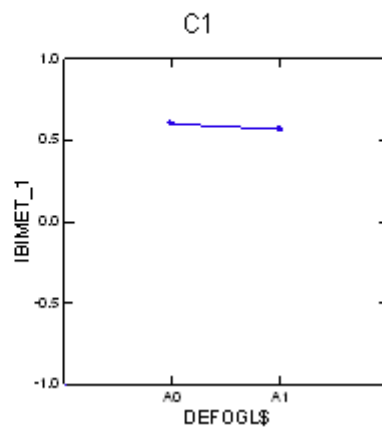
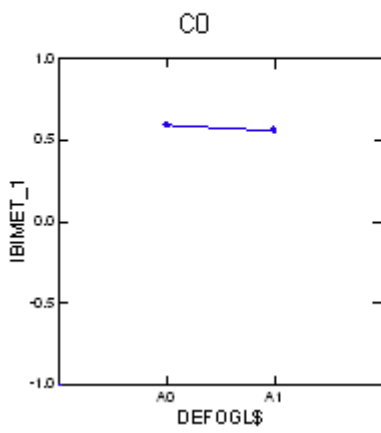
Least Squares Means



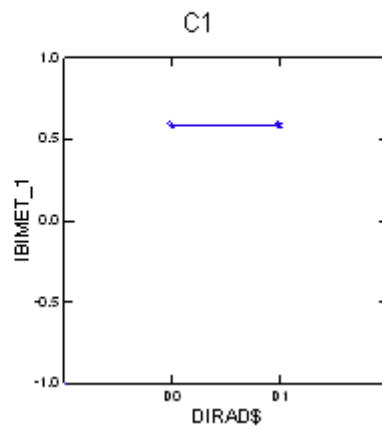
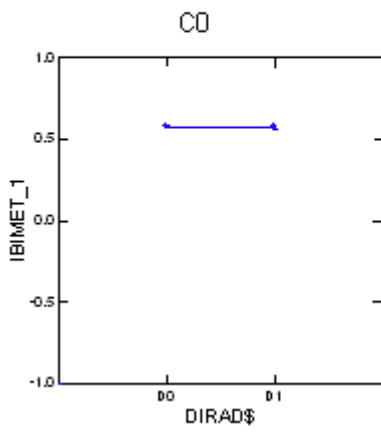
Least Squares Means



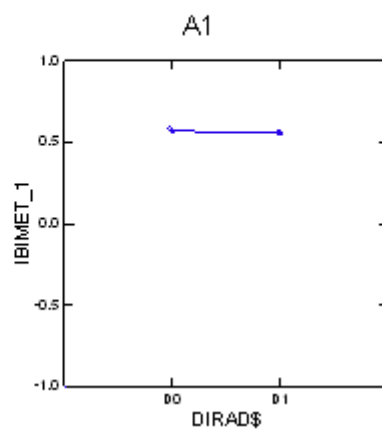
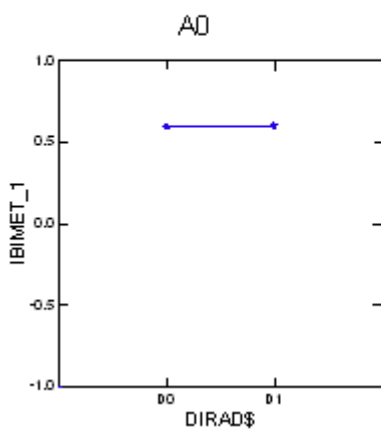
Least Squares Means



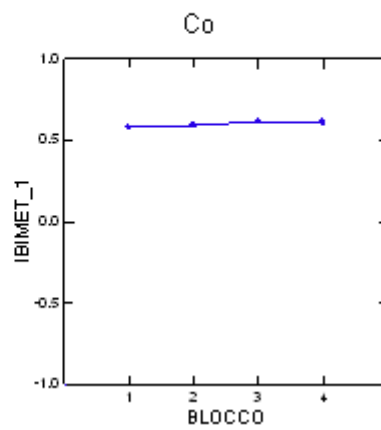
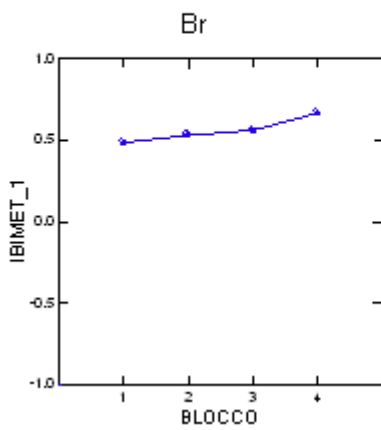
Least Squares Means



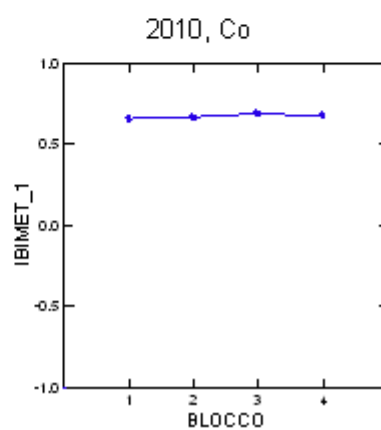
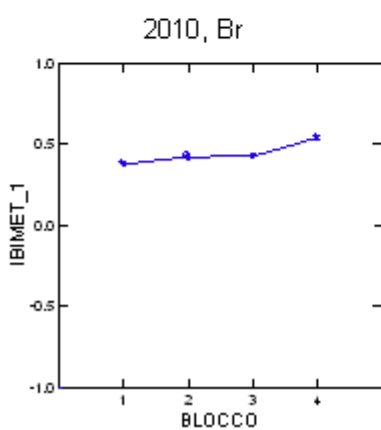
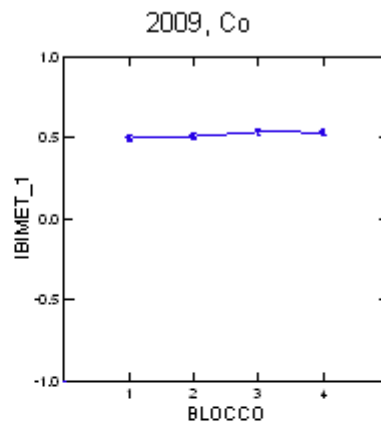
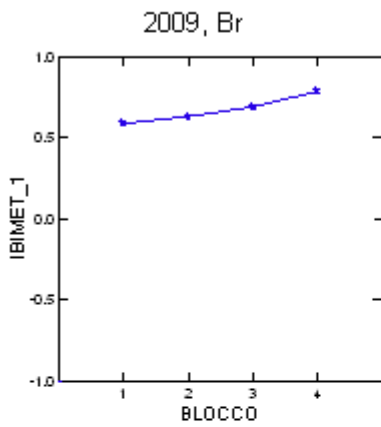
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.923
First Order Autocorrelation | 0.027

Information Criteria

AIC | -621.172
AIC (Corrected) | -603.417
Schwarz's BIC | -538.463

▼ General Linear Model

IBIMET2 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

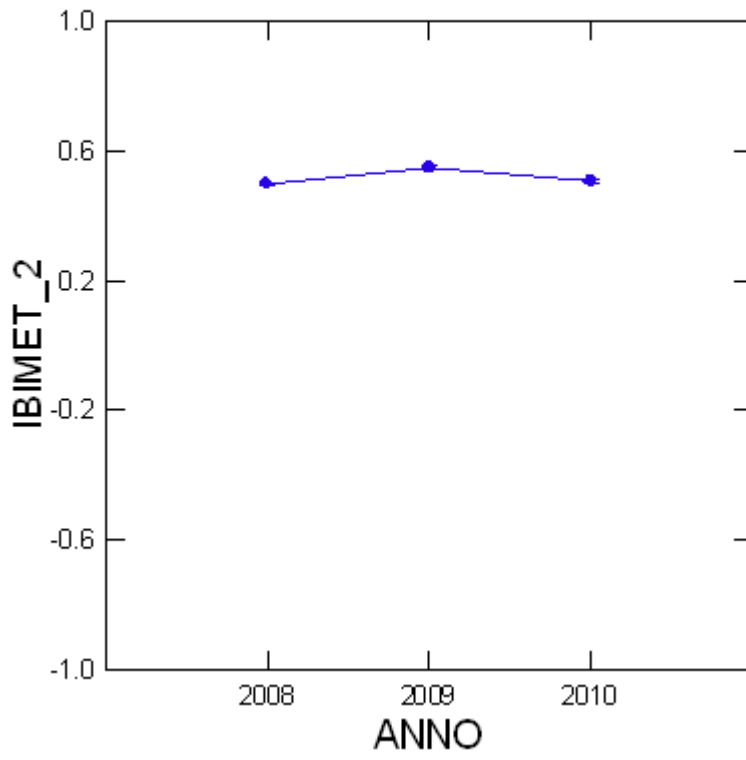
Variables	Levels			
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		
DIRAD\$ (2 levels)	D0	D1		

Dependent Variable	IBIMET_2
N	192
Multiple R	0.990
Squared Multiple R	0.979

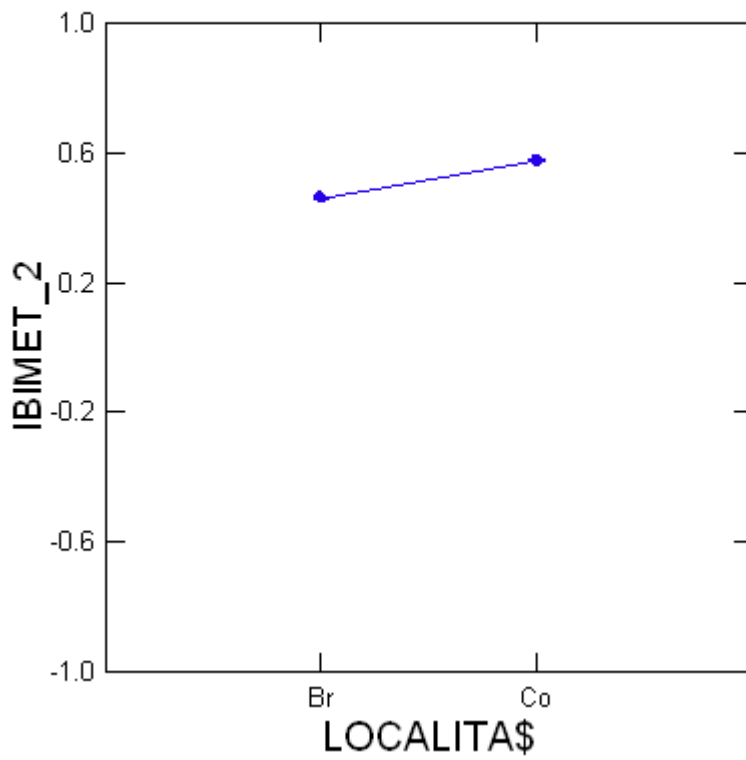
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.057	2	0.028	93.758	0.000
LOCALITA\$	0.492	1	0.492	1,624.094	0.000
N_GEMME\$	0.001	1	0.001	3.974	0.048
DEFOGL\$	0.000	1	0.000	0.494	0.483
DIRAD\$	0.001	1	0.001	2.208	0.139
LOCALITA\$*ANNO	0.397	2	0.199	656.560	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.470	0.626
DEFOGL\$*ANNO	0.000	2	0.000	0.529	0.590
DIRAD\$*ANNO	0.000	2	0.000	0.114	0.892
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.322	0.571
DEFOGL\$*LOCALITA\$	0.001	1	0.001	1.947	0.165
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.108	0.743
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.017	0.897
DIRAD\$*N_GEMME\$	0.001	1	0.001	1.976	0.162
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.655	0.420
BLOCCO(LOCALITA\$)	0.776	6	0.129	427.216	0.000
BLOCCO*ANNO(LOCALITA\$)	0.023	12	0.002	6.287	0.000
Error	0.046	153	0.000		

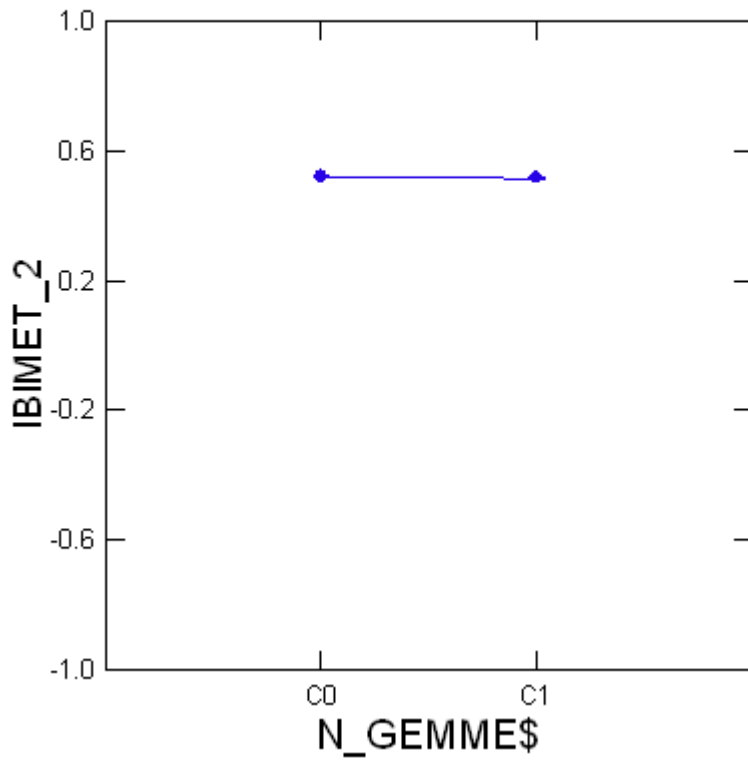
Least Squares Means



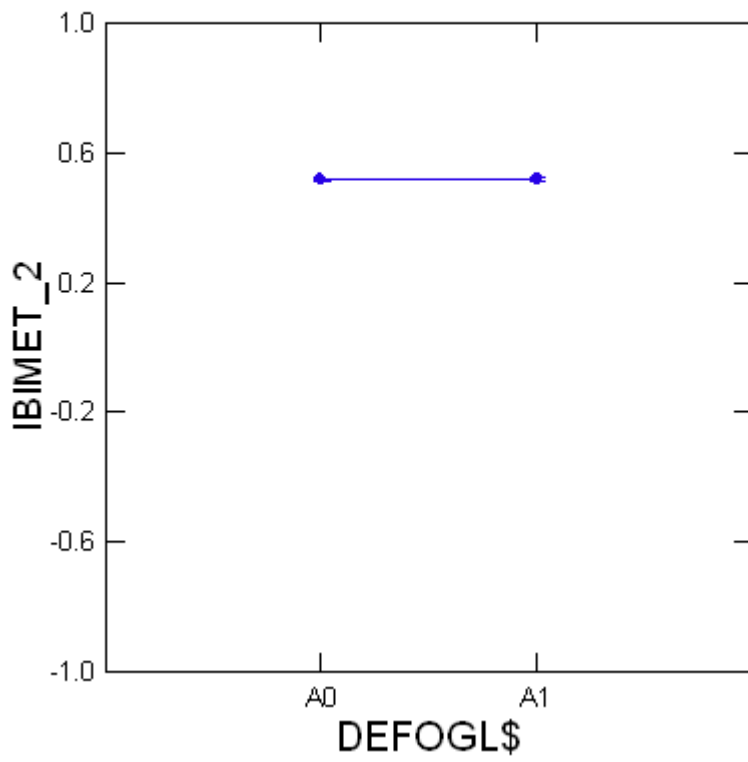
Least Squares Means



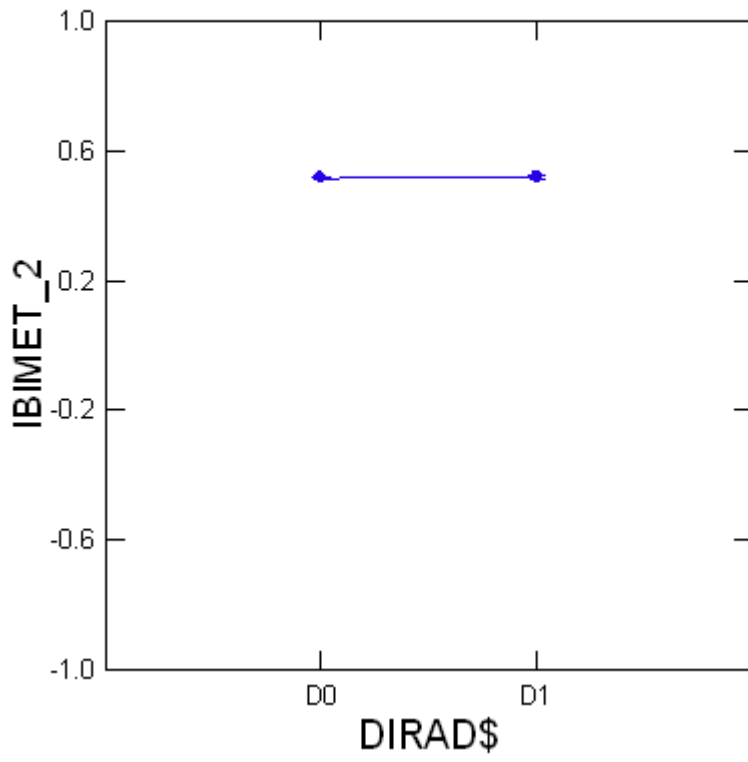
Least Squares Means



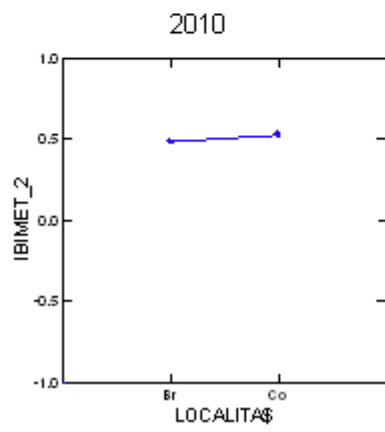
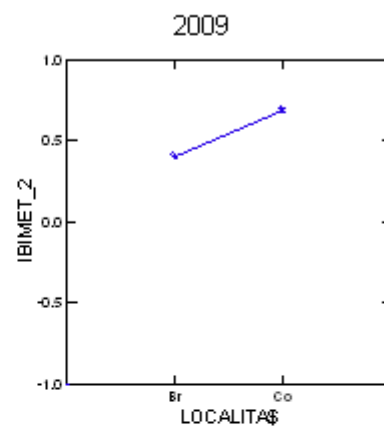
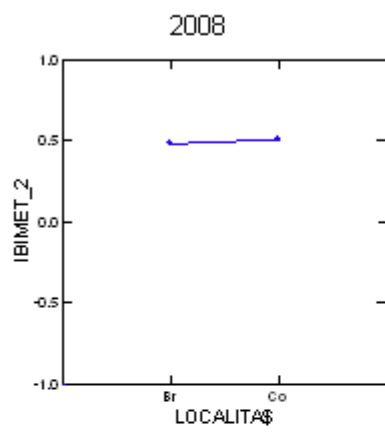
Least Squares Means



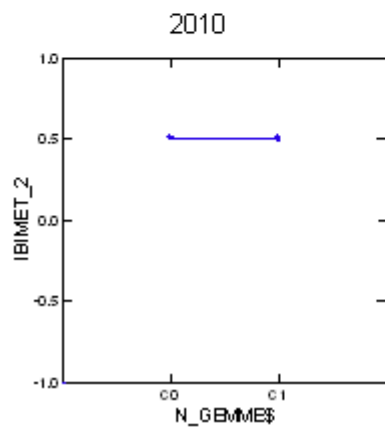
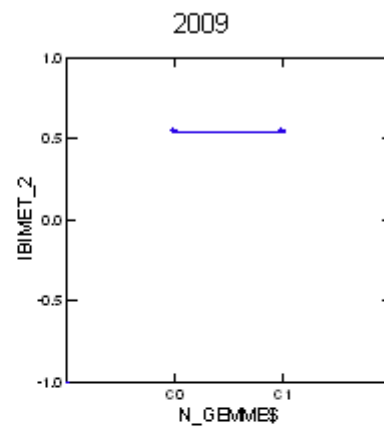
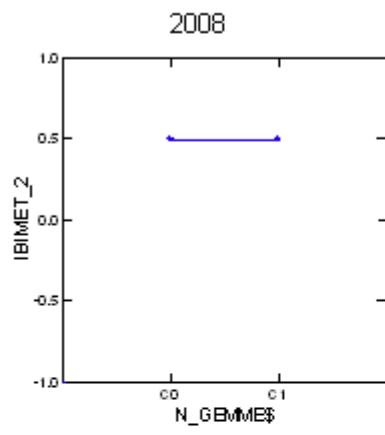
Least Squares Means



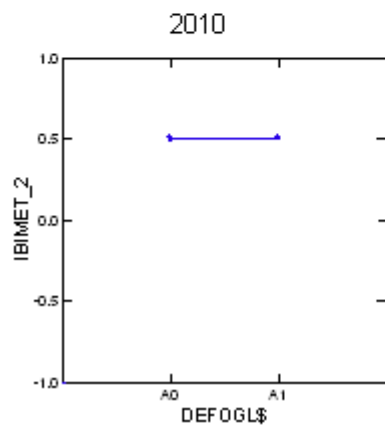
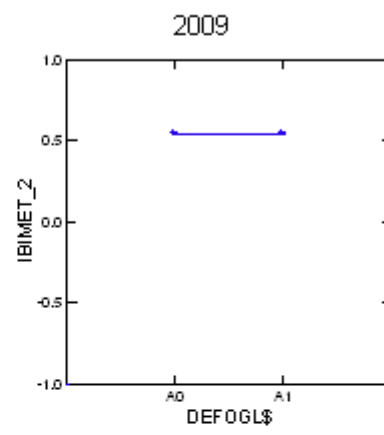
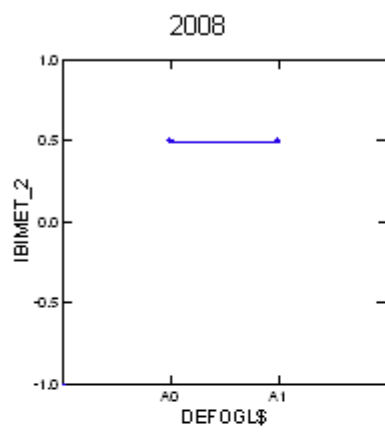
Least Squares Means



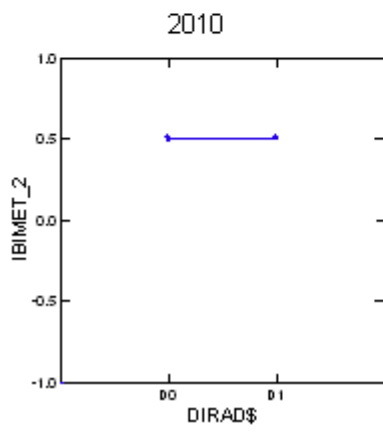
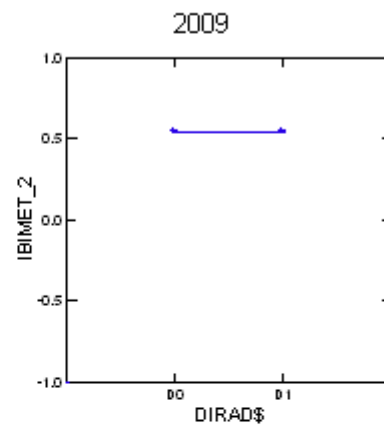
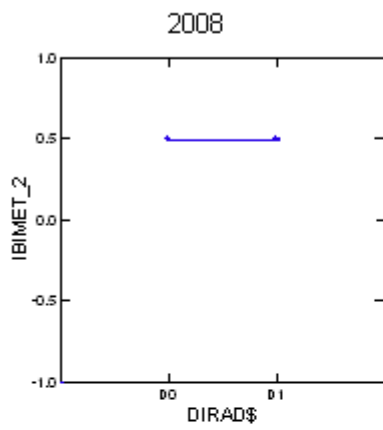
Least Squares Means



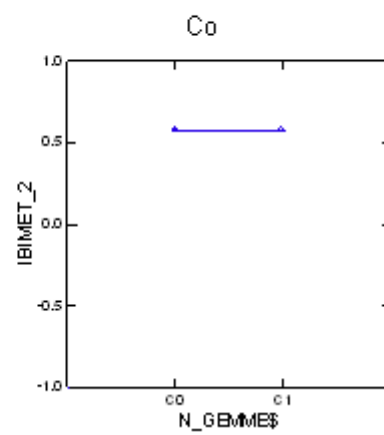
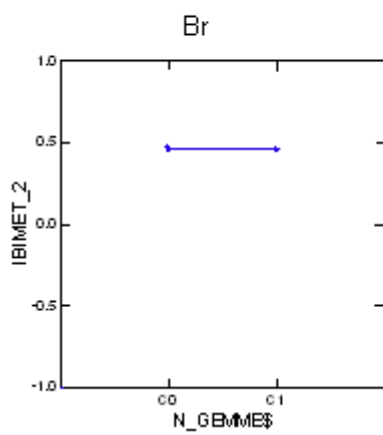
Least Squares Means



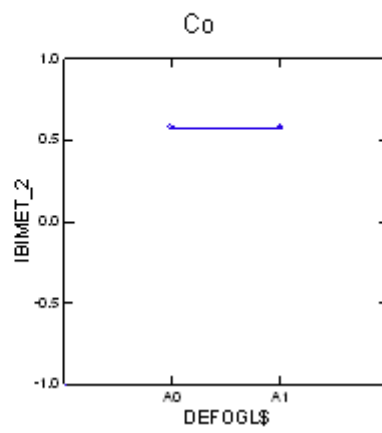
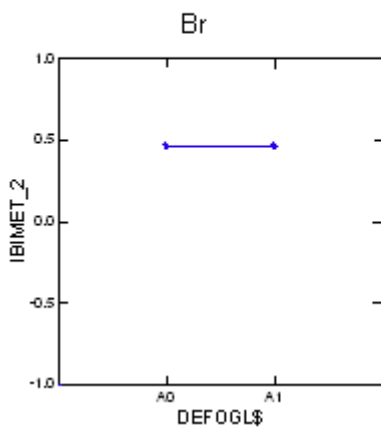
Least Squares Means



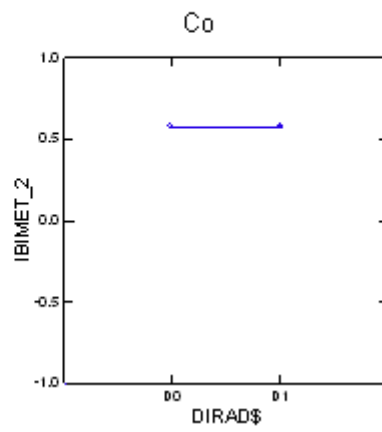
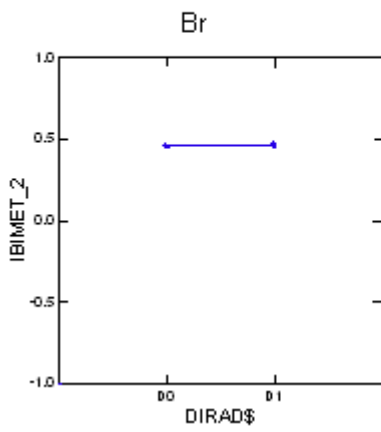
Least Squares Means



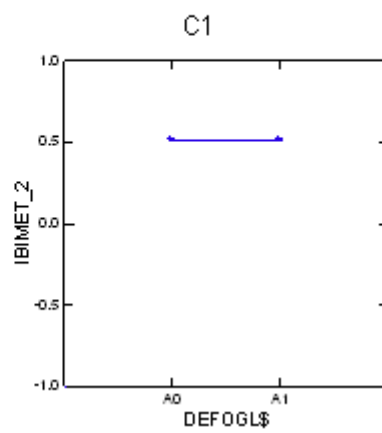
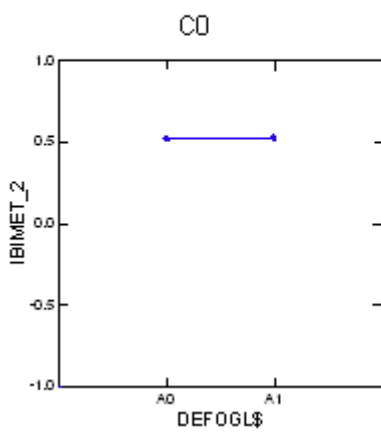
Least Squares Means



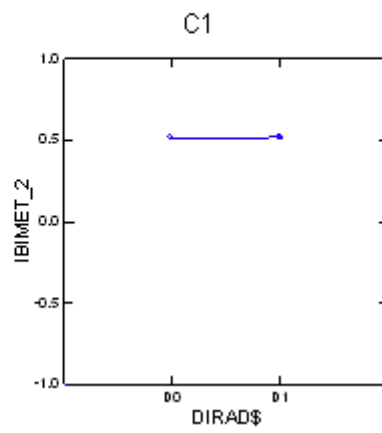
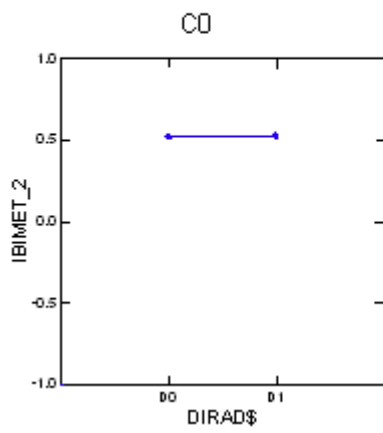
Least Squares Means



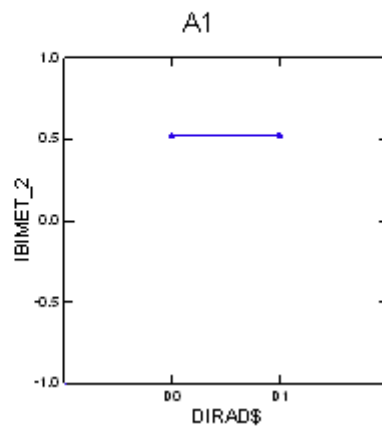
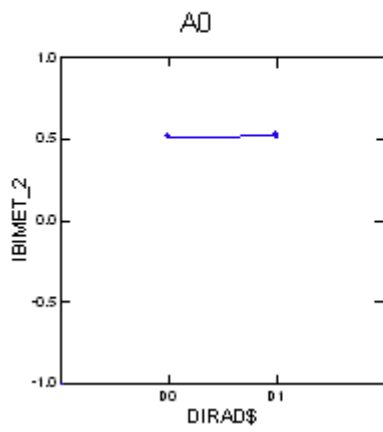
Least Squares Means



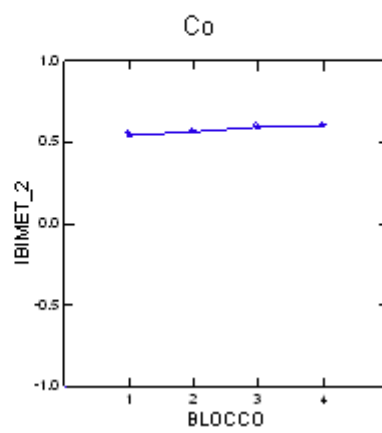
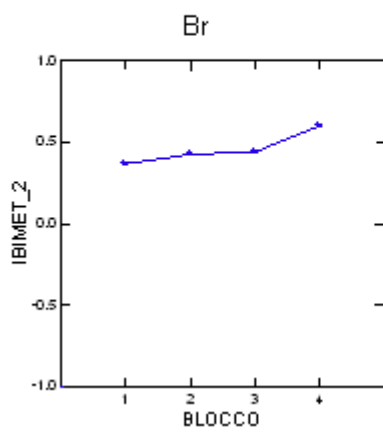
Least Squares Means



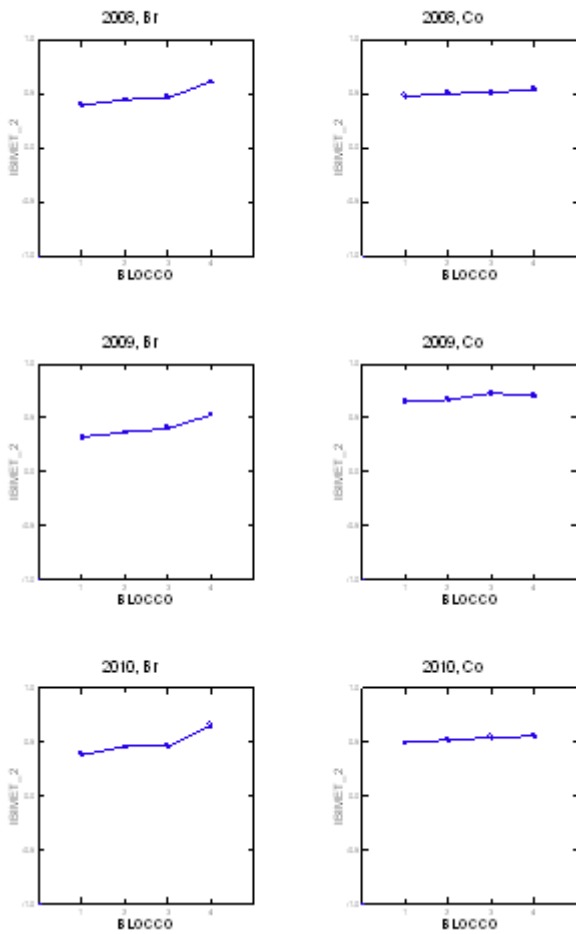
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.404
 First Order Autocorrelation | 0.285

Information Criteria

AIC | -974.498
 AIC (Corrected) | -952.776
 Schwarz's BIC | -844.198

▼ General Linear Model

IBIMET3 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels			
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		

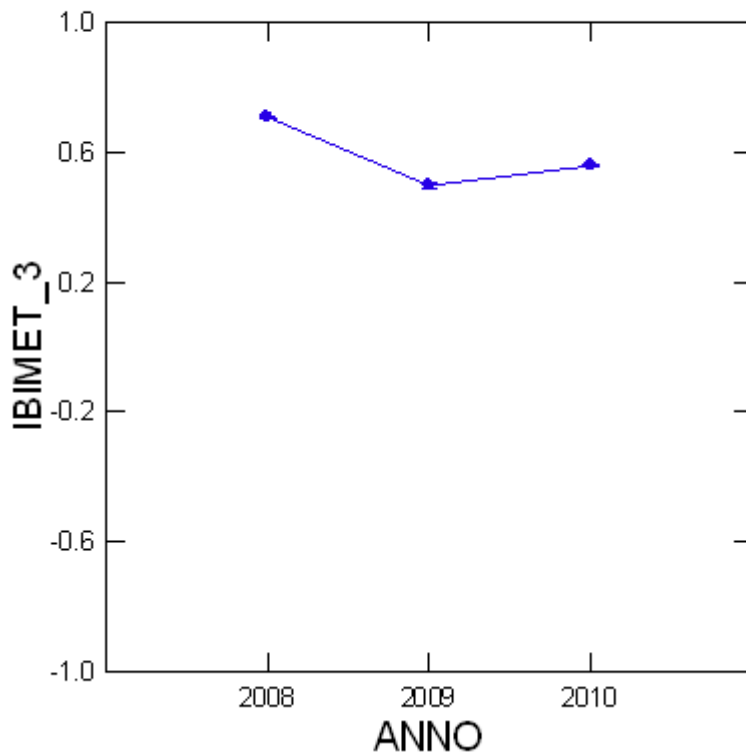
DEFOGL\$ (2 levels)	A0	A1
DIRAD\$ (2 levels)	D0	D1

Dependent Variable	IBIMET_3
N	192
Multiple R	0.996
Squared Multiple R	0.991

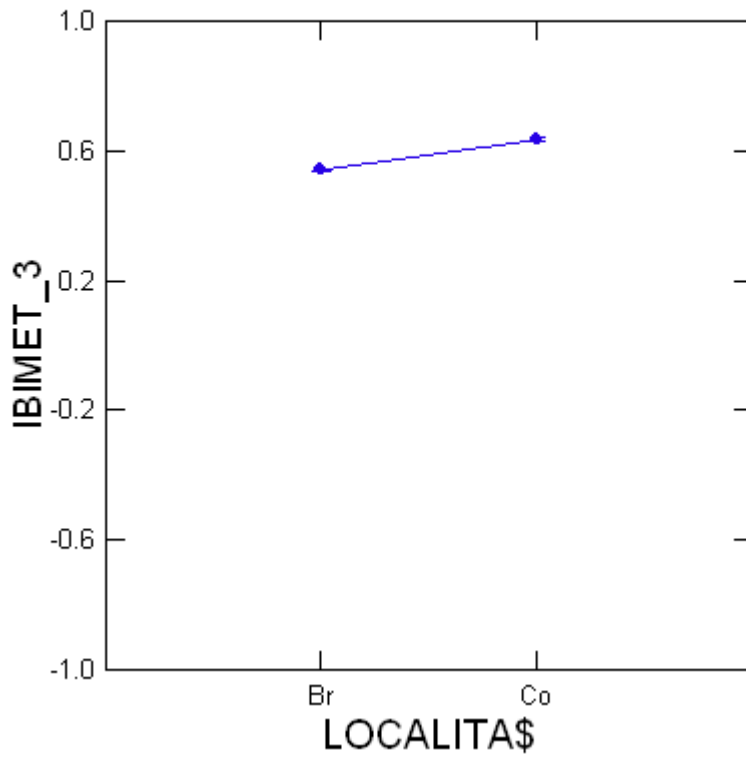
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	1.192	2	0.596	1,551.528	0.000
LOCALITA\$	0.311	1	0.311	809.963	0.000
N_GEMME\$	0.002	1	0.002	5.931	0.016
DEFOGL\$	0.001	1	0.001	2.137	0.146
DIRAD\$	0.002	1	0.002	4.476	0.036
LOCALITA\$*ANNO	3.624	2	1.812	4,715.609	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.614	0.542
DEFOGL\$*ANNO	0.001	2	0.000	1.174	0.312
DIRAD\$*ANNO	0.001	2	0.000	0.795	0.453
N_GEMME\$*LOCALITA\$	0.001	1	0.001	2.875	0.092
DEFOGL\$*LOCALITA\$	0.001	1	0.001	2.883	0.092
DIRAD\$*LOCALITA\$	0.001	1	0.001	2.583	0.110
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.215	0.644
DIRAD\$*N_GEMME\$	0.000	1	0.000	1.238	0.268
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.002	0.965
BLOCCO(LOCALITA\$)	0.726	6	0.121	314.995	0.000
BLOCCO*ANNO(LOCALITA\$)	0.056	12	0.005	12.057	0.000
Error	0.059	153	0.000		

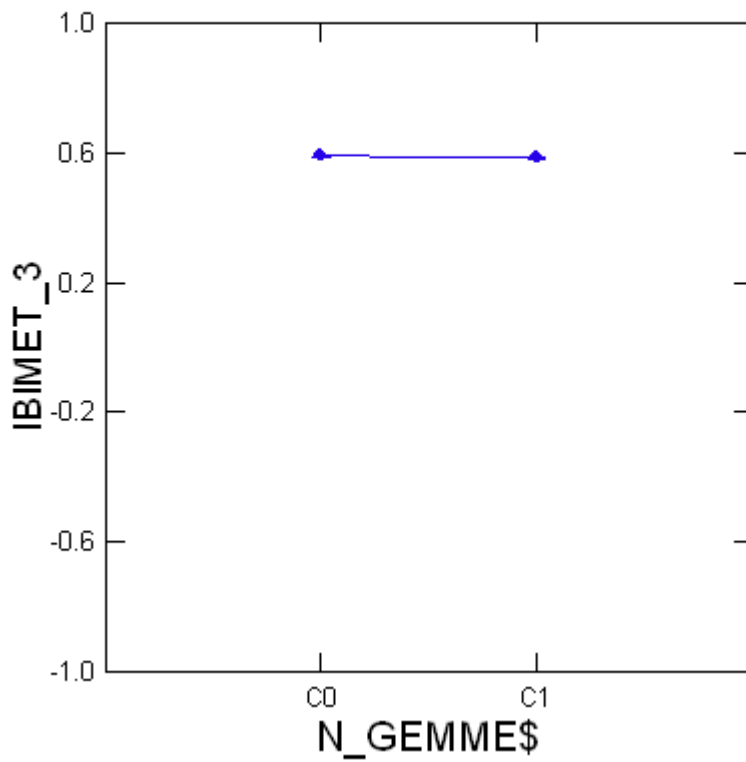
Least Squares Means



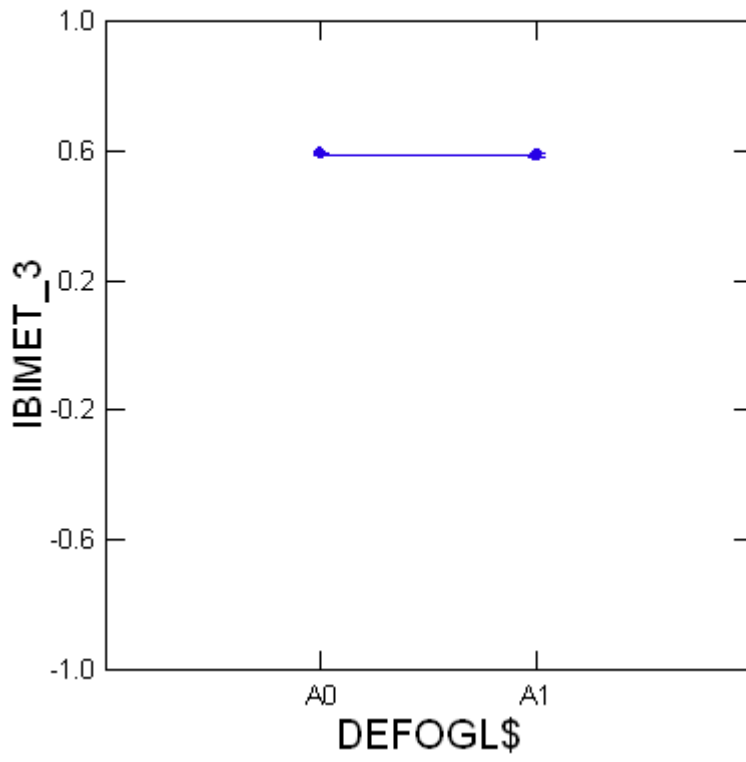
Least Squares Means



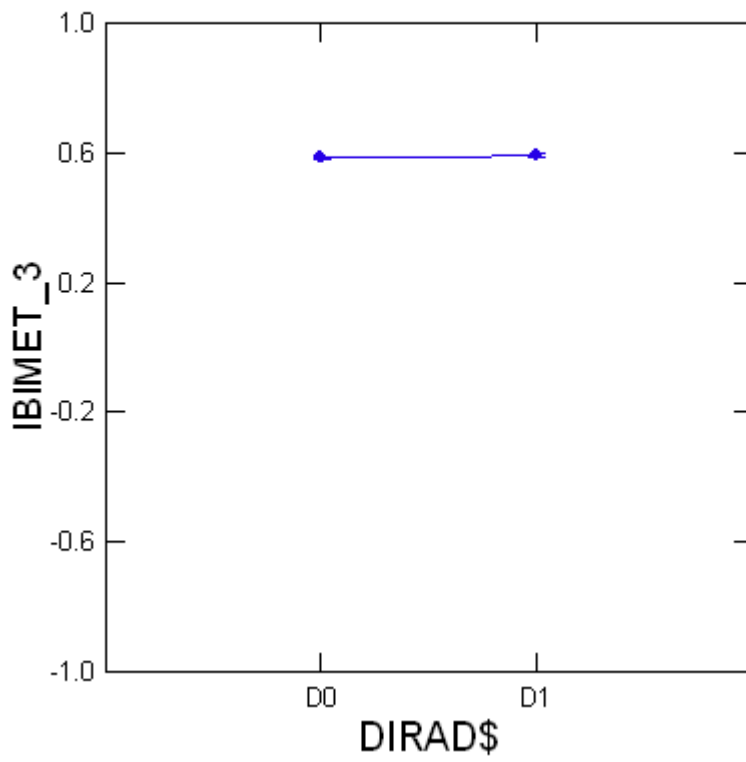
Least Squares Means



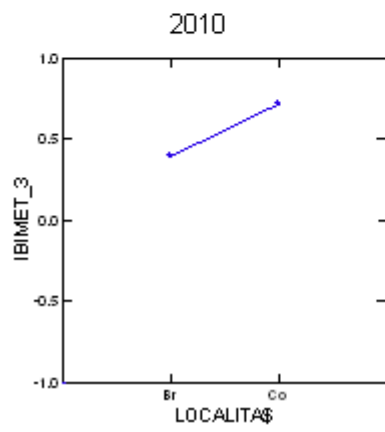
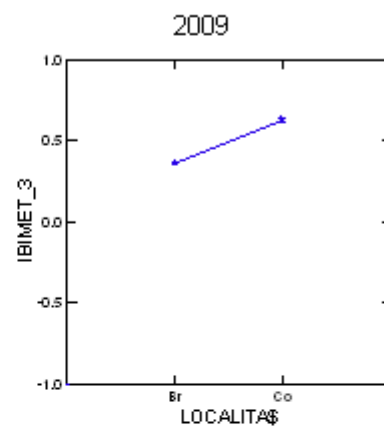
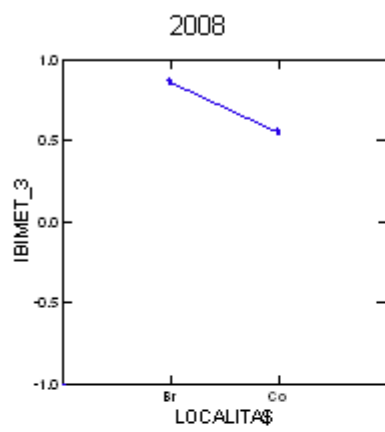
Least Squares Means



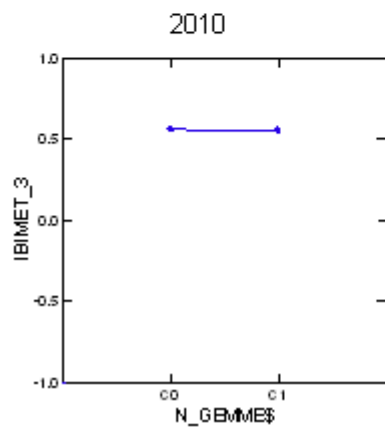
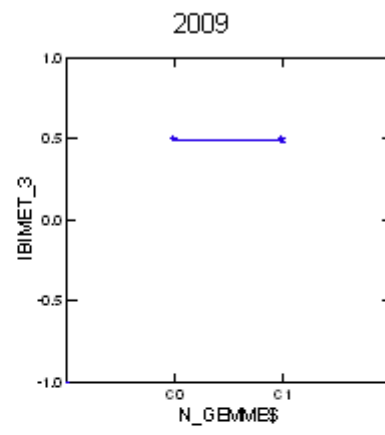
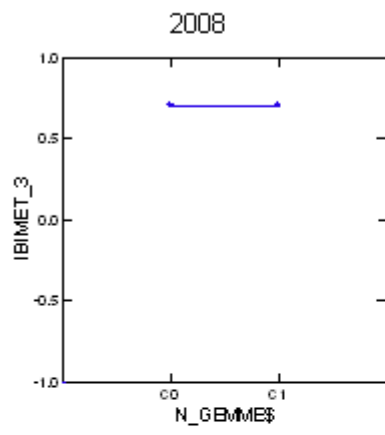
Least Squares Means



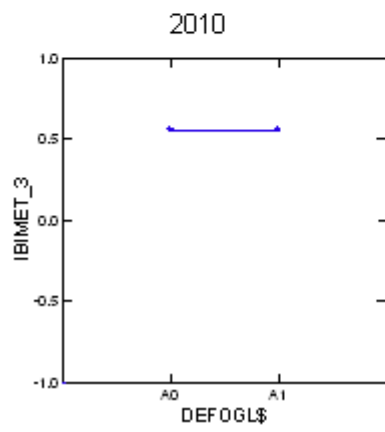
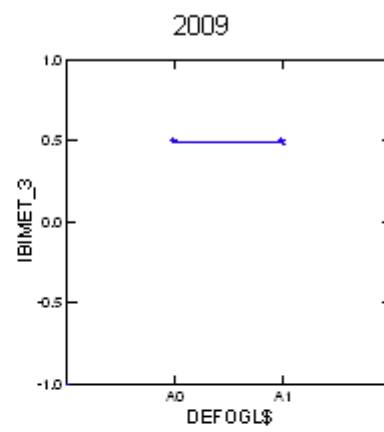
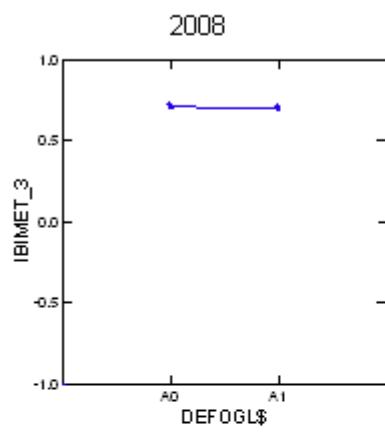
Least Squares Means



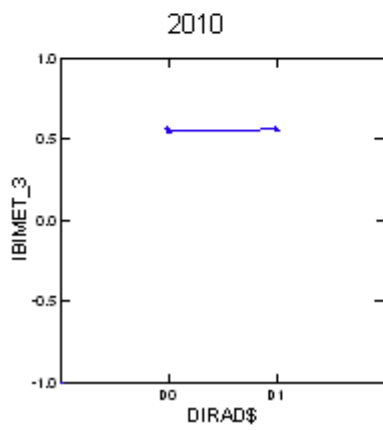
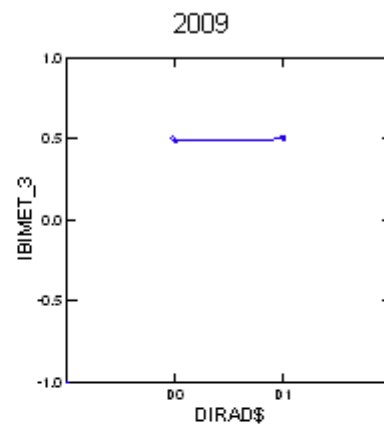
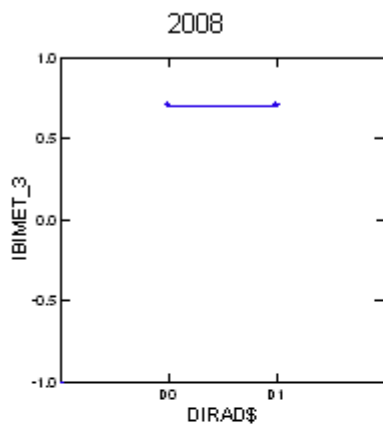
Least Squares Means



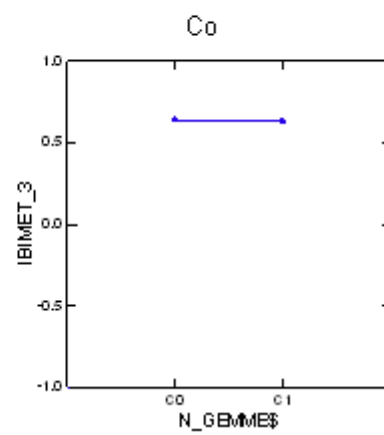
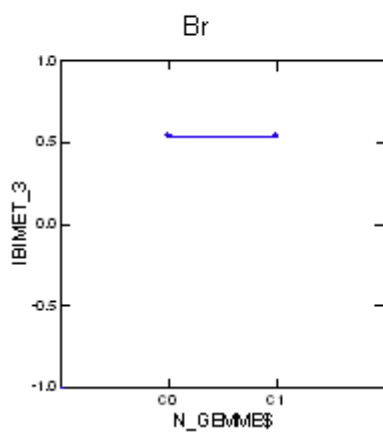
Least Squares Means



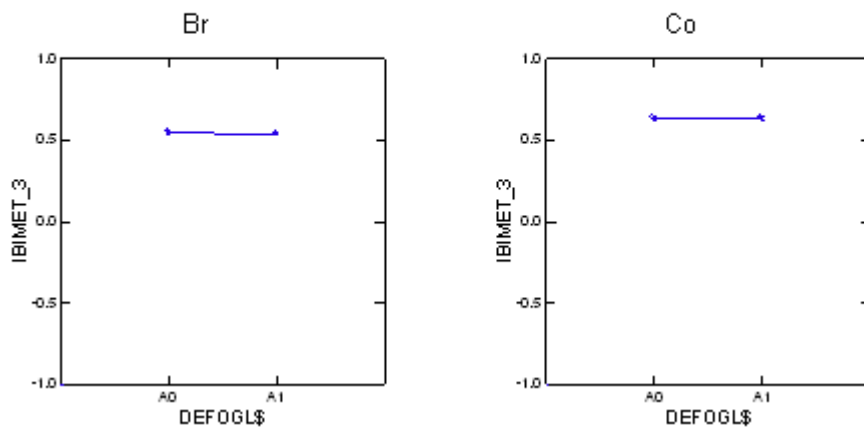
Least Squares Means



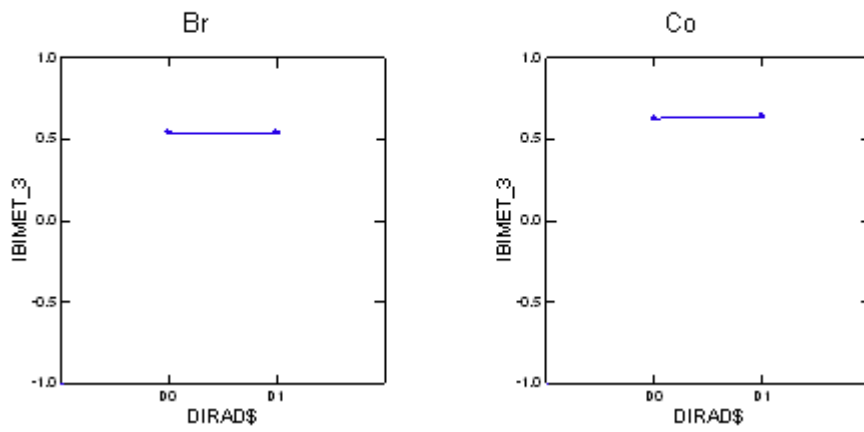
Least Squares Means



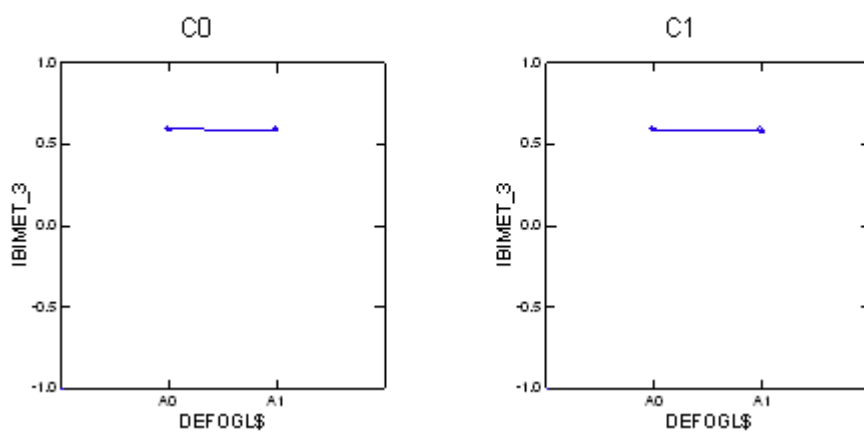
Least Squares Means



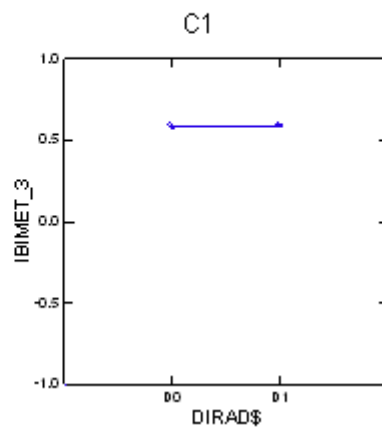
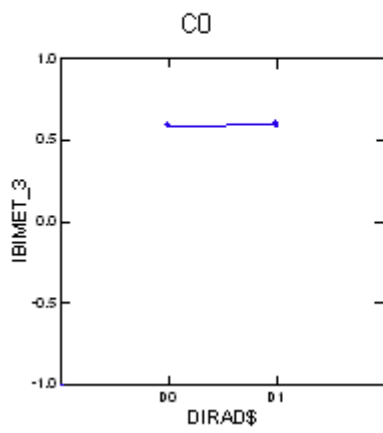
Least Squares Means



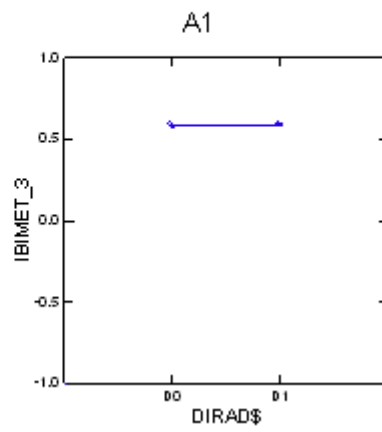
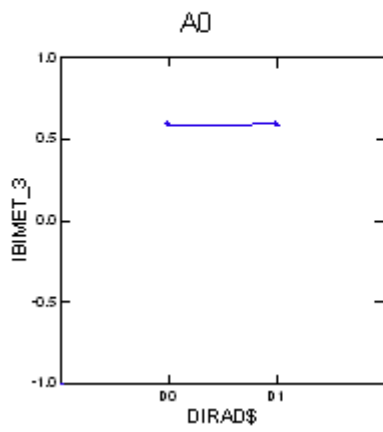
Least Squares Means



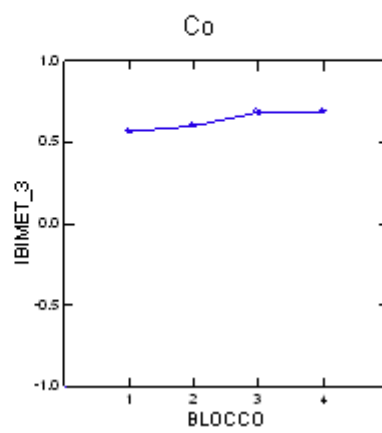
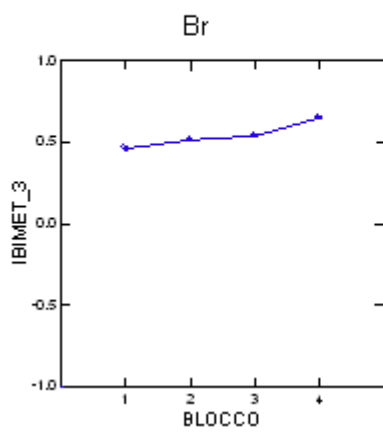
Least Squares Means



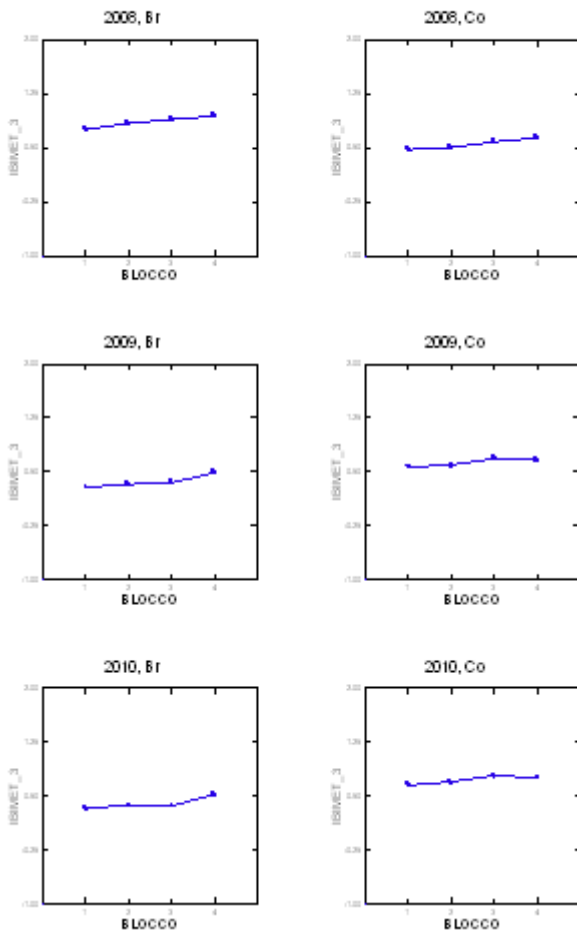
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.586
 First Order Autocorrelation | 0.163

Information Criteria

AIC | -928.634
 AIC (Corrected) | -906.912
 Schwarz's BIC | -798.334

▼ General Linear Model

IASMA1 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels			
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		

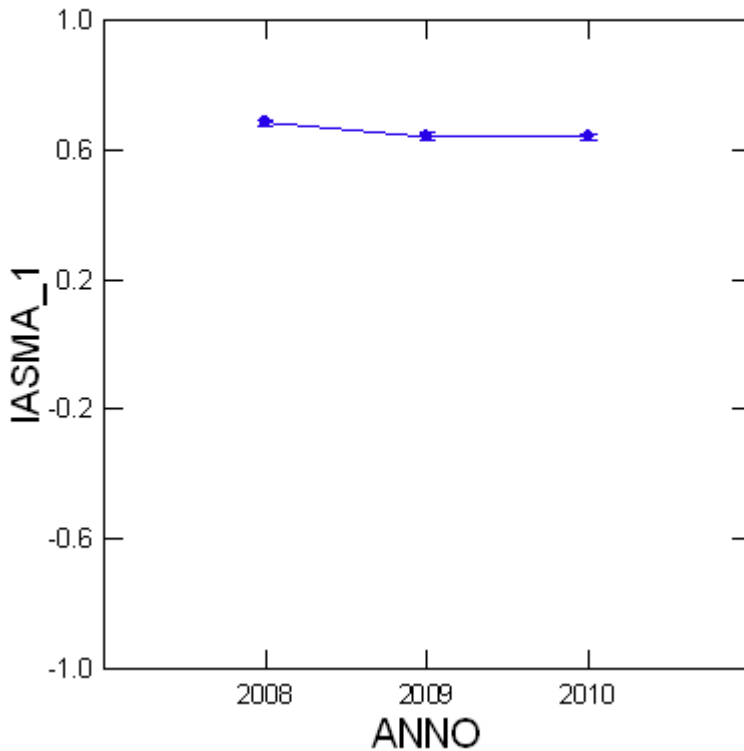
DIRAD\$ (2 levels) | D0 D1

Dependent Variable | IASMA_1
 N | 192
 Multiple R | 0.888
 Squared Multiple R | 0.789

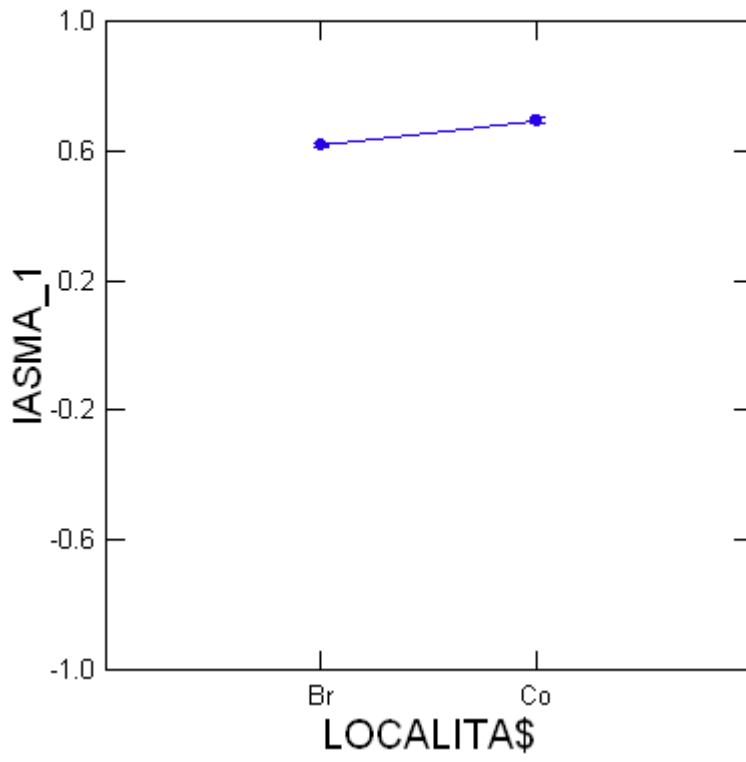
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.077	2	0.038	28.438	0.000
LOCALITA\$	0.202	1	0.202	150.110	0.000
N_GEMME\$	0.011	1	0.011	8.363	0.004
DEFOGL\$	0.018	1	0.018	13.551	0.000
DIRAD\$	0.003	1	0.003	2.401	0.123
LOCALITA\$*ANNO	0.046	2	0.023	17.182	0.000
N_GEMME\$*ANNO	0.014	2	0.007	5.319	0.006
DEFOGL\$*ANNO	0.001	2	0.000	0.366	0.694
DIRAD\$*ANNO	0.004	2	0.002	1.609	0.204
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.246	0.621
DEFOGL\$*LOCALITA\$	0.010	1	0.010	7.171	0.008
DIRAD\$*LOCALITA\$	0.003	1	0.003	1.903	0.170
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.113	0.737
DIRAD\$*N_GEMME\$	0.002	1	0.002	1.316	0.253
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.033	0.855
BLOCCO(LOCALITA\$)	0.109	6	0.018	13.552	0.000
BLOCCO*ANNO(LOCALITA\$)	0.165	12	0.014	10.198	0.000
Error	0.206	153	0.001		

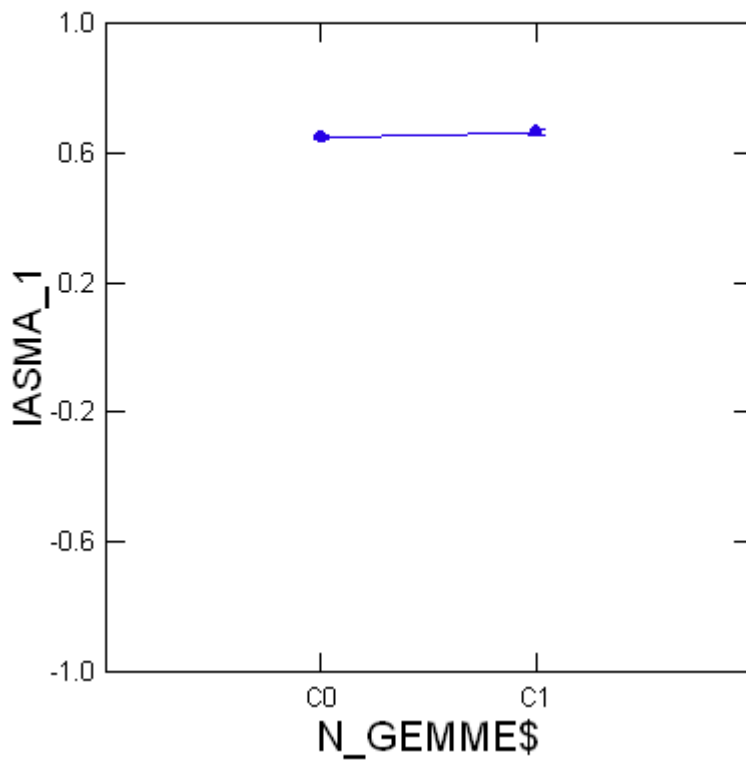
Least Squares Means



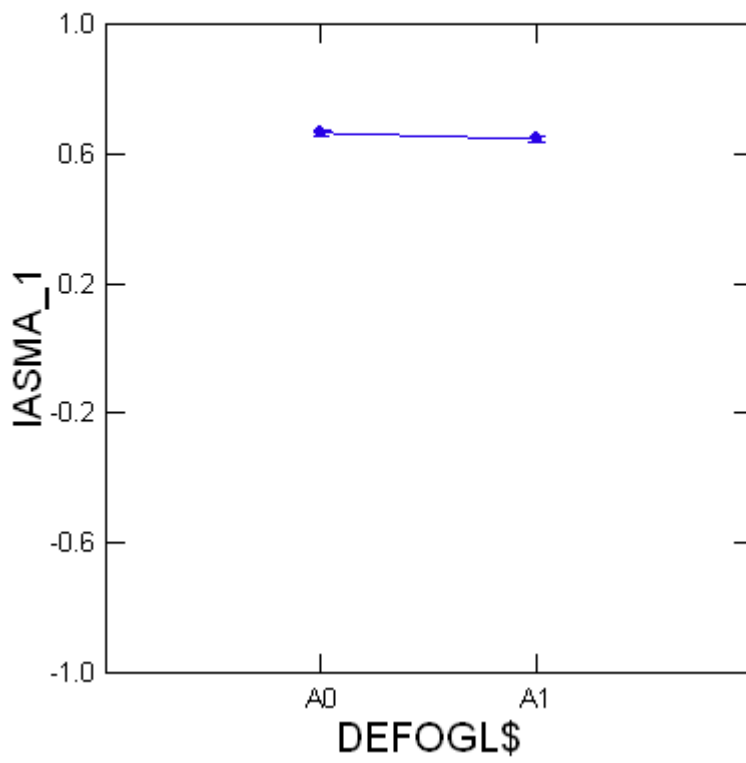
Least Squares Means



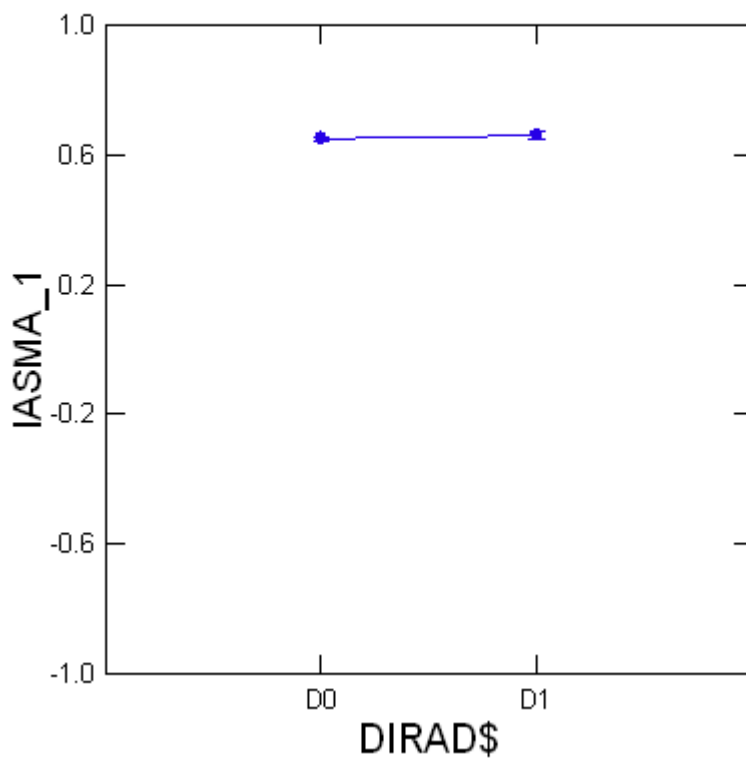
Least Squares Means



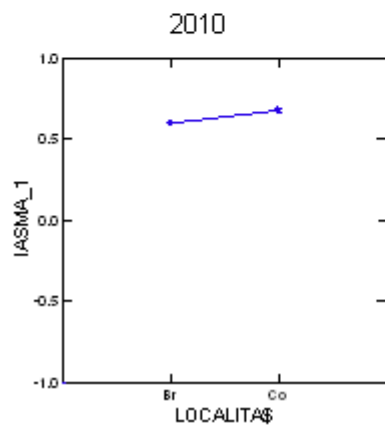
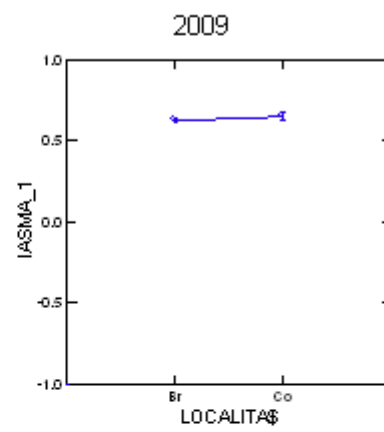
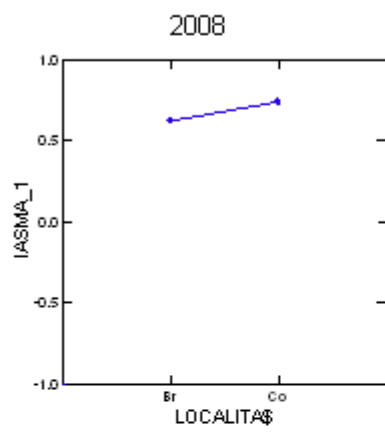
Least Squares Means



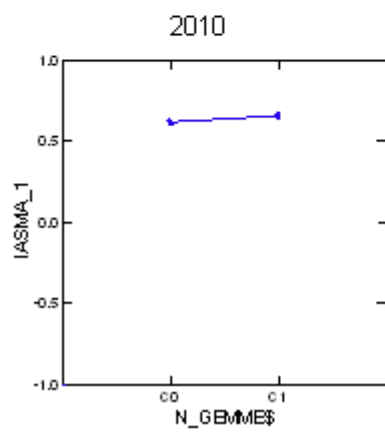
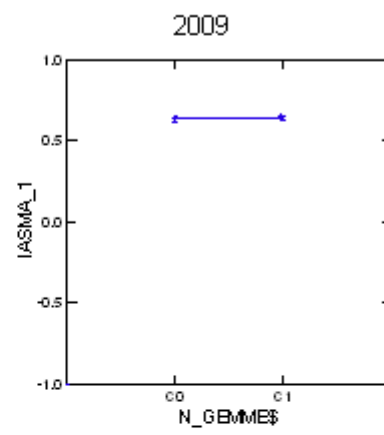
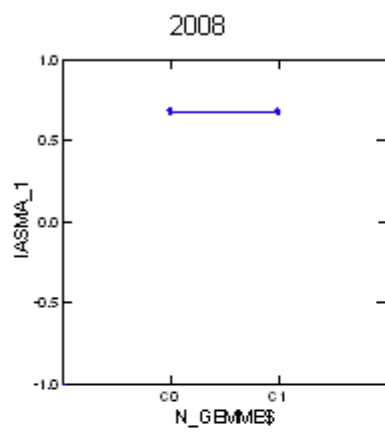
Least Squares Means



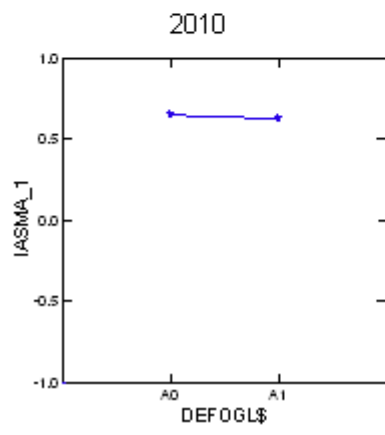
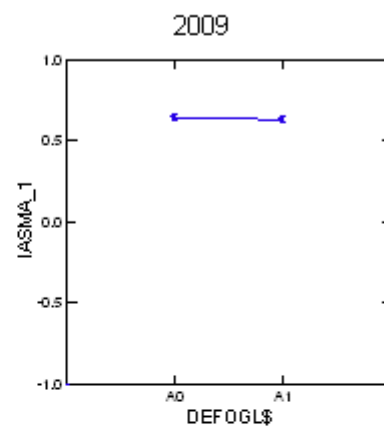
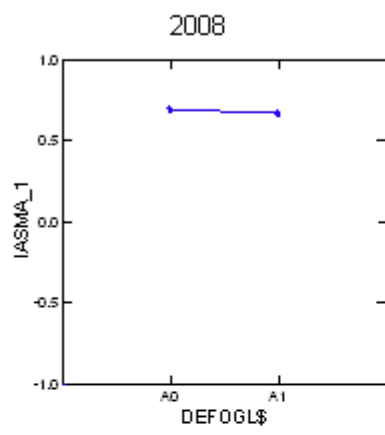
Least Squares Means



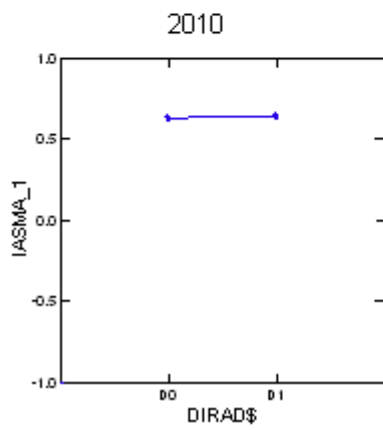
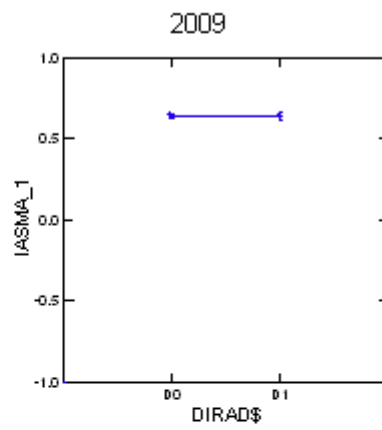
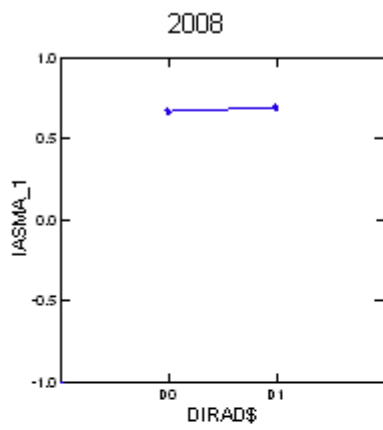
Least Squares Means



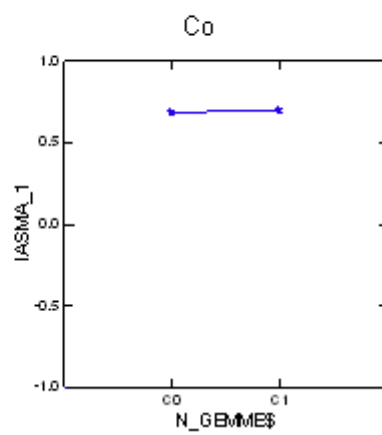
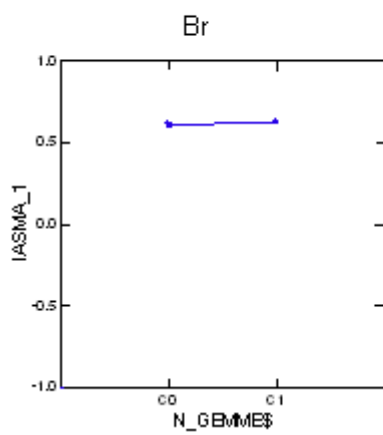
Least Squares Means



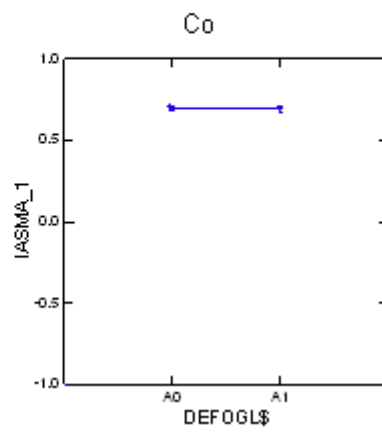
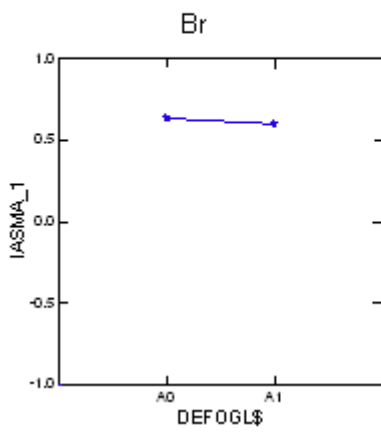
Least Squares Means



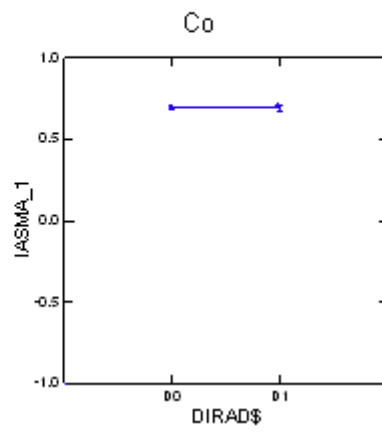
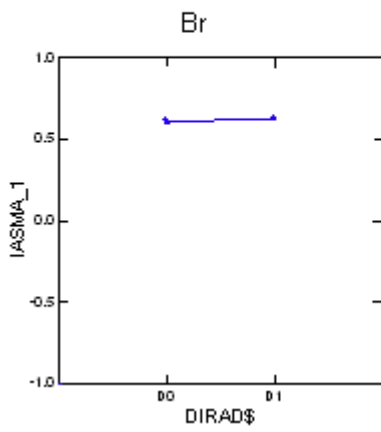
Least Squares Means



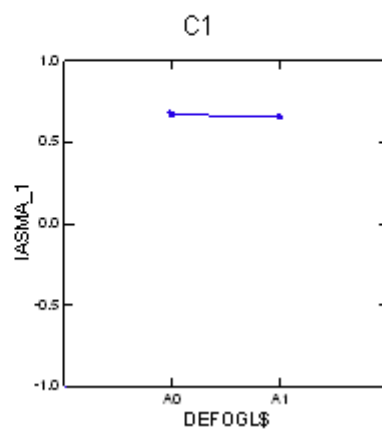
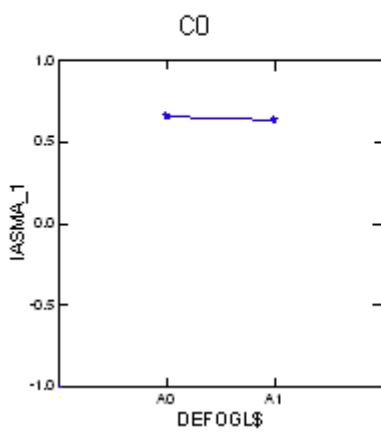
Least Squares Means



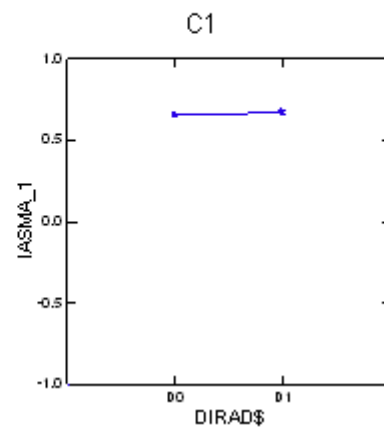
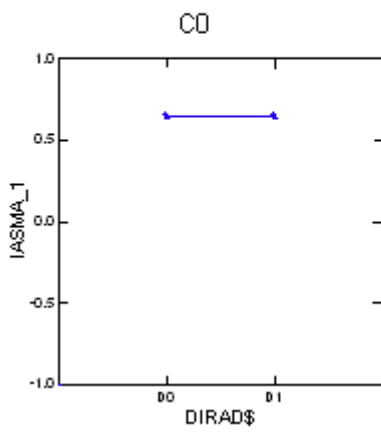
Least Squares Means



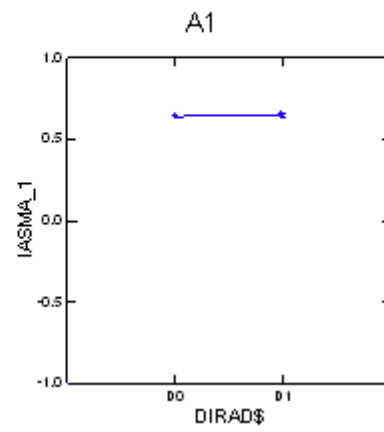
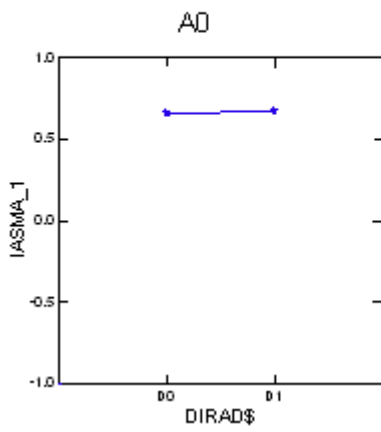
Least Squares Means



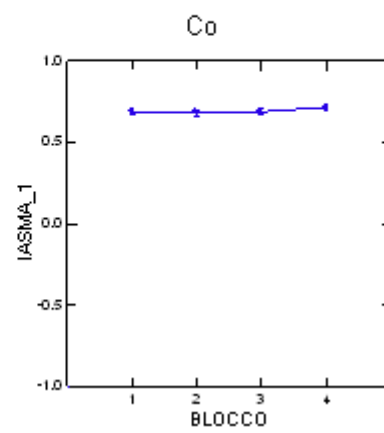
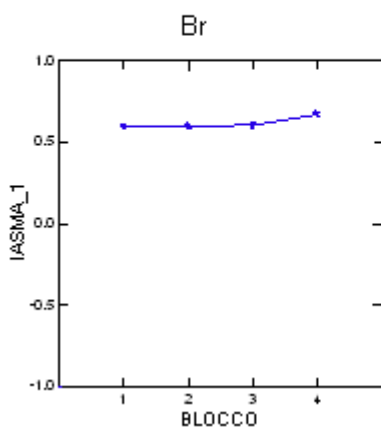
Least Squares Means



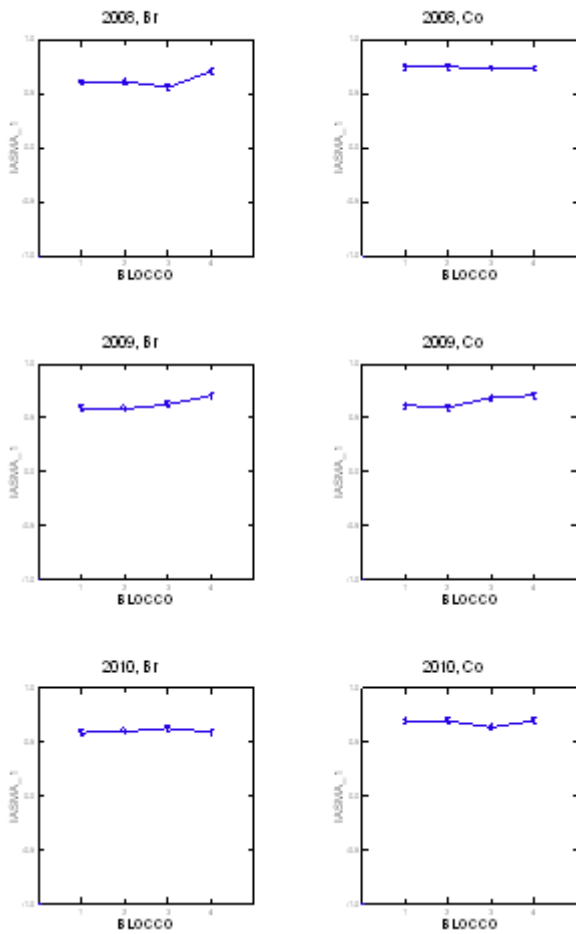
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 73 is an Outlier (Studentized Residual : -5.194)

Durbin-Watson D Statistic | 1.482
 First Order Autocorrelation | 0.257

Information Criteria

AIC | -687.964
 AIC (Corrected) | -666.242
 Schwarz's BIC | -557.664

▼ General Linear Model

IASMA2 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables | Levels

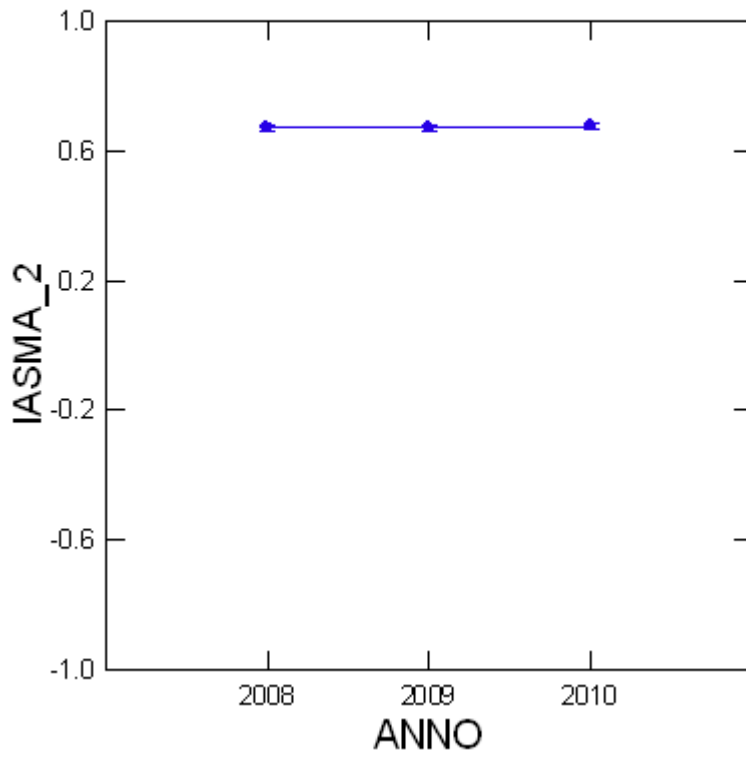
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		
DIRAD\$ (2 levels)	D0	D1		

Dependent Variable	IASMA_2
N	192
Multiple R	0.854
Squared Multiple R	0.730

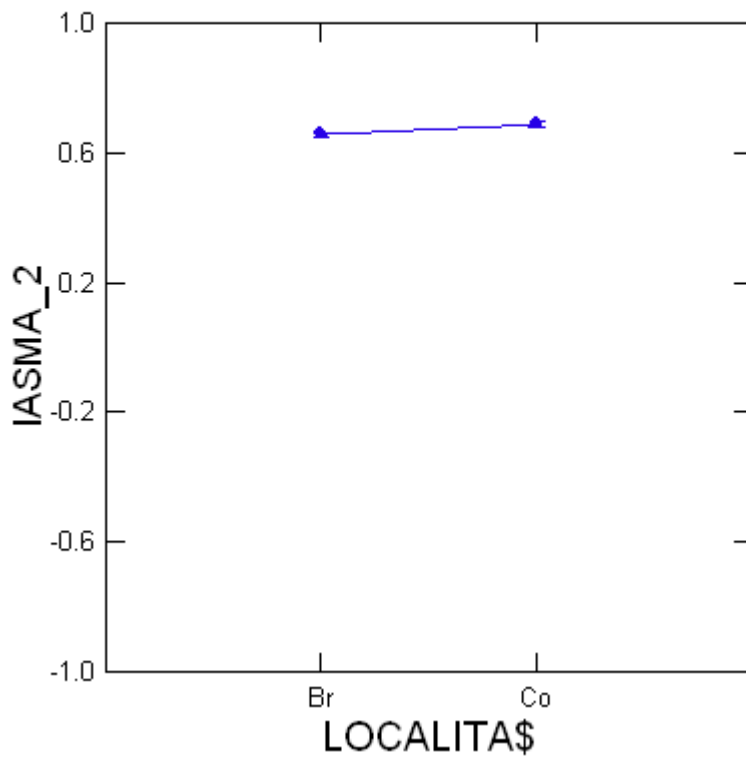
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.001	2	0.001	0.664	0.516
LOCALITA\$	0.034	1	0.034	38.618	0.000
N_GEMME\$	0.003	1	0.003	3.573	0.061
DEFOGL\$	0.000	1	0.000	0.368	0.545
DIRAD\$	0.003	1	0.003	3.222	0.075
LOCALITA\$*ANNO	0.045	2	0.023	26.060	0.000
N_GEMME\$*ANNO	0.002	2	0.001	1.258	0.287
DEFOGL\$*ANNO	0.003	2	0.002	1.880	0.156
DIRAD\$*ANNO	0.003	2	0.001	1.505	0.225
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.008	0.929
DEFOGL\$*LOCALITA\$	0.003	1	0.003	3.480	0.064
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.428	0.514
DEFOGL\$*N_GEMME\$	0.002	1	0.002	2.248	0.136
DIRAD\$*N_GEMME\$	0.001	1	0.001	0.698	0.405
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.191	0.663
BLOCCO(LOCALITA\$)	0.218	6	0.036	41.673	0.000
BLOCCO*ANNO(LOCALITA\$)	0.034	12	0.003	3.243	0.000
Error	0.134	153	0.001		

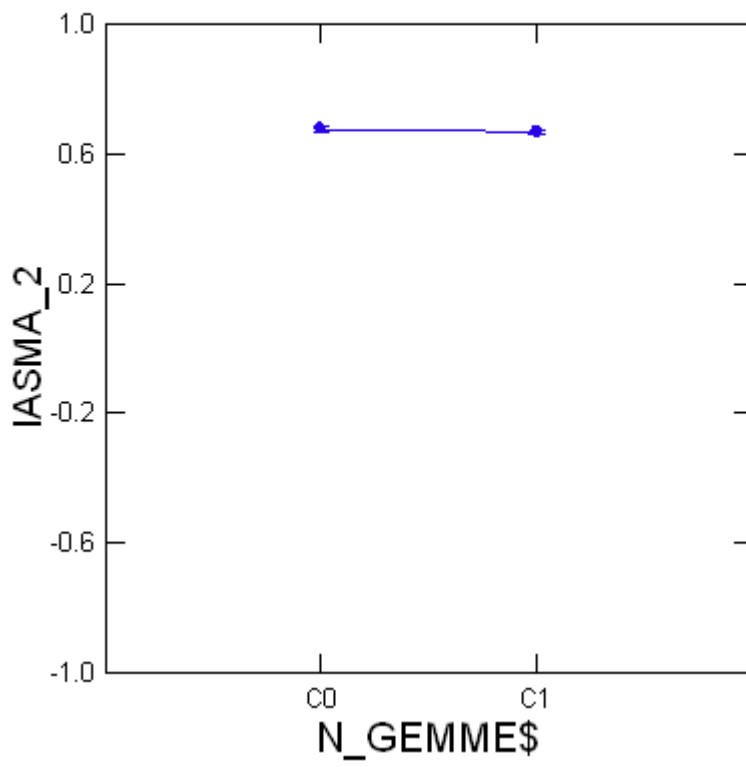
Least Squares Means



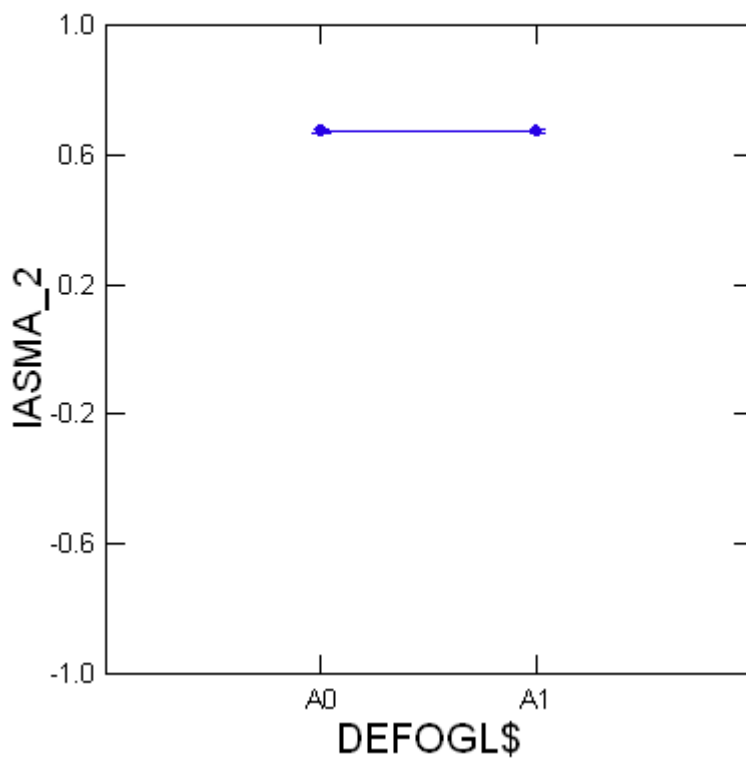
Least Squares Means



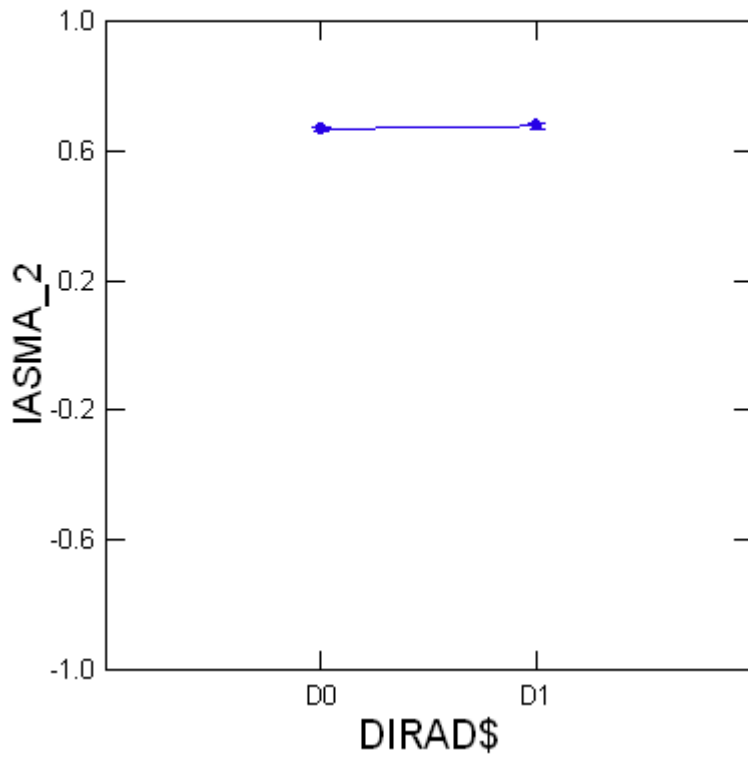
Least Squares Means



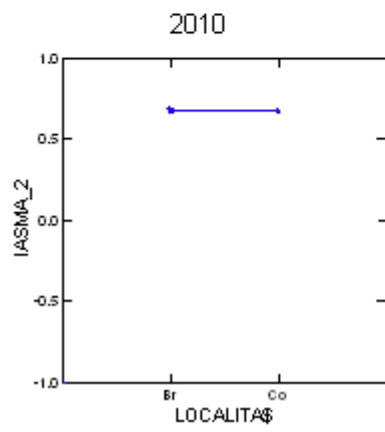
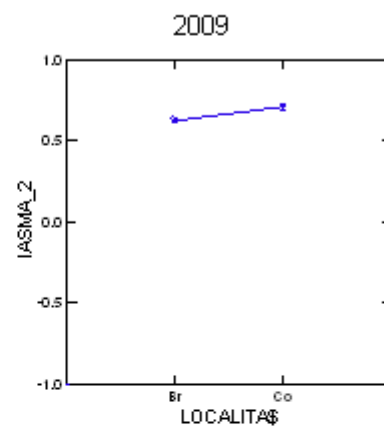
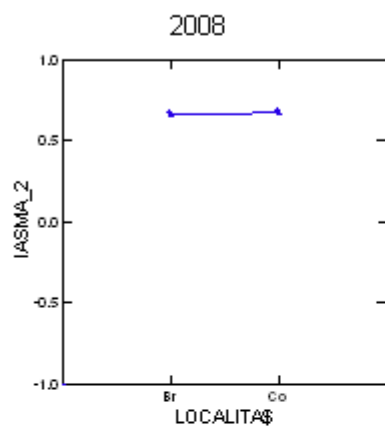
Least Squares Means



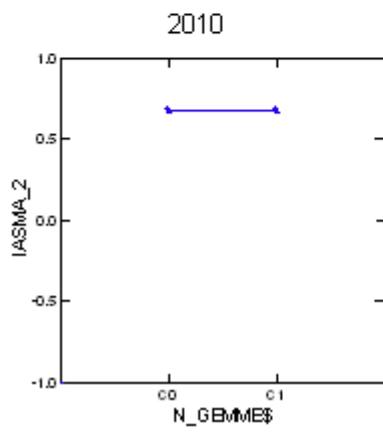
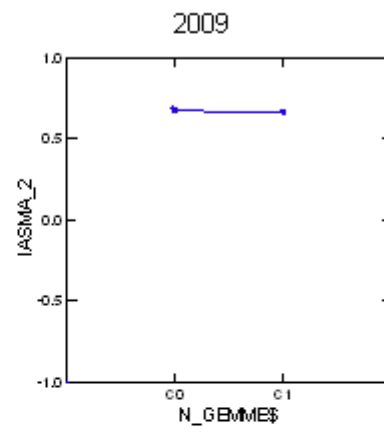
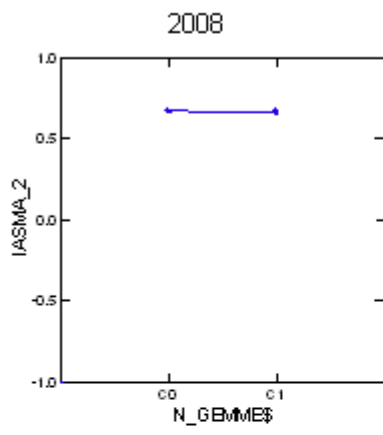
Least Squares Means



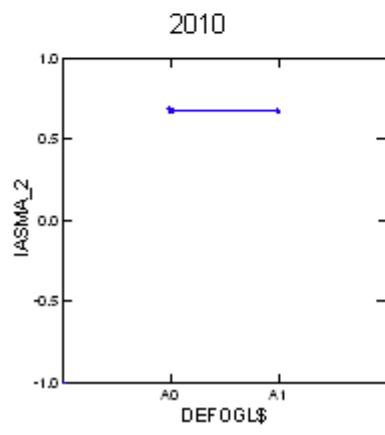
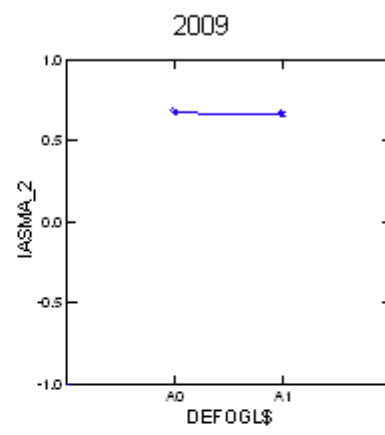
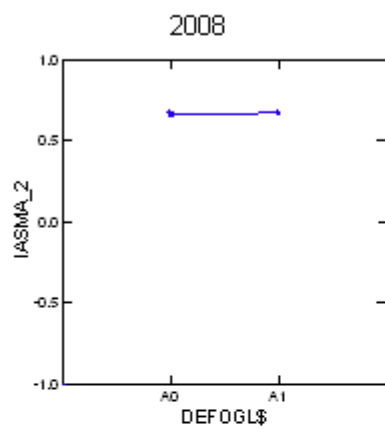
Least Squares Means



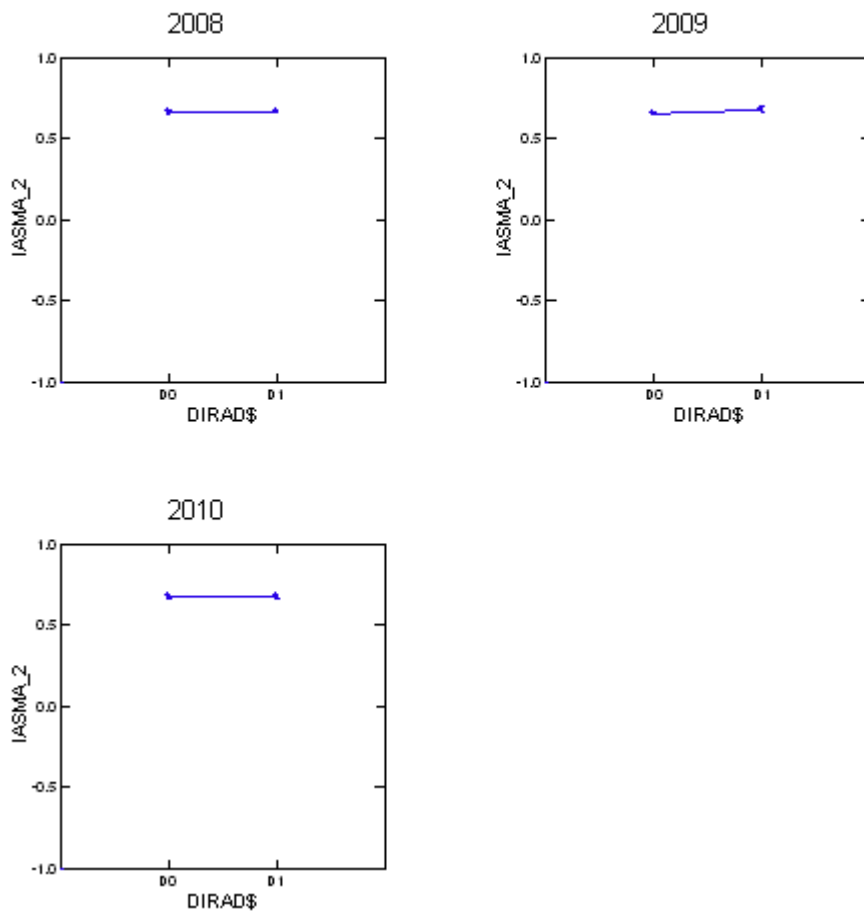
Least Squares Means



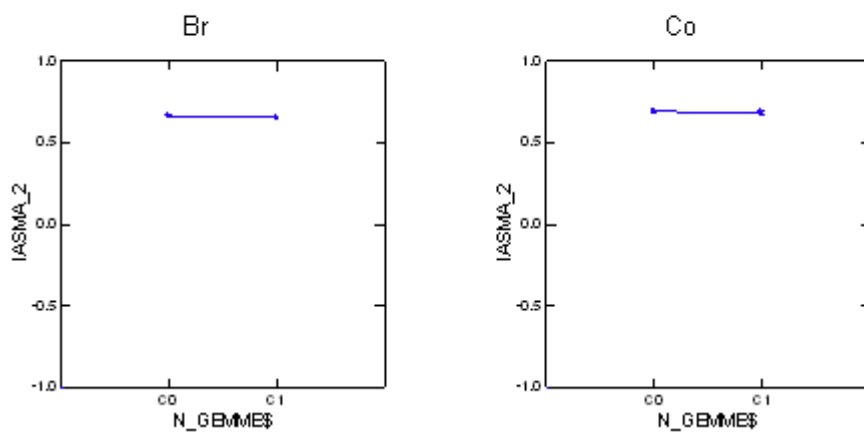
Least Squares Means



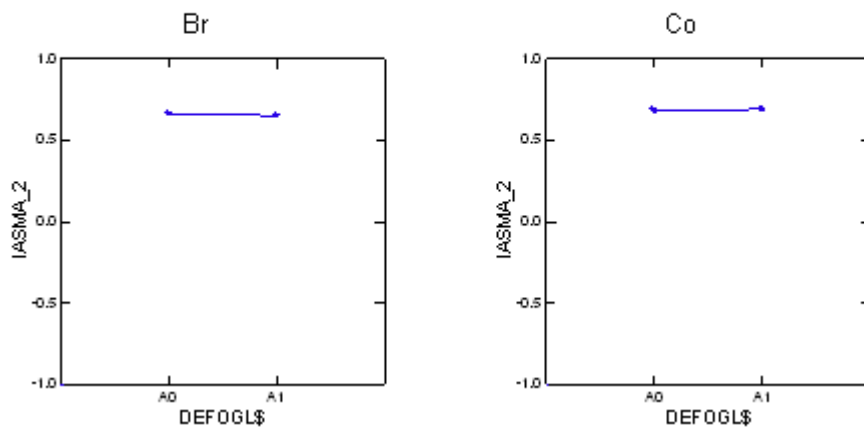
Least Squares Means



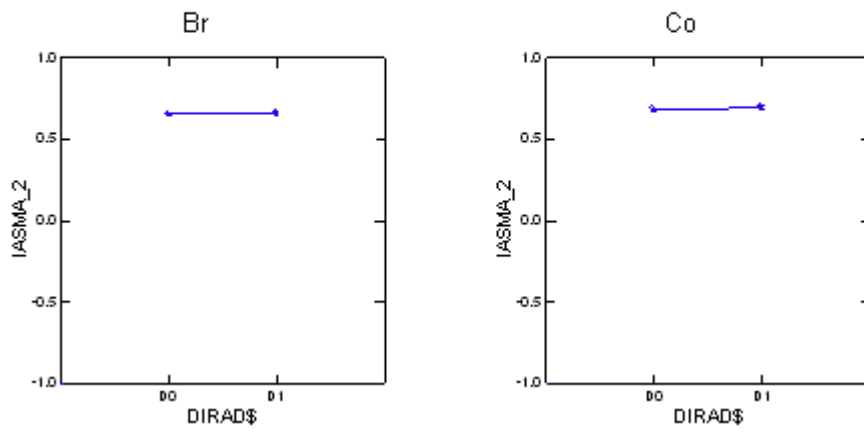
Least Squares Means



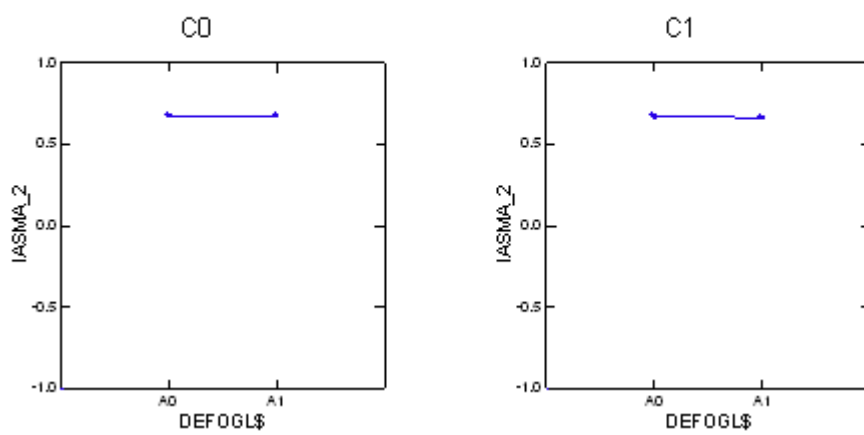
Least Squares Means



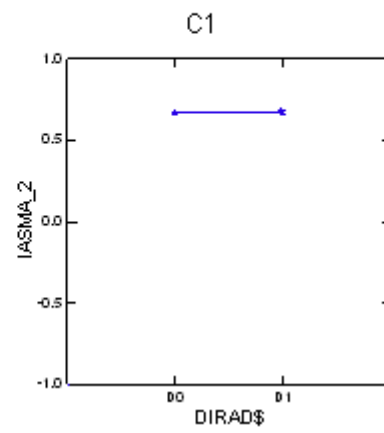
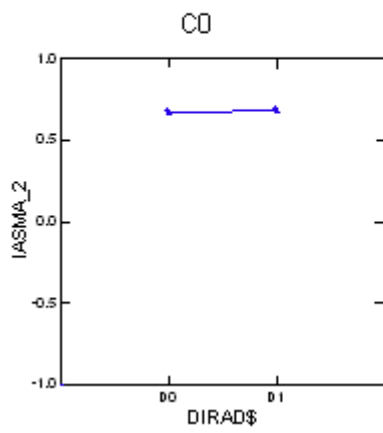
Least Squares Means



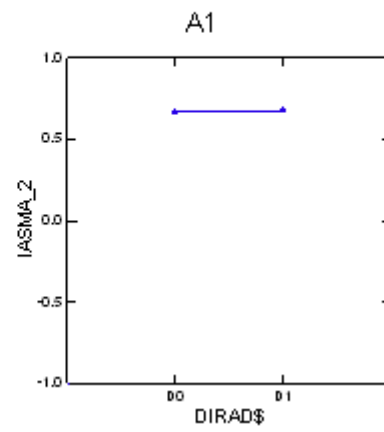
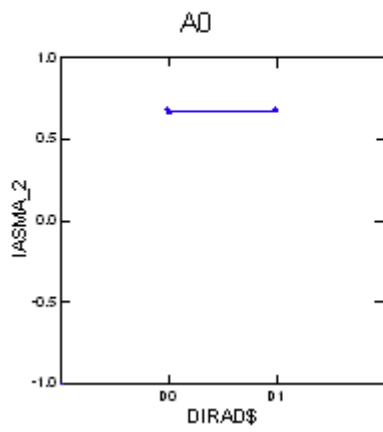
Least Squares Means



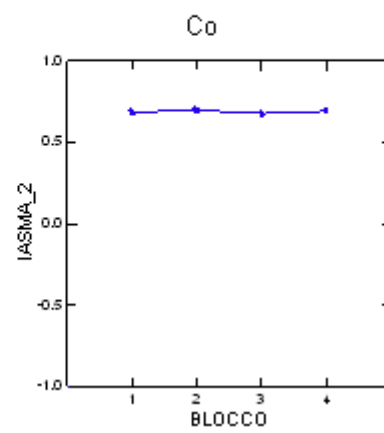
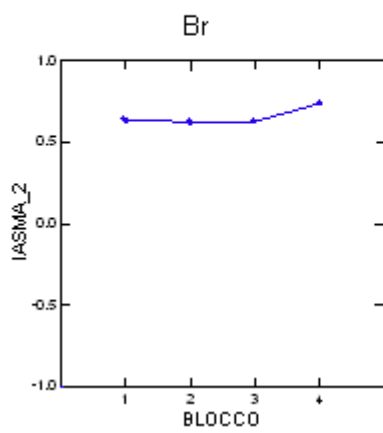
Least Squares Means



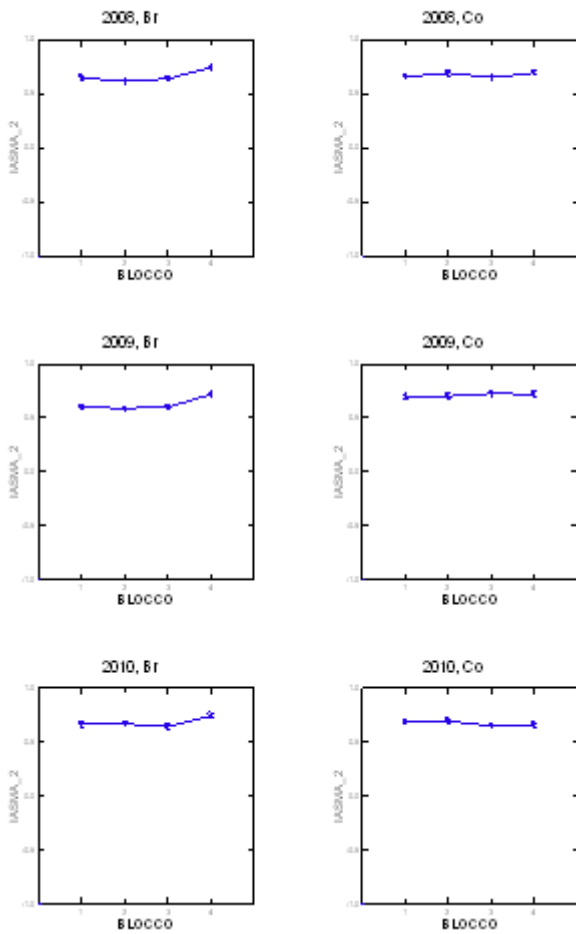
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 212 is an Outlier (Studentized Residual : -3.682)

Durbin-Watson D Statistic | 1.836
 First Order Autocorrelation | 0.081

Information Criteria

AIC | -771.131
 AIC (Corrected) | -749.410
 Schwarz's BIC | -640.832

▼ General Linear Model

IASMA3 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables		Levels
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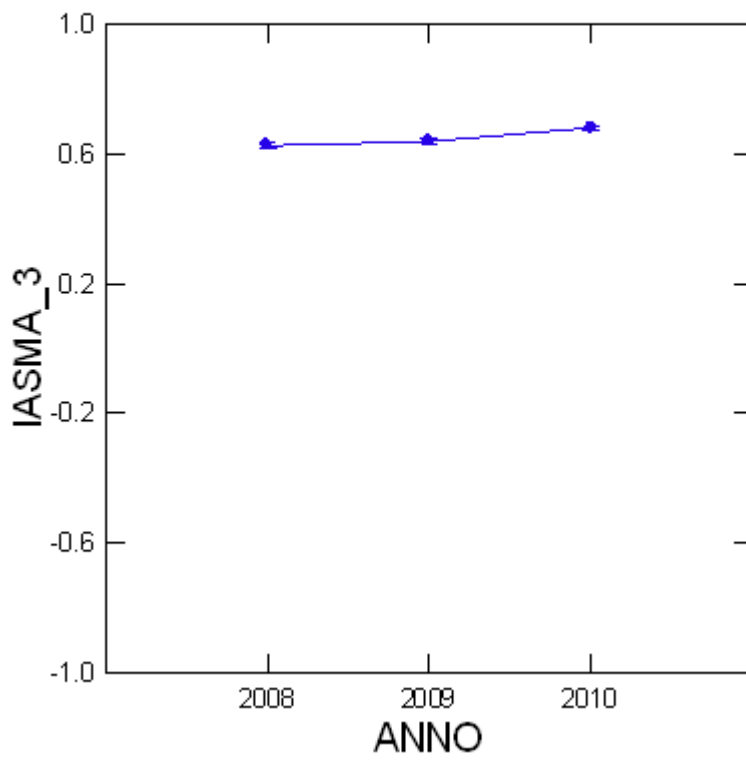
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		
DIRAD\$ (2 levels)	D0	D1		

Dependent Variable	IASMA_3
N	192
Multiple R	0.935
Squared Multiple R	0.874

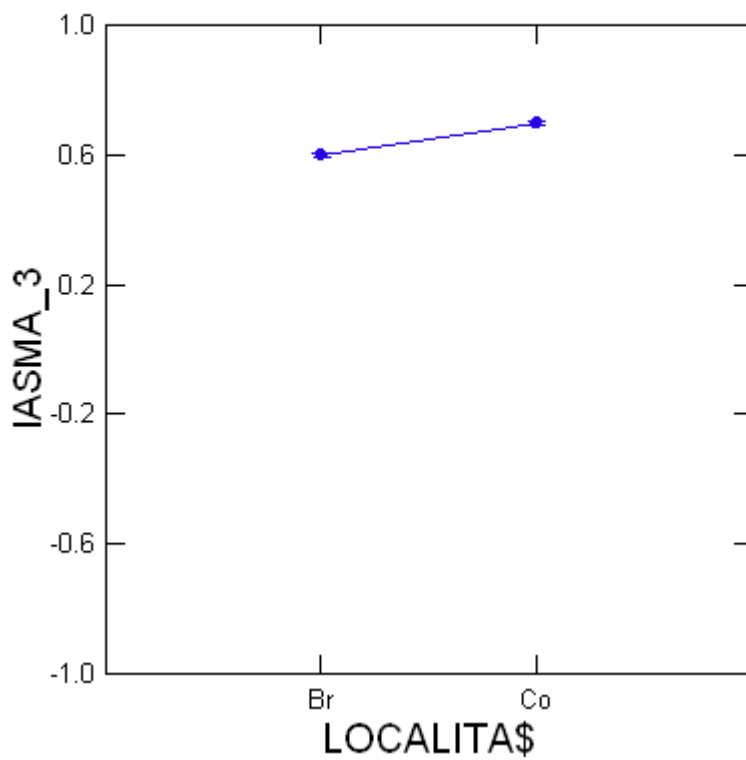
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.102	2	0.051	58.193	0.000
LOCALITA\$	0.346	1	0.346	392.787	0.000
N_GEMME\$	0.001	1	0.001	1.209	0.273
DEFOGL\$	0.000	1	0.000	0.348	0.556
DIRAD\$	0.008	1	0.008	8.854	0.003
LOCALITA\$*ANNO	0.037	2	0.018	20.975	0.000
N_GEMME\$*ANNO	0.002	2	0.001	1.292	0.278
DEFOGL\$*ANNO	0.002	2	0.001	0.935	0.395
DIRAD\$*ANNO	0.001	2	0.001	0.572	0.566
N_GEMME\$*LOCALITA\$	0.001	1	0.001	1.088	0.299
DEFOGL\$*LOCALITA\$	0.000	1	0.000	0.000	0.999
DIRAD\$*LOCALITA\$	0.003	1	0.003	3.119	0.079
DEFOGL\$*N_GEMME\$	0.001	1	0.001	0.875	0.351
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.535	0.466
DIRAD\$*DEFOGL\$	0.001	1	0.001	1.241	0.267
BLOCCO(LOCALITA\$)	0.224	6	0.037	42.310	0.000
BLOCCO*ANNO(LOCALITA\$)	0.117	12	0.010	11.076	0.000
Error	0.135	153	0.001		

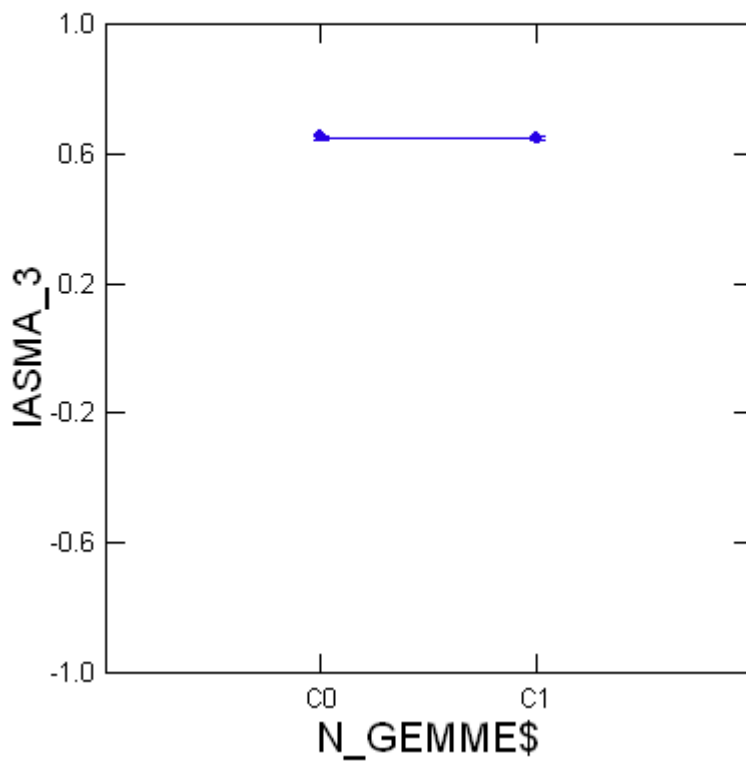
Least Squares Means



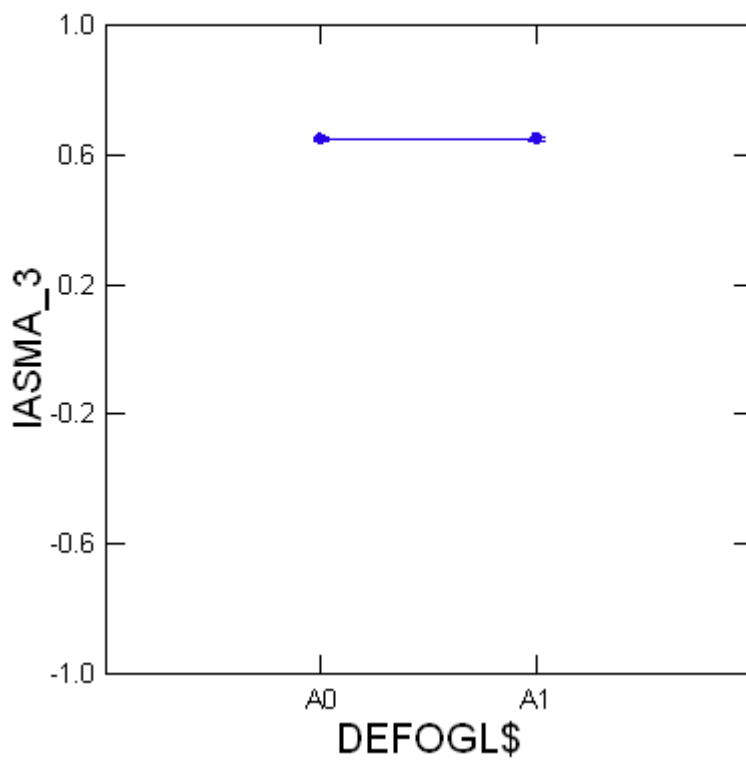
Least Squares Means



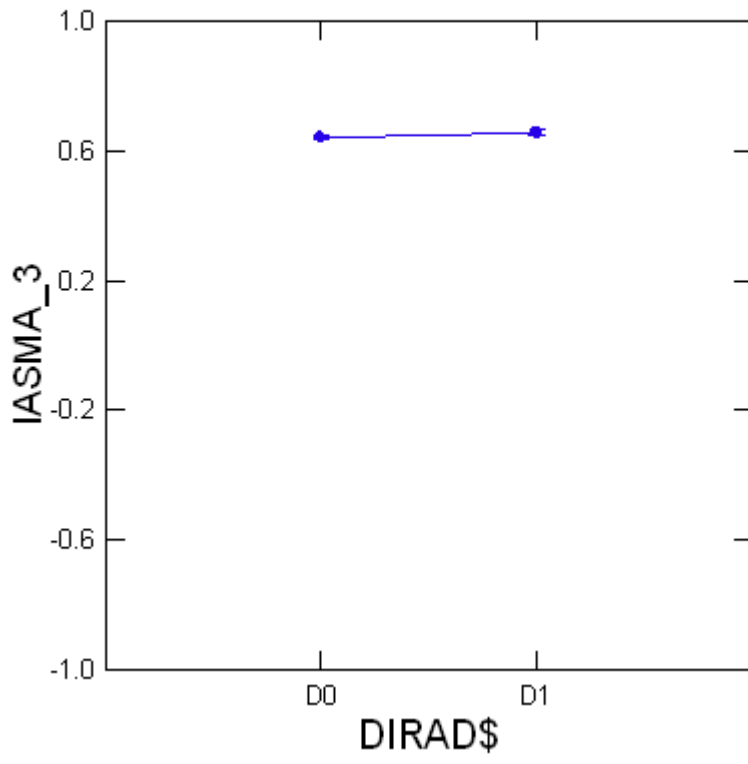
Least Squares Means



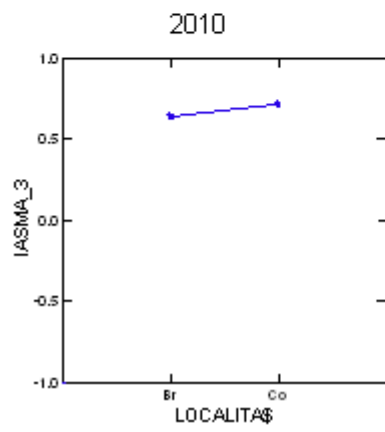
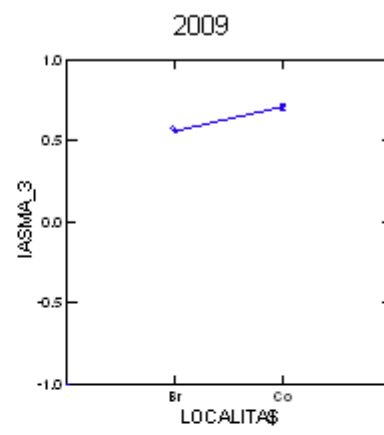
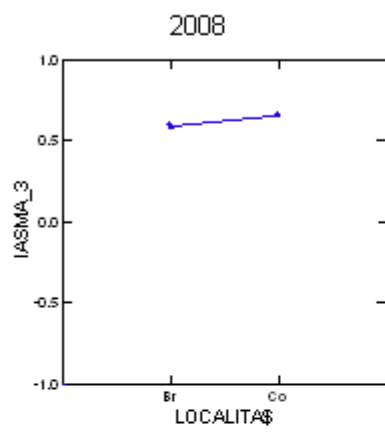
Least Squares Means



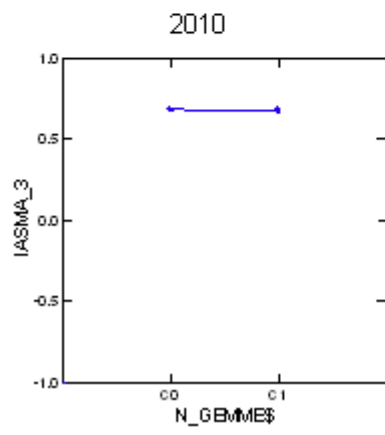
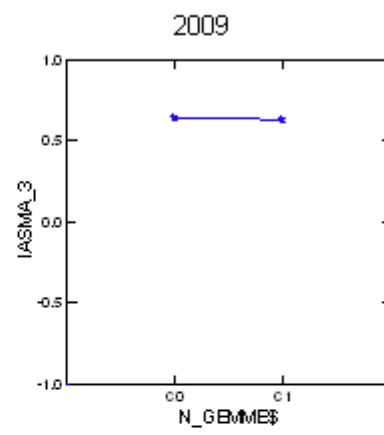
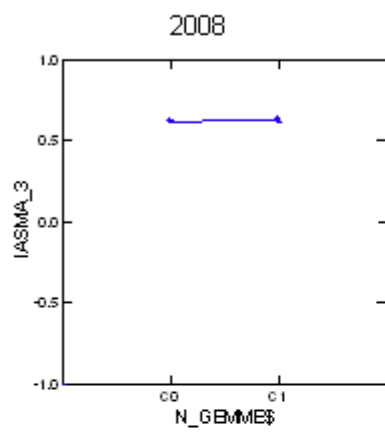
Least Squares Means



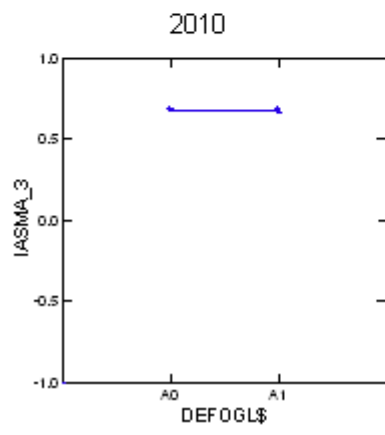
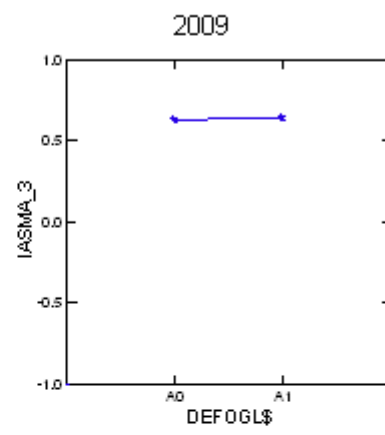
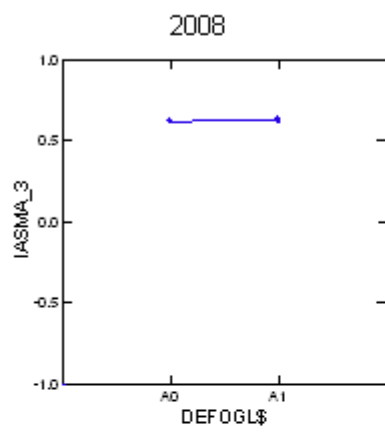
Least Squares Means



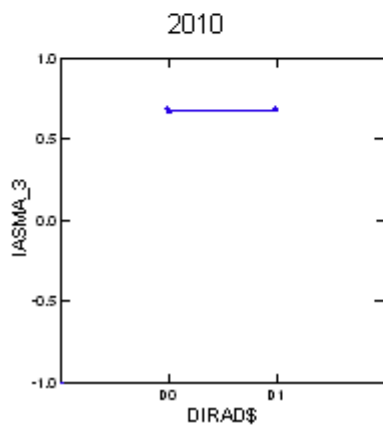
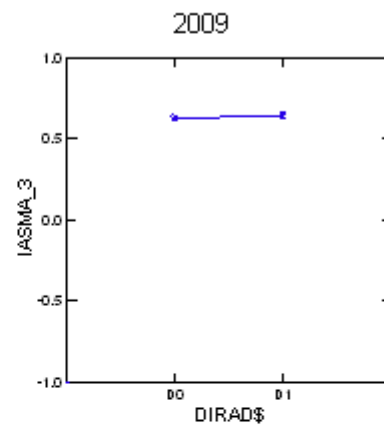
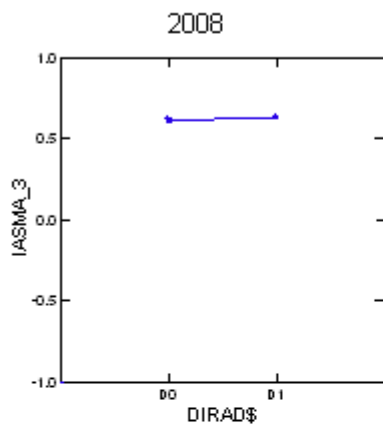
Least Squares Means



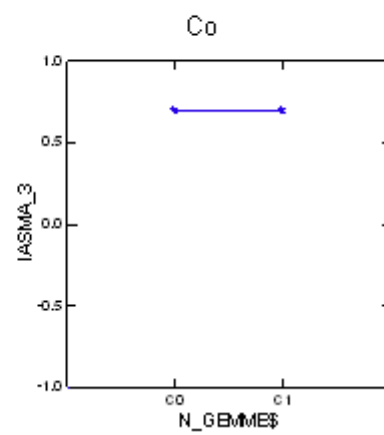
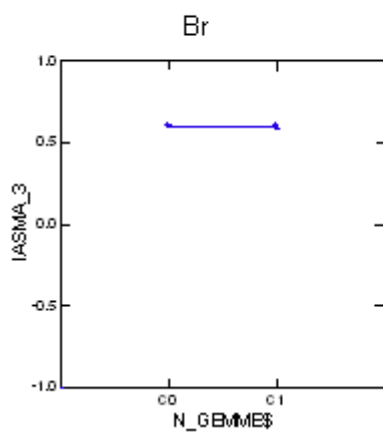
Least Squares Means



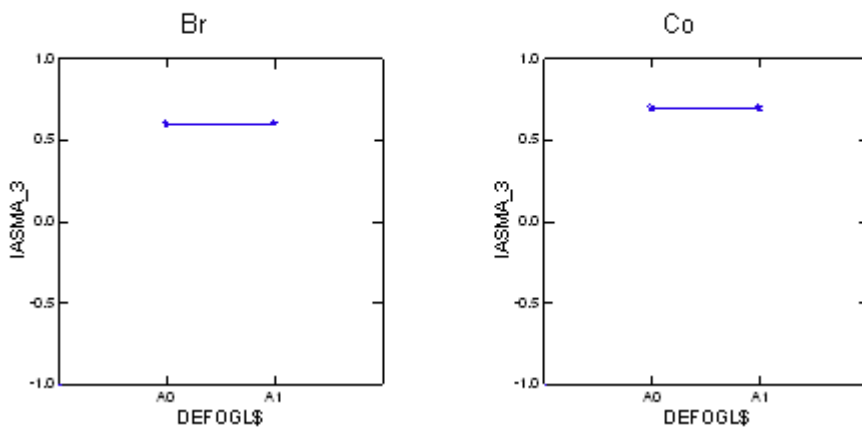
Least Squares Means



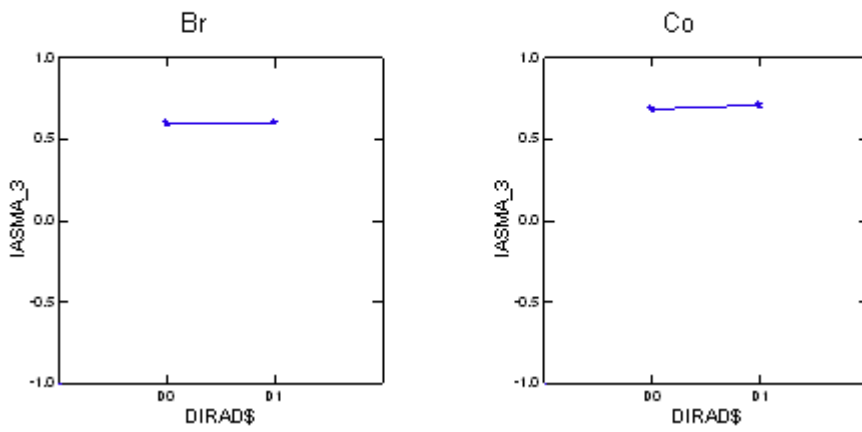
Least Squares Means



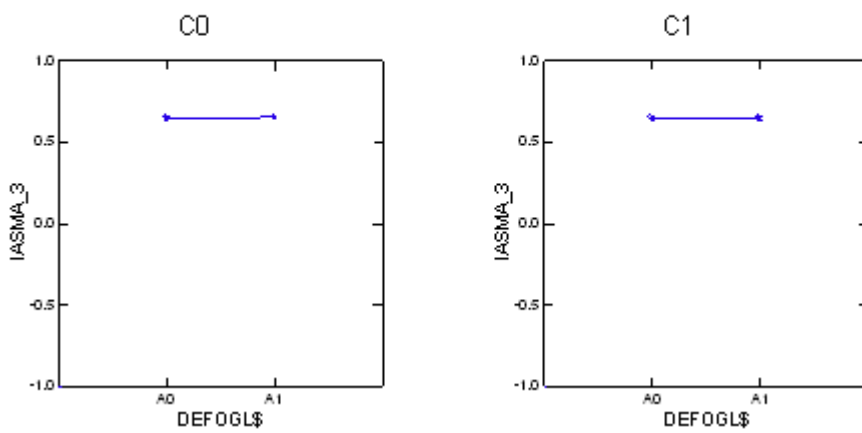
Least Squares Means



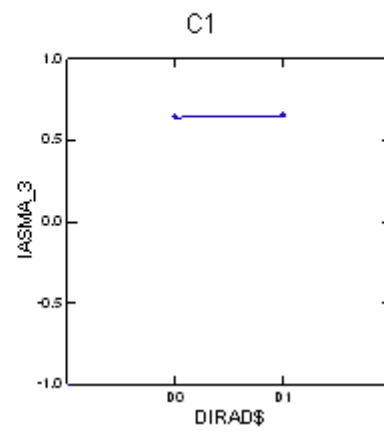
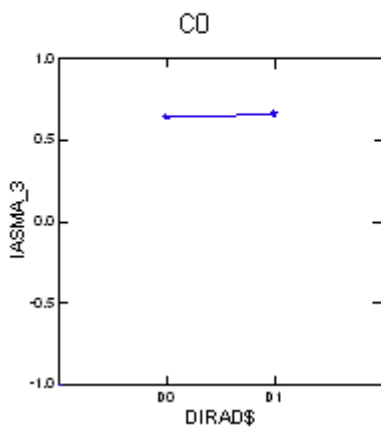
Least Squares Means



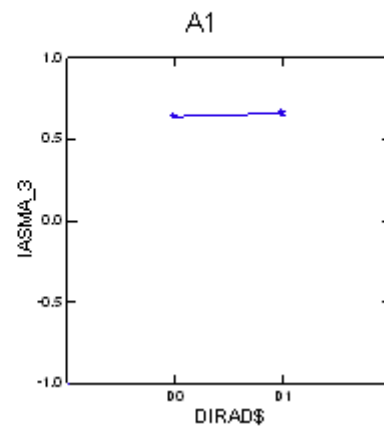
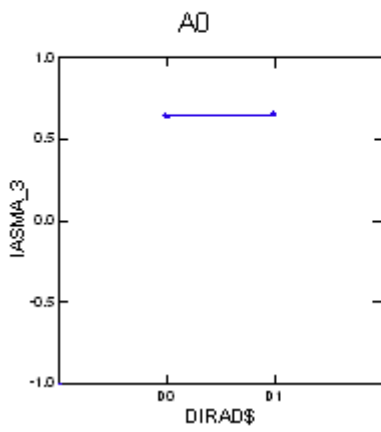
Least Squares Means



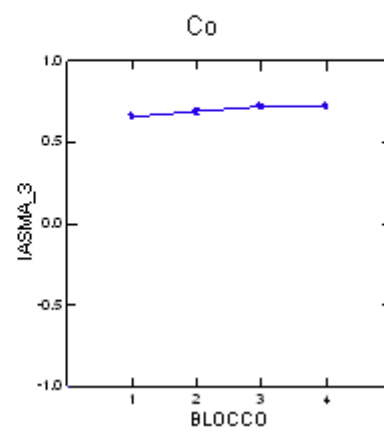
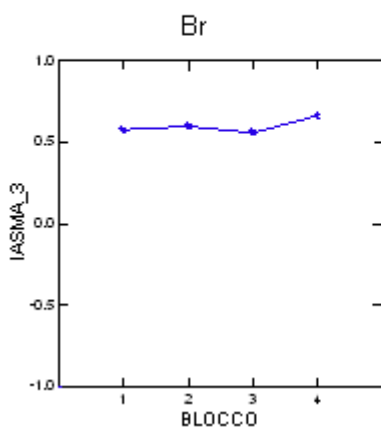
Least Squares Means



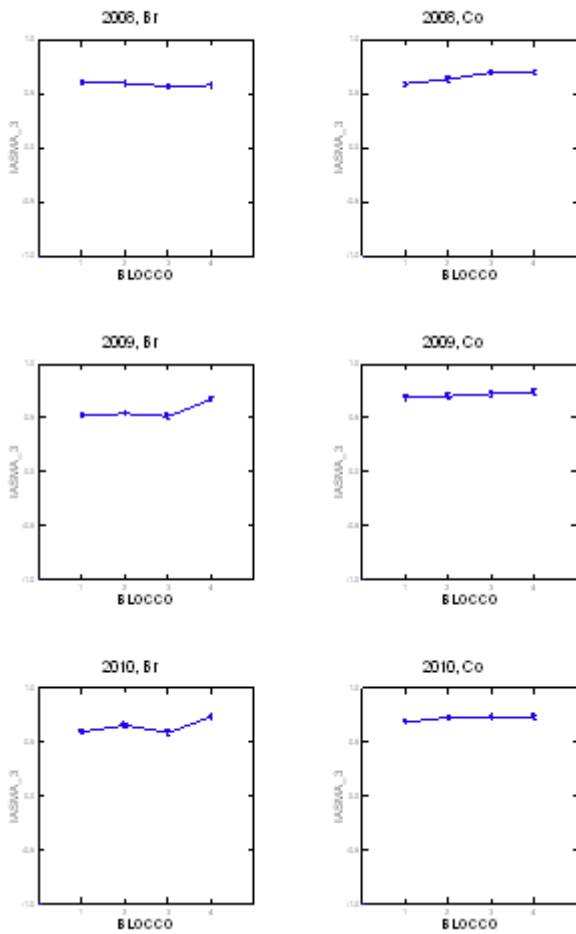
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.857
 First Order Autocorrelation | 0.068

Information Criteria

AIC | -769.454
 AIC (Corrected) | -747.732
 Schwarz's BIC | -639.154

▼ General Linear Model

DIPROVE1 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels			
ANNO (3 levels)	2,008.000	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		

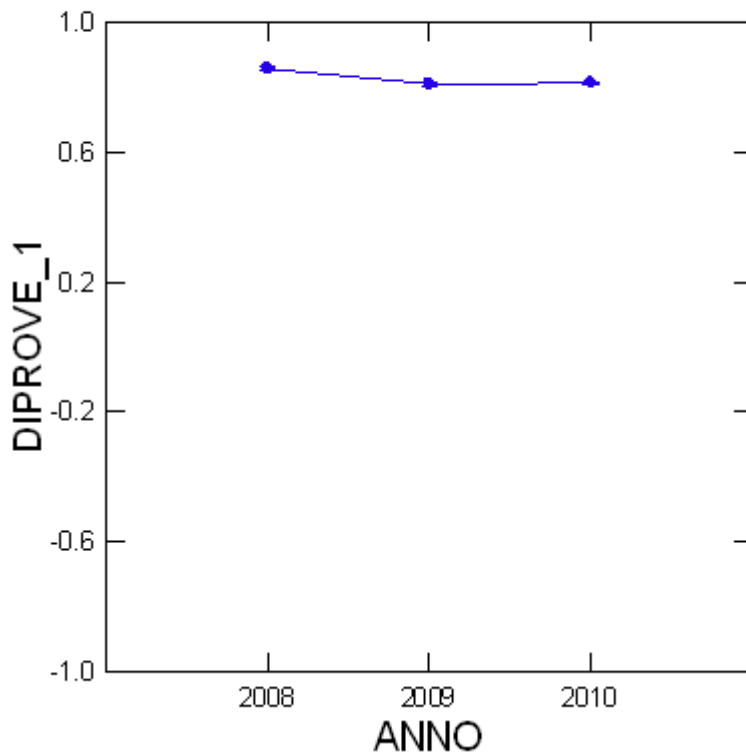
DEFOGL\$ (2 levels)	A0	A1
DIRAD\$ (2 levels)	D0	D1

Dependent Variable	DIPROVE_1
N	192
Multiple R	0.934
Squared Multiple R	0.873

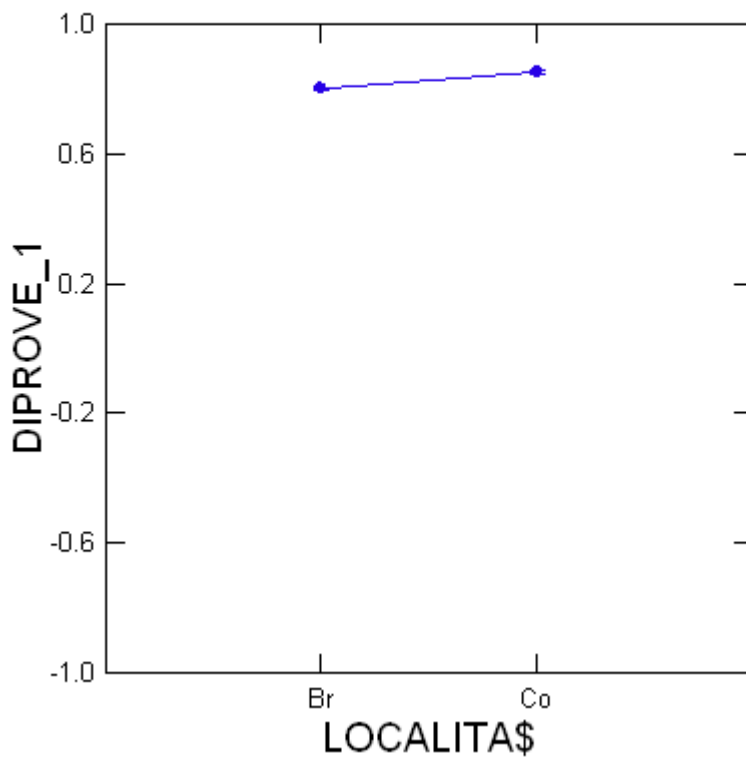
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.079	2	0.039	116.702	0.000
LOCALITA\$	0.094	1	0.094	278.378	0.000
N_GEMME\$	0.006	1	0.006	18.034	0.000
DEFOGL\$	0.037	1	0.037	108.738	0.000
DIRAD\$	0.001	1	0.001	1.490	0.224
LOCALITA\$*ANNO	0.027	2	0.014	40.341	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.382	0.683
DEFOGL\$*ANNO	0.012	2	0.006	17.568	0.000
DIRAD\$*ANNO	0.000	2	0.000	0.298	0.743
N_GEMME\$*LOCALITA\$	0.003	1	0.003	8.747	0.004
DEFOGL\$*LOCALITA\$	0.011	1	0.011	32.846	0.000
DIRAD\$*LOCALITA\$	0.000	1	0.000	1.093	0.297
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.333	0.565
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.000	0.994
DIRAD\$*DEFOGL\$	0.002	1	0.002	4.663	0.032
BLOCCO(LOCALITA\$)	0.024	6	0.004	11.682	0.000
BLOCCO*ANNO(LOCALITA\$)	0.014	12	0.001	3.513	0.000
Error	0.052	153	0.000		

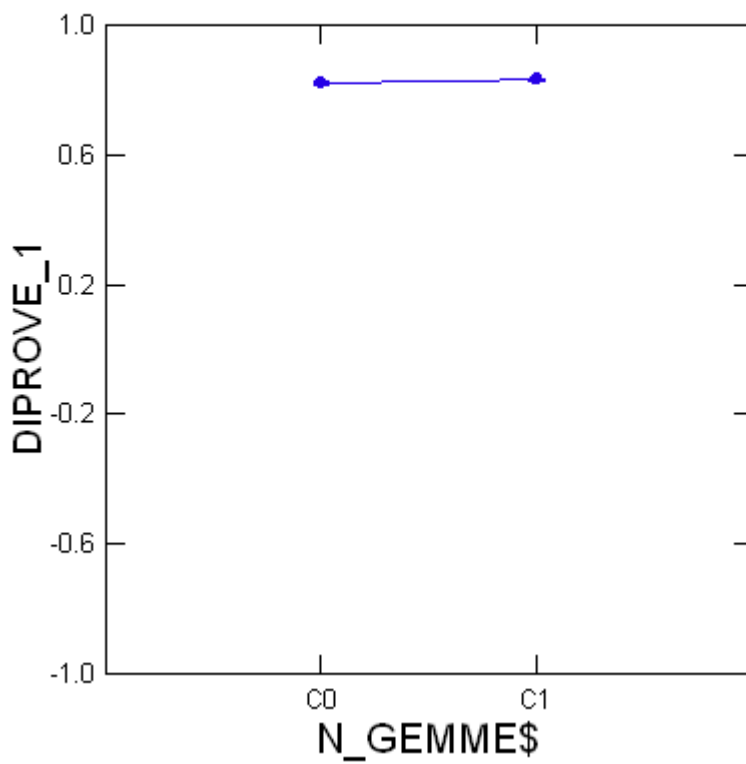
Least Squares Means



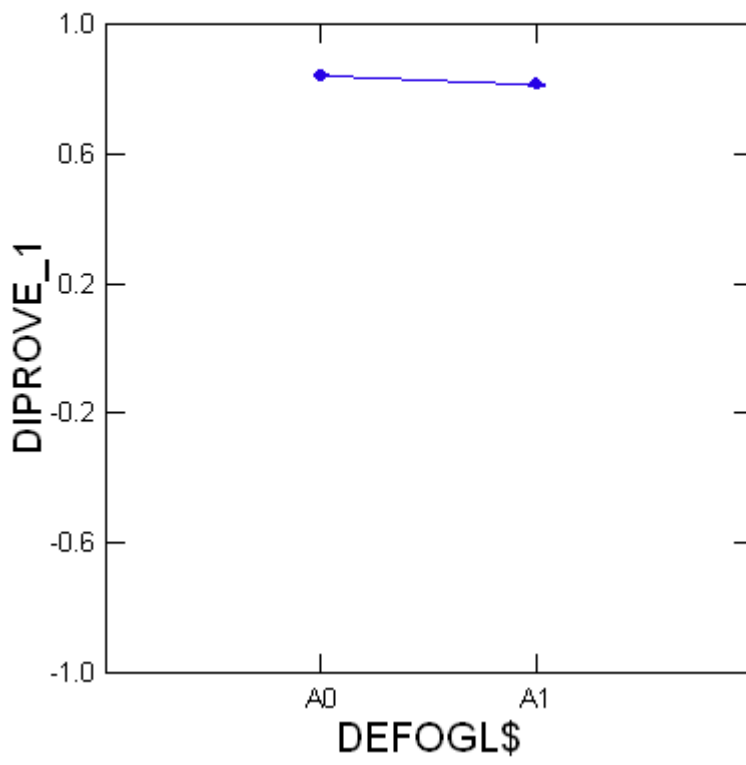
Least Squares Means



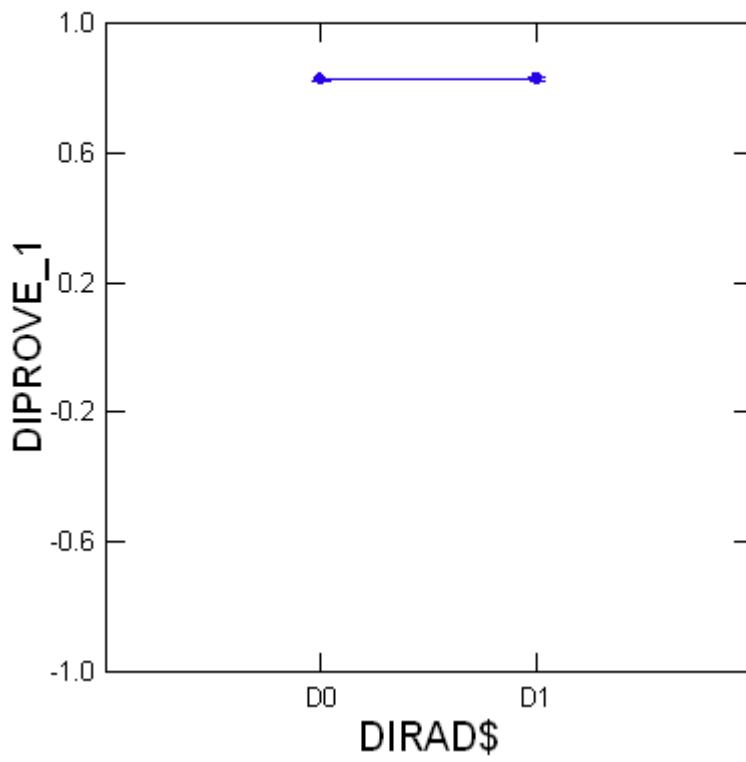
Least Squares Means



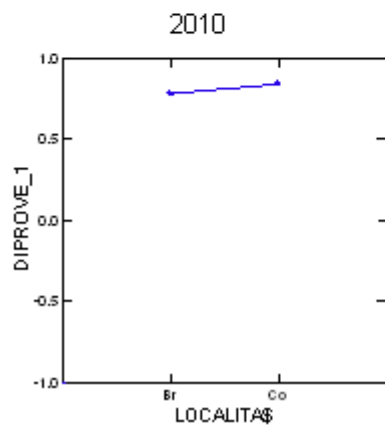
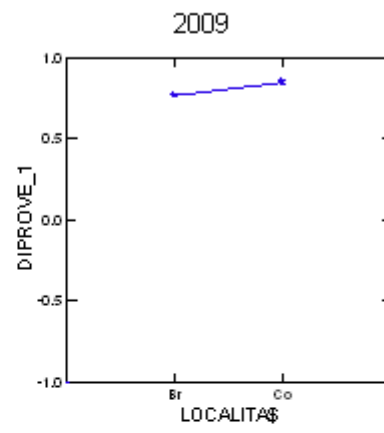
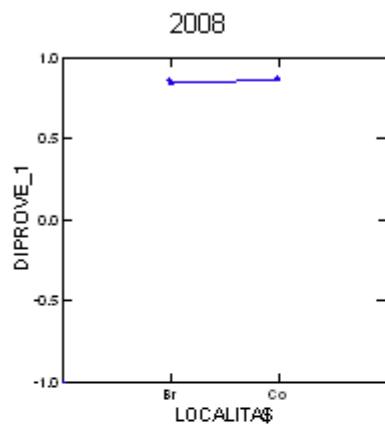
Least Squares Means



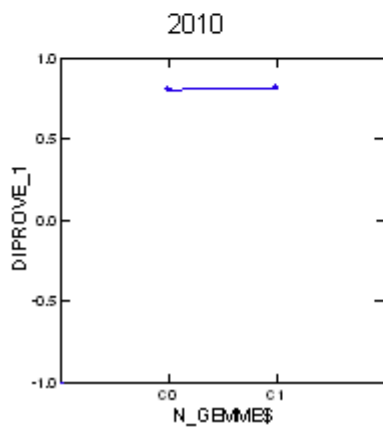
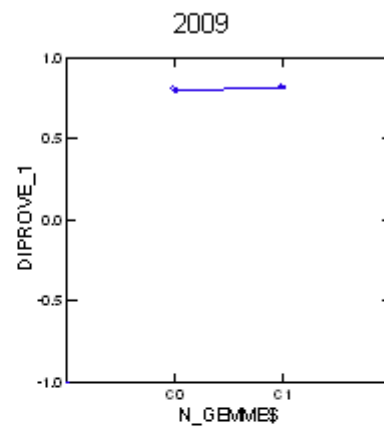
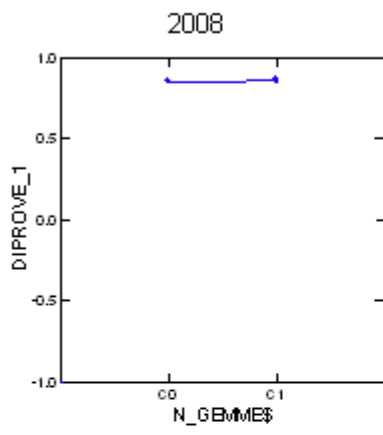
Least Squares Means



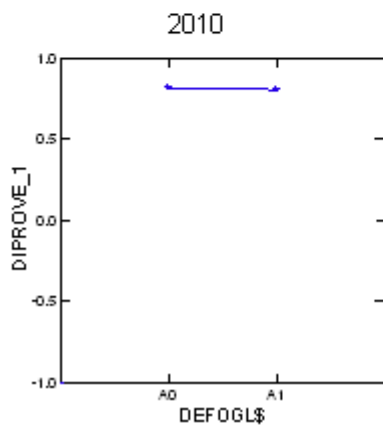
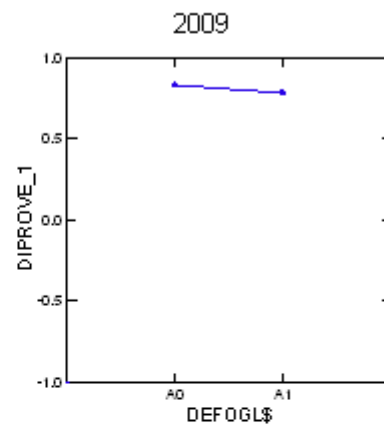
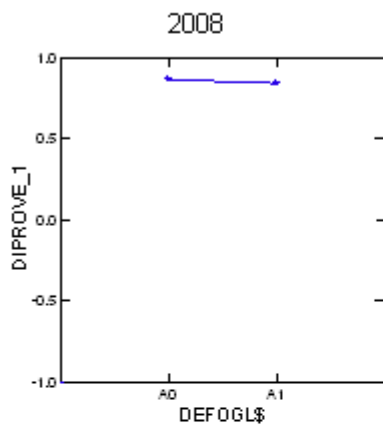
Least Squares Means



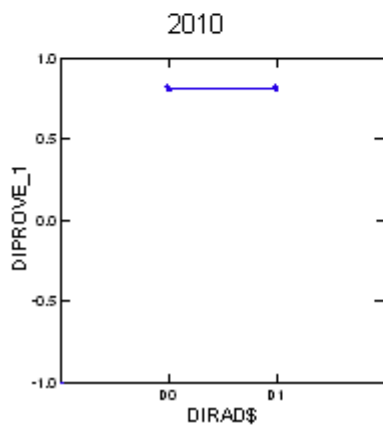
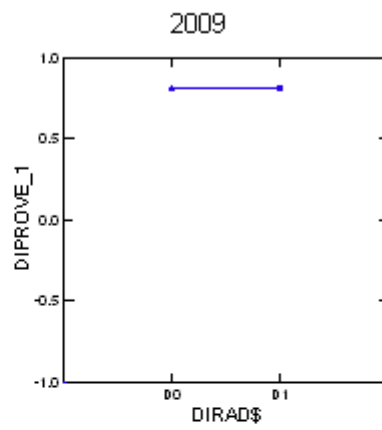
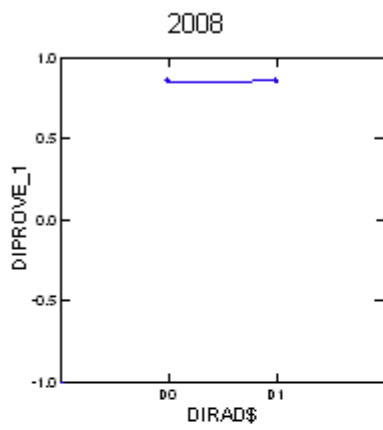
Least Squares Means



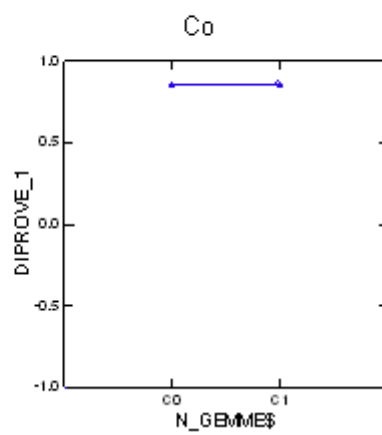
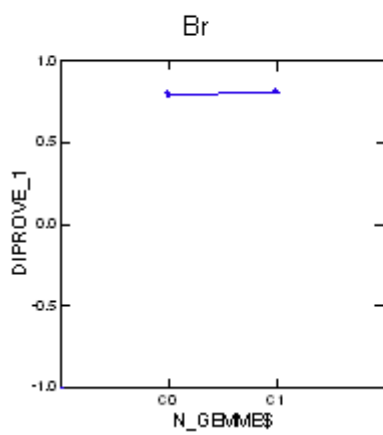
Least Squares Means



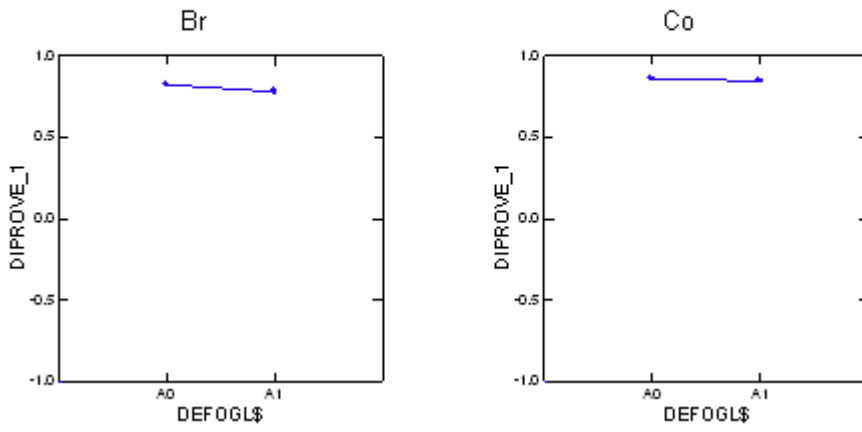
Least Squares Means



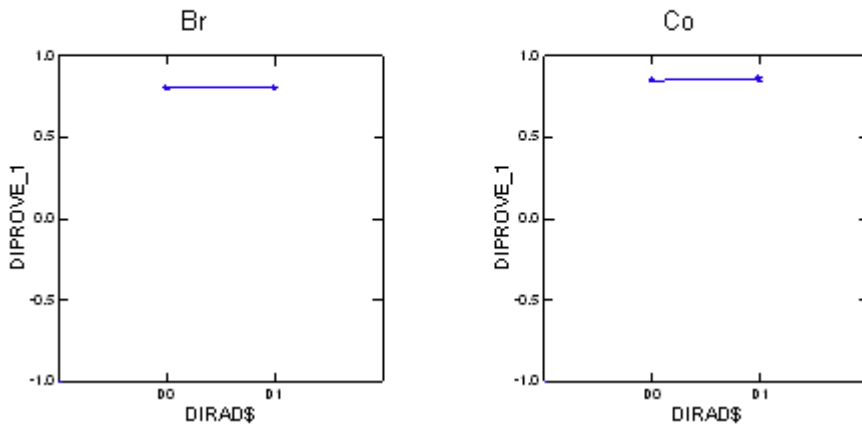
Least Squares Means



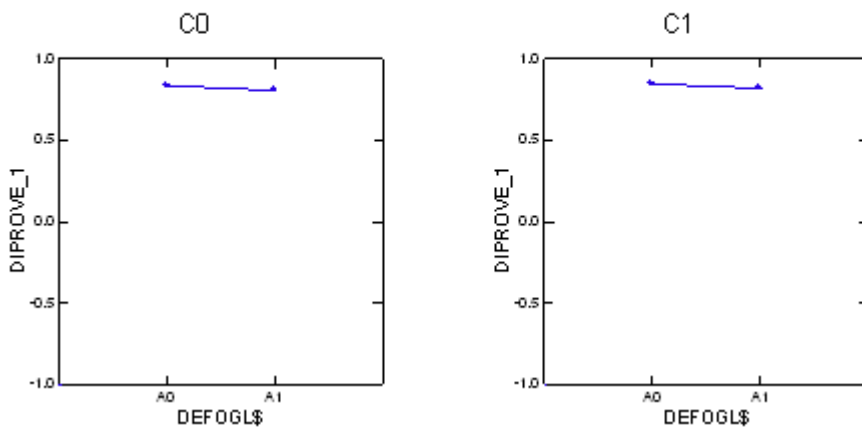
Least Squares Means



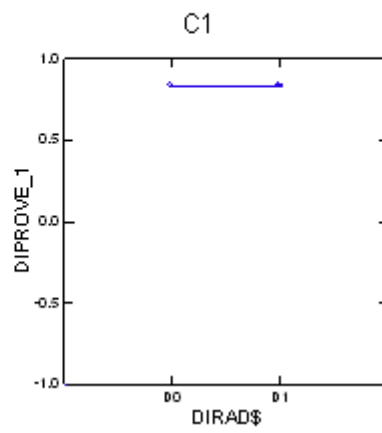
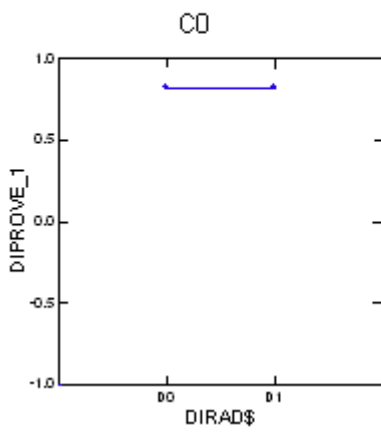
Least Squares Means



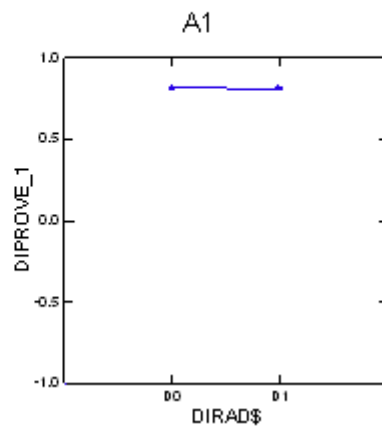
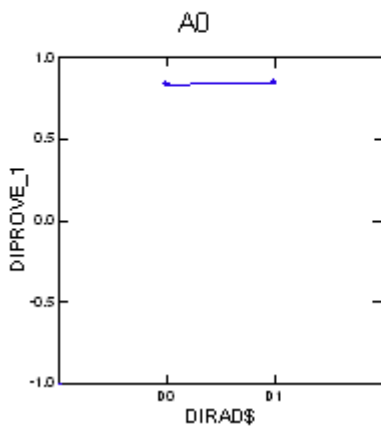
Least Squares Means



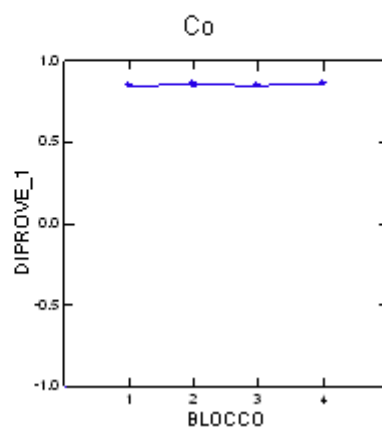
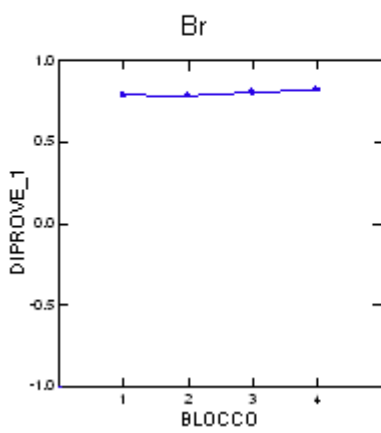
Least Squares Means



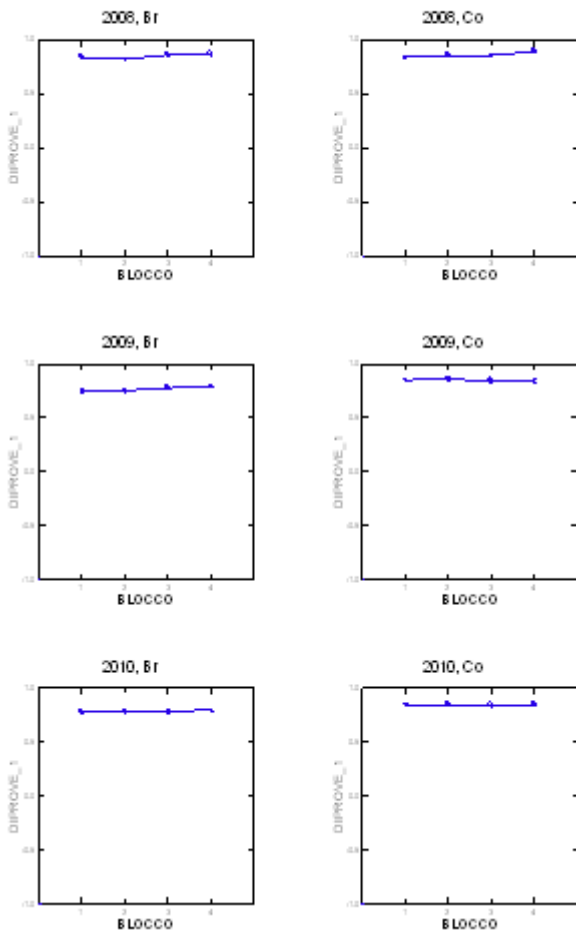
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 49 is an Outlier (Studentized Residual : -6.300)

Durbin-Watson D Statistic | 1.981
 First Order Autocorrelation | 0.009

Information Criteria

AIC | -953.517
 AIC (Corrected) | -931.795
 Schwarz's BIC | -823.217

▼ General Linear Model

DIPROVE2 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000

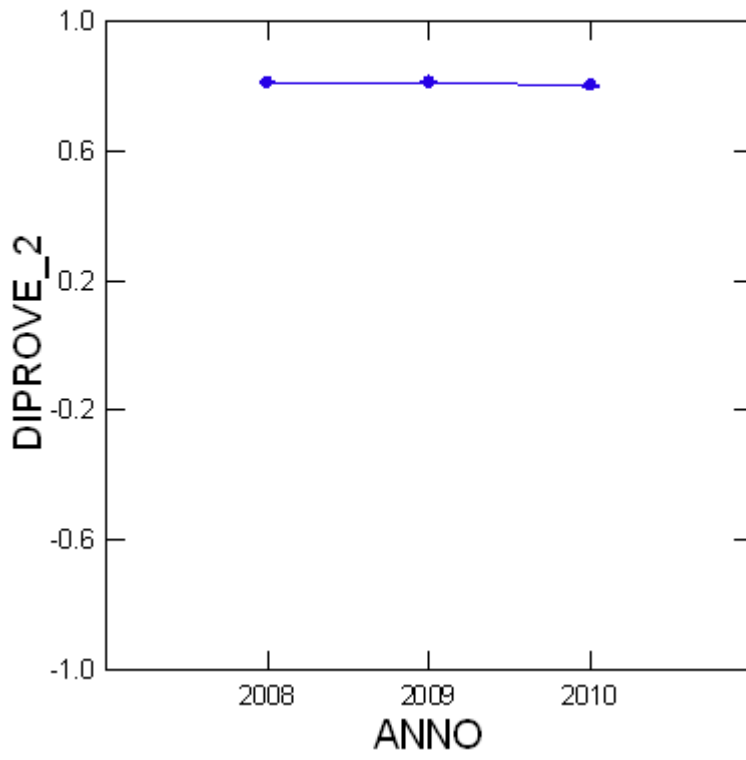
LOCALITA\$ (2 levels)	Br	Co		
BLOCCO (4 levels)	1.000	2.000	3.000	4.000
N_GEMME\$ (2 levels)	C0	C1		
DEFOGL\$ (2 levels)	A0	A1		
DIRAD\$ (2 levels)	D0	D1		

Dependent Variable	DIPROVE_2
N	192
Multiple R	0.857
Squared Multiple R	0.734

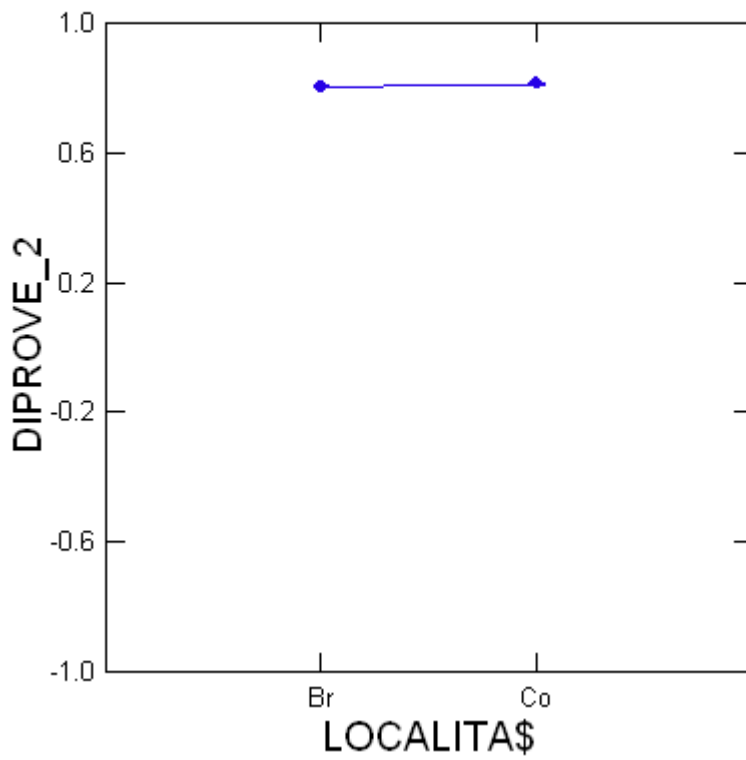
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.003	2	0.002	7.468	0.001
LOCALITA\$	0.003	1	0.003	12.289	0.001
N_GEMME\$	0.000	1	0.000	1.333	0.250
DEFOGL\$	0.005	1	0.005	24.655	0.000
DIRAD\$	0.000	1	0.000	0.592	0.443
LOCALITA\$*ANNO	0.014	2	0.007	34.064	0.000
N_GEMME\$*ANNO	0.001	2	0.000	2.418	0.093
DEFOGL\$*ANNO	0.002	2	0.001	4.916	0.009
DIRAD\$*ANNO	0.000	2	0.000	0.977	0.379
N_GEMME\$*LOCALITA\$	0.000	1	0.000	1.859	0.175
DEFOGL\$*LOCALITA\$	0.002	1	0.002	10.225	0.002
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.415	0.520
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.546	0.461
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.196	0.658
DIRAD\$*DEFOGL\$	0.002	1	0.002	9.183	0.003
BLOCCO (LOCALITA\$)	0.032	6	0.005	26.201	0.000
BLOCCO*ANNO (LOCALITA\$)	0.007	12	0.001	3.014	0.001
Error	0.031	153	0.000		

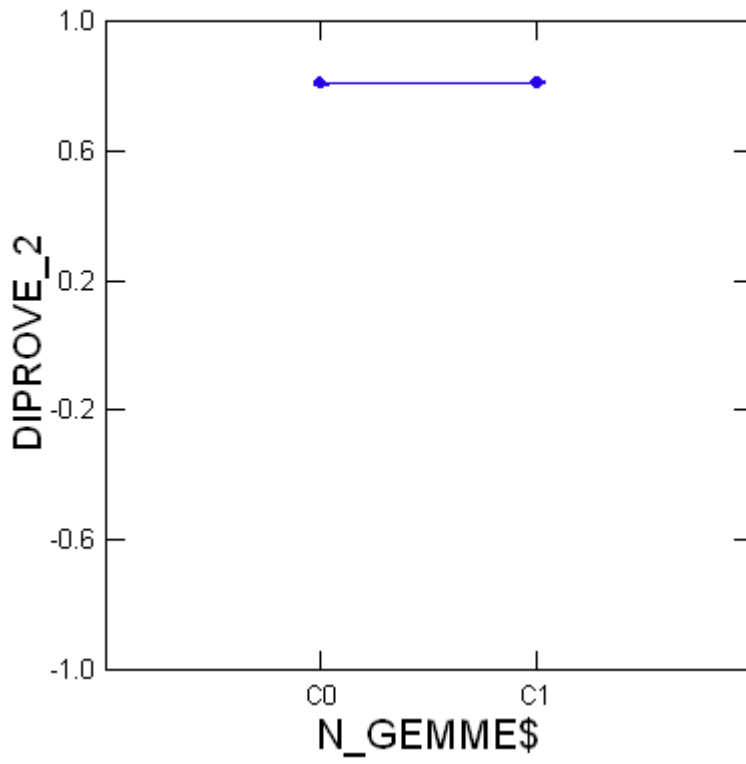
Least Squares Means



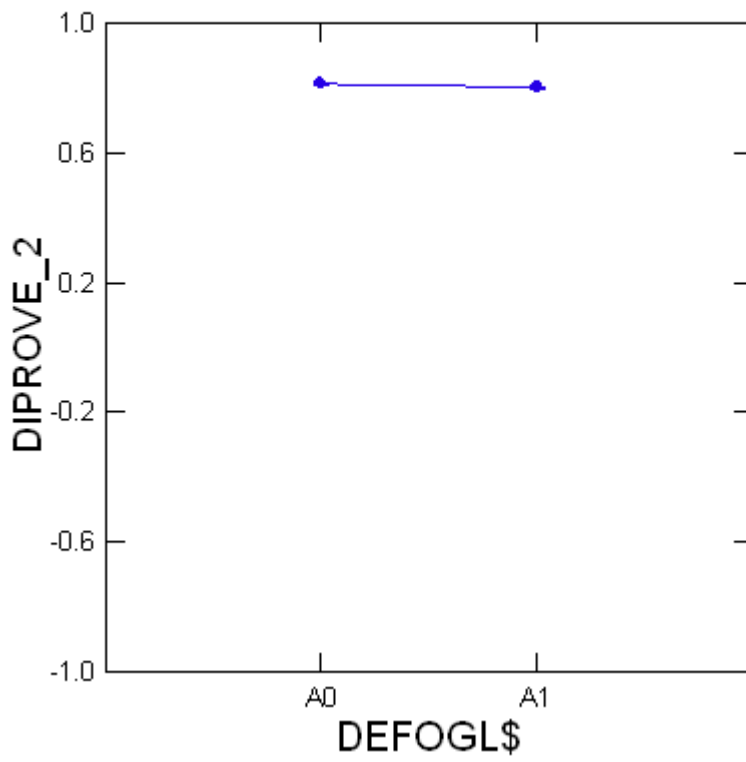
Least Squares Means



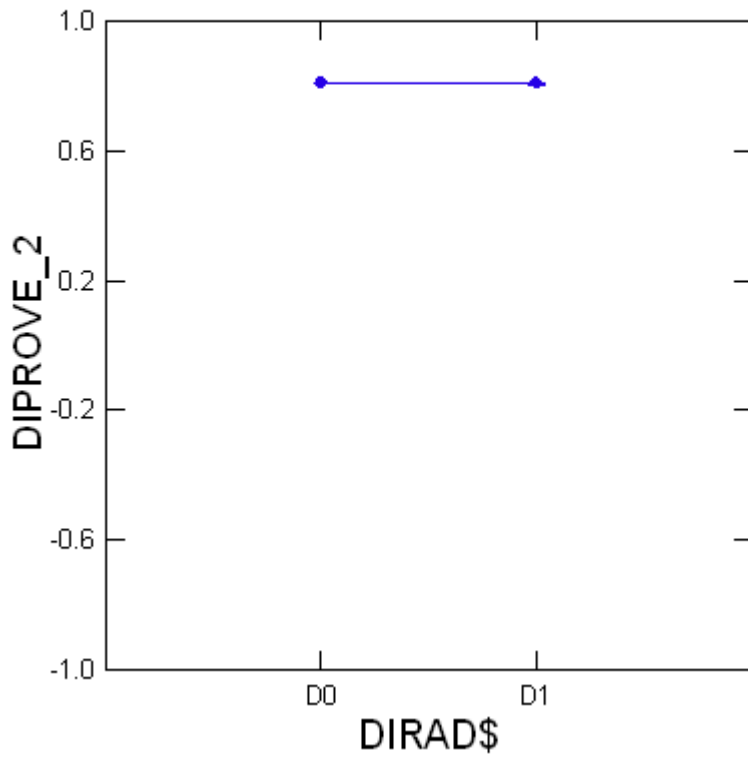
Least Squares Means



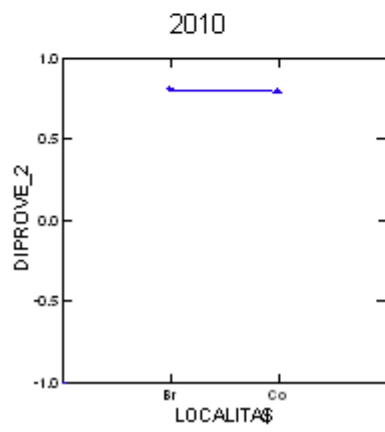
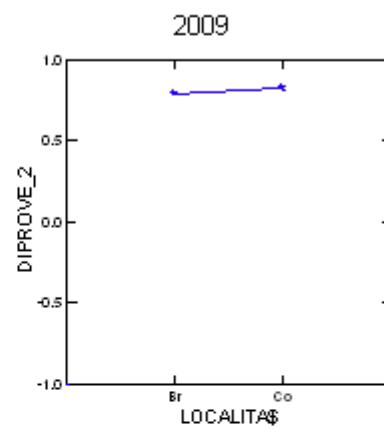
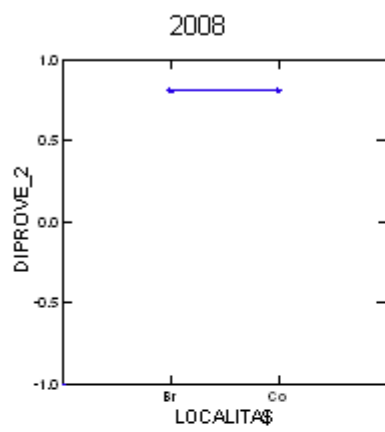
Least Squares Means



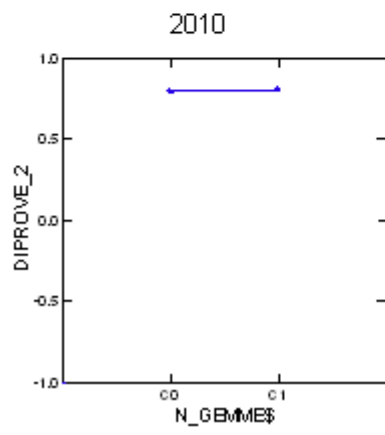
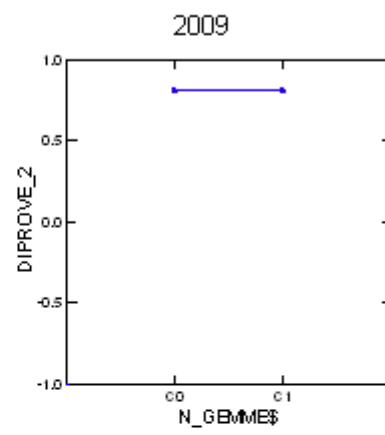
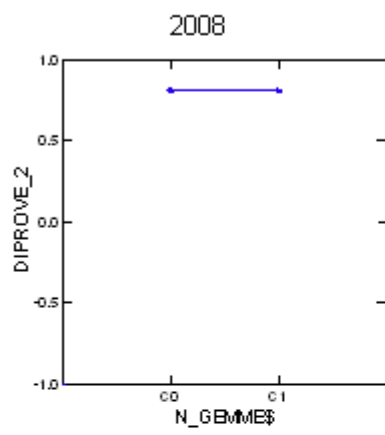
Least Squares Means



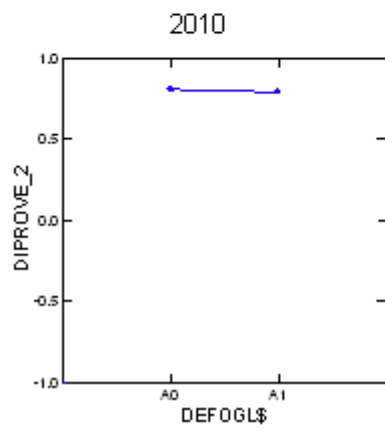
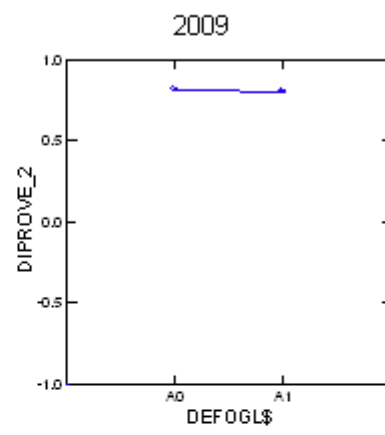
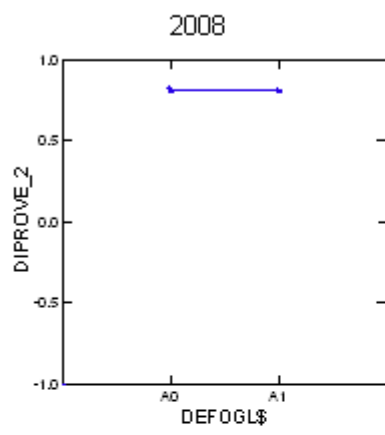
Least Squares Means



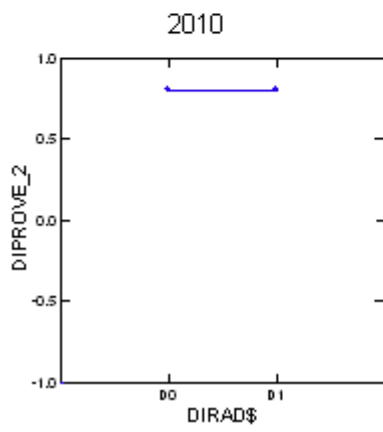
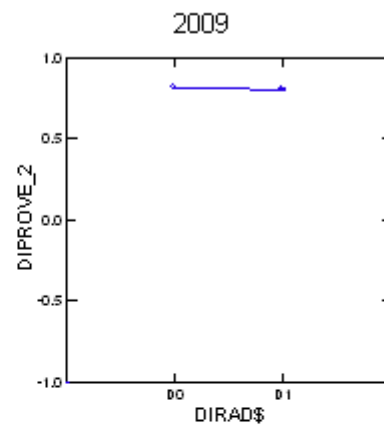
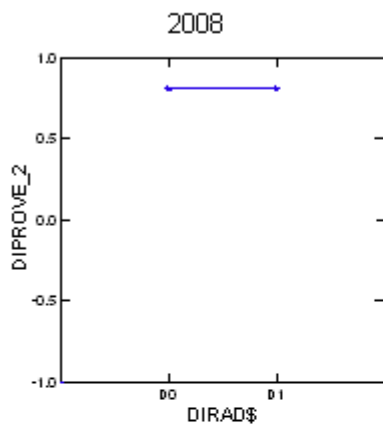
Least Squares Means



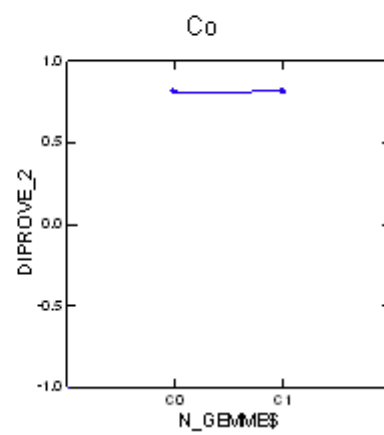
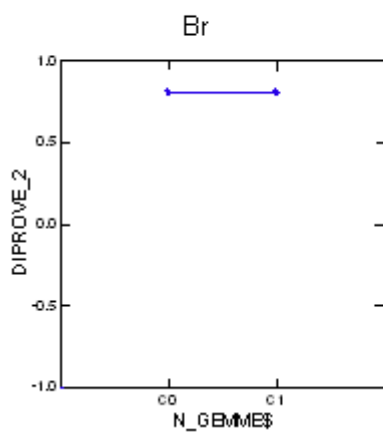
Least Squares Means



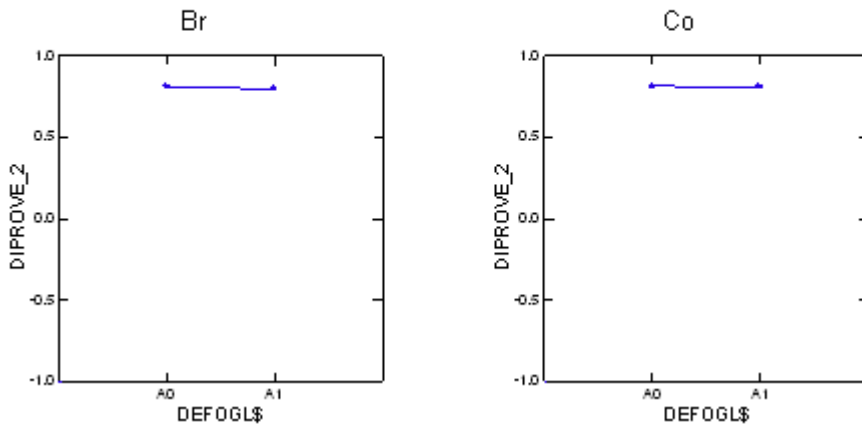
Least Squares Means



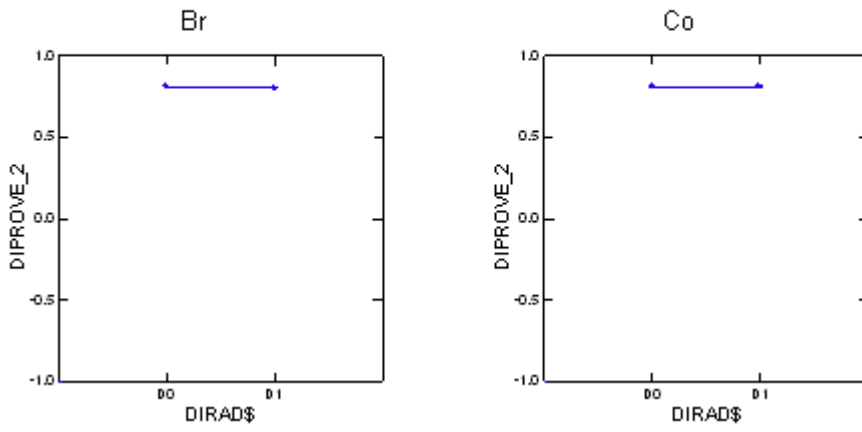
Least Squares Means



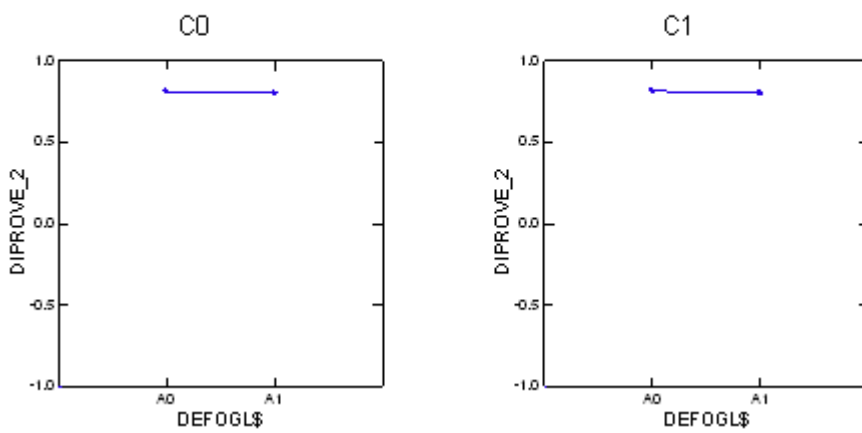
Least Squares Means



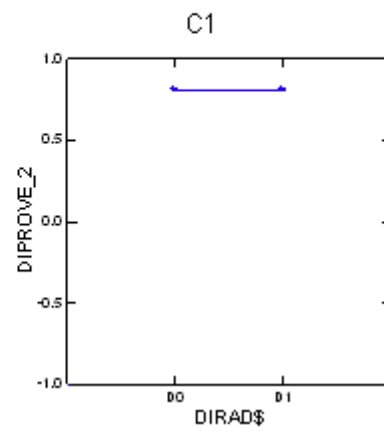
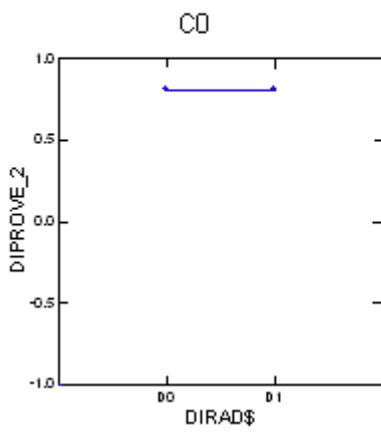
Least Squares Means



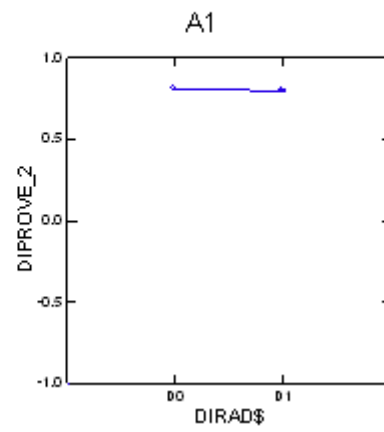
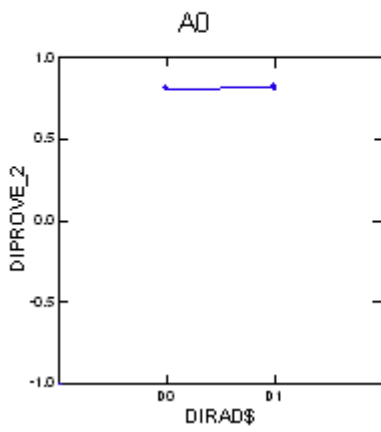
Least Squares Means



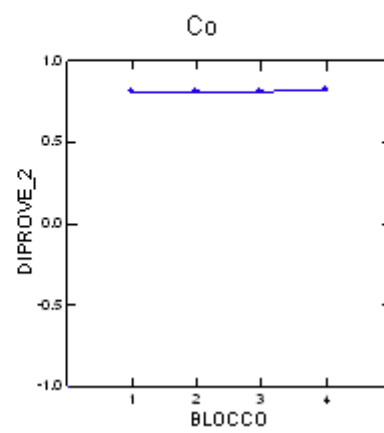
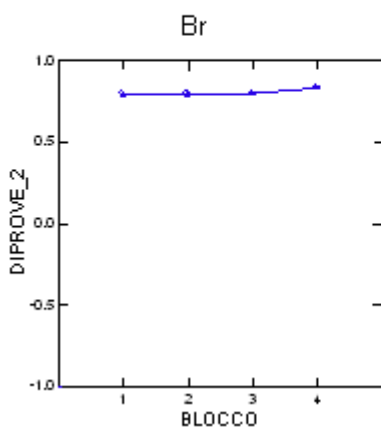
Least Squares Means



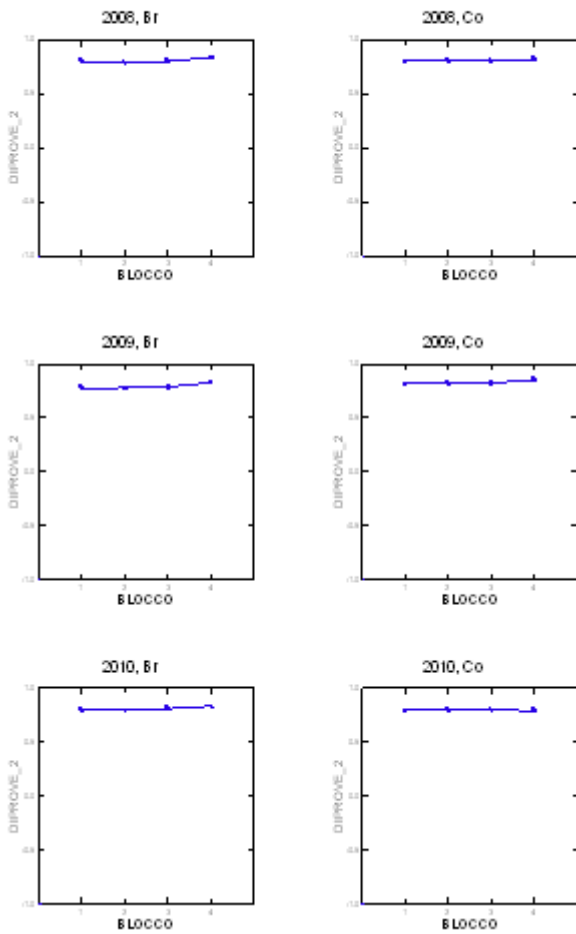
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 48 is an Outlier (Studentized Residual : -5.115)
 Case 49 is an Outlier (Studentized Residual : -3.837)

Durbin-Watson D Statistic | 1.625
 First Order Autocorrelation | 0.187

Information Criteria

AIC | -1,050.574
 AIC (Corrected) | -1,028.852
 Schwarz's BIC | -920.274

▼ General Linear Model

DIPROVE3 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Sangiovese')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels
-----+-----	

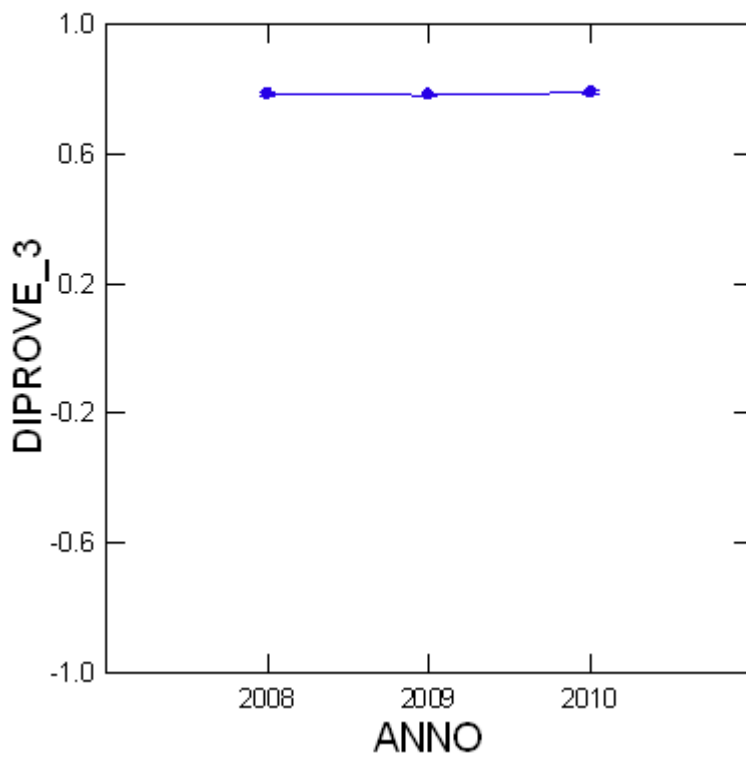
ANNO (3 levels)		2,008.000	2,009.000	2,010.000		
LOCALITA\$ (2 levels)		Br	Co			
BLOCCO (4 levels)		1.000	2.000	3.000	4.000	
N_GEMME\$ (2 levels)		C0	C1			
DEFOGL\$ (2 levels)		A0	A1			
DIRAD\$ (2 levels)		D0	D1			

Dependent Variable		DIPROVE_3
N		192
Multiple R		0.940
Squared Multiple R		0.884

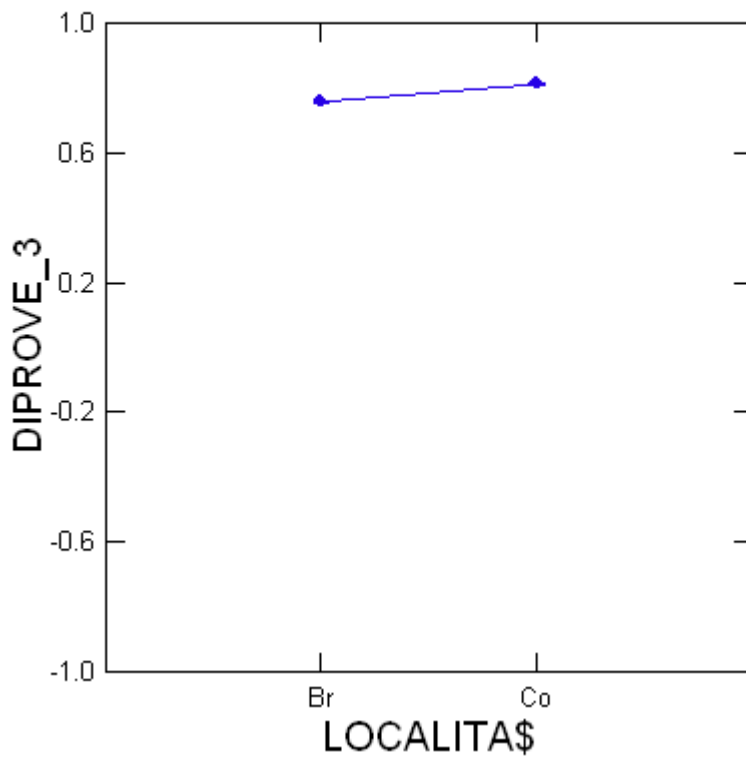
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.002	2	0.001	3.259	0.041
LOCALITA\$	0.106	1	0.106	395.054	0.000
N_GEMME\$	0.000	1	0.000	0.185	0.668
DEFOGL\$	0.000	1	0.000	1.251	0.265
DIRAD\$	0.000	1	0.000	0.031	0.861
LOCALITA\$*ANNO	0.019	2	0.010	36.072	0.000
N_GEMME\$*ANNO	0.001	2	0.001	1.971	0.143
DEFOGL\$*ANNO	0.002	2	0.001	4.491	0.013
DIRAD\$*ANNO	0.000	2	0.000	0.327	0.721
N_GEMME\$*LOCALITA\$	0.001	1	0.001	3.686	0.057
DEFOGL\$*LOCALITA\$	0.002	1	0.002	7.169	0.008
DIRAD\$*LOCALITA\$	0.001	1	0.001	3.174	0.077
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.121	0.728
DIRAD\$*N_GEMME\$	0.001	1	0.001	3.235	0.074
DIRAD\$*DEFOGL\$	0.001	1	0.001	3.795	0.053
BLOCCO (LOCALITA\$)	0.133	6	0.022	82.473	0.000
BLOCCO*ANNO (LOCALITA\$)	0.008	12	0.001	2.381	0.008
Error	0.041	153	0.000		

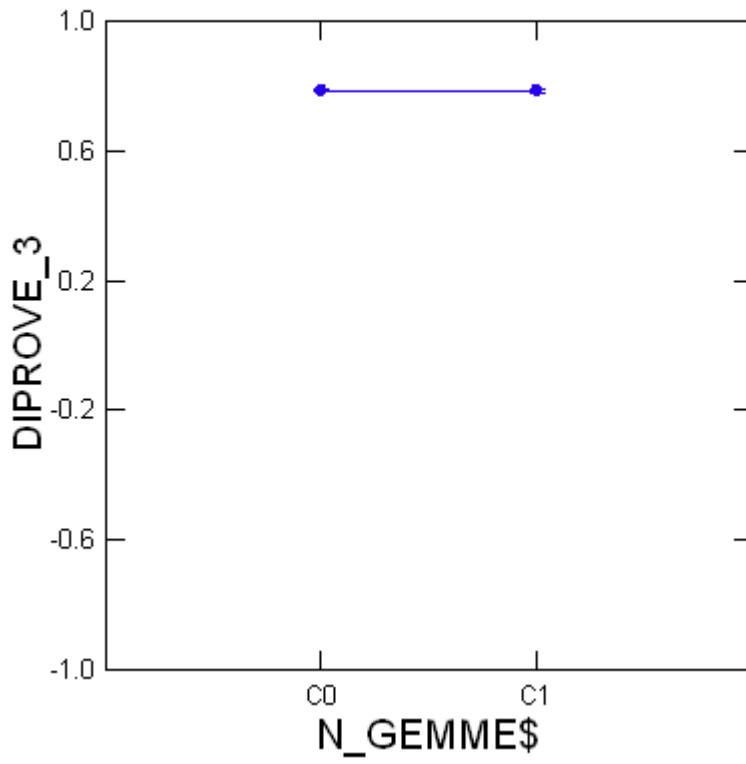
Least Squares Means



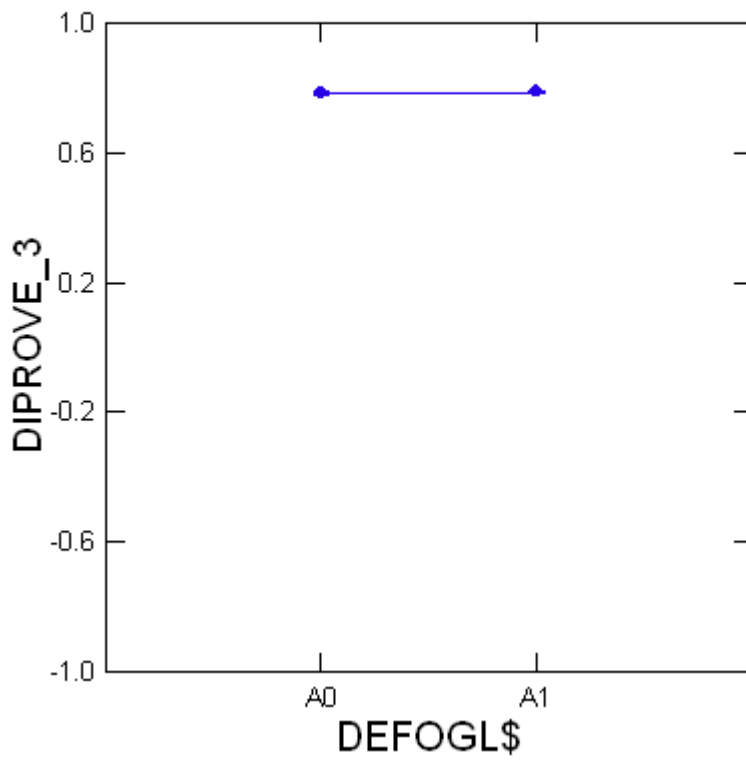
Least Squares Means



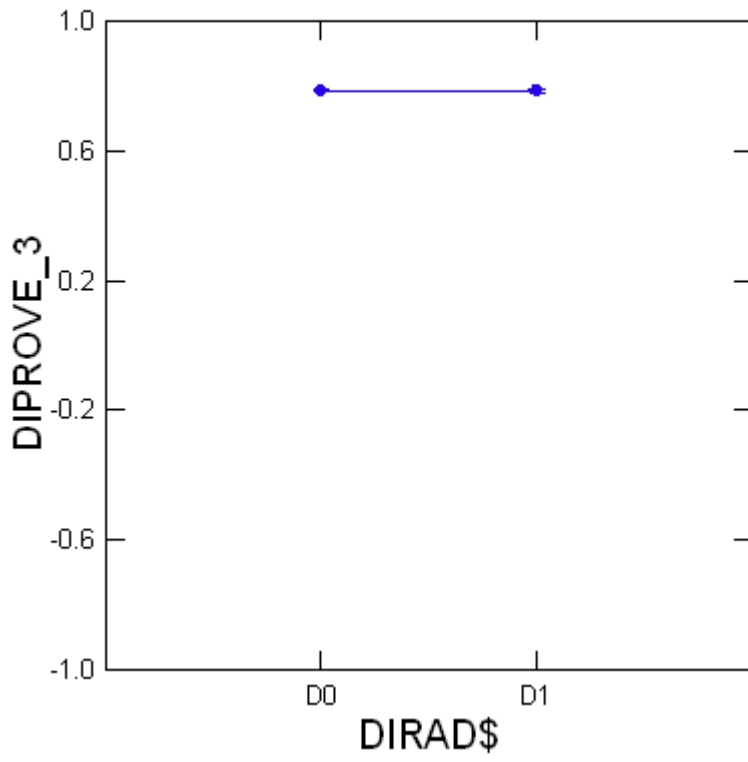
Least Squares Means



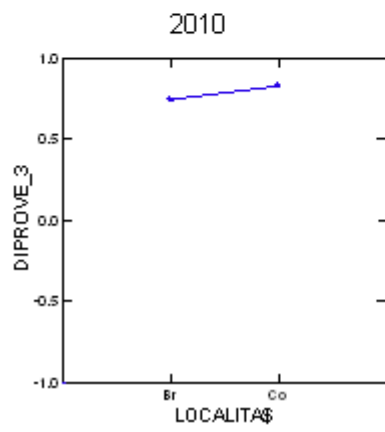
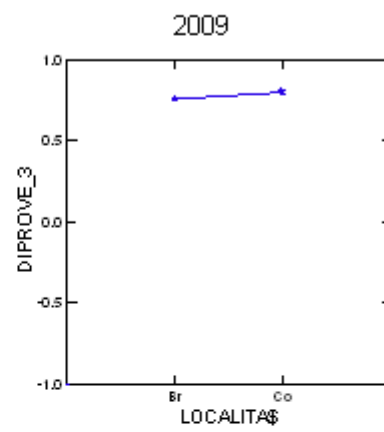
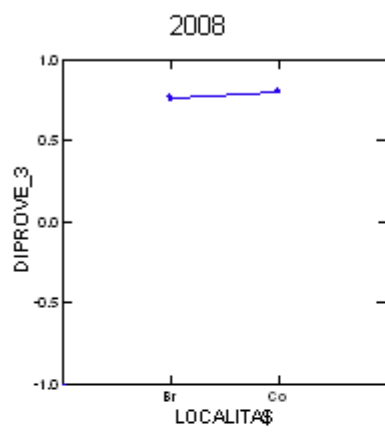
Least Squares Means



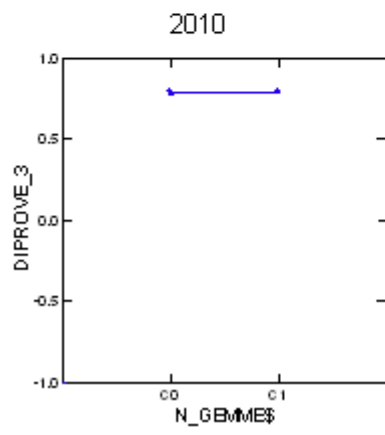
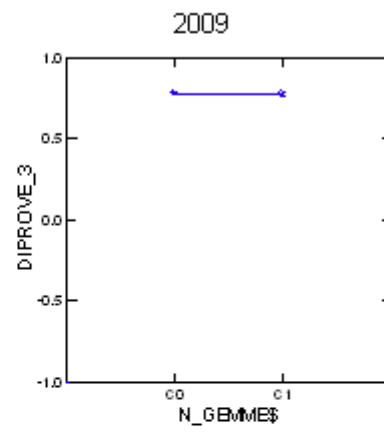
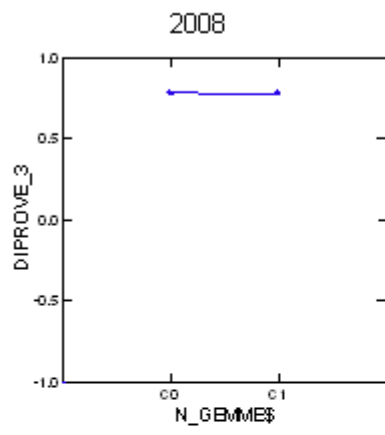
Least Squares Means



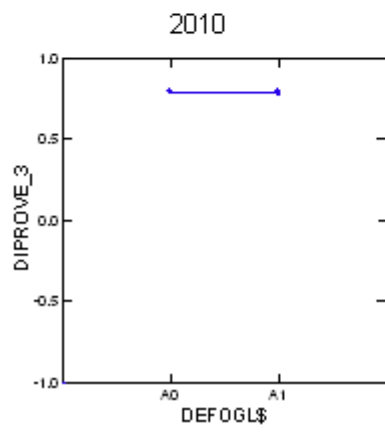
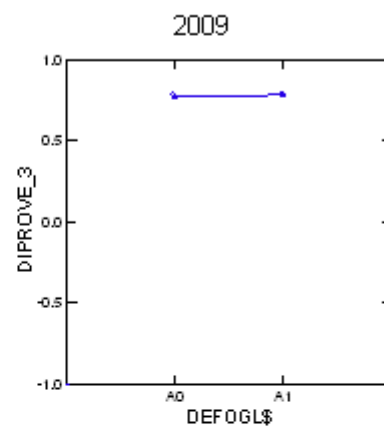
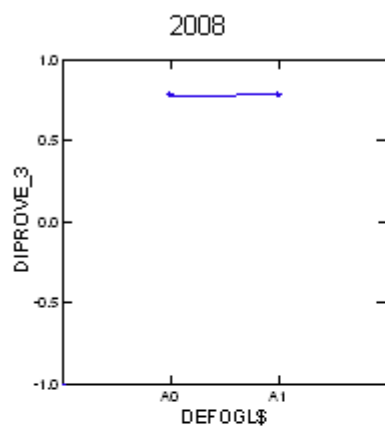
Least Squares Means



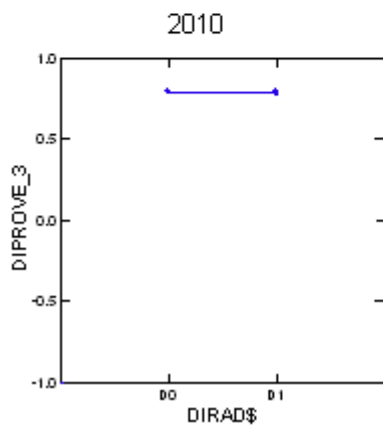
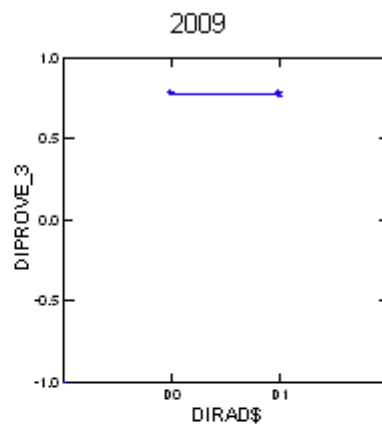
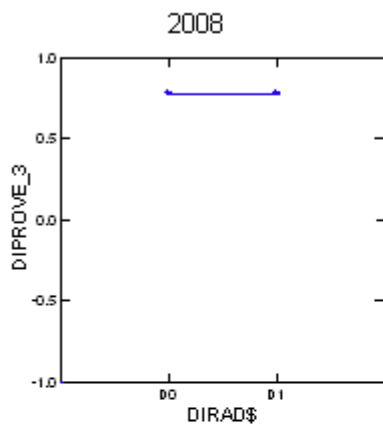
Least Squares Means



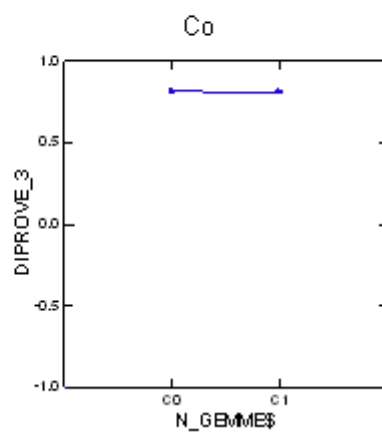
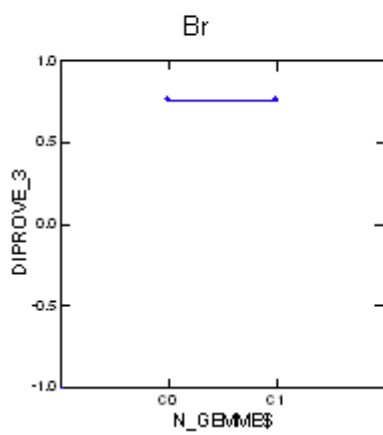
Least Squares Means



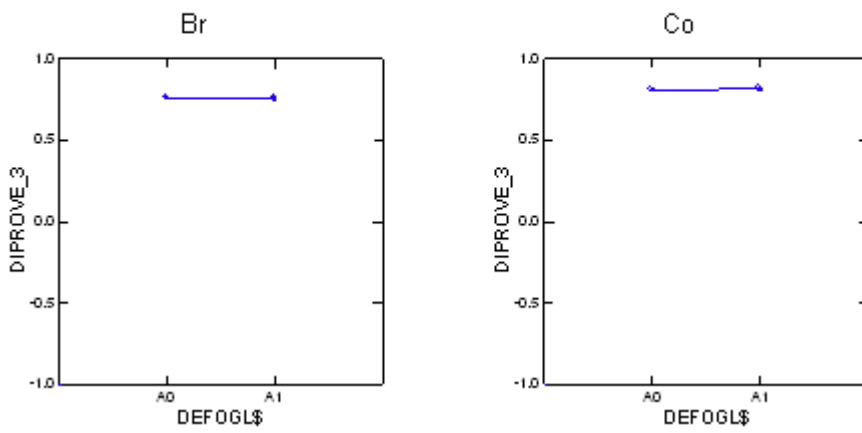
Least Squares Means



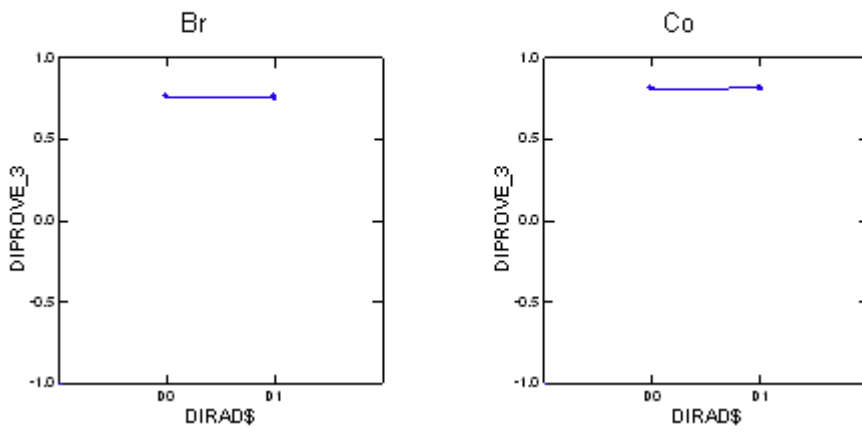
Least Squares Means



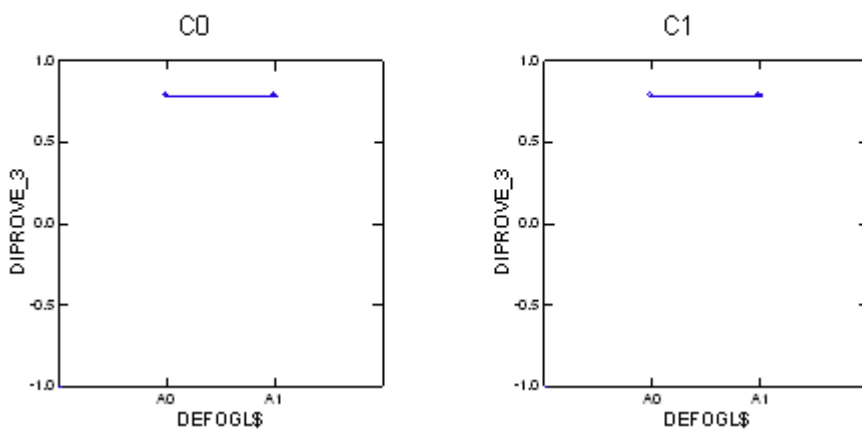
Least Squares Means



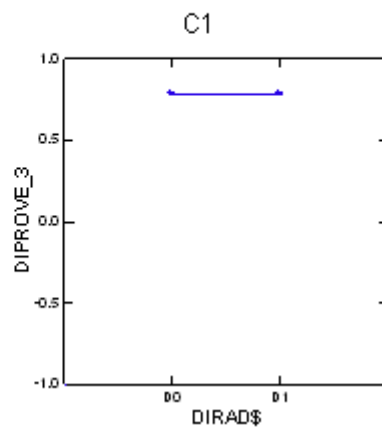
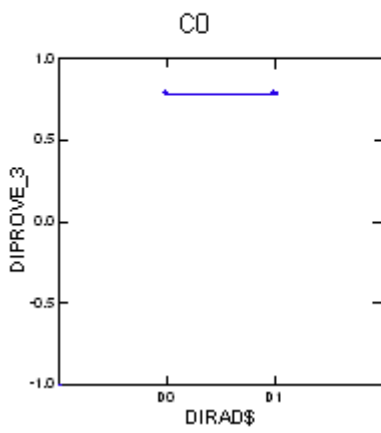
Least Squares Means



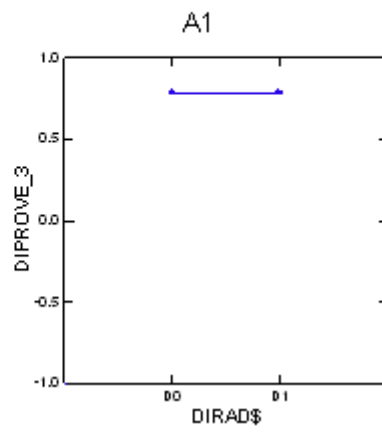
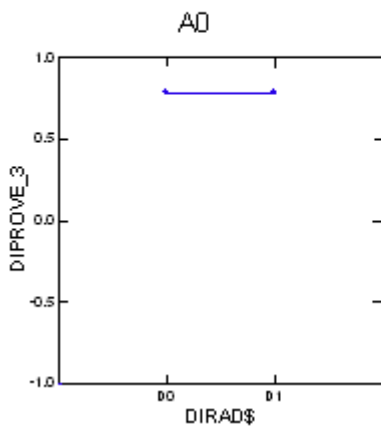
Least Squares Means



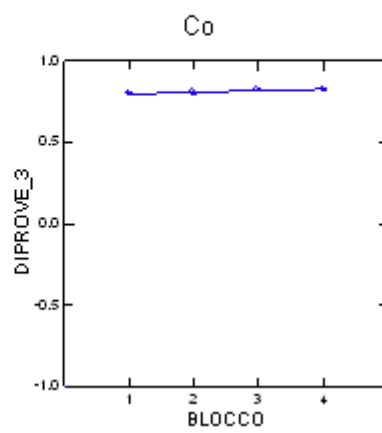
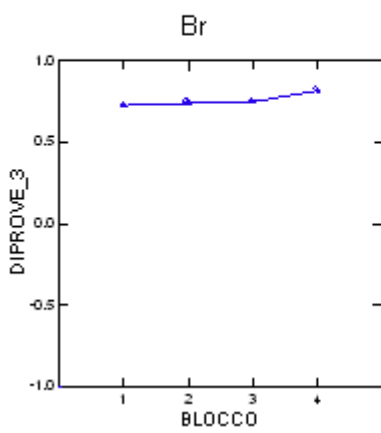
Least Squares Means



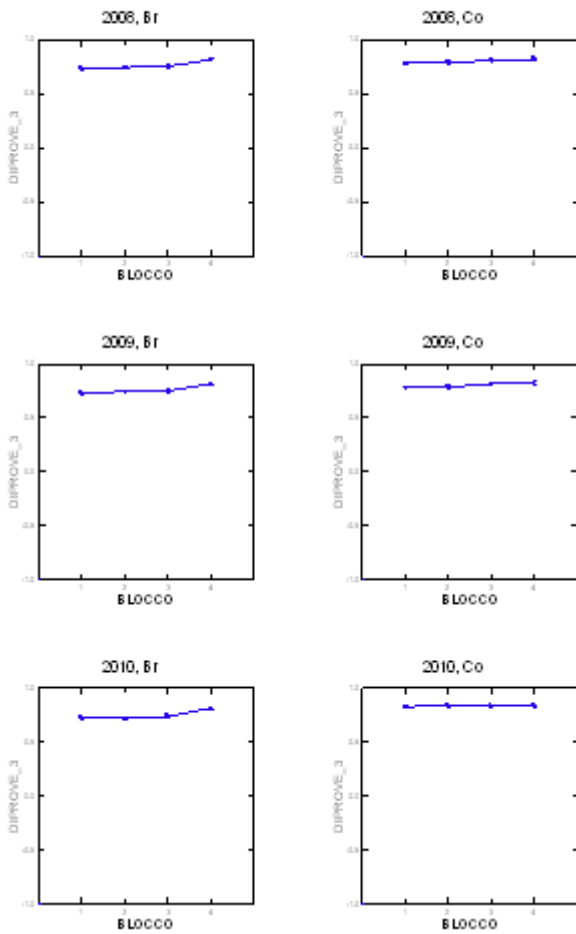
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 56 is an Outlier (Studentized Residual : 3.649)
Case 81 is an Outlier (Studentized Residual : -4.417)
Case 88 is an Outlier (Studentized Residual : 3.707)

Durbin-Watson D Statistic | 1.357
First Order Autocorrelation | 0.321

Information Criteria

AIC | -997.659
AIC (Corrected) | -975.937
Schwarz's BIC | -867.359

Allegato 2.2 – Anova Multiscala Cabernet

ANOVA PER VIGNETI CABERNET SAUVIGNON (CACCIAGRANDE E DONNA OLIMPIA) NEGLI ANNI 2008-2009-2010

▼ General Linear Model IBIMET 1 2009-2010

Data for the following results were selected according to
SELECT (VARIETA\$ = 'Cabernet') AND (ANNO <> 2008)

Effects coding used for categorical variables in model.
The categorical values encountered during processing are

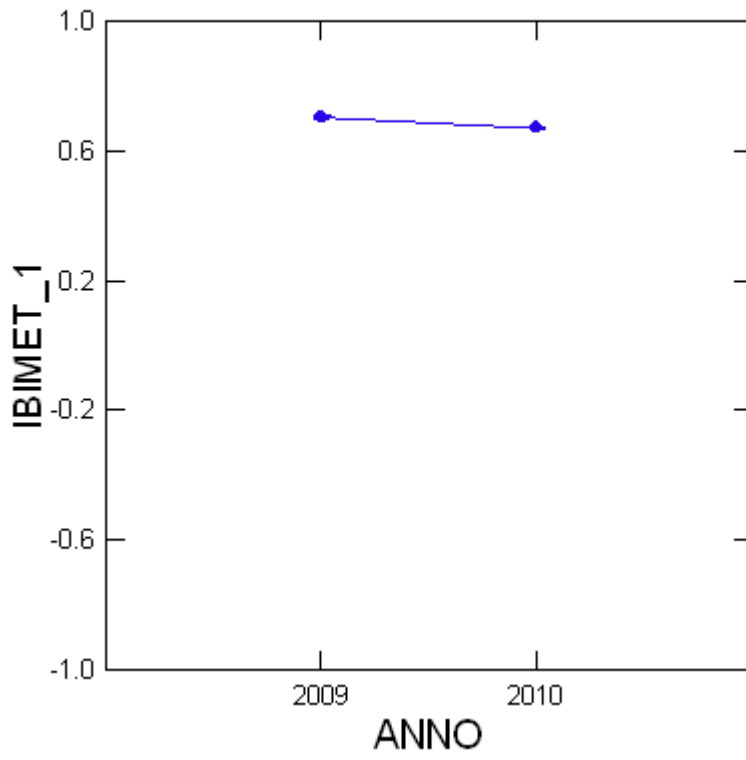
Variables	Levels		
ANNO (2 levels)	2,009.000	2,010.000	
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	IBIMET_1
N	96
Multiple R	0.979
Squared Multiple R	0.959

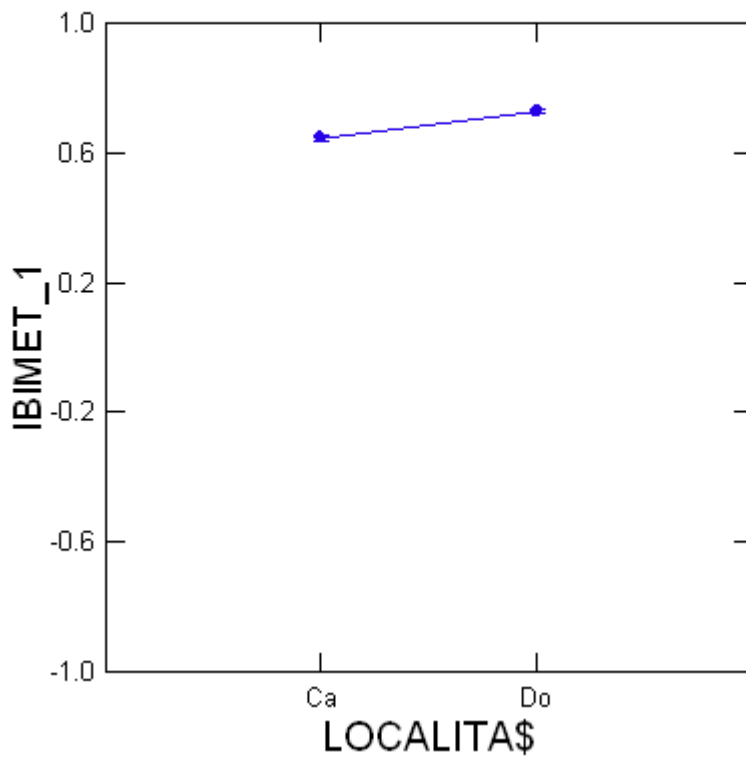
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.015	1	0.015	57.882	0.000
LOCALITA\$	0.095	1	0.095	357.575	0.000
N_GEMME\$	0.000	1	0.000	1.260	0.265
DEFOGL\$	0.006	1	0.006	21.270	0.000
DIRAD\$	0.001	1	0.001	2.820	0.097
LOCALITA\$*ANNO	0.026	1	0.026	100.018	0.000
N_GEMME\$*ANNO	0.000	1	0.000	0.185	0.668
DEFOGL\$*ANNO	0.002	1	0.002	7.555	0.008
DIRAD\$*ANNO	0.000	1	0.000	0.000	0.997
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.803	0.373
DEFOGL\$*LOCALITA\$	0.000	1	0.000	0.264	0.609
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.824	0.367
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.897	0.347
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.877	0.352
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.209	0.649
BLOCCO(LOCALITA\$)	0.158	4	0.040	149.122	0.000
BLOCCO*ANNO(LOCALITA\$)	0.024	4	0.006	22.574	0.000
Error	0.019	72	0.000		

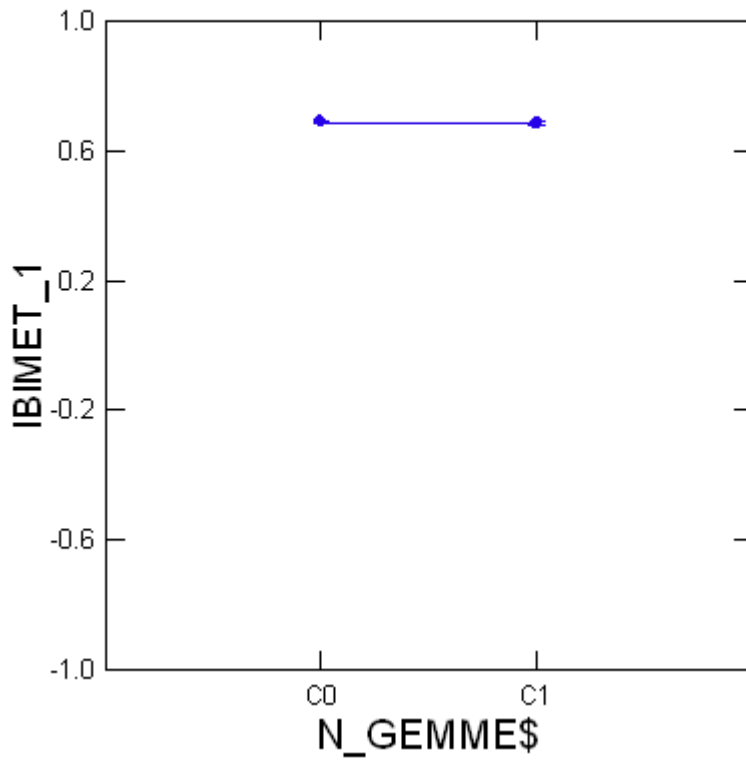
Least Squares Means



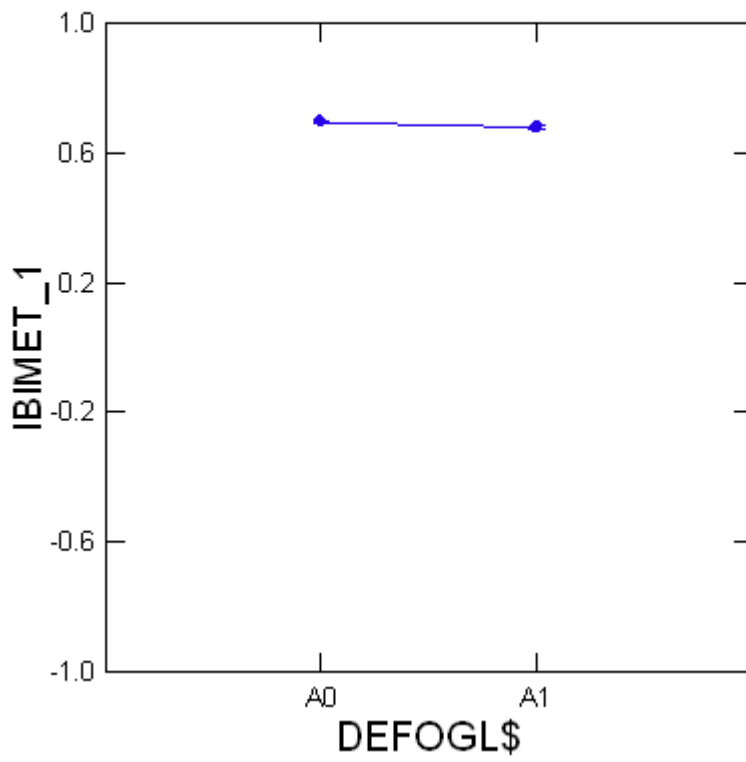
Least Squares Means



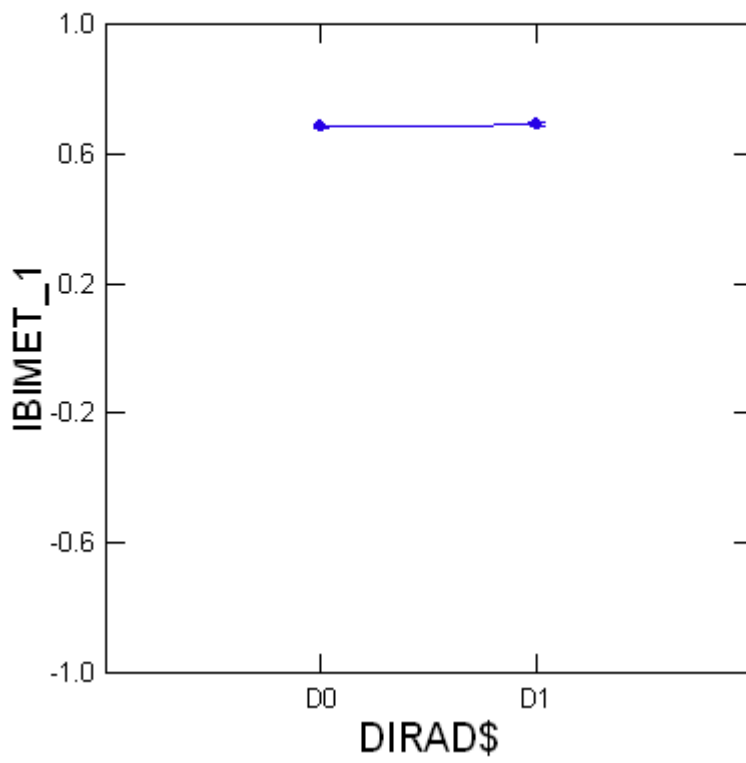
Least Squares Means



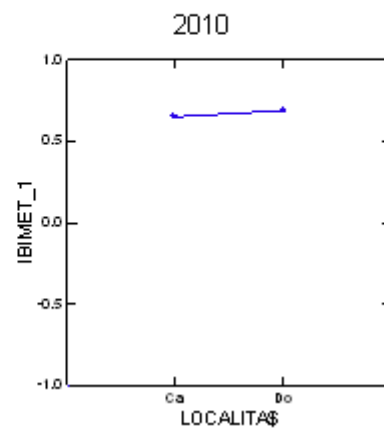
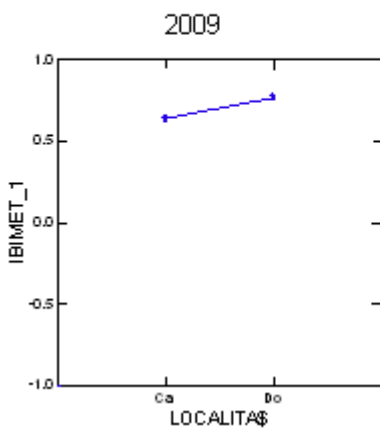
Least Squares Means



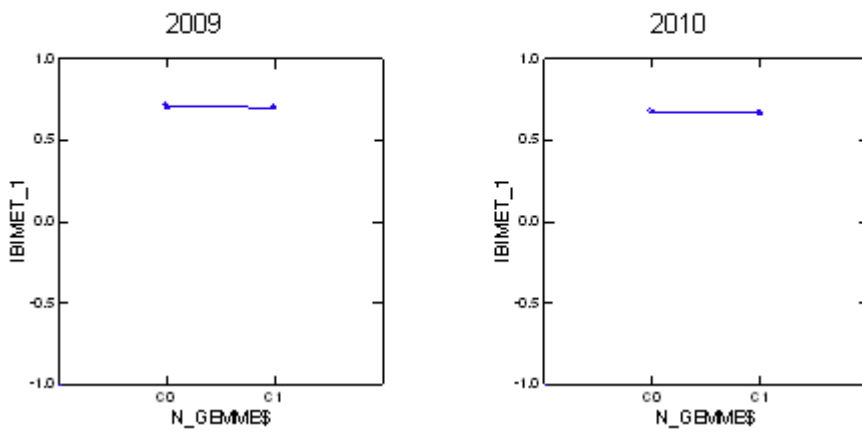
Least Squares Means



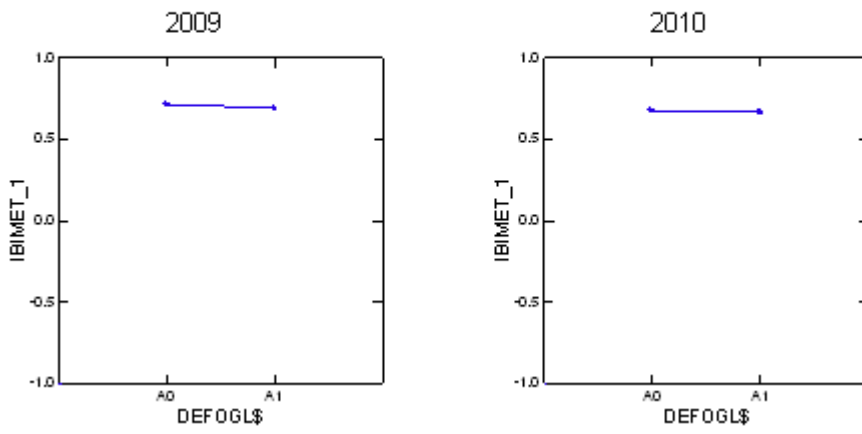
Least Squares Means



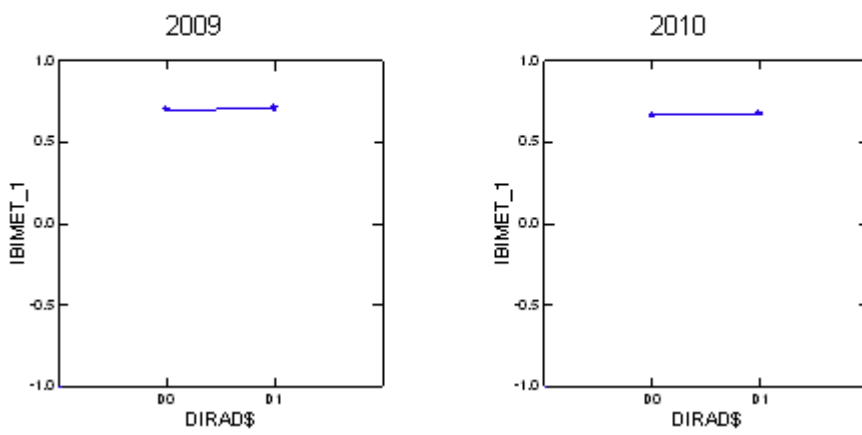
Least Squares Means



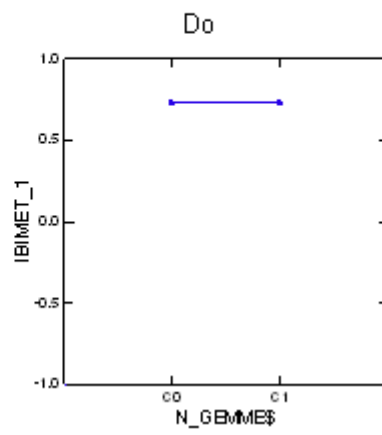
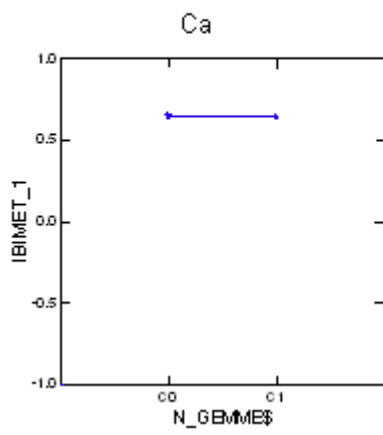
Least Squares Means



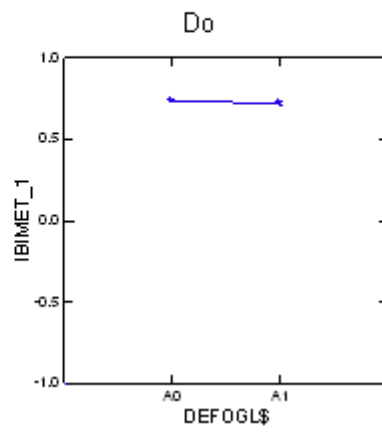
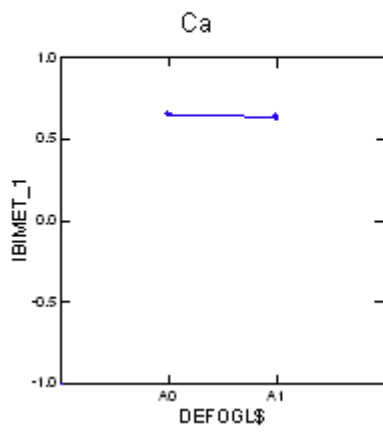
Least Squares Means



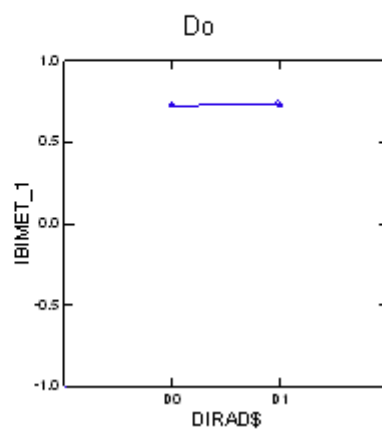
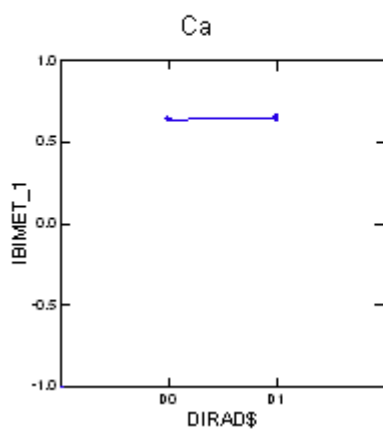
Least Squares Means



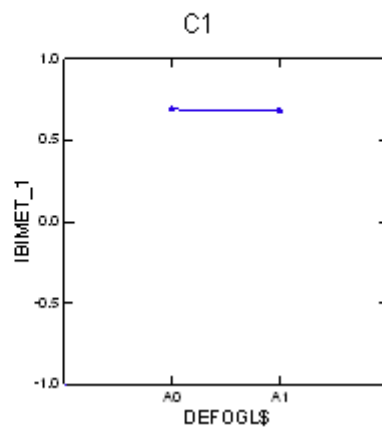
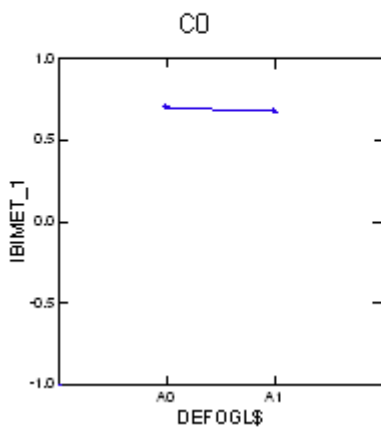
Least Squares Means



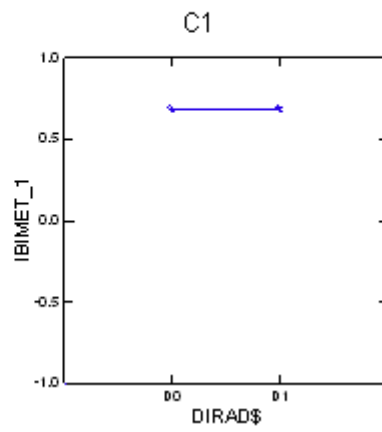
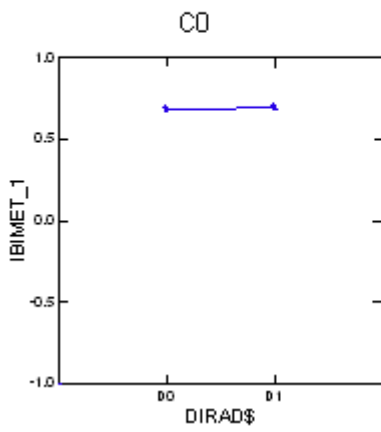
Least Squares Means



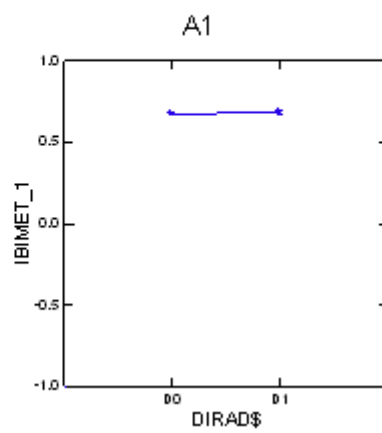
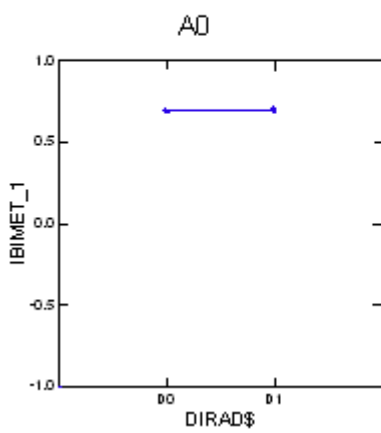
Least Squares Means



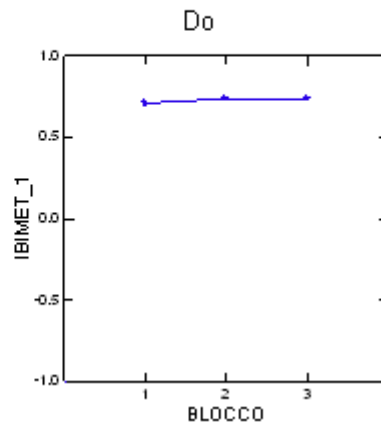
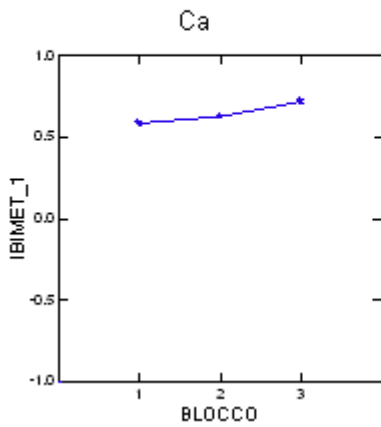
Least Squares Means



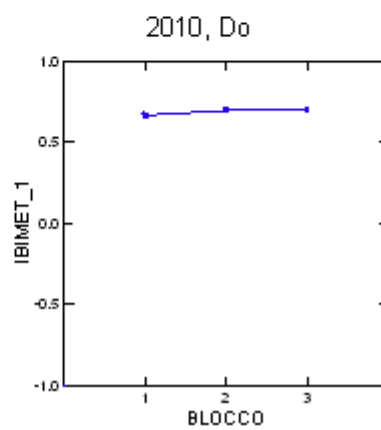
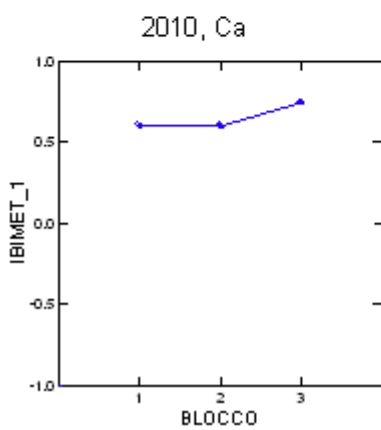
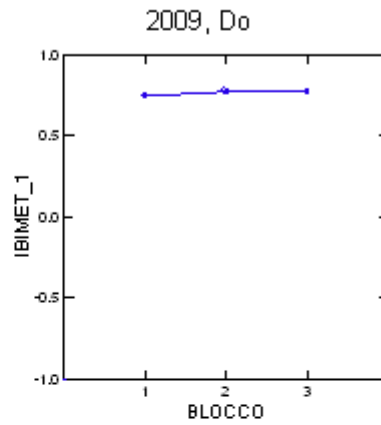
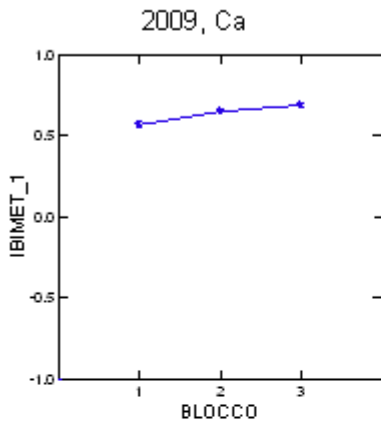
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.899
First Order Autocorrelation | 0.048

Information Criteria

AIC | -495.846
AIC (Corrected) | -477.274
Schwarz's BIC | -431.737

▼ General Linear Model
IBIMET2 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

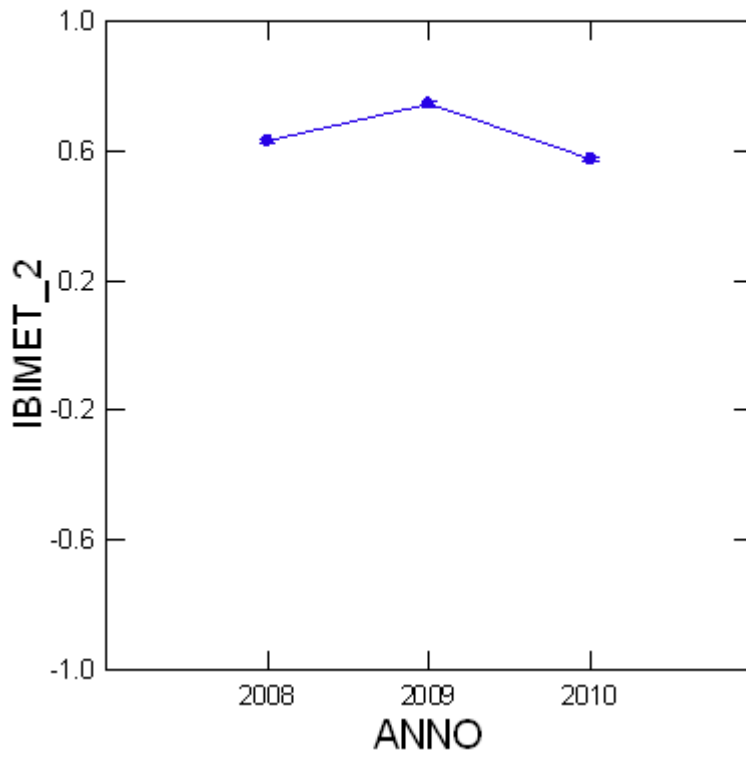
Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	IBIMET_2
N	144
Multiple R	0.992
Squared Multiple R	0.984

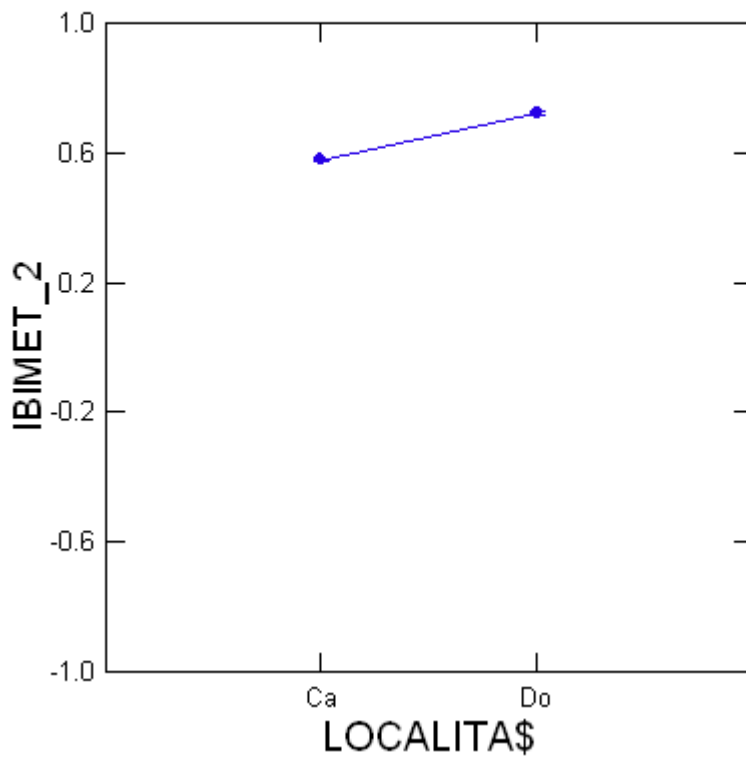
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.465	2	0.232	753.339	0.000
LOCALITA\$	0.572	1	0.572	1,853.964	0.000
N_GEMME\$	0.006	1	0.006	17.851	0.000
DEFOGL\$	0.000	1	0.000	0.352	0.554
DIRAD\$	0.000	1	0.000	0.029	0.864
LOCALITA\$*ANNO	0.243	2	0.122	394.122	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.274	0.761
DEFOGL\$*ANNO	0.001	2	0.001	1.977	0.143
DIRAD\$*ANNO	0.000	2	0.000	0.256	0.775
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.016	0.901
DEFOGL\$*LOCALITA\$	0.001	1	0.001	3.785	0.054
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.375	0.541
DEFOGL\$*N_GEMME\$	0.000	1	0.000	1.028	0.313
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.047	0.829
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.727	0.396
BLOCCO(LOCALITA\$)	0.225	4	0.056	182.201	0.000
BLOCCO*ANNO(LOCALITA\$)	0.025	8	0.003	10.130	0.000
Error	0.034	111	0.000		

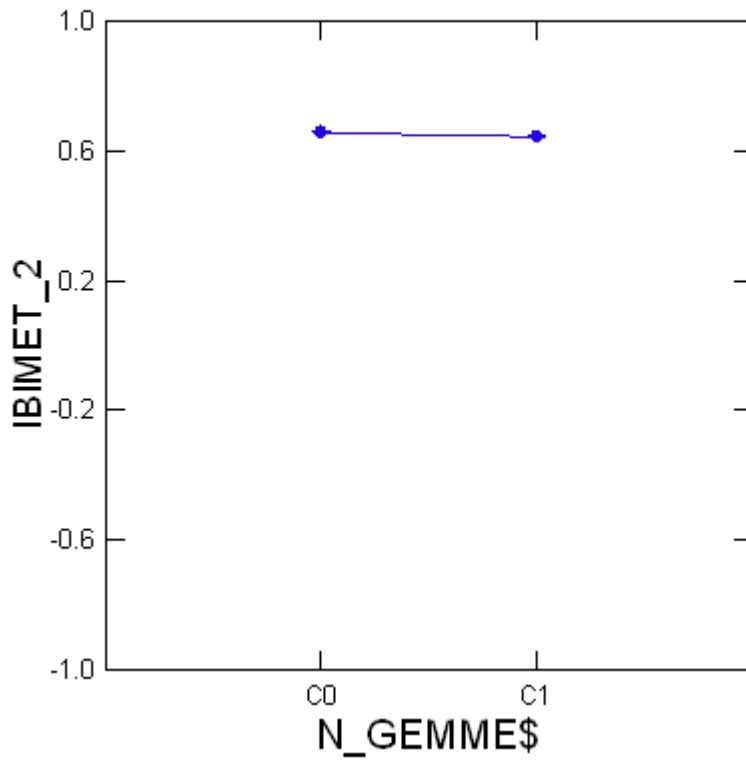
Least Squares Means



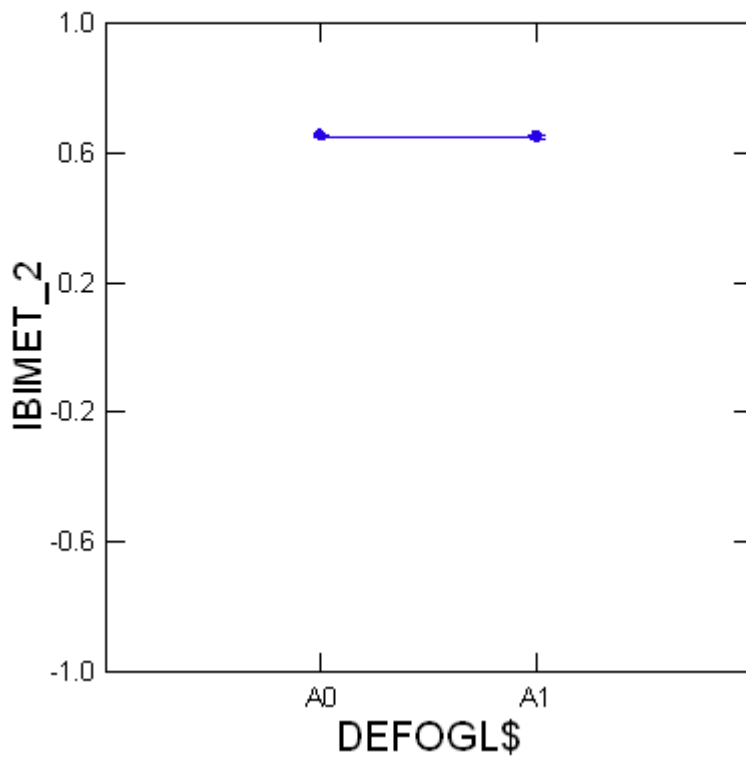
Least Squares Means



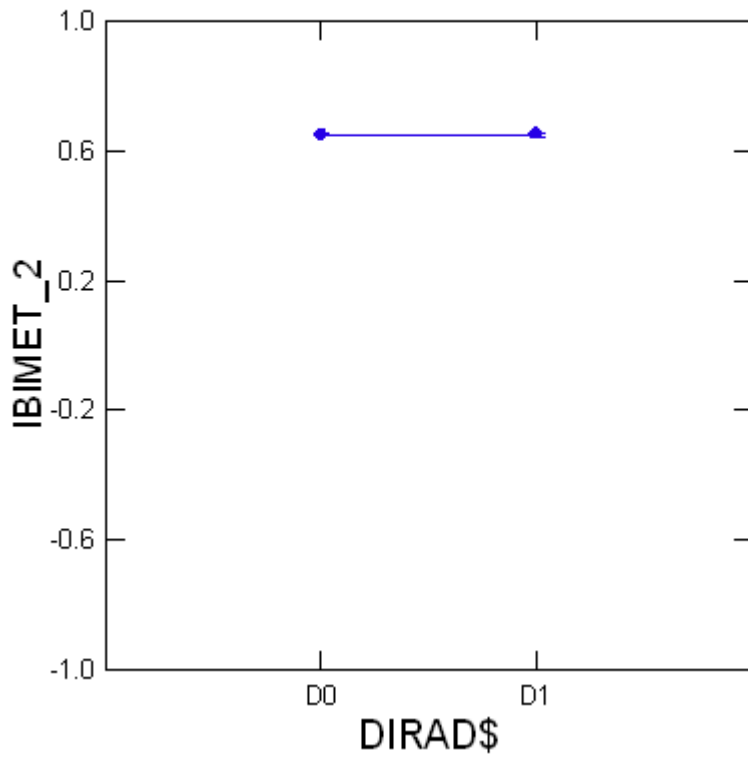
Least Squares Means



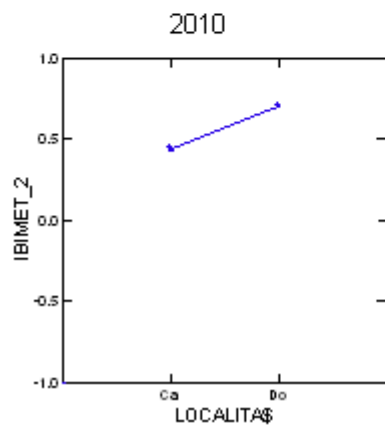
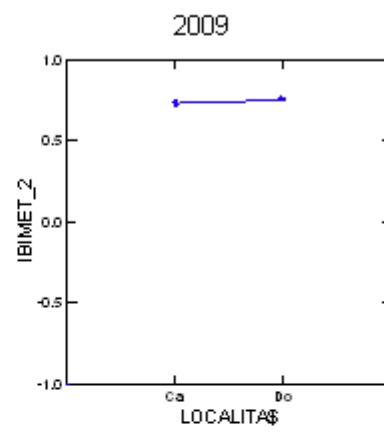
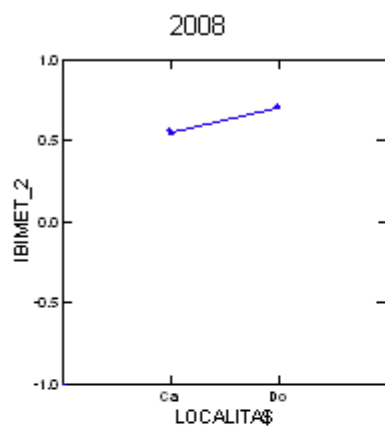
Least Squares Means



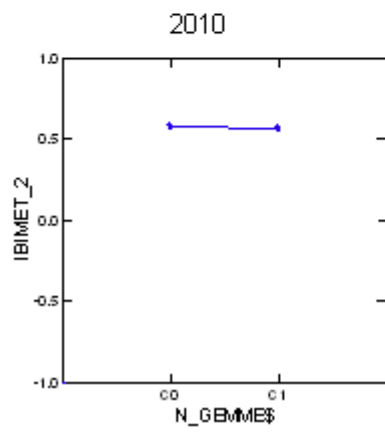
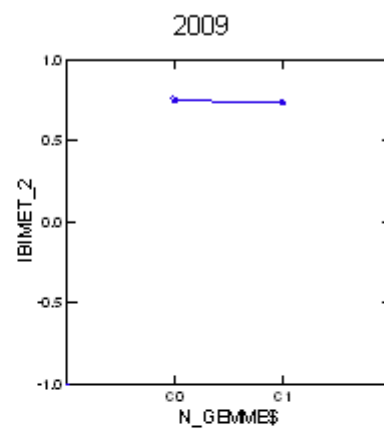
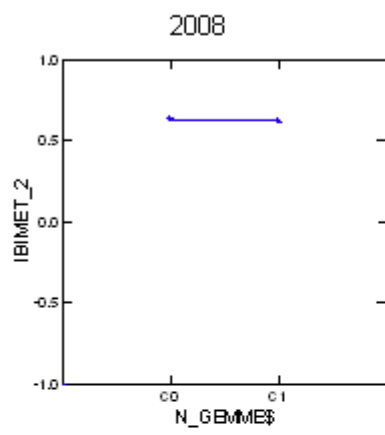
Least Squares Means



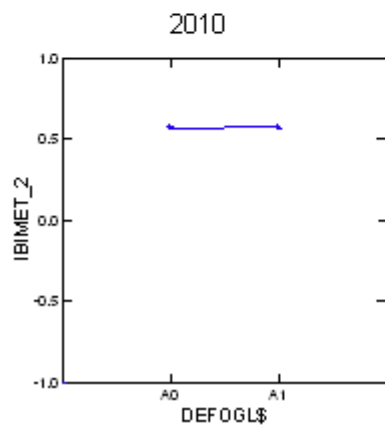
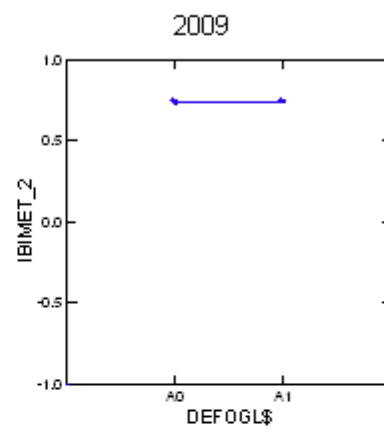
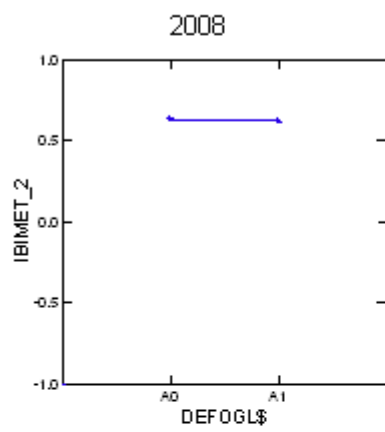
Least Squares Means



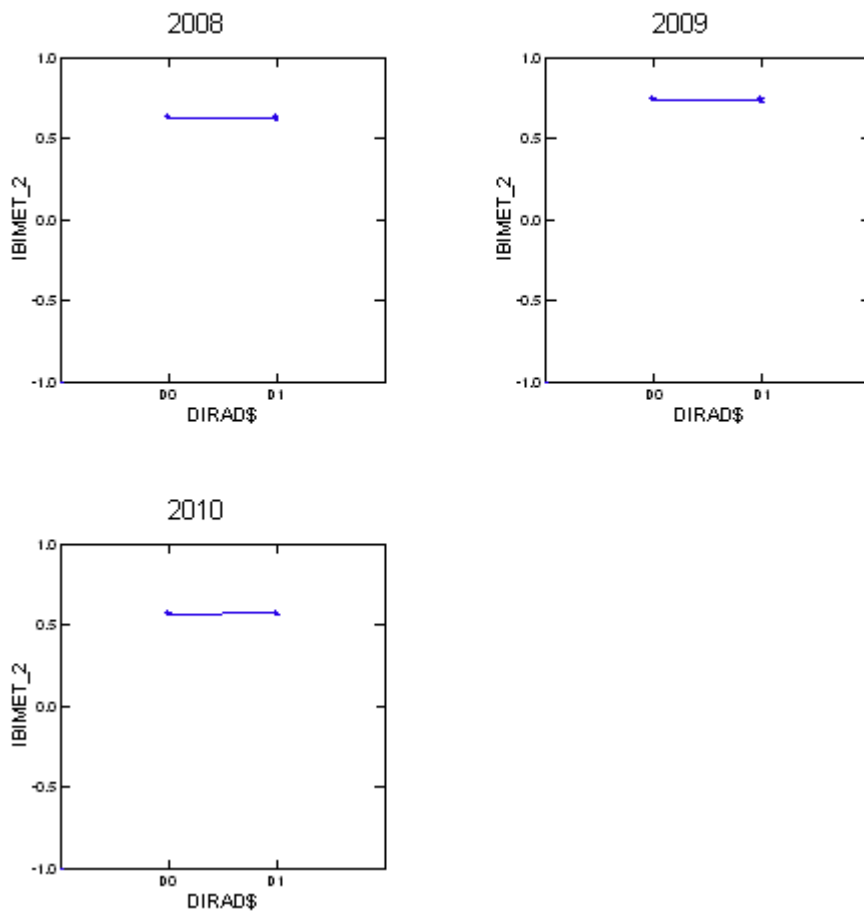
Least Squares Means



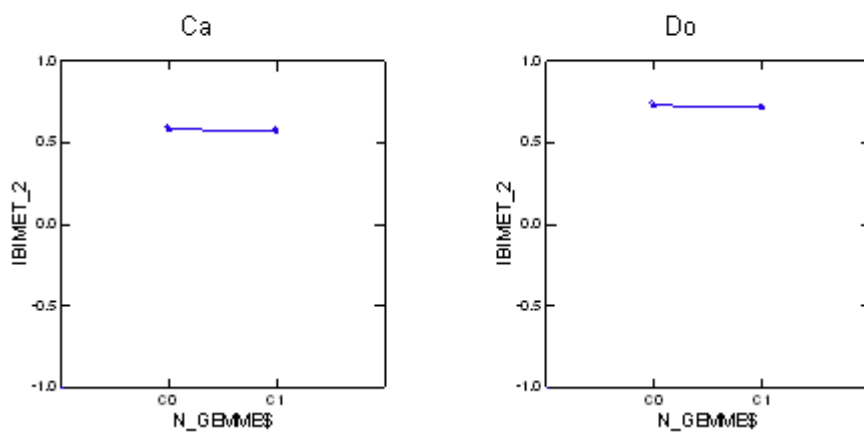
Least Squares Means



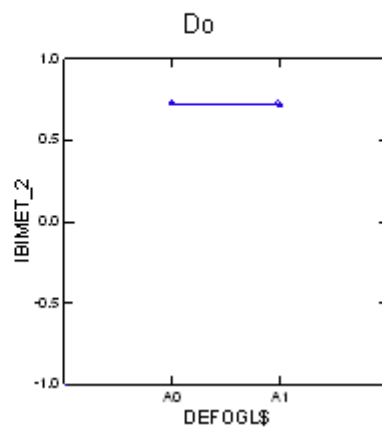
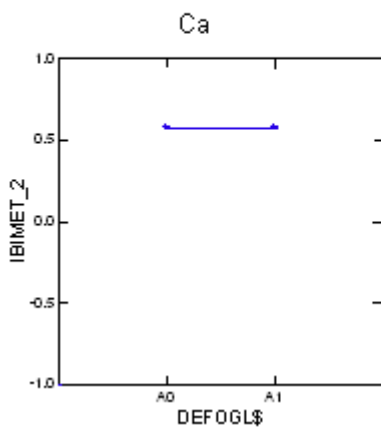
Least Squares Means



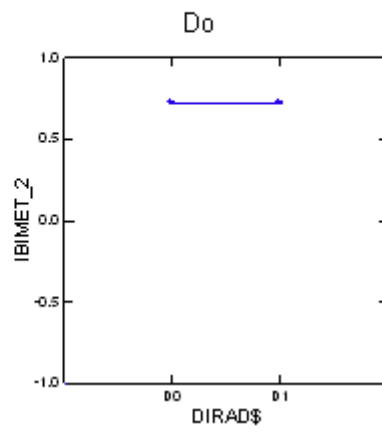
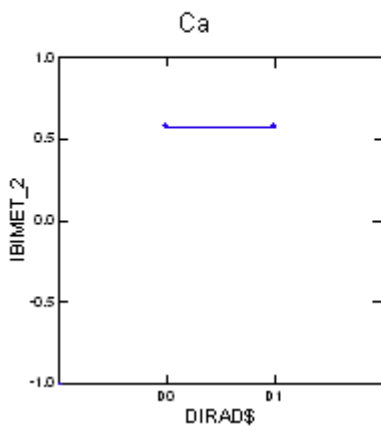
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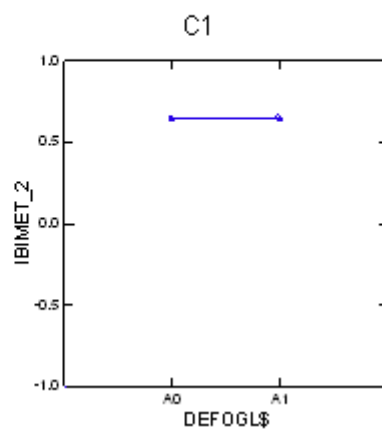
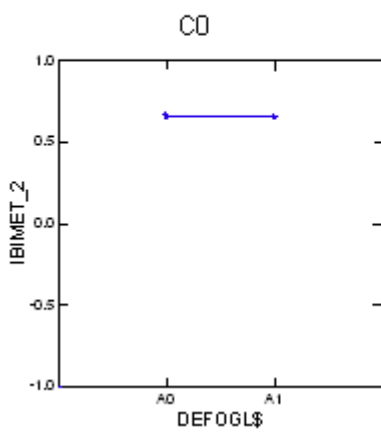
Least Squares Means



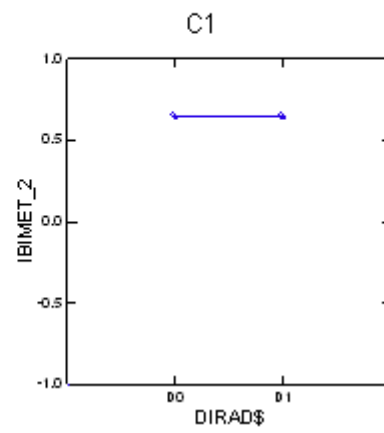
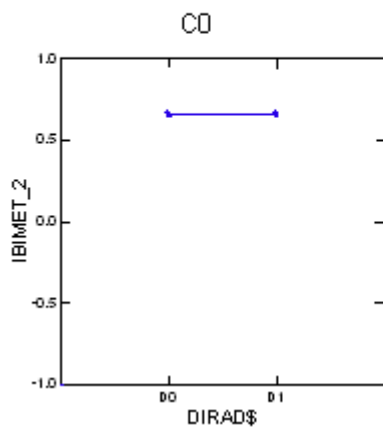
Least Squares Means



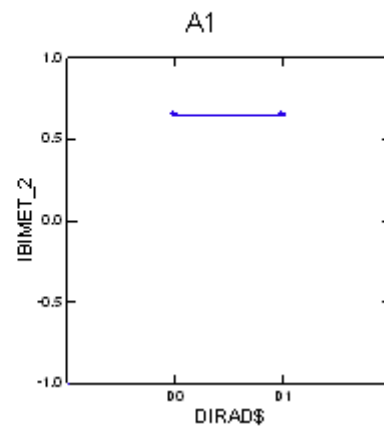
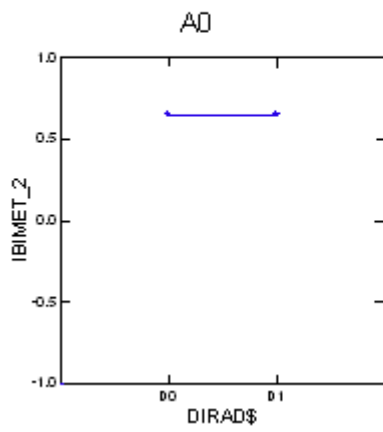
Least Squares Means



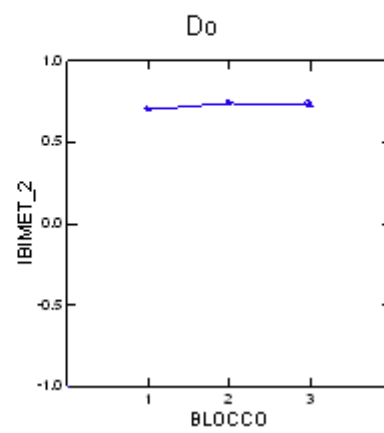
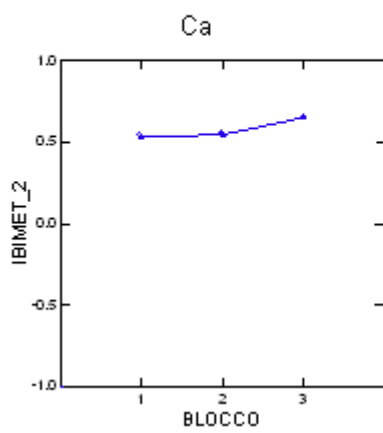
Least Squares Means



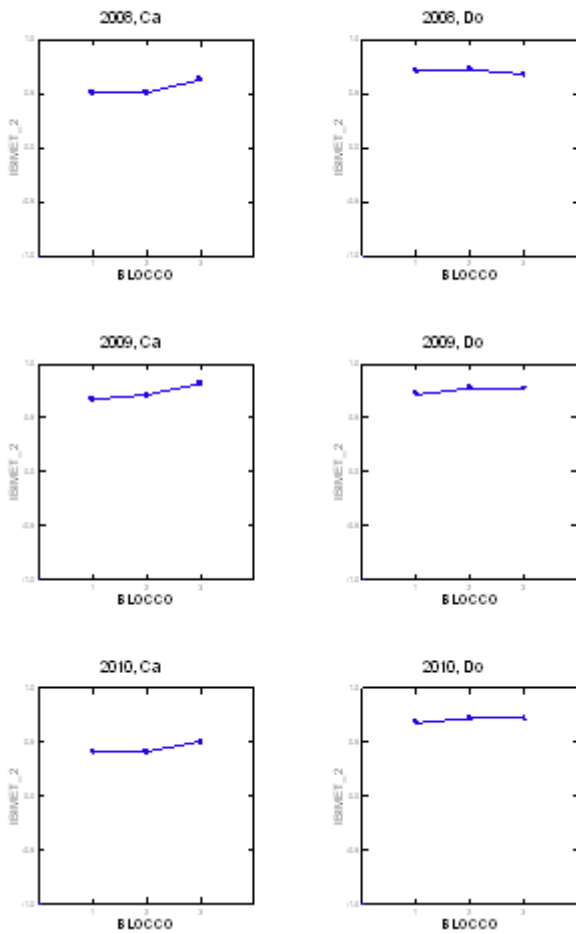
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 113 is an Outlier (Studentized Residual : -3.584)
 Case 118 is an Outlier (Studentized Residual : 3.504)

Durbin-Watson D Statistic | 1.181
 First Order Autocorrelation | 0.405

Information Criteria

AIC | -724.891
 AIC (Corrected) | -703.056
 Schwarz's BIC | -623.917

▼ General Linear Model IBIMET3 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels
ANNO (3 levels)	2,008.000 2,009.000 2,010.000

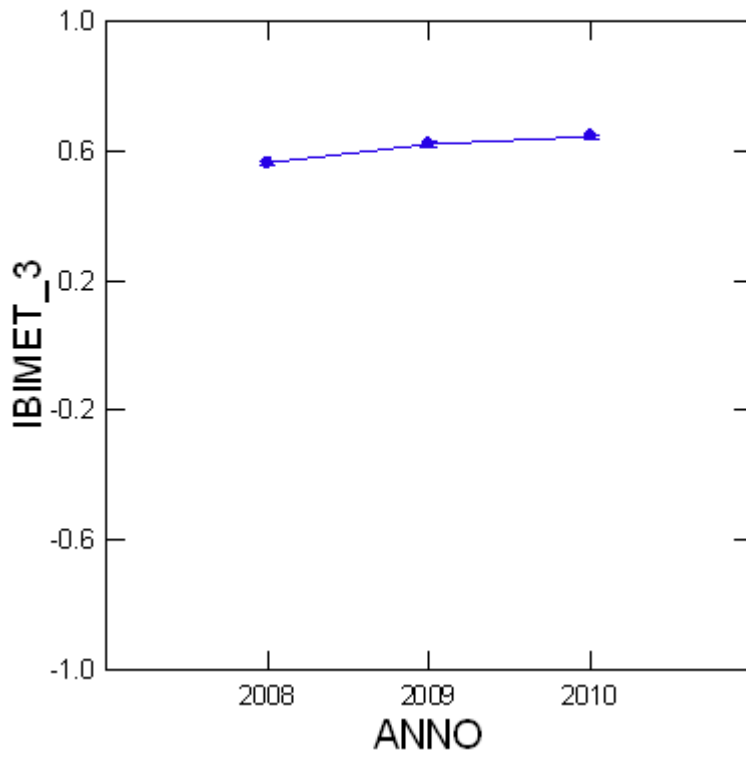
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	IBIMET_3
N	144
Multiple R	0.978
Squared Multiple R	0.956

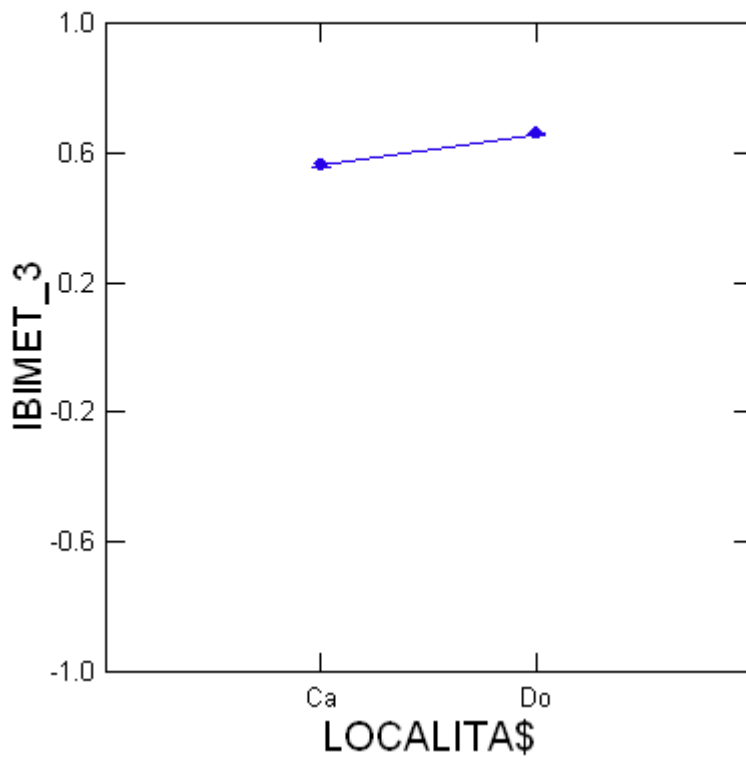
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.163	2	0.082	174.616	0.000
LOCALITA\$	0.239	1	0.239	512.030	0.000
N_GEMME\$	0.005	1	0.005	10.727	0.001
DEFOGL\$	0.000	1	0.000	0.807	0.371
DIRAD\$	0.002	1	0.002	4.966	0.028
LOCALITA\$*ANNO	0.258	2	0.129	275.576	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.074	0.929
DEFOGL\$*ANNO	0.001	2	0.001	1.325	0.270
DIRAD\$*ANNO	0.000	2	0.000	0.176	0.839
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.159	0.690
DEFOGL\$*LOCALITA\$	0.000	1	0.000	0.286	0.594
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.382	0.538
DEFOGL\$*N_GEMME\$	0.001	1	0.001	2.107	0.149
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.801	0.373
DIRAD\$*DEFOGL\$	0.001	1	0.001	2.955	0.088
BLOCCO (LOCALITA\$)	0.267	4	0.067	142.776	0.000
BLOCCO*ANNO (LOCALITA\$)	0.026	8	0.003	7.024	0.000
Error	0.052	111	0.000		

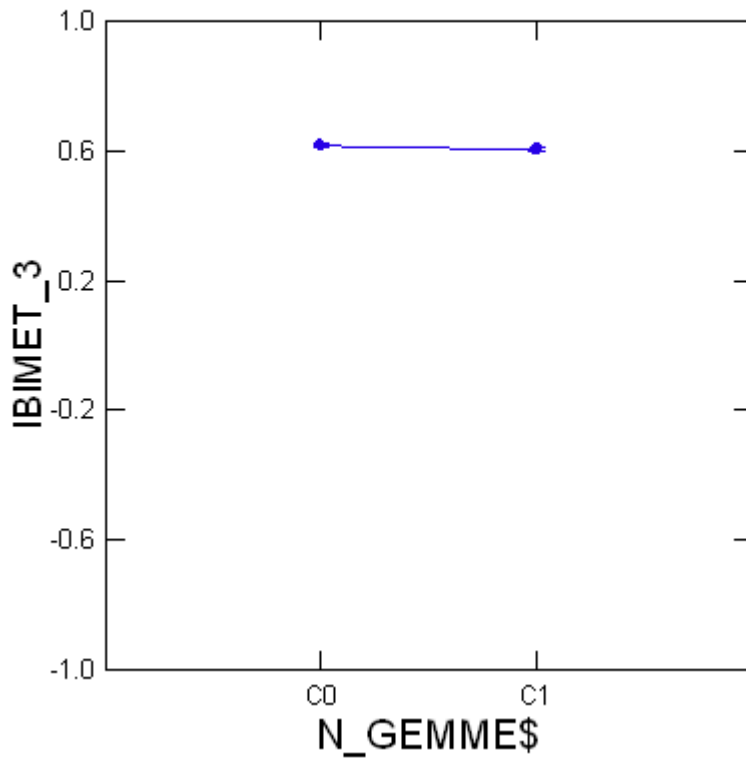
Least Squares Means



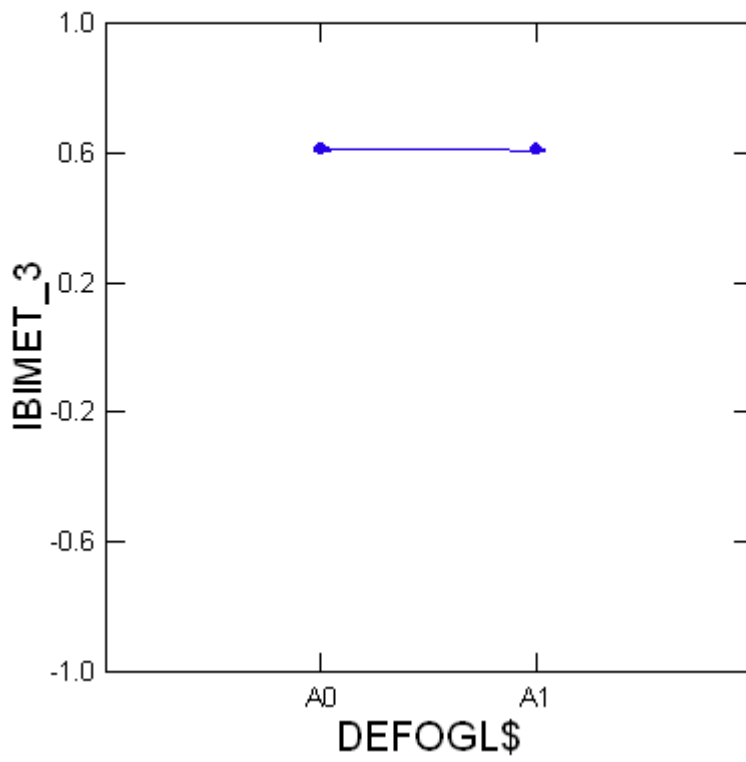
Least Squares Means



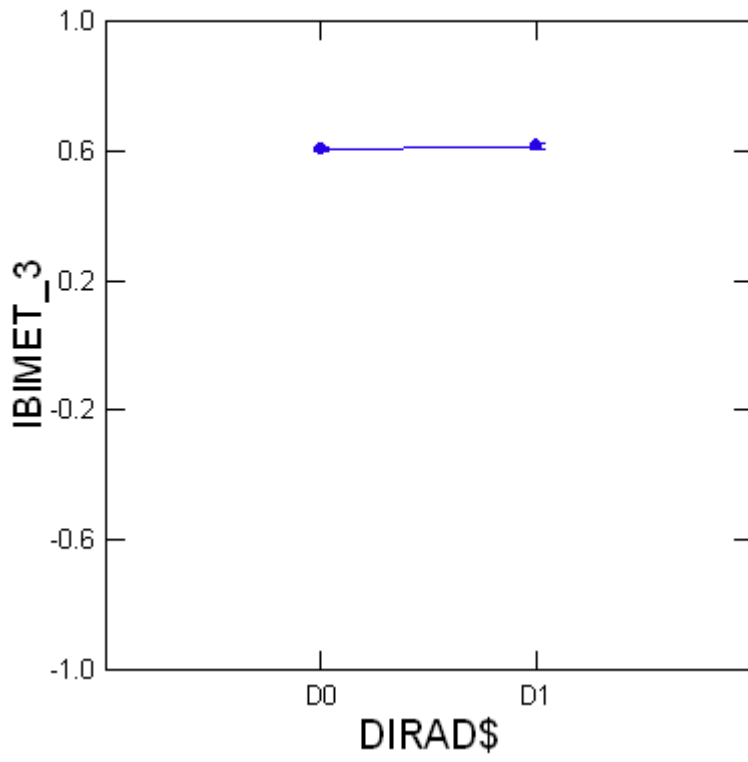
Least Squares Means



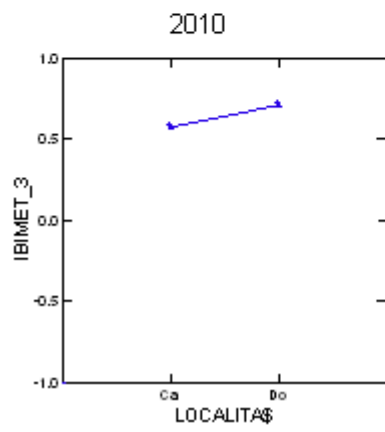
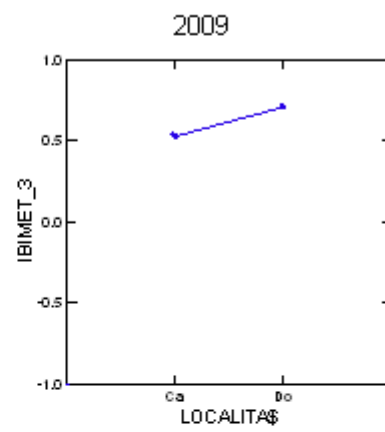
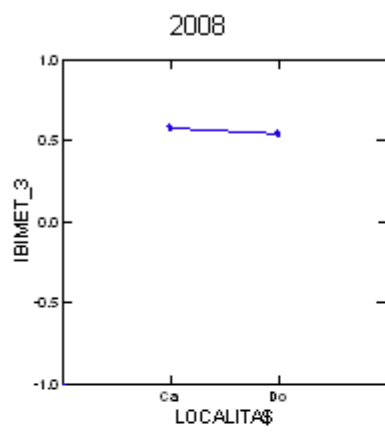
Least Squares Means



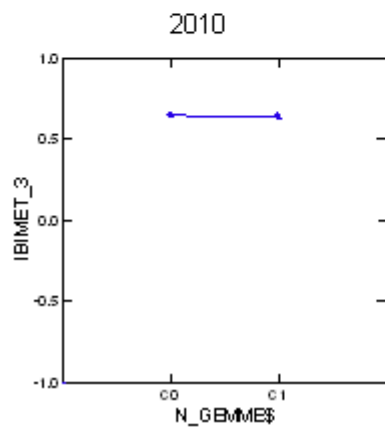
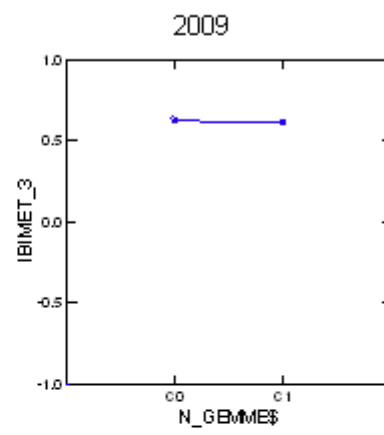
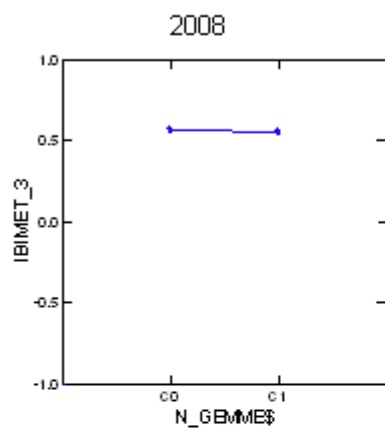
Least Squares Means



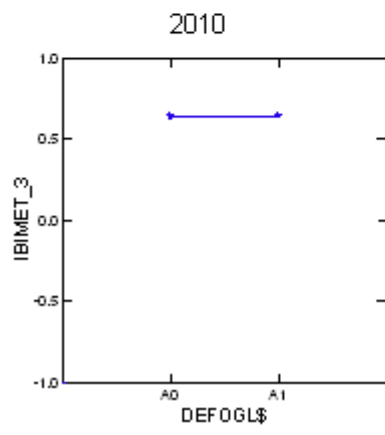
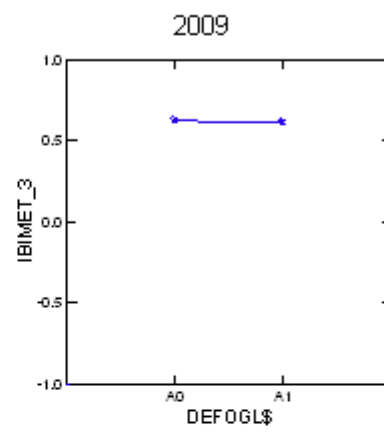
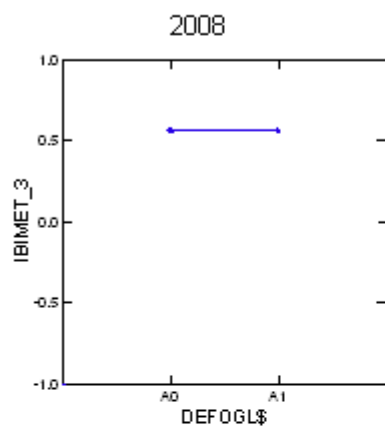
Least Squares Means



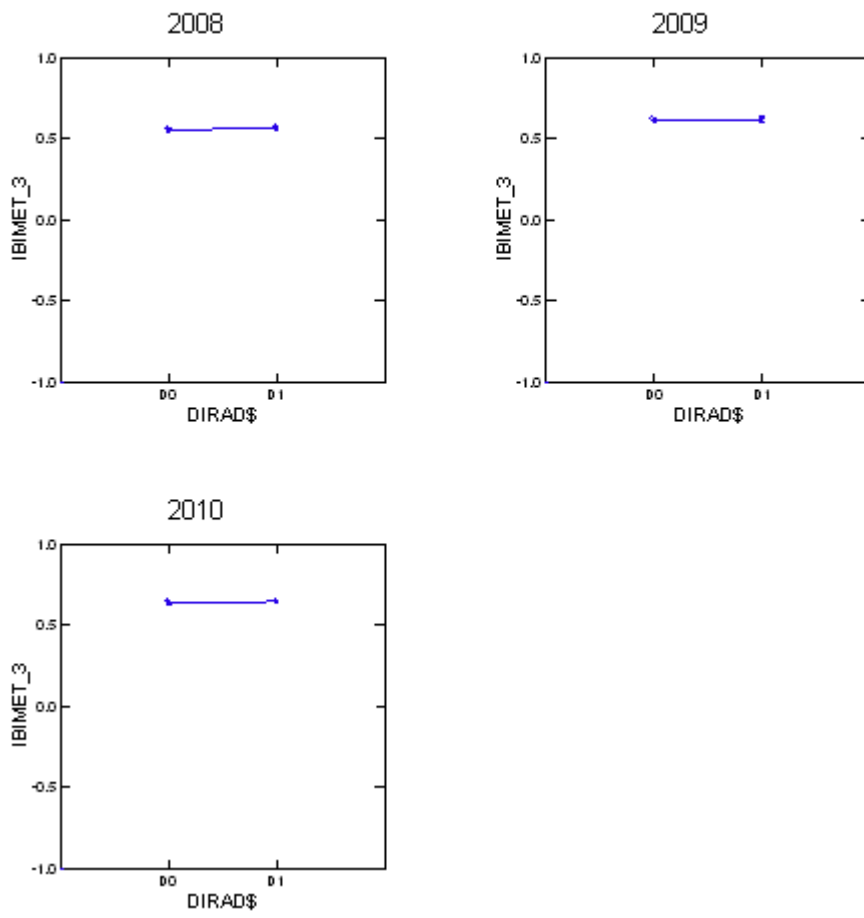
Least Squares Means



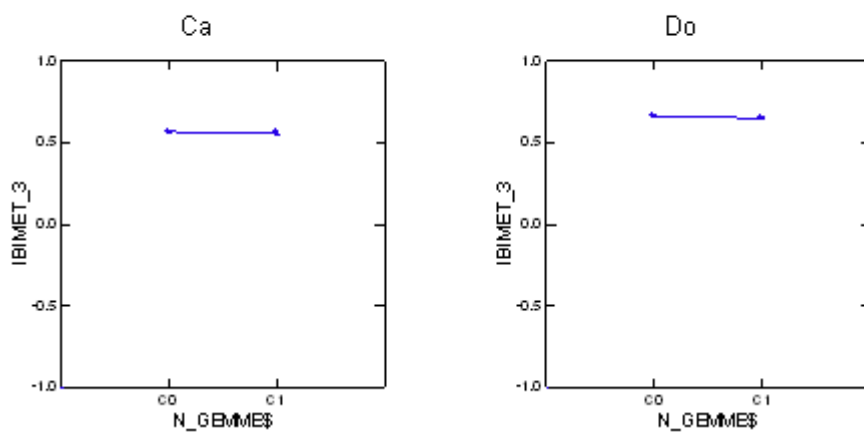
Least Squares Means



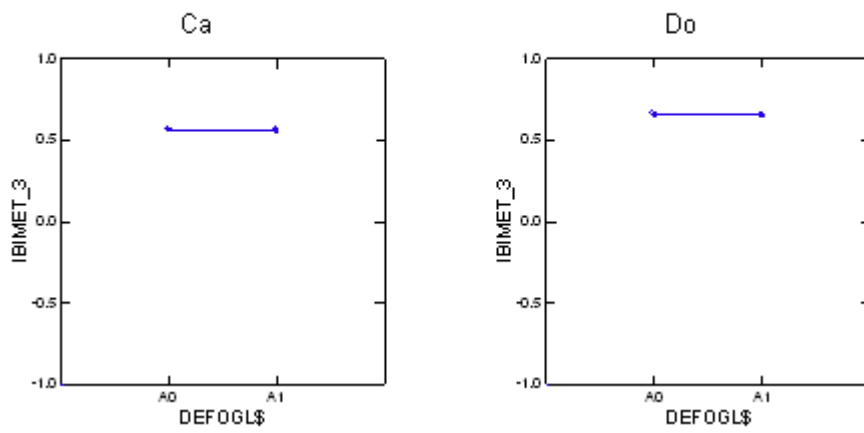
Least Squares Means



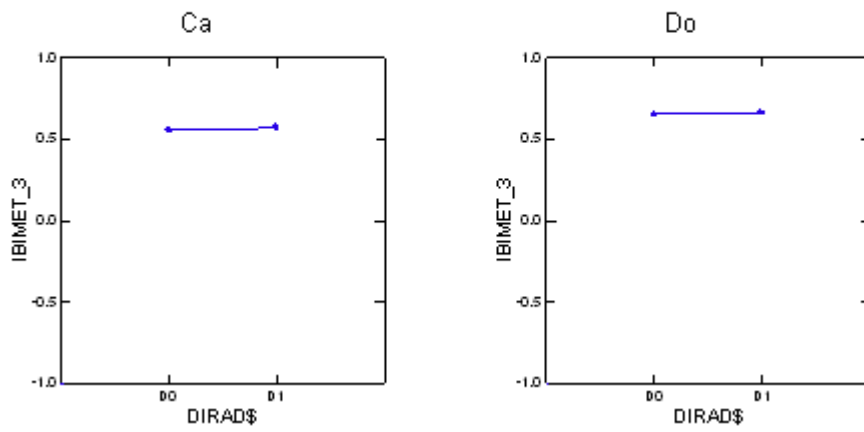
Least Squares Means



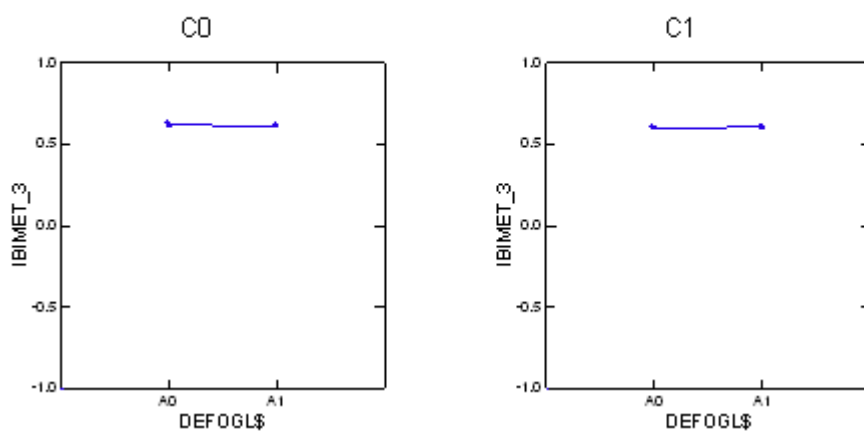
Least Squares Means



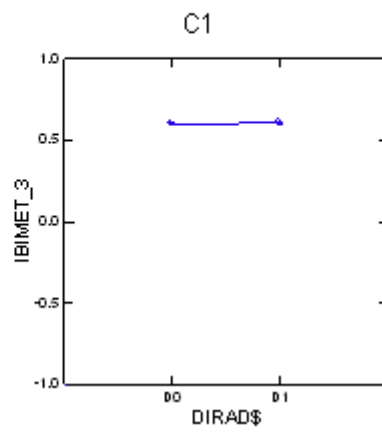
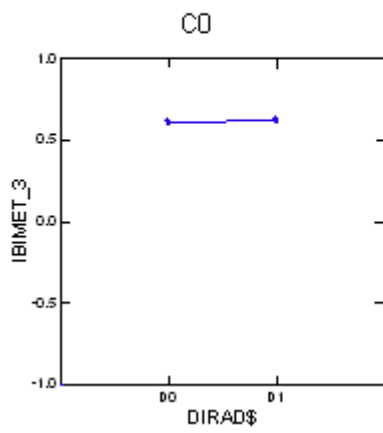
Least Squares Means



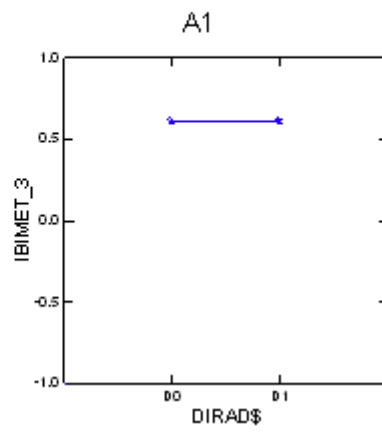
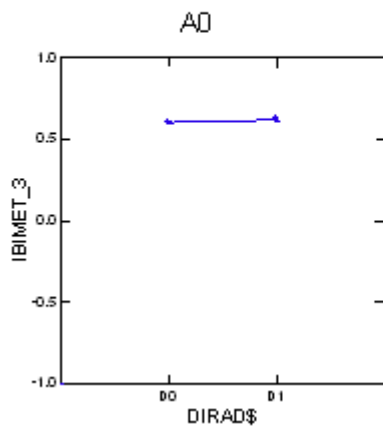
Least Squares Means



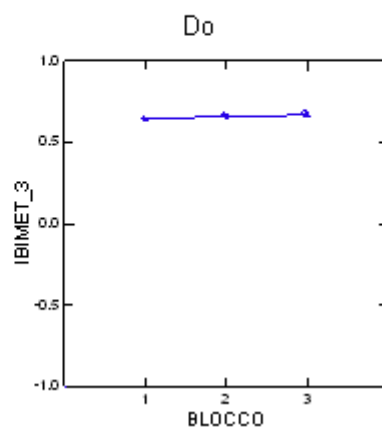
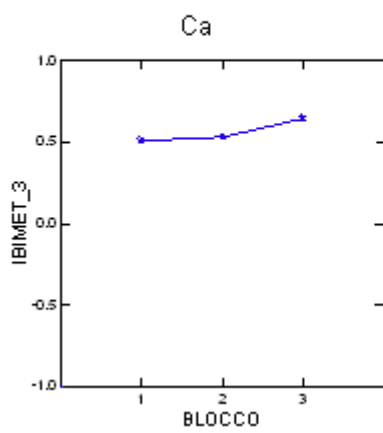
Least Squares Means



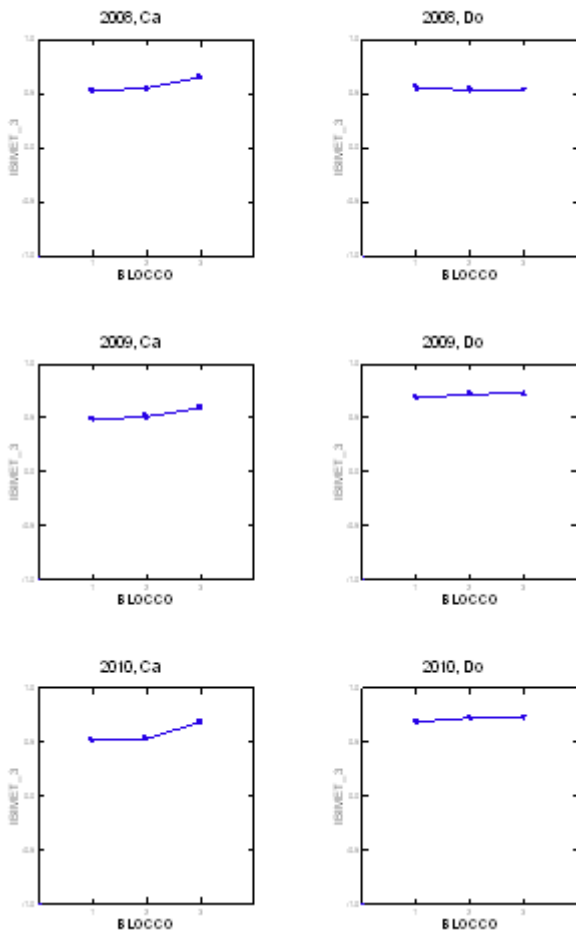
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.271
 First Order Autocorrelation | 0.360

Information Criteria

AIC | -665.037
 AIC (Corrected) | -643.203
 Schwarz's BIC | -564.064

▼ General Linear Model IASMA1 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

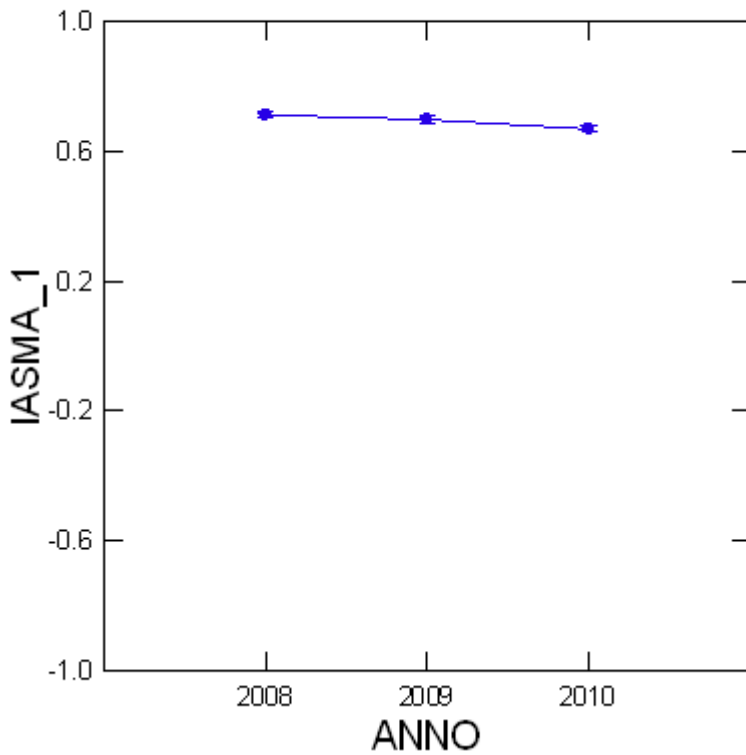
Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable | IASMA_1
 N | 144
 Multiple R | 0.808
 Squared Multiple R | 0.653

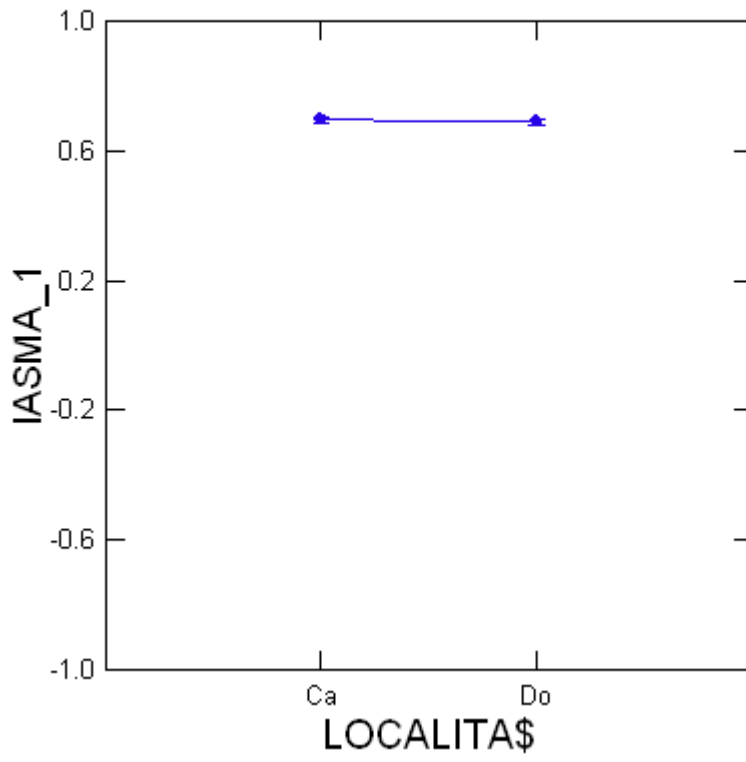
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.047	2	0.024	20.819	0.000
LOCALITA\$	0.002	1	0.002	1.565	0.214
N_GEMME\$	0.007	1	0.007	6.163	0.015
DEFOGL\$	0.006	1	0.006	4.961	0.028
DIRAD\$	0.000	1	0.000	0.226	0.636
LOCALITA\$*ANNO	0.053	2	0.026	23.248	0.000
N_GEMME\$*ANNO	0.004	2	0.002	1.763	0.176
DEFOGL\$*ANNO	0.001	2	0.001	0.520	0.596
DIRAD\$*ANNO	0.000	2	0.000	0.004	0.996
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.131	0.718
DEFOGL\$*LOCALITA\$	0.000	1	0.000	0.087	0.769
DIRAD\$*LOCALITA\$	0.005	1	0.005	4.270	0.041
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.130	0.719
DIRAD\$*N_GEMME\$	0.002	1	0.002	2.161	0.144
DIRAD\$*DEFOGL\$	0.002	1	0.002	2.118	0.148
BLOCCO(LOCALITA\$)	0.039	4	0.010	8.616	0.000
BLOCCO*ANNO(LOCALITA\$)	0.024	8	0.003	2.706	0.009
Error	0.126	111	0.001		

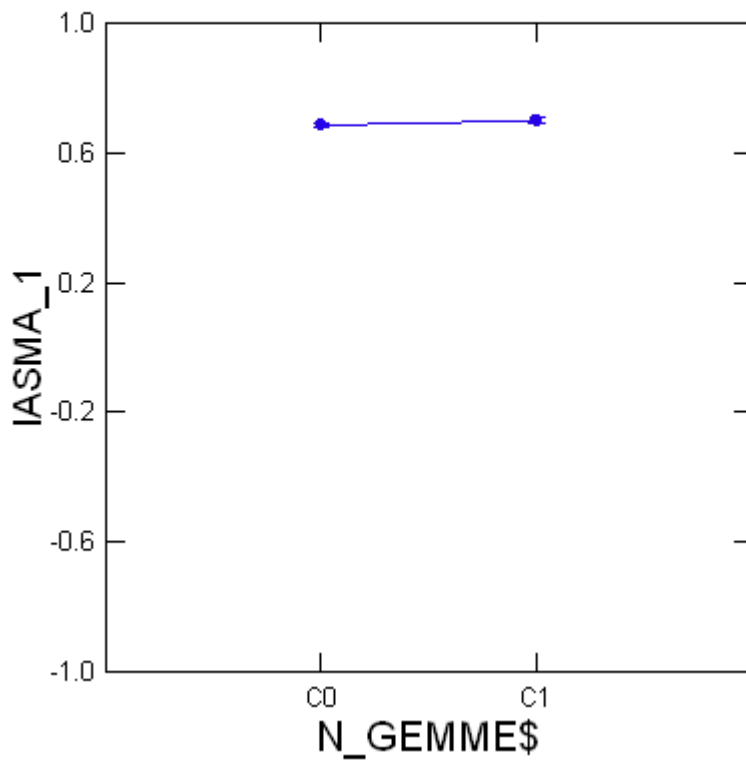
Least Squares Means



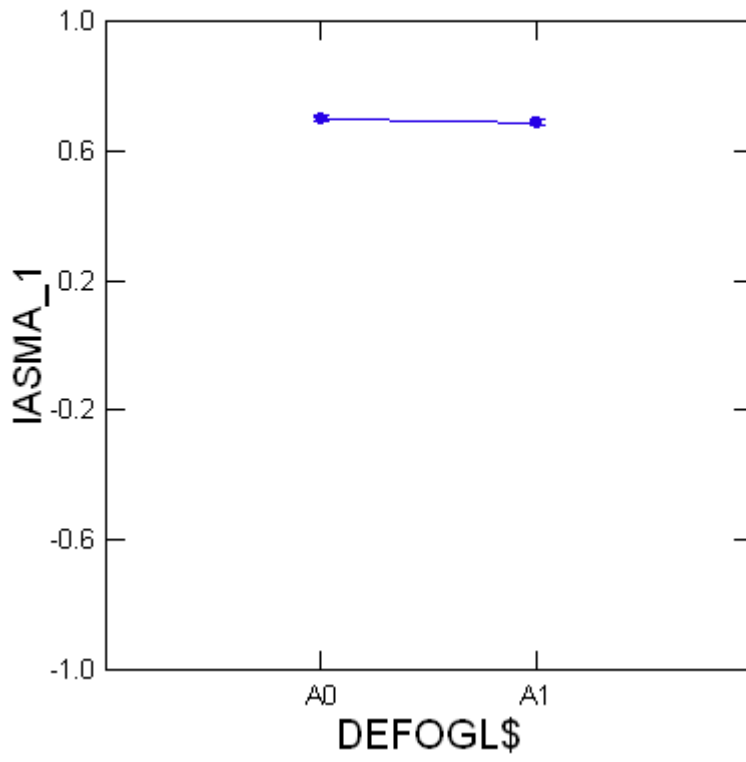
Least Squares Means



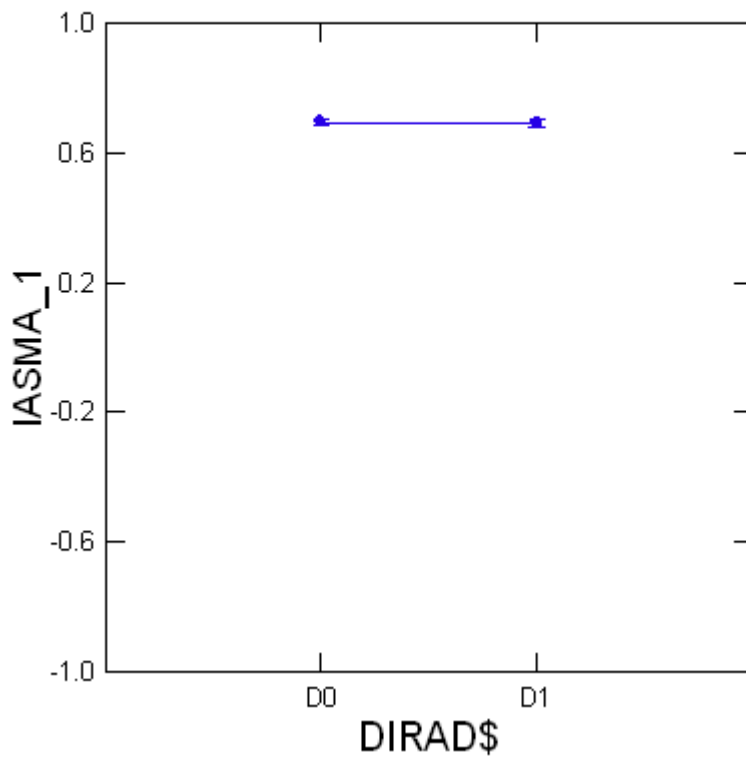
Least Squares Means



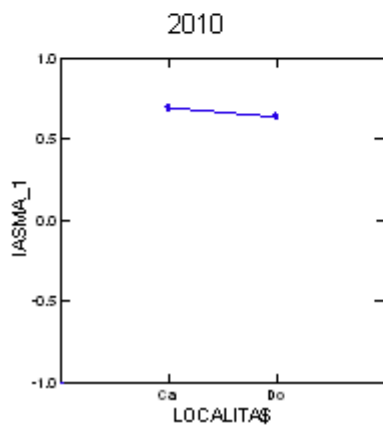
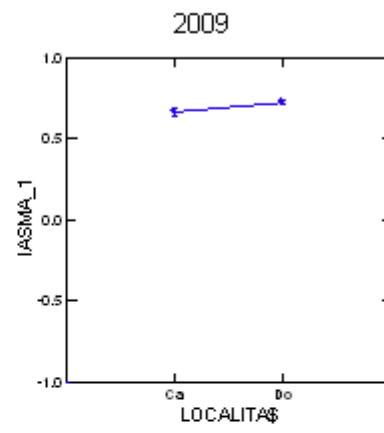
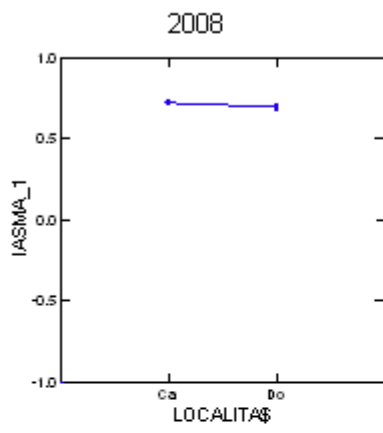
Least Squares Means



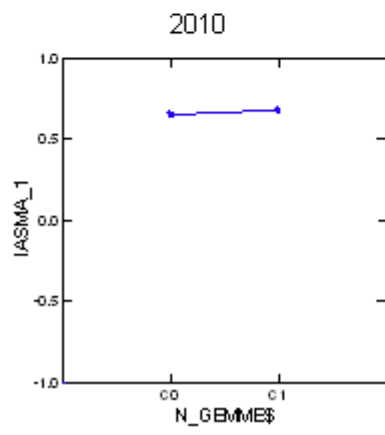
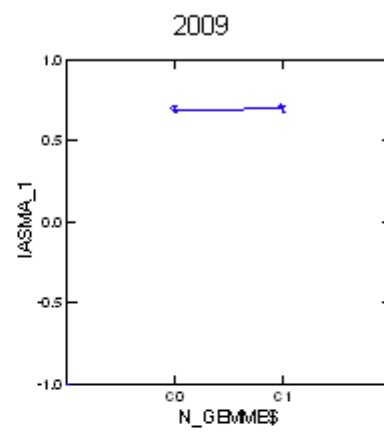
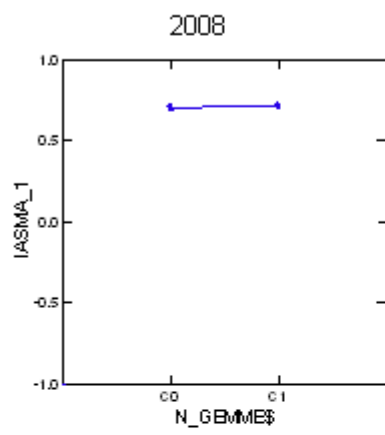
Least Squares Means



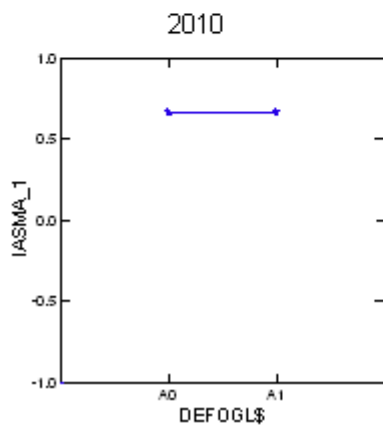
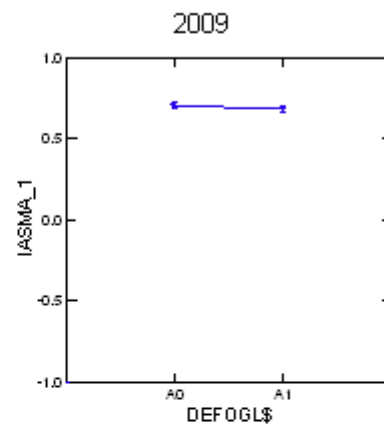
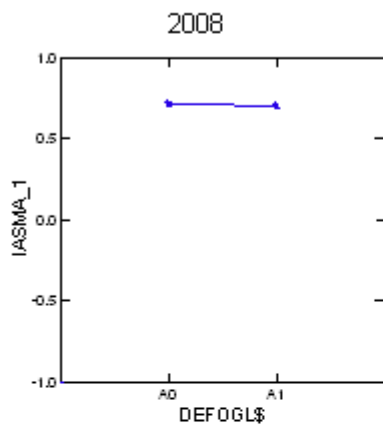
Least Squares Means



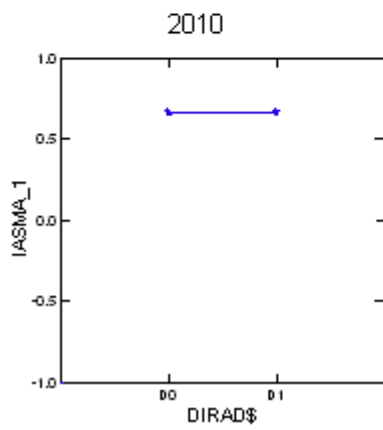
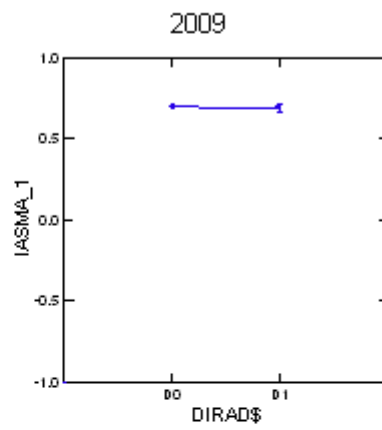
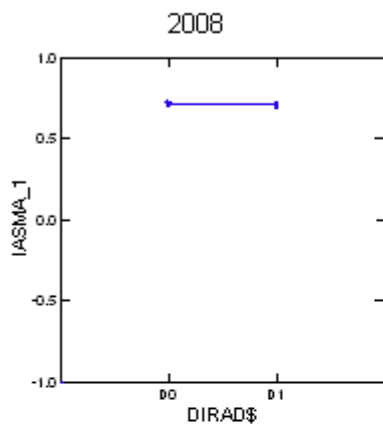
Least Squares Means



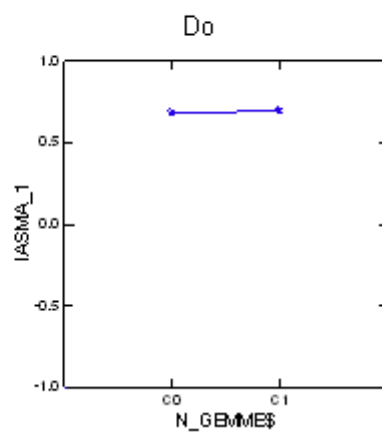
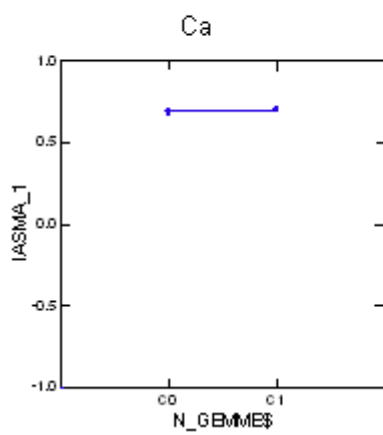
Least Squares Means



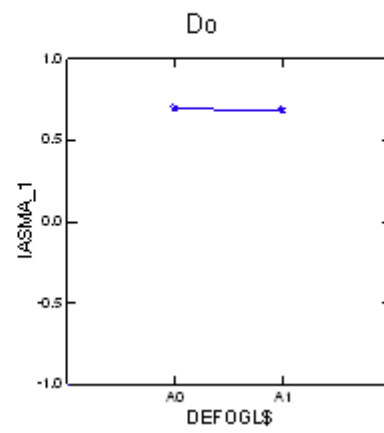
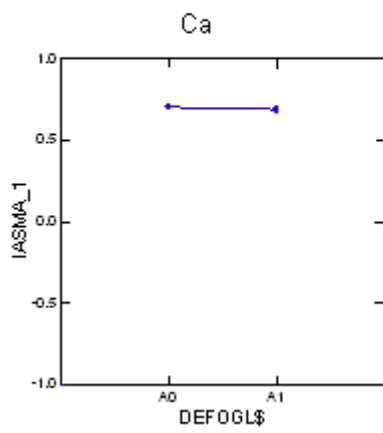
Least Squares Means



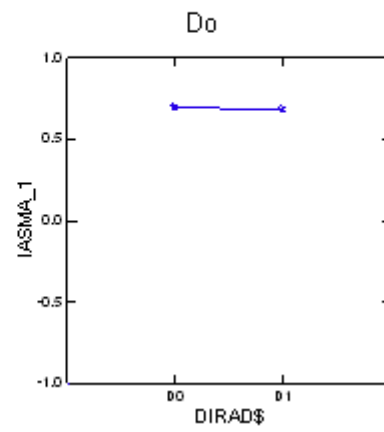
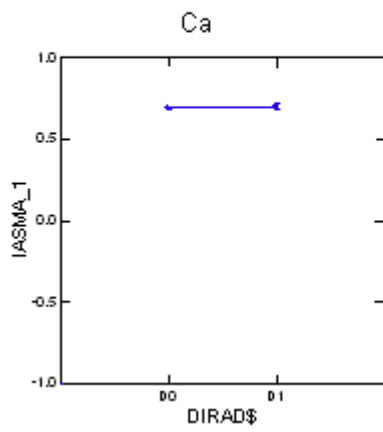
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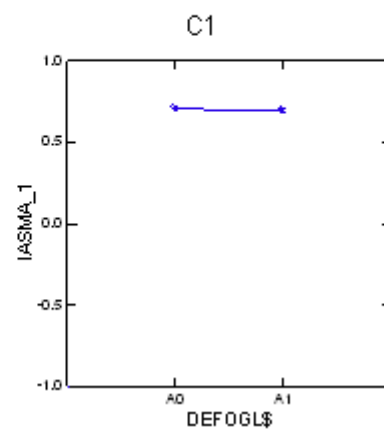
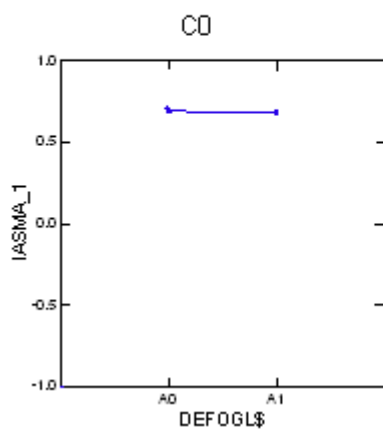
Least Squares Means



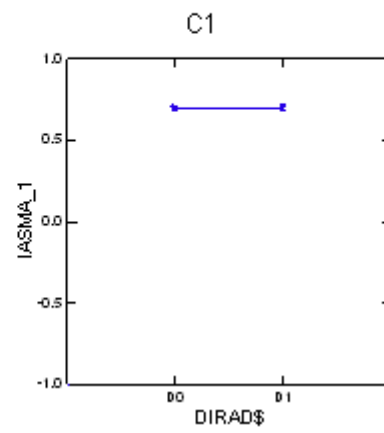
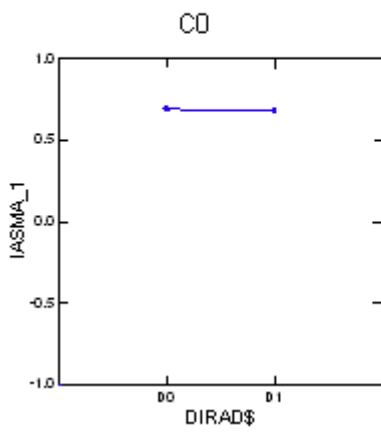
Least Squares Means



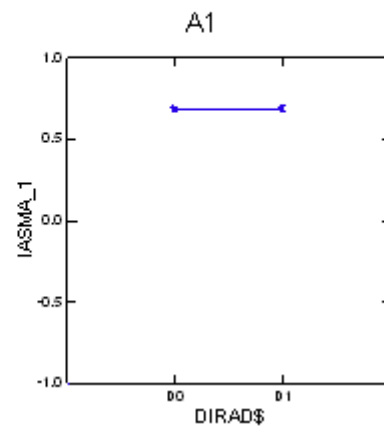
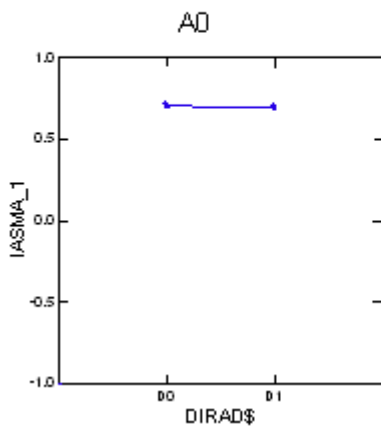
Least Squares Means



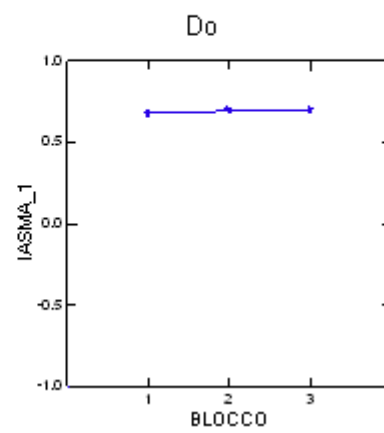
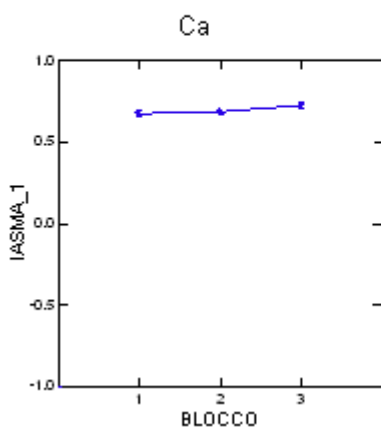
Least Squares Means



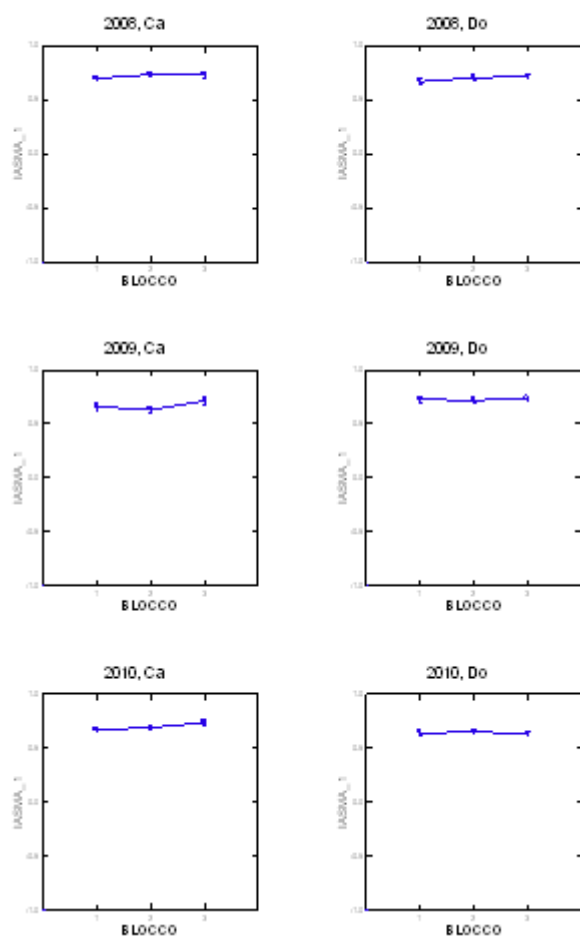
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 266 is an Outlier (Studentized Residual : -4.355)

Durbin-Watson D Statistic | 1.916
 First Order Autocorrelation | 0.022

Information Criteria

AIC | -537.762
 AIC (Corrected) | -515.927
 Schwarz's BIC | -436.788

▼ General Linear Model IASMA2 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	

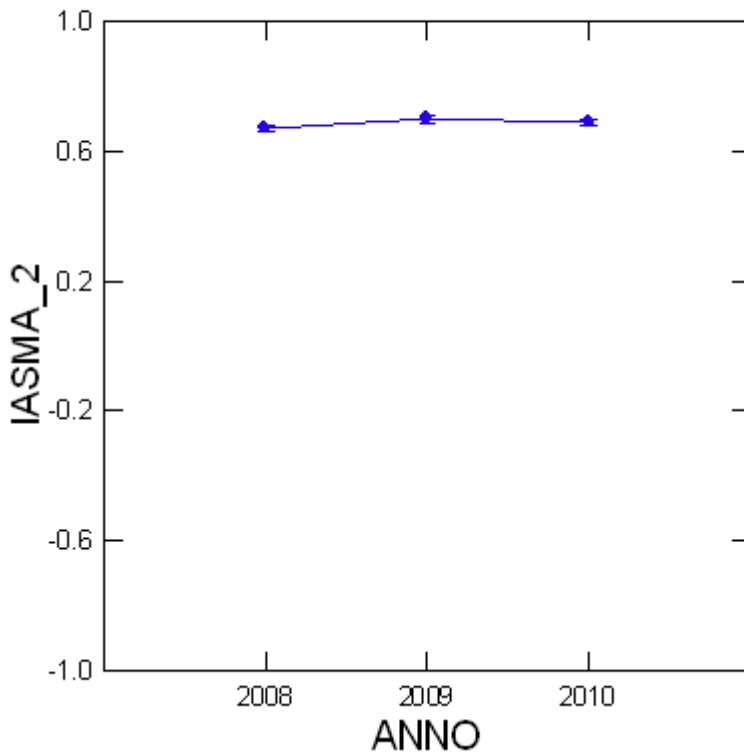
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	IASMA_2
N	144
Multiple R	0.841
Squared Multiple R	0.707

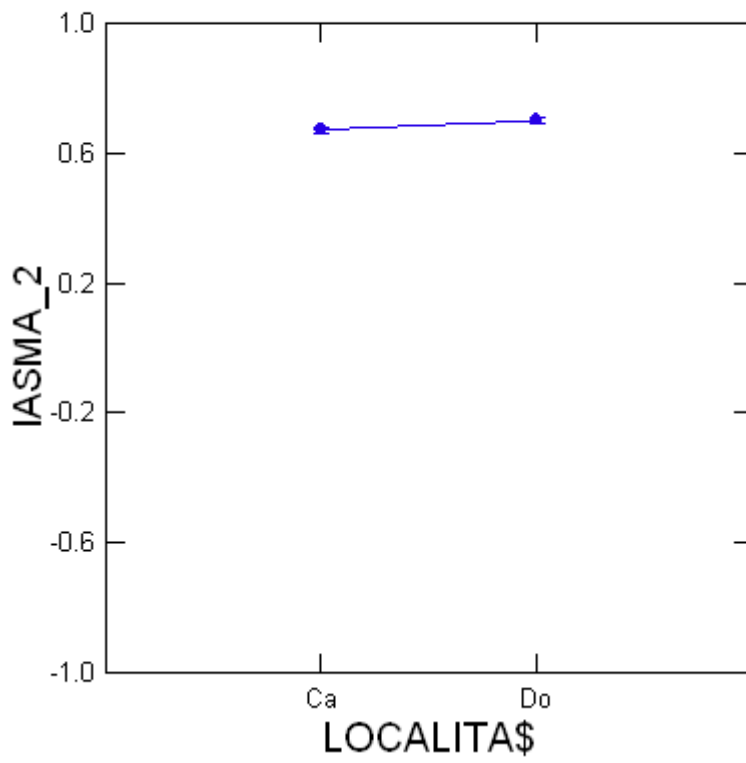
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.017	2	0.008	9.077	0.000
LOCALITA\$	0.021	1	0.021	22.409	0.000
N_GEMME\$	0.003	1	0.003	2.952	0.089
DEFOGL\$	0.000	1	0.000	0.375	0.541
DIRAD\$	0.000	1	0.000	0.076	0.783
LOCALITA\$*ANNO	0.051	2	0.026	27.701	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.101	0.904
DEFOGL\$*ANNO	0.005	2	0.003	2.776	0.067
DIRAD\$*ANNO	0.000	2	0.000	0.009	0.991
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.513	0.475
DEFOGL\$*LOCALITA\$	0.002	1	0.002	1.922	0.168
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.112	0.738
DEFOGL\$*N_GEMME\$	0.000	1	0.000	0.191	0.663
DIRAD\$*N_GEMME\$	0.002	1	0.002	2.488	0.118
DIRAD\$*DEFOGL\$	0.004	1	0.004	4.842	0.030
BLOCCO(LOCALITA\$)	0.086	4	0.022	23.421	0.000
BLOCCO*ANNO(LOCALITA\$)	0.045	8	0.006	6.052	0.000
Error	0.102	111	0.001		

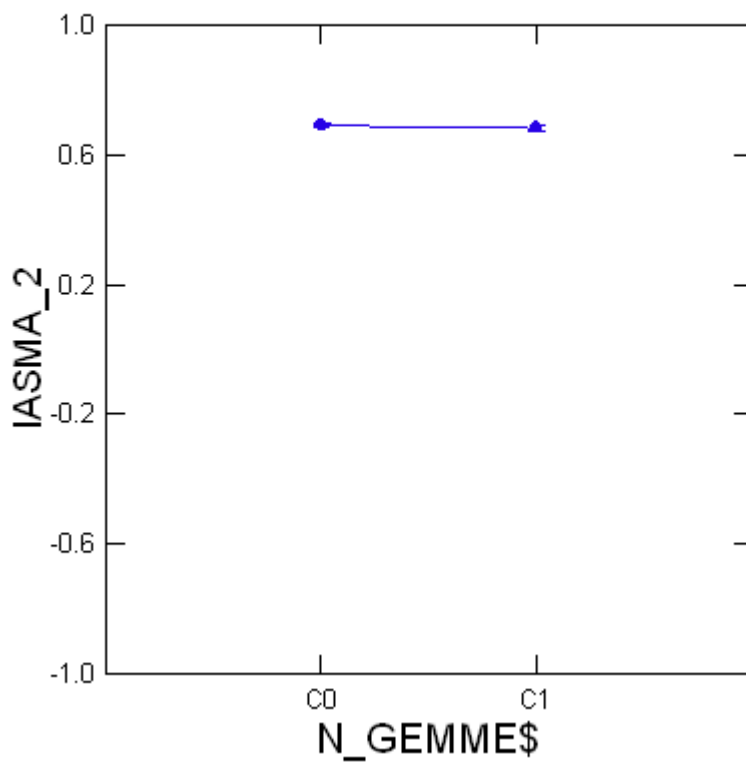
Least Squares Means



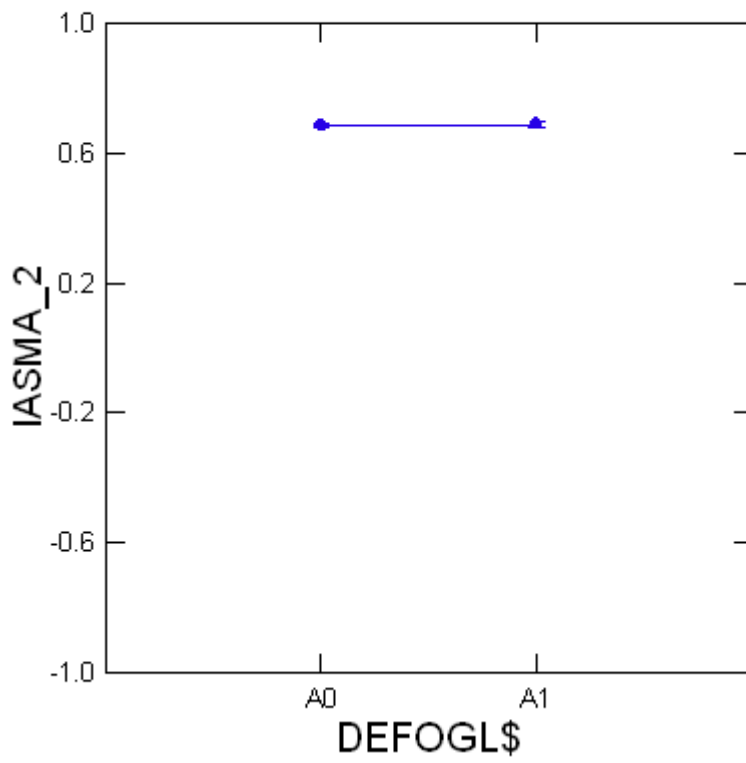
Least Squares Means



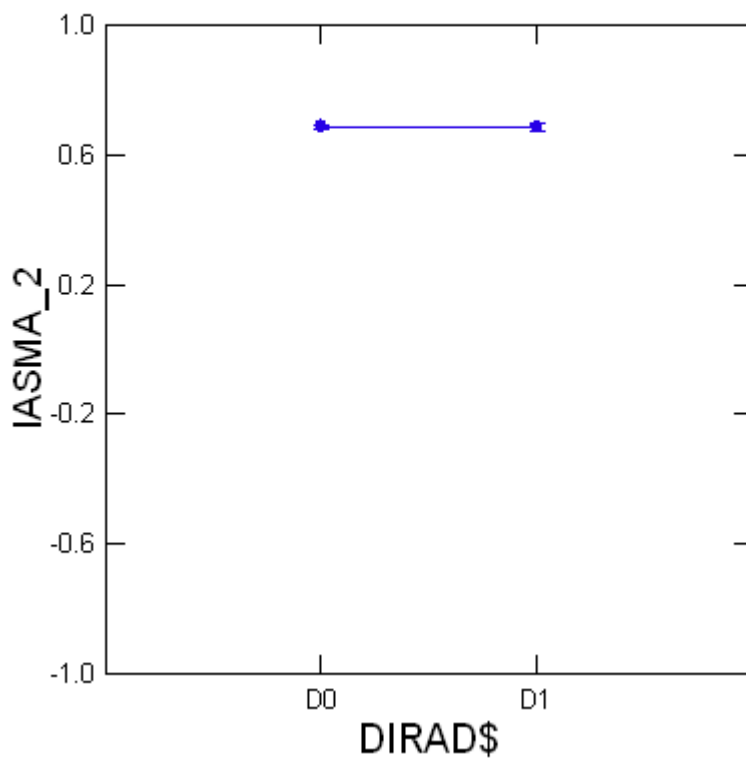
Least Squares Means



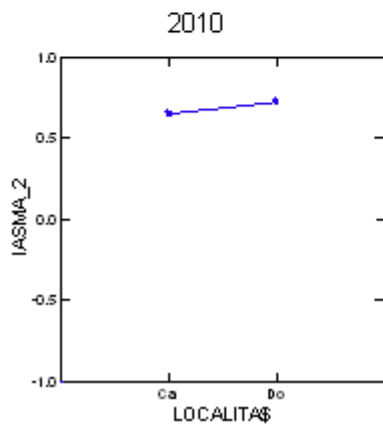
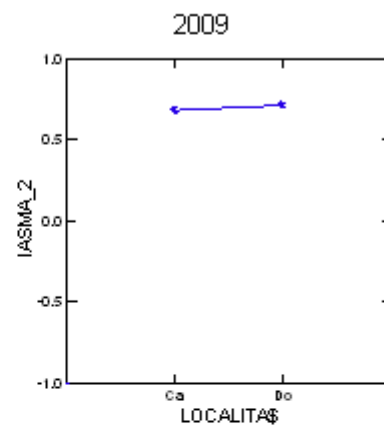
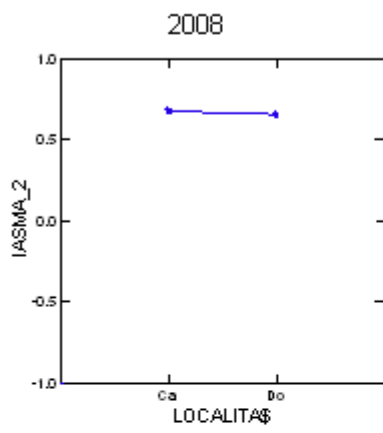
Least Squares Means



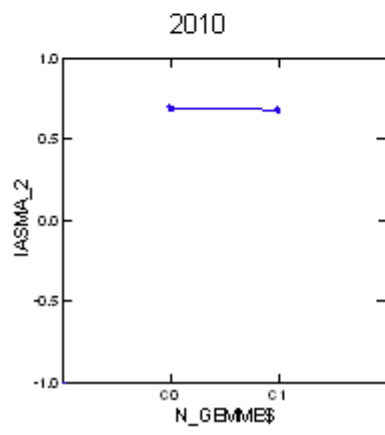
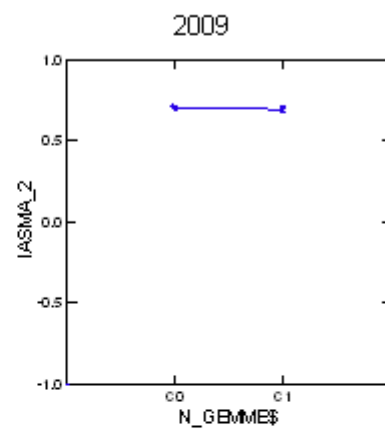
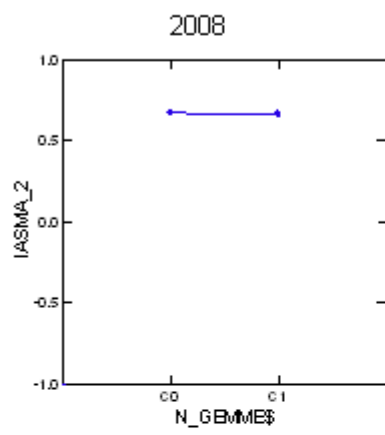
Least Squares Means



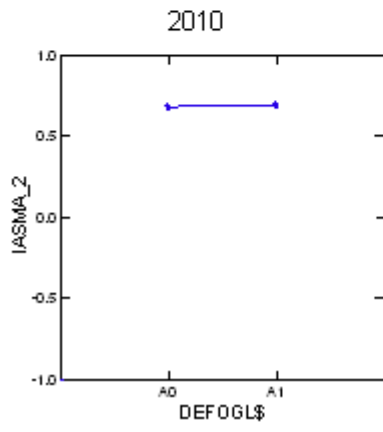
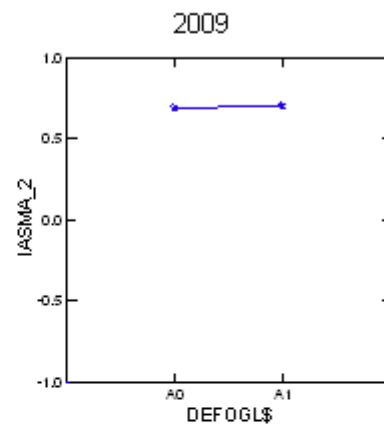
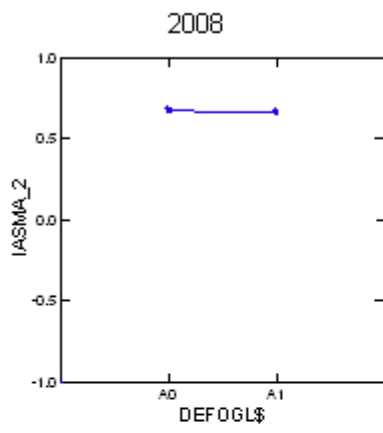
Least Squares Means



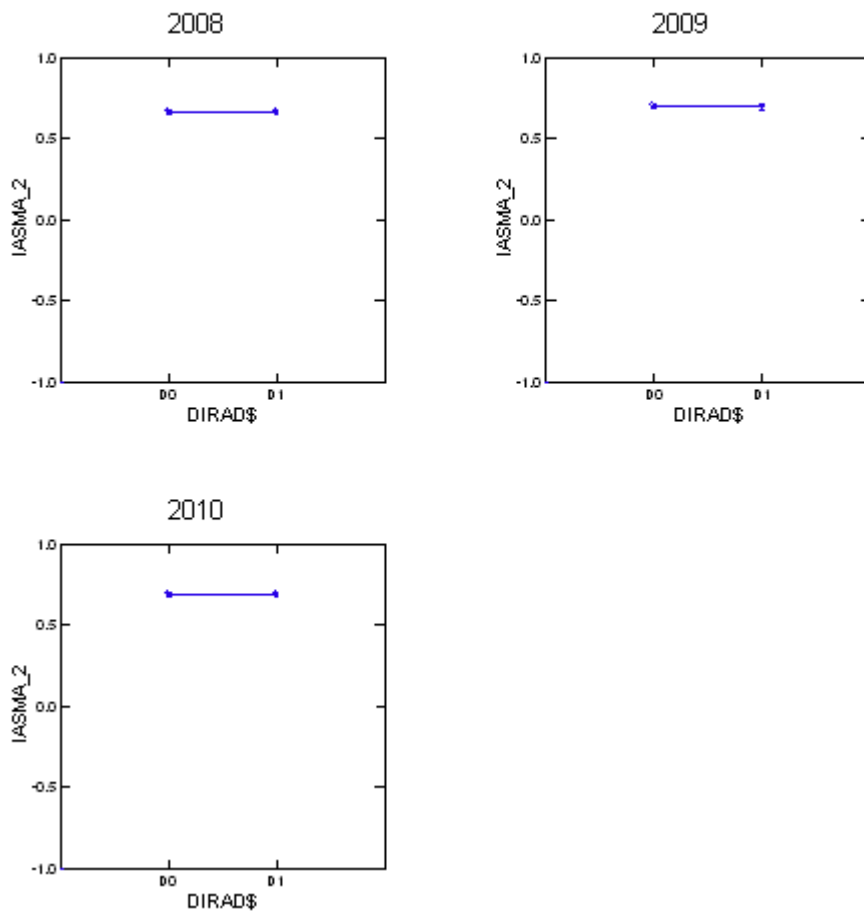
Least Squares Means



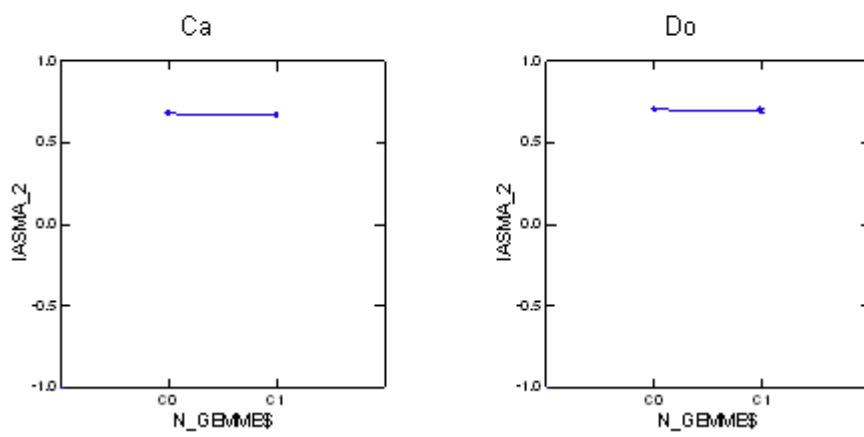
Least Squares Means



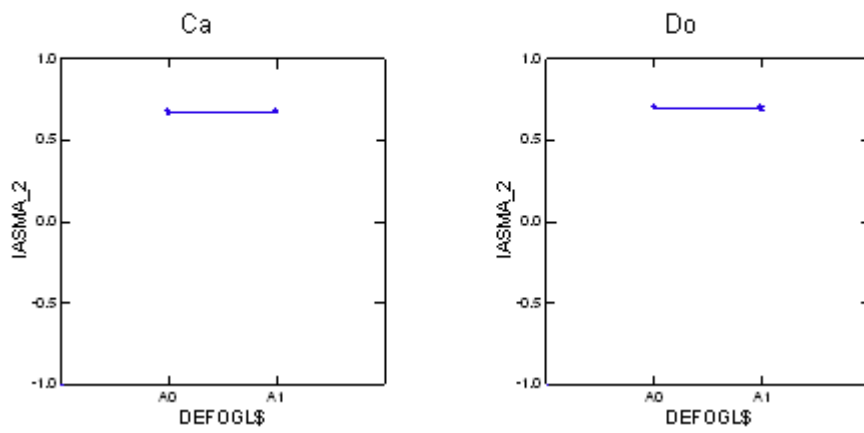
Least Squares Means



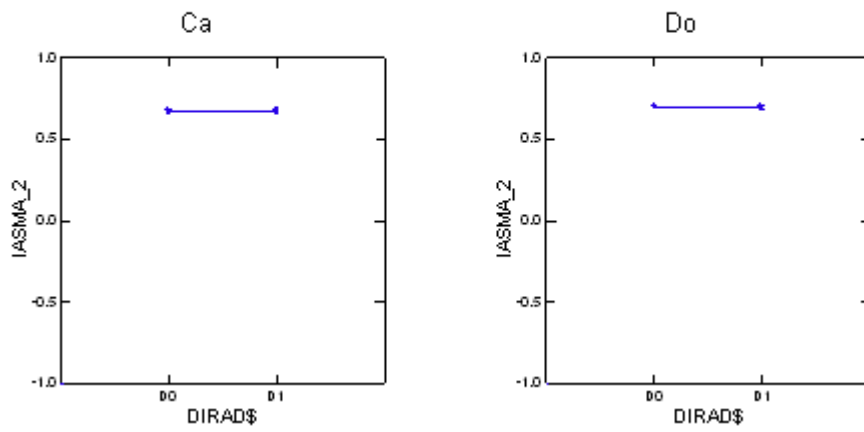
Least Squares Means



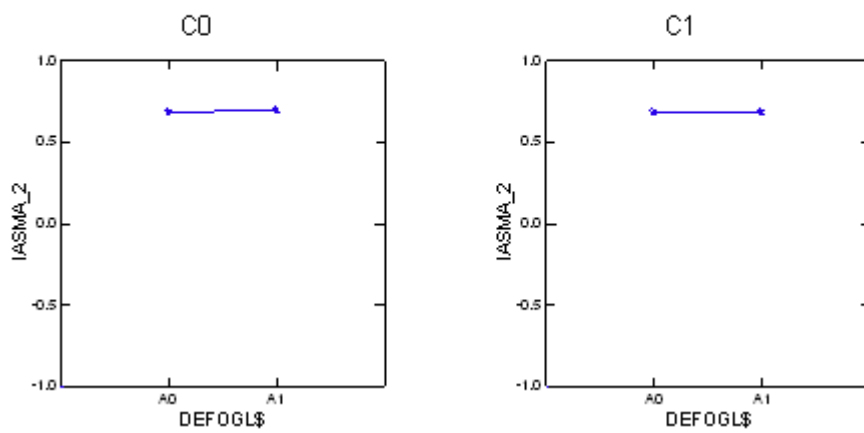
Least Squares Means



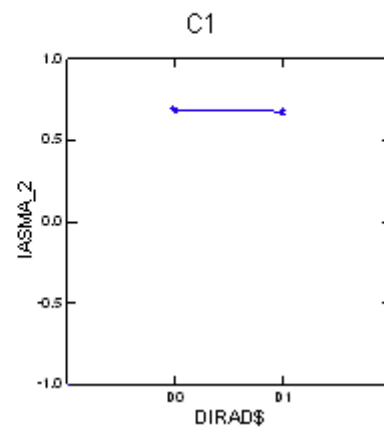
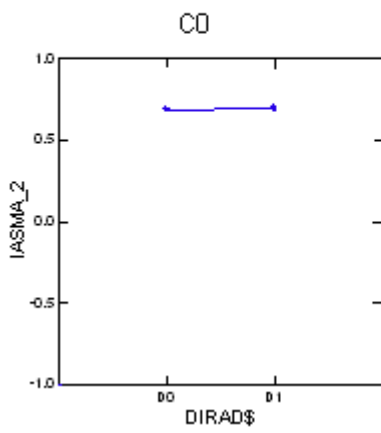
Least Squares Means



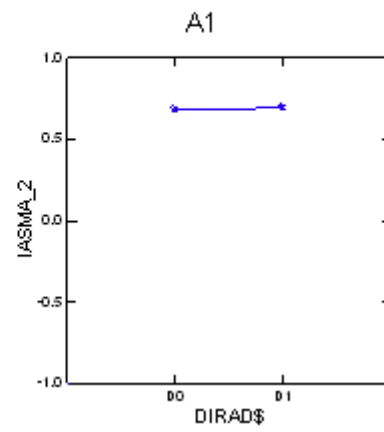
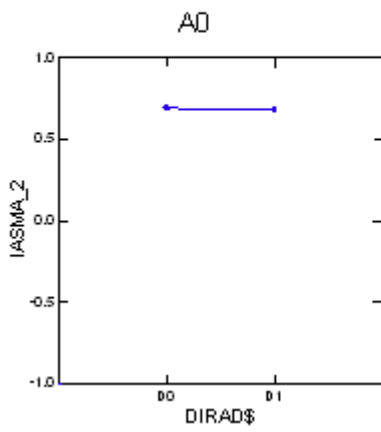
Least Squares Means



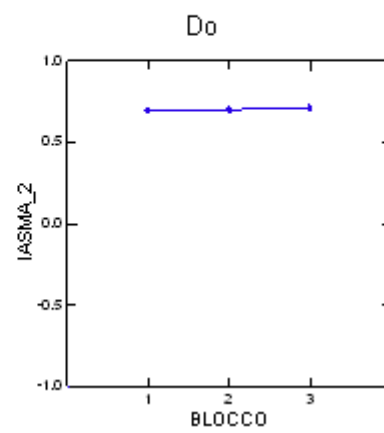
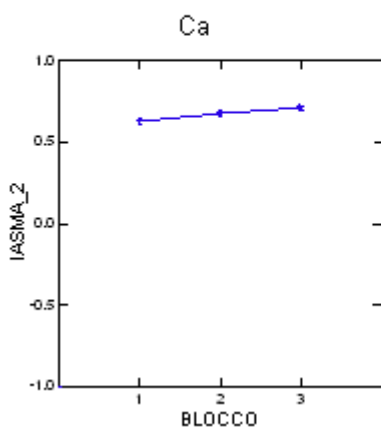
Least Squares Means



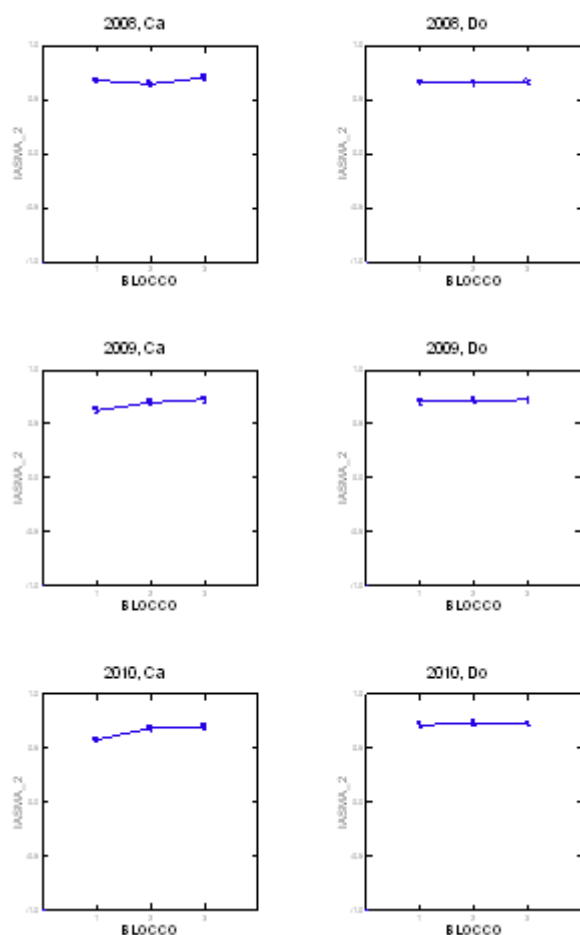
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 2.090
 First Order Autocorrelation | -0.049

Information Criteria

AIC | -567.198
 AIC (Corrected) | -545.364
 Schwarz's BIC | -466.225

▼ General Linear Model

IASMA3 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

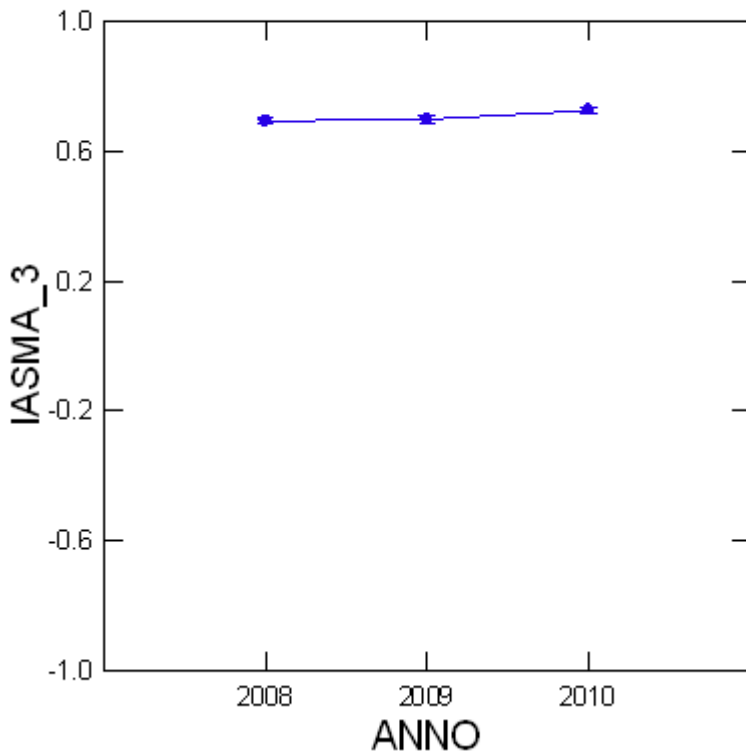
Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable | IASMA_3
 N | 144
 Multiple R | 0.764
 Squared Multiple R | 0.583

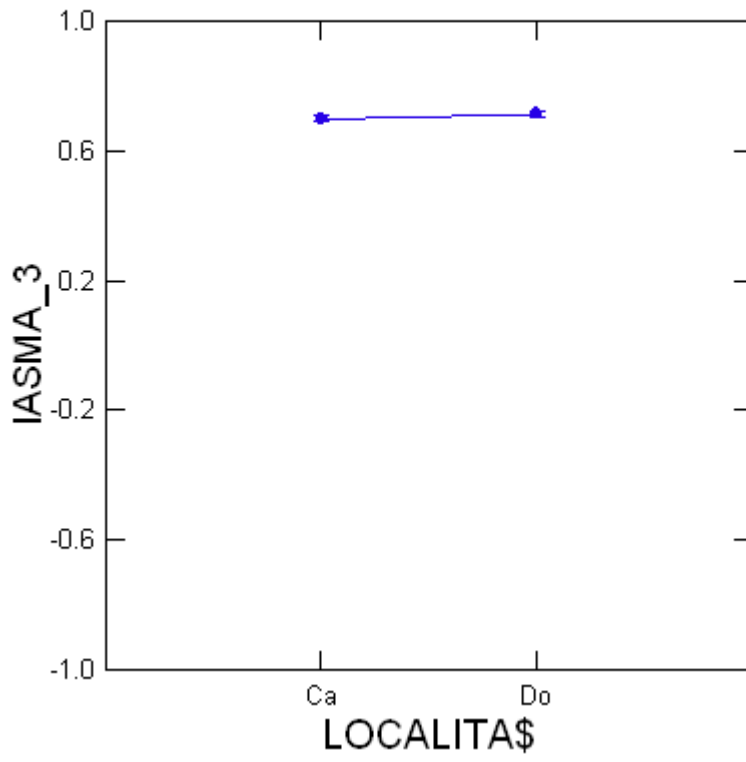
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.028	2	0.014	15.217	0.000
LOCALITA\$	0.006	1	0.006	6.909	0.010
N_GEMME\$	0.000	1	0.000	0.504	0.479
DEFOGL\$	0.000	1	0.000	0.455	0.501
DIRAD\$	0.000	1	0.000	0.470	0.495
LOCALITA\$*ANNO	0.015	2	0.007	8.120	0.001
N_GEMME\$*ANNO	0.001	2	0.001	0.699	0.499
DEFOGL\$*ANNO	0.000	2	0.000	0.099	0.906
DIRAD\$*ANNO	0.000	2	0.000	0.150	0.860
N_GEMME\$*LOCALITA\$	0.001	1	0.001	1.455	0.230
DEFOGL\$*LOCALITA\$	0.001	1	0.001	0.639	0.426
DIRAD\$*LOCALITA\$	0.001	1	0.001	1.049	0.308
DEFOGL\$*N_GEMME\$	0.001	1	0.001	0.921	0.339
DIRAD\$*N_GEMME\$	0.001	1	0.001	1.511	0.222
DIRAD\$*DEFOGL\$	0.003	1	0.003	3.335	0.071
BLOCCO(LOCALITA\$)	0.051	4	0.013	13.938	0.000
BLOCCO*ANNO(LOCALITA\$)	0.022	8	0.003	3.026	0.004
Error	0.101	111	0.001		

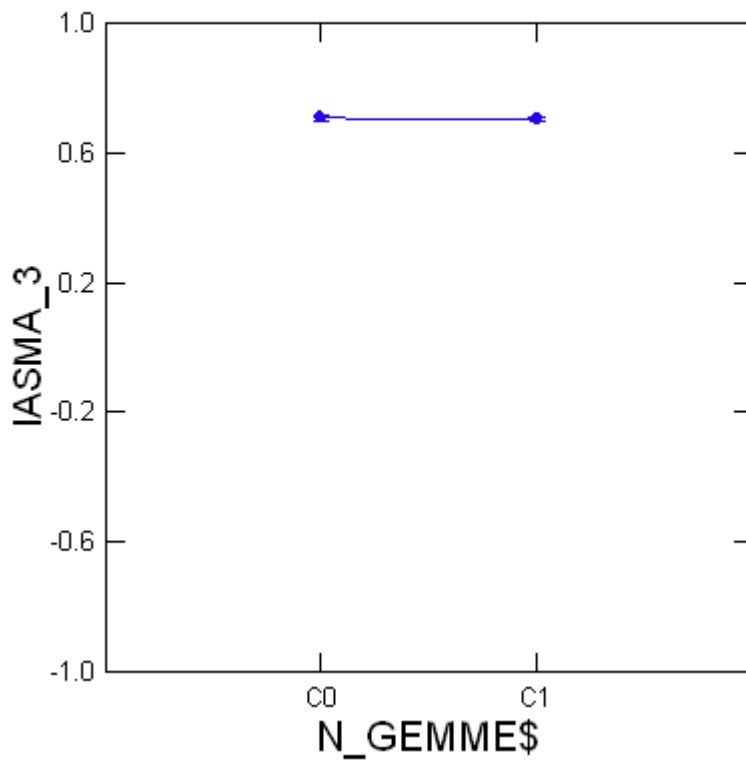
Least Squares Means



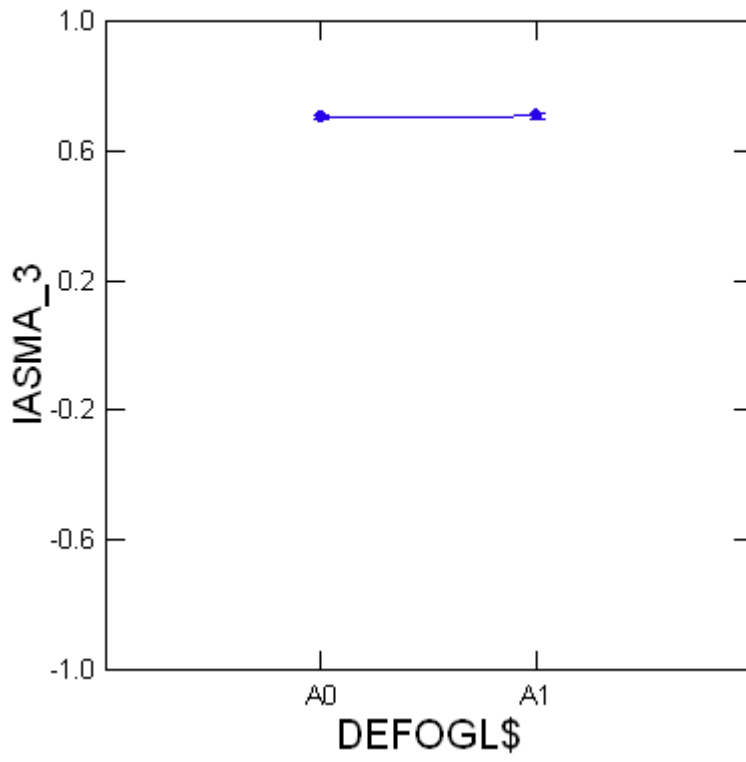
Least Squares Means



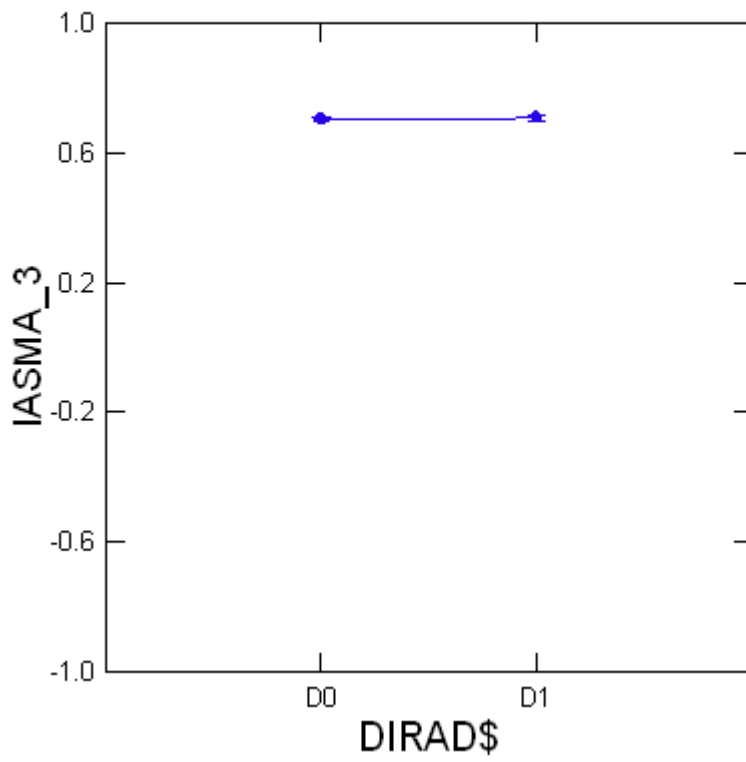
Least Squares Means



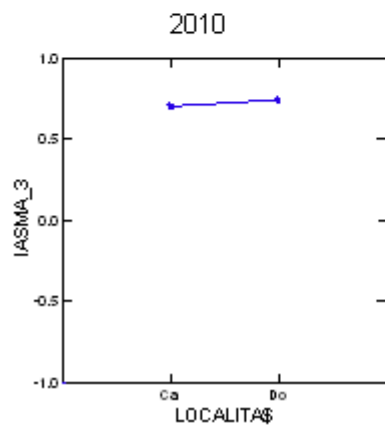
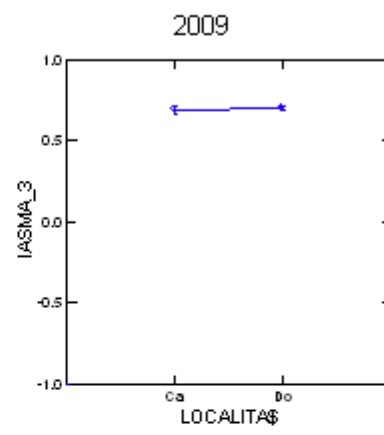
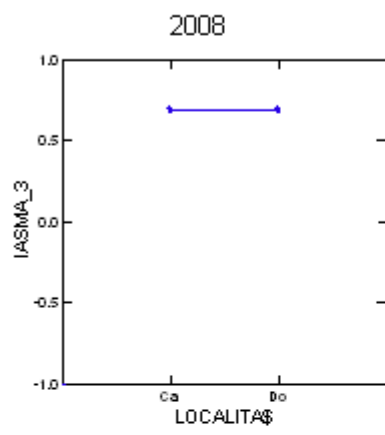
Least Squares Means



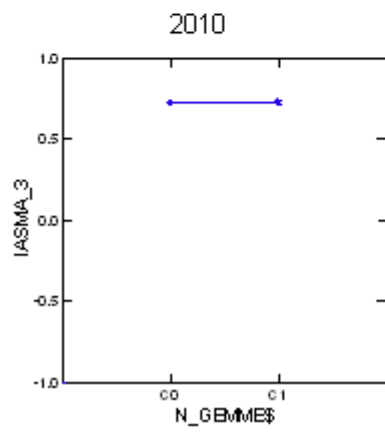
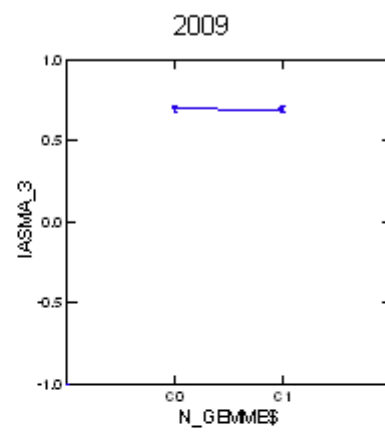
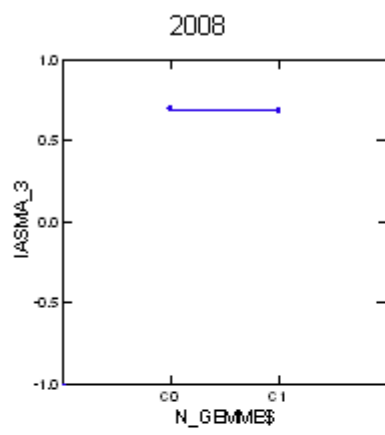
Least Squares Means



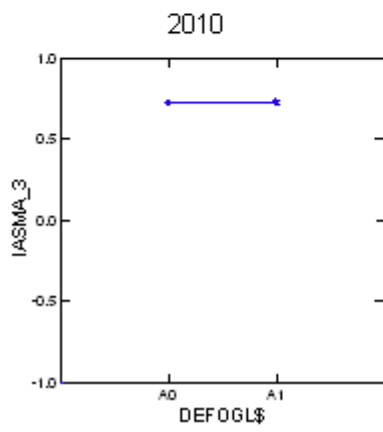
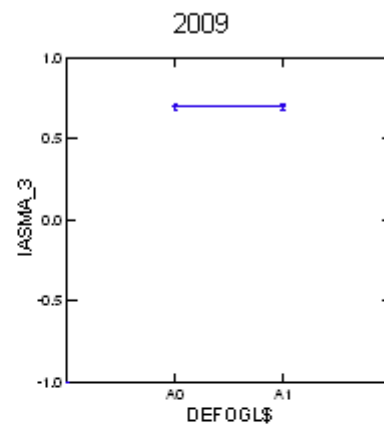
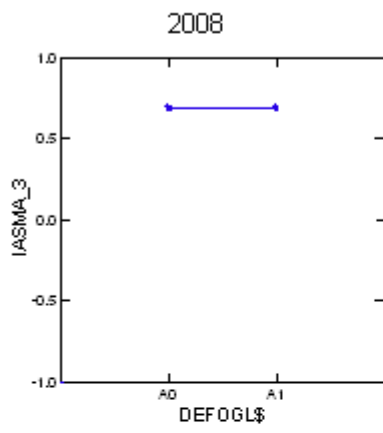
Least Squares Means



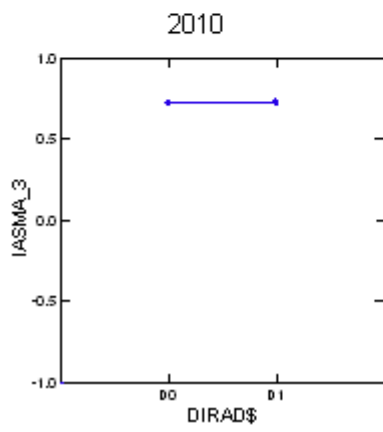
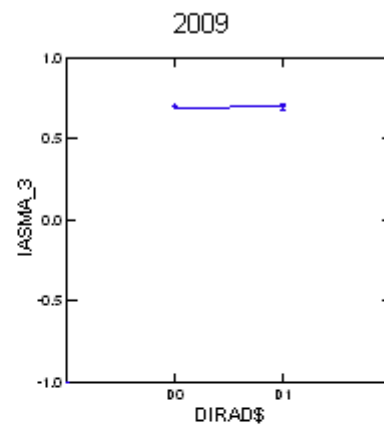
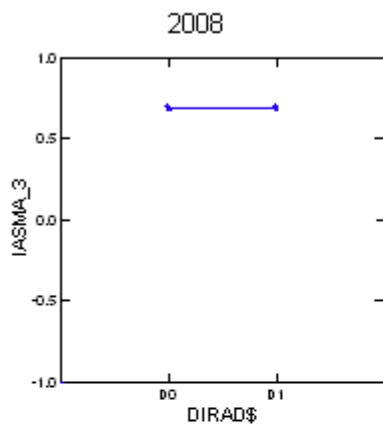
Least Squares Means



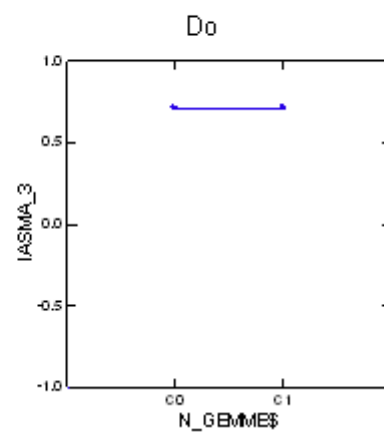
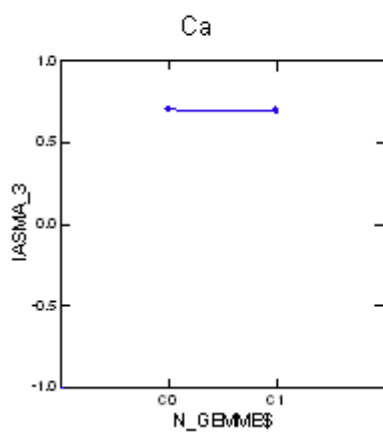
Least Squares Means



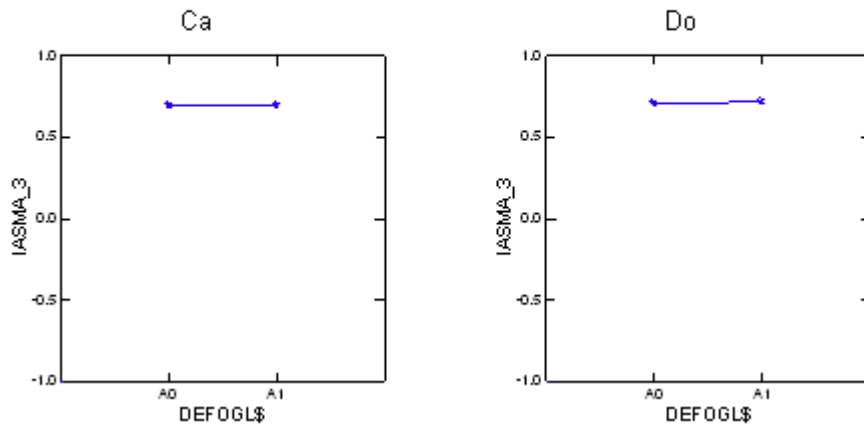
Least Squares Means



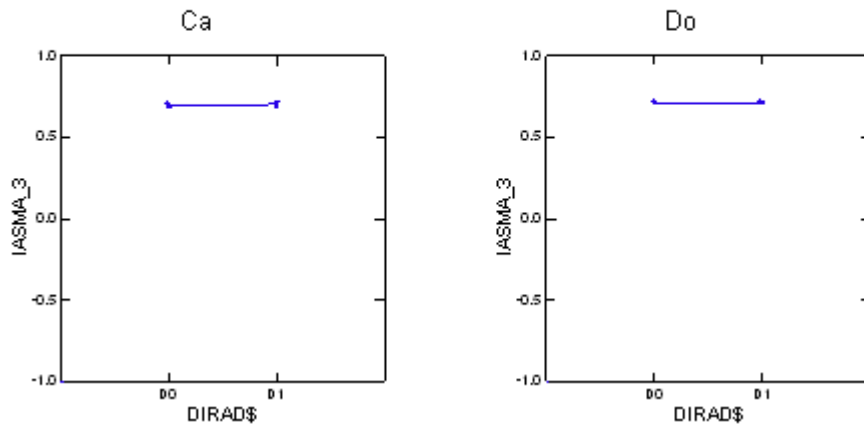
Least Squares Means



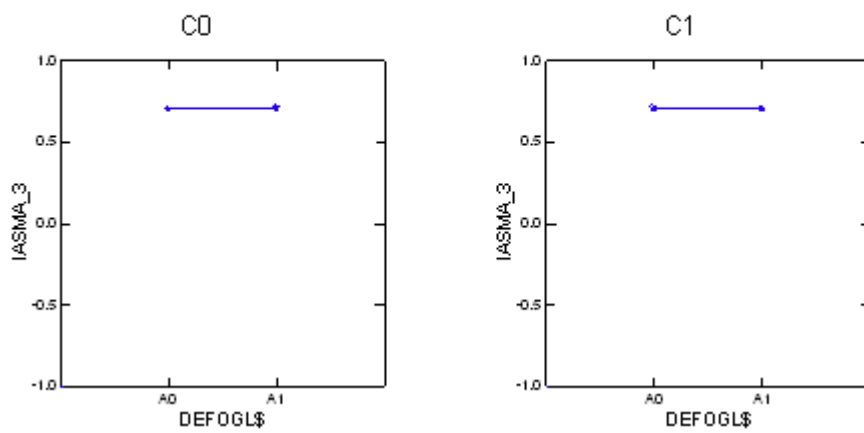
Least Squares Means



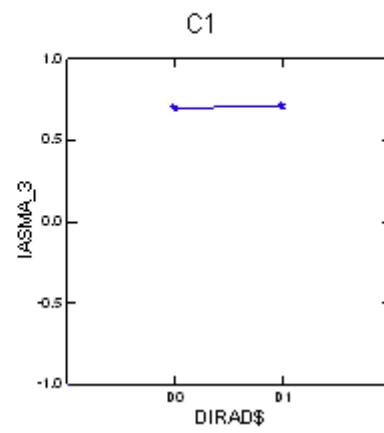
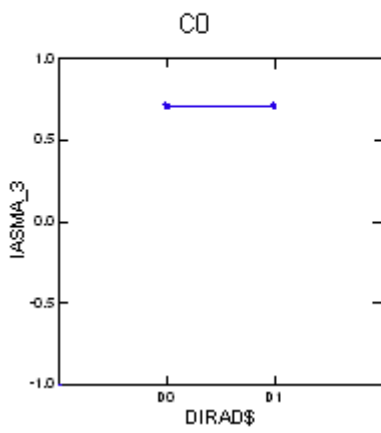
Least Squares Means



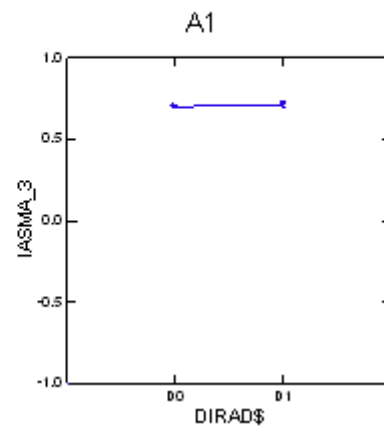
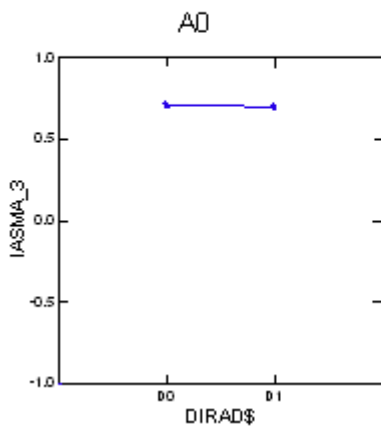
Least Squares Means



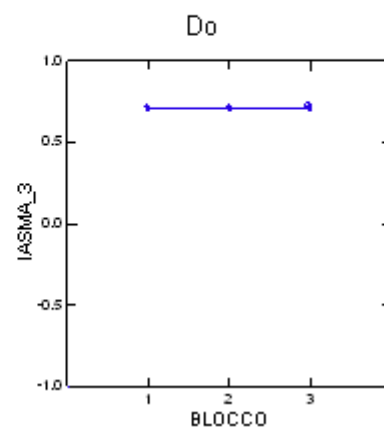
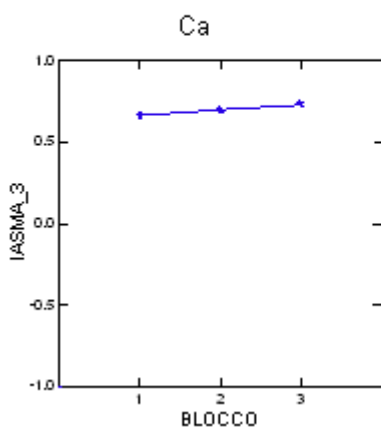
Least Squares Means



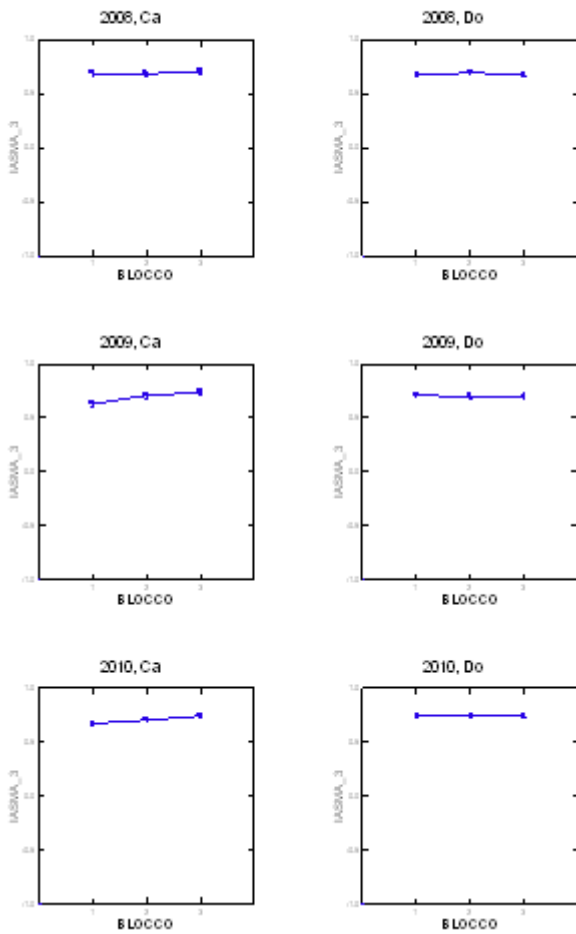
Least Squares Means



Least Squares Means



Least Squares Means



Durbin-Watson D Statistic | 1.788
 First Order Autocorrelation | 0.106

Information Criteria

AIC | -569.259
 AIC (Corrected) | -547.424
 Schwarz's BIC | -468.285

▼ General Linear Model DIPROVE1 2008- 2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

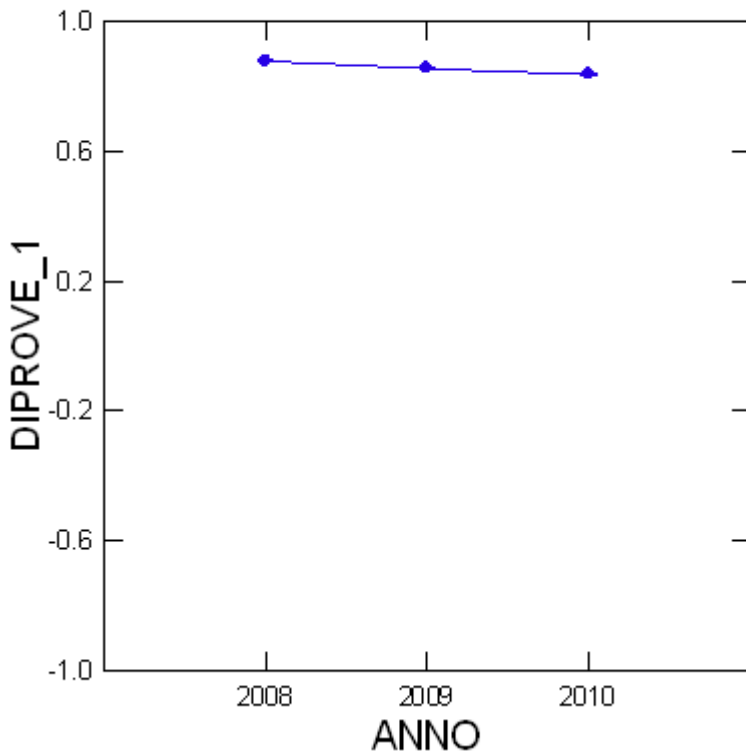
Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable | DIPROVE_1
 N | 144
 Multiple R | 0.972
 Squared Multiple R | 0.945

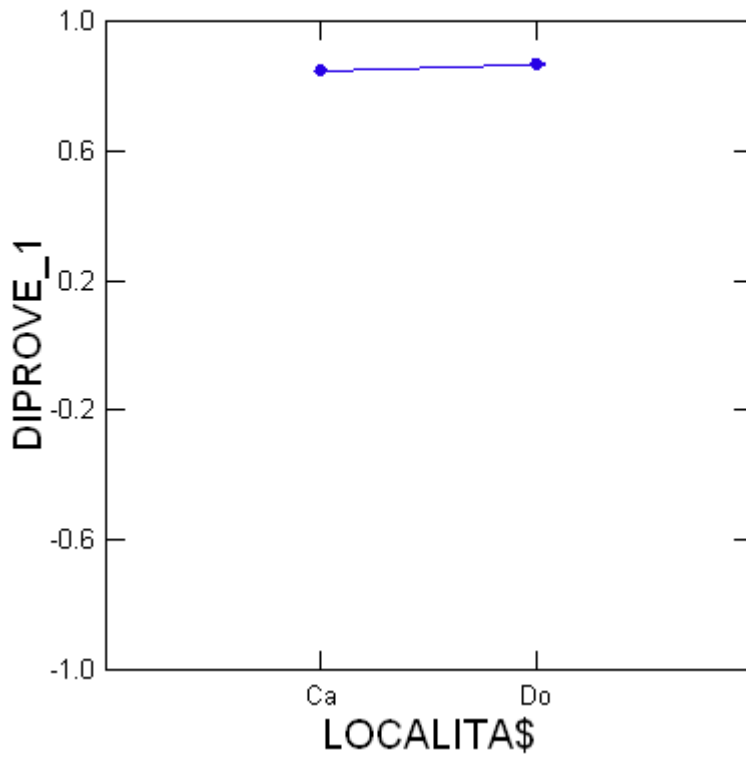
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.042	2	0.021	330.491	0.000
LOCALITA\$	0.011	1	0.011	178.366	0.000
N_GEMME\$	0.003	1	0.003	46.304	0.000
DEFOGL\$	0.012	1	0.012	190.957	0.000
DIRAD\$	0.000	1	0.000	3.127	0.080
LOCALITA\$*ANNO	0.019	2	0.009	144.728	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.740	0.479
DEFOGL\$*ANNO	0.005	2	0.003	41.187	0.000
DIRAD\$*ANNO	0.000	2	0.000	1.048	0.354
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.766	0.383
DEFOGL\$*LOCALITA\$	0.001	1	0.001	14.093	0.000
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.502	0.480
DEFOGL\$*N_GEMME\$	0.000	1	0.000	4.866	0.029
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.581	0.448
DIRAD\$*DEFOGL\$	0.000	1	0.000	2.868	0.093
BLOCCO (LOCALITA\$)	0.002	4	0.001	8.370	0.000
BLOCCO*ANNO (LOCALITA\$)	0.003	8	0.000	6.688	0.000
Error	0.007	111	0.000		

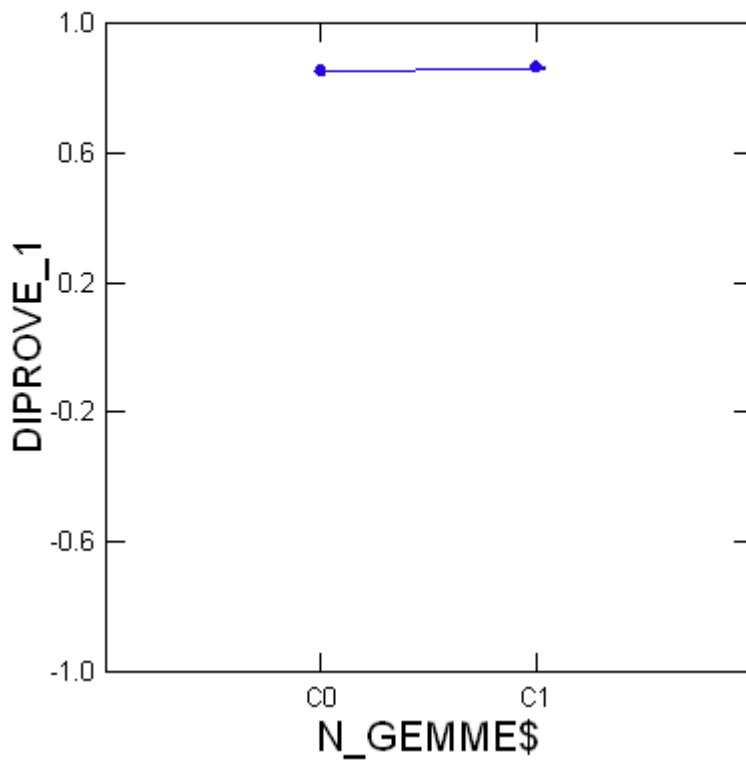
Least Squares Means



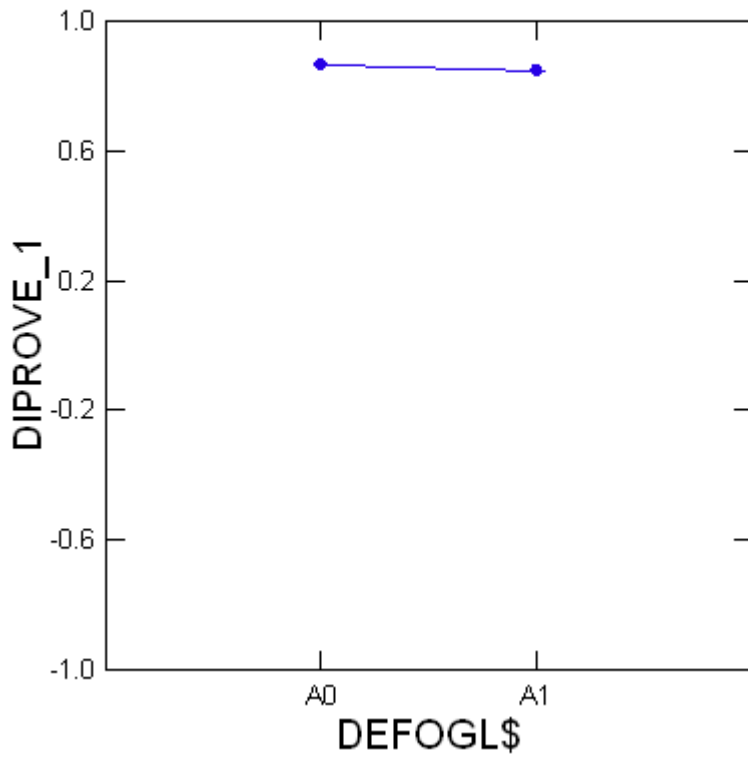
Least Squares Means



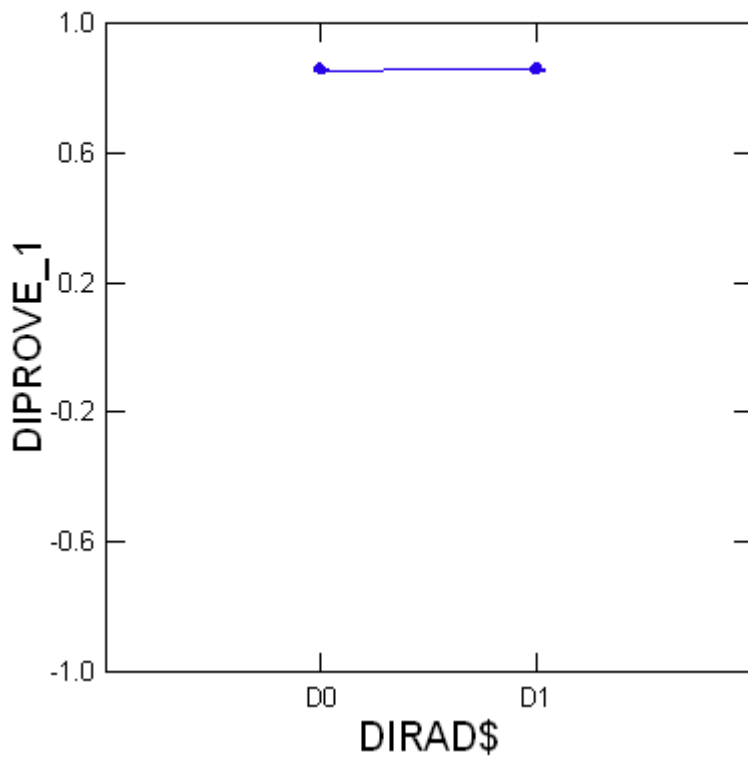
Least Squares Means



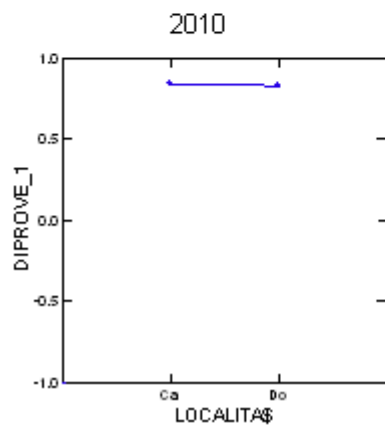
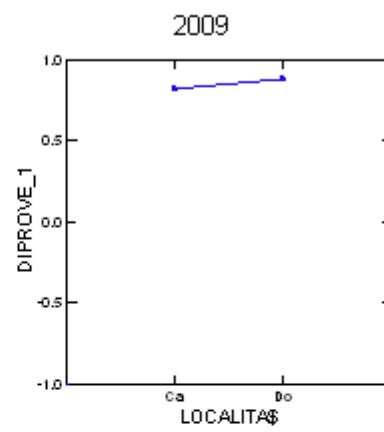
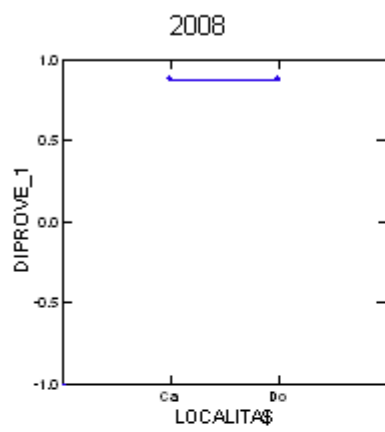
Least Squares Means



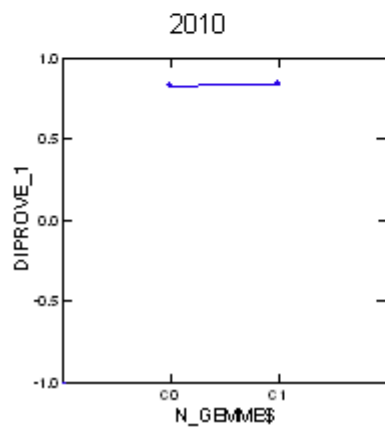
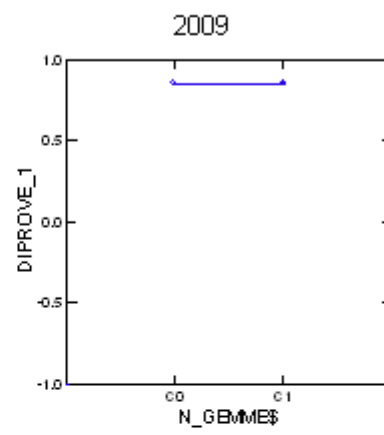
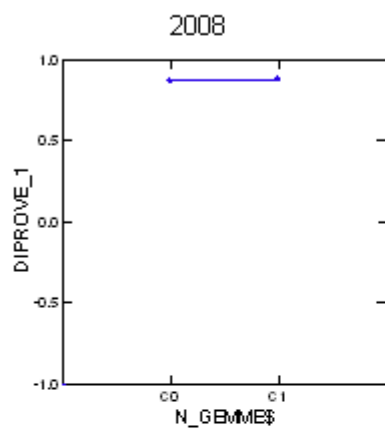
Least Squares Means



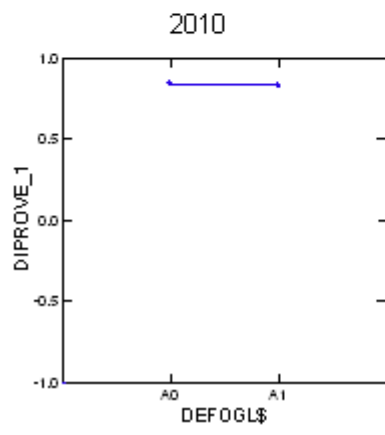
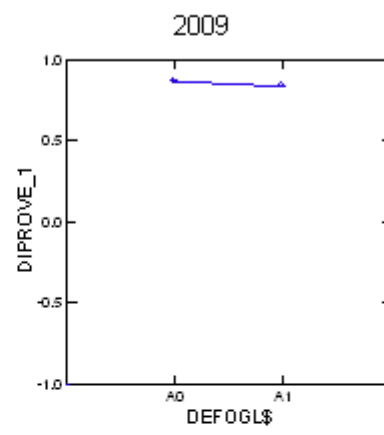
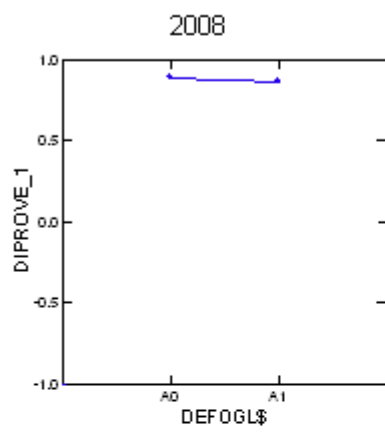
Least Squares Means



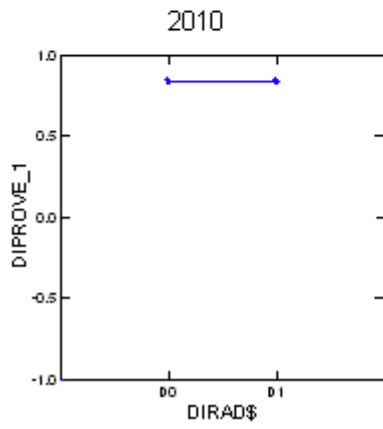
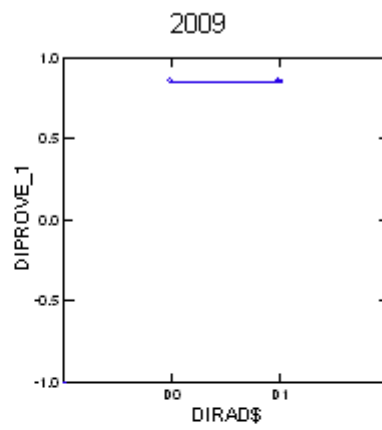
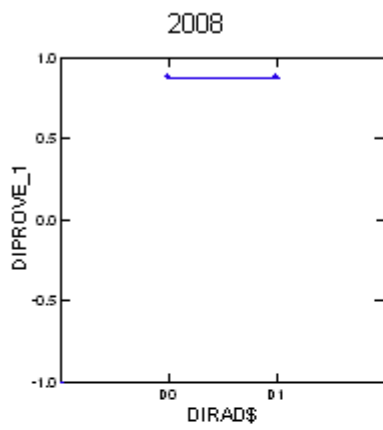
Least Squares Means



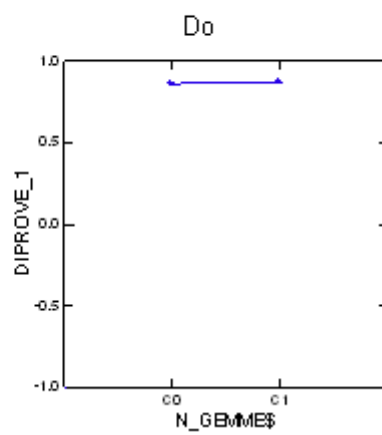
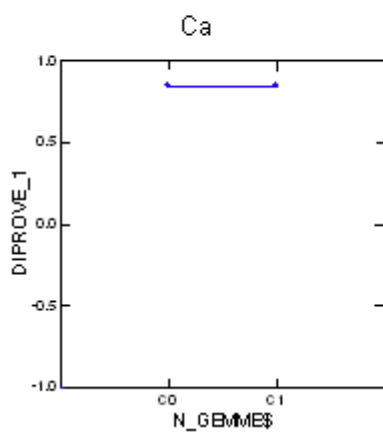
Least Squares Means



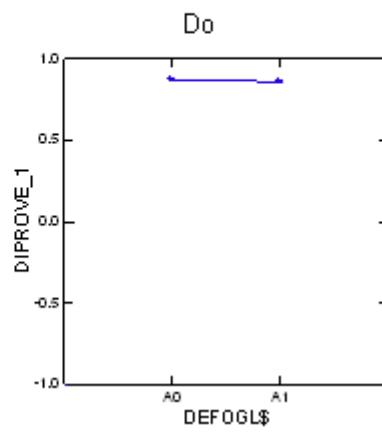
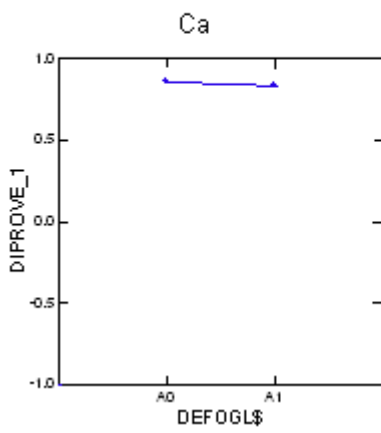
Least Squares Means



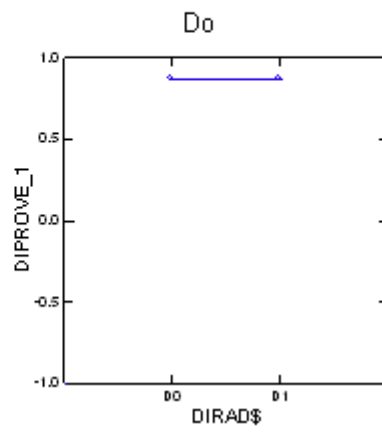
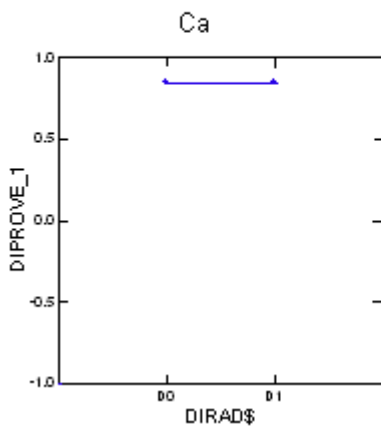
Least Squares Means



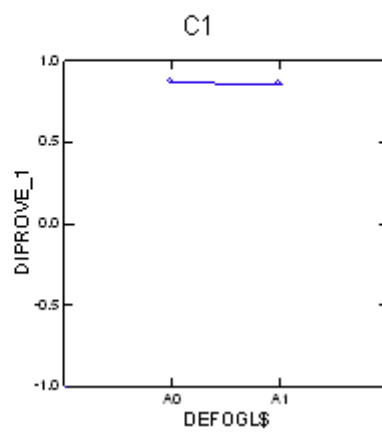
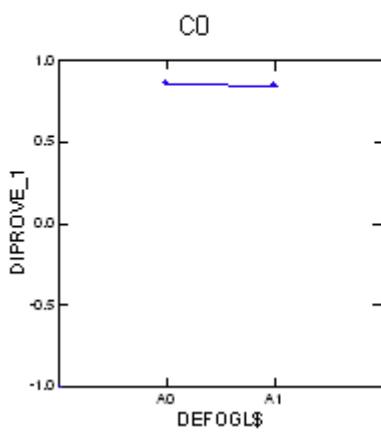
Least Squares Means



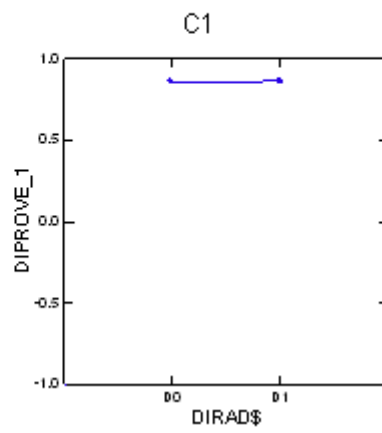
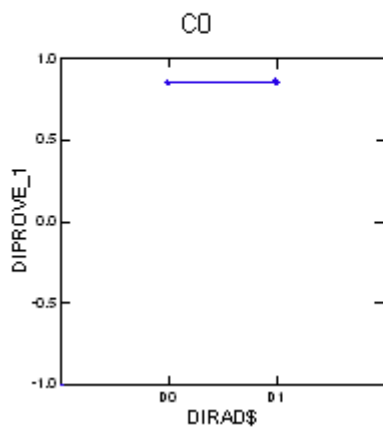
Least Squares Means



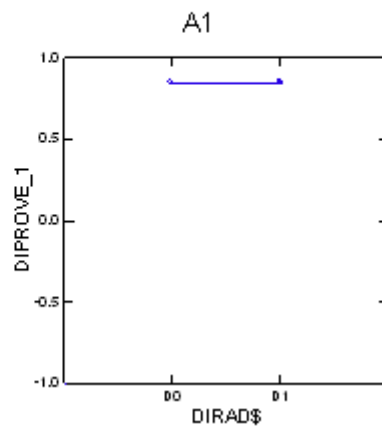
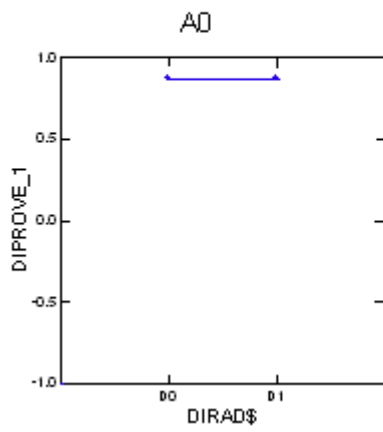
Least Squares Means



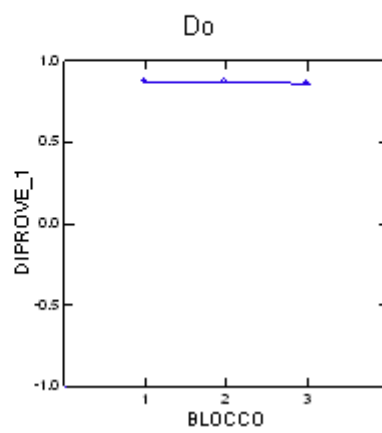
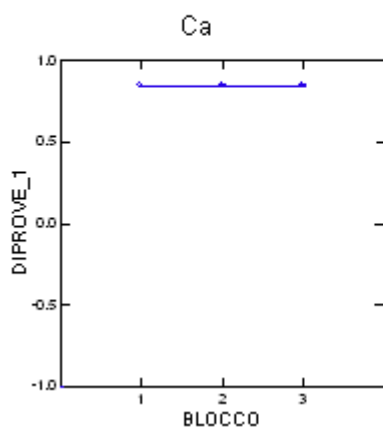
Least Squares Means



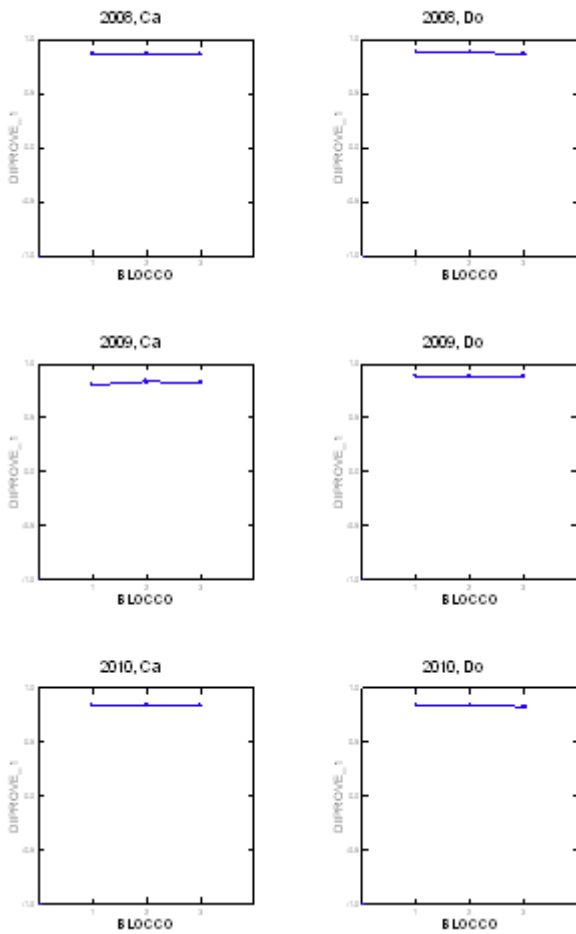
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 274 is an Outlier (Studentized Residual : -3.625)

Durbin-Watson D Statistic | 1.573
 First Order Autocorrelation | 0.213

Information Criteria

AIC | -951.078
 AIC (Corrected) | -929.243
 Schwarz's BIC | -850.105

▼ General Linear Model DIPROVE2 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	

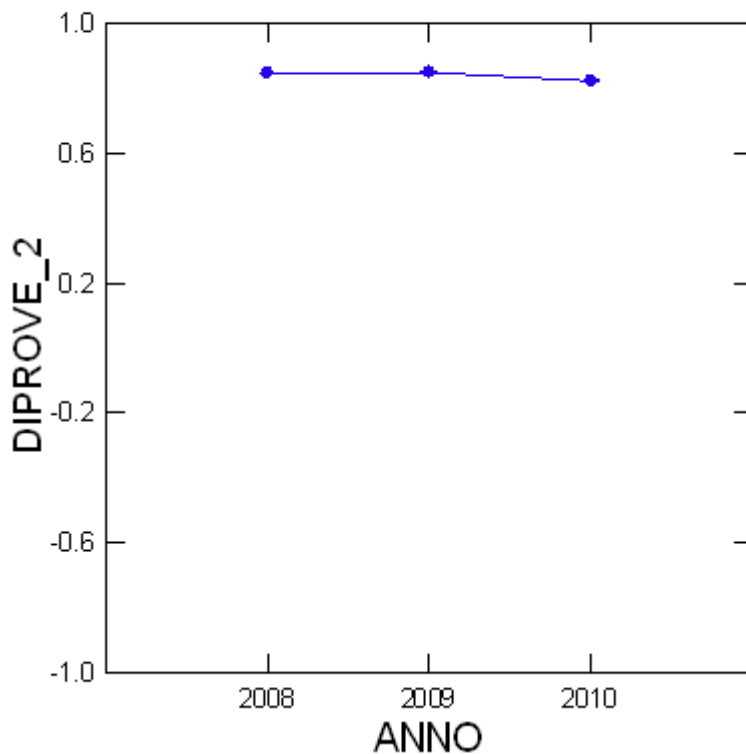
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	DIPROVE_2
N	144
Multiple R	0.945
Squared Multiple R	0.894

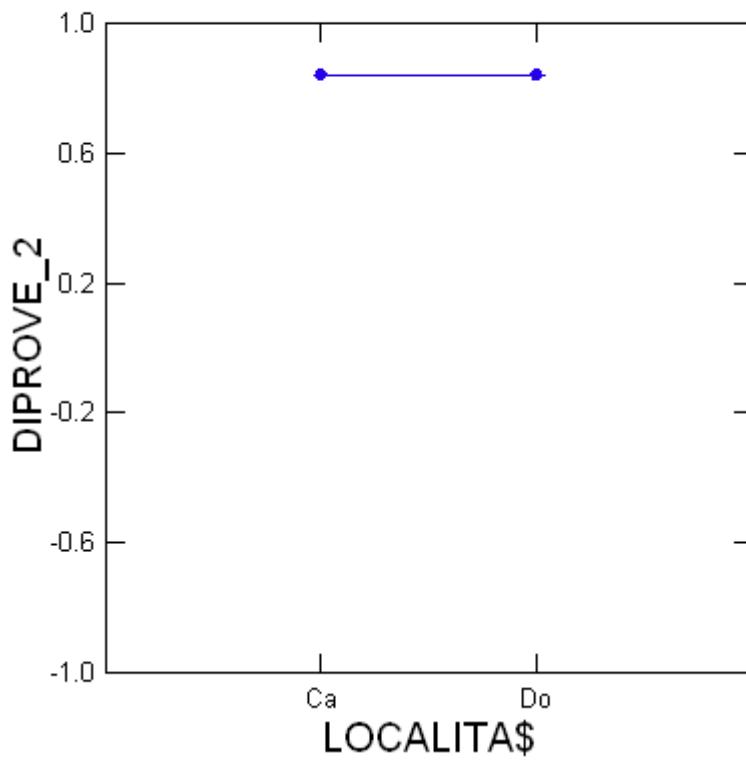
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.017	2	0.009	162.559	0.000
LOCALITA\$	0.000	1	0.000	0.280	0.598
N_GEMME\$	0.000	1	0.000	3.828	0.053
DEFOGL\$	0.001	1	0.001	19.444	0.000
DIRAD\$	0.000	1	0.000	0.105	0.747
LOCALITA\$*ANNO	0.022	2	0.011	205.048	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.767	0.467
DEFOGL\$*ANNO	0.000	2	0.000	1.142	0.323
DIRAD\$*ANNO	0.000	2	0.000	0.492	0.613
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.531	0.468
DEFOGL\$*LOCALITA\$	0.000	1	0.000	1.566	0.213
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.579	0.448
DEFOGL\$*N_GEMME\$	0.000	1	0.000	1.919	0.169
DIRAD\$*N_GEMME\$	0.000	1	0.000	0.511	0.476
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.212	0.646
BLOCCO(LOCALITA\$)	0.001	4	0.000	6.638	0.000
BLOCCO*ANNO(LOCALITA\$)	0.002	8	0.000	4.258	0.000
Error	0.006	111	0.000		

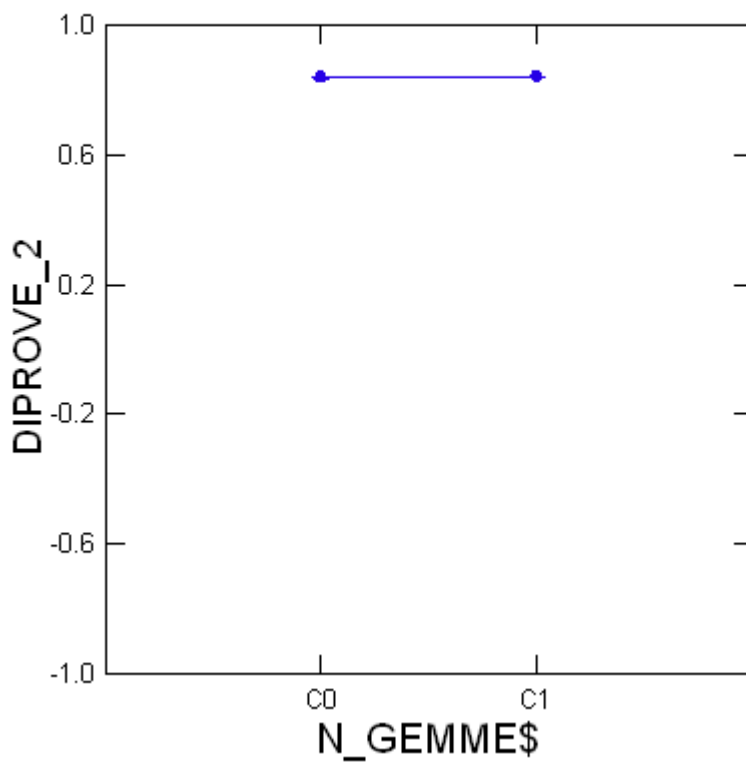
Least Squares Means



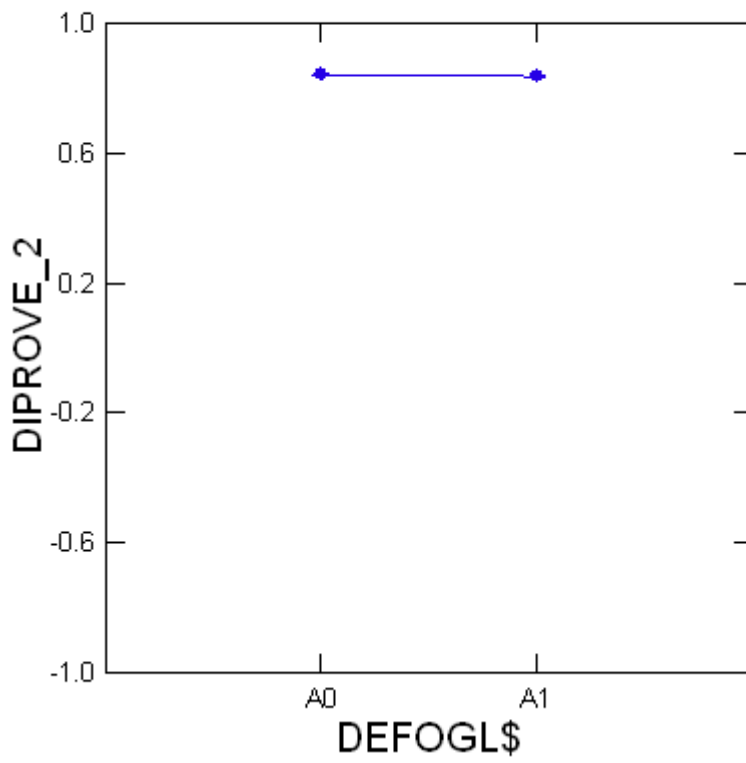
Least Squares Means



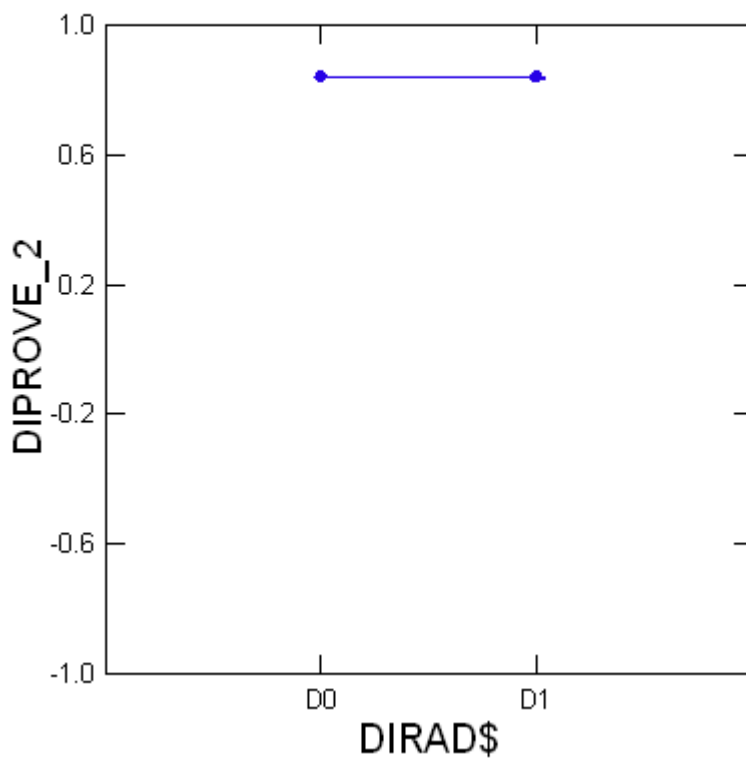
Least Squares Means



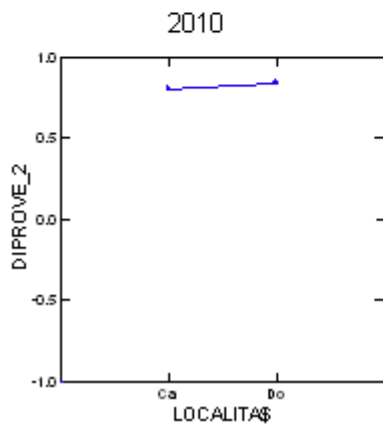
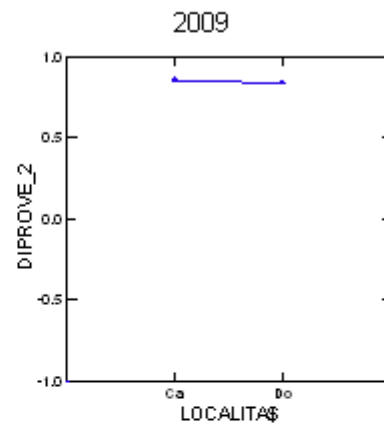
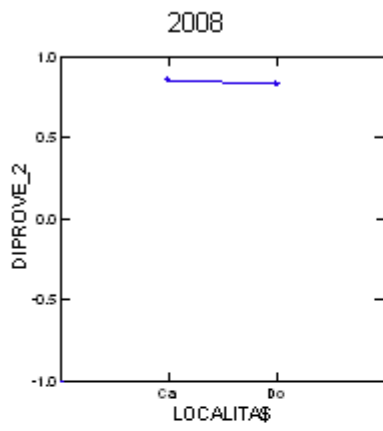
Least Squares Means



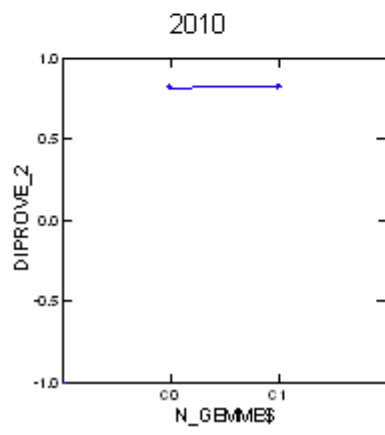
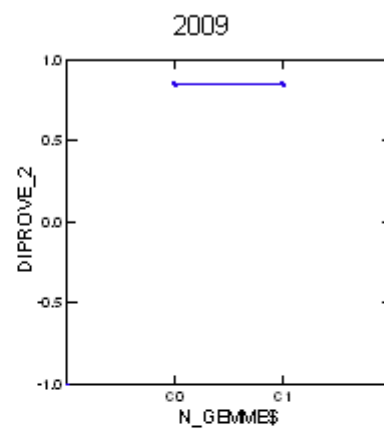
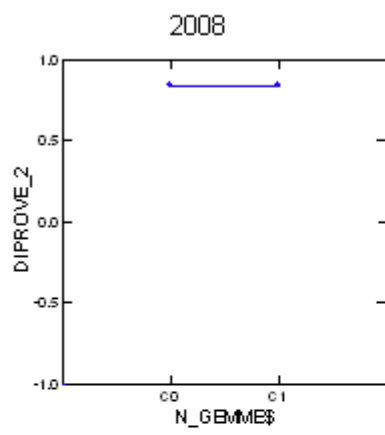
Least Squares Means



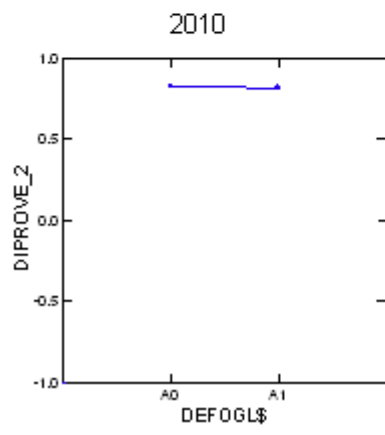
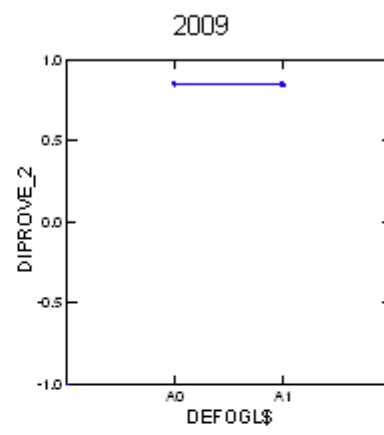
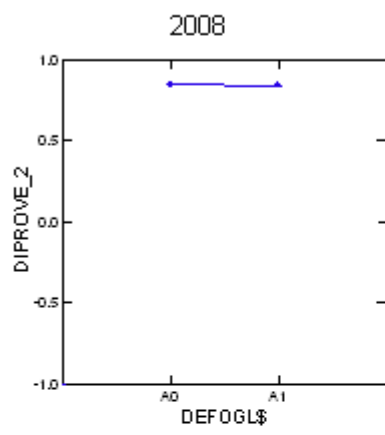
Least Squares Means



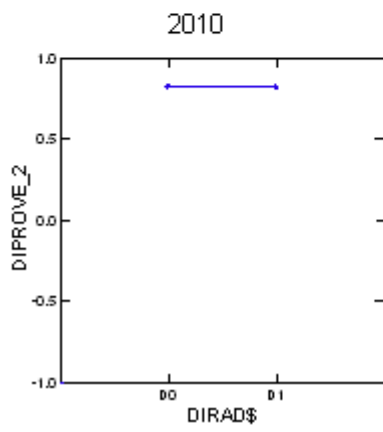
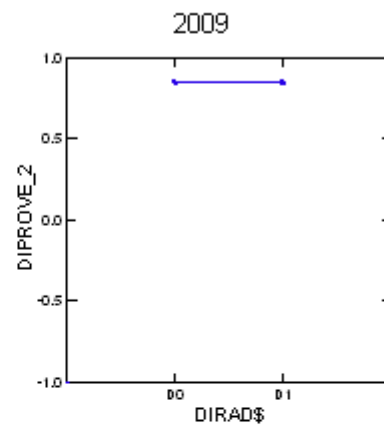
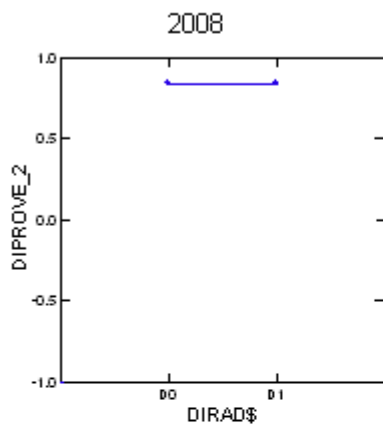
Least Squares Means



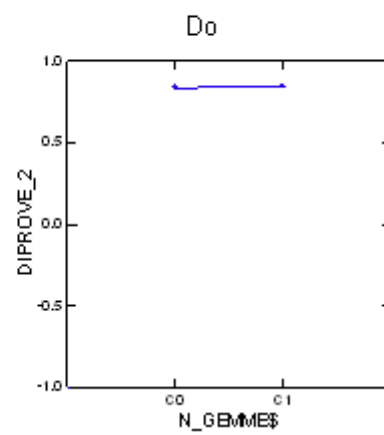
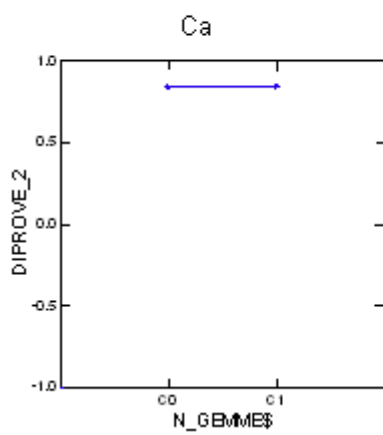
Least Squares Means



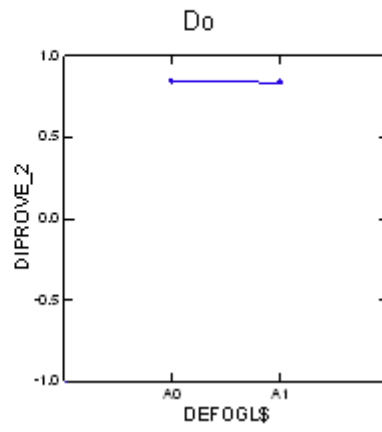
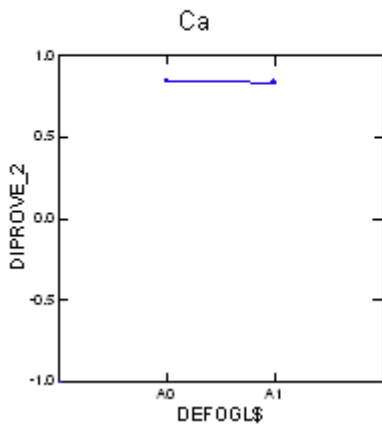
Least Squares Means



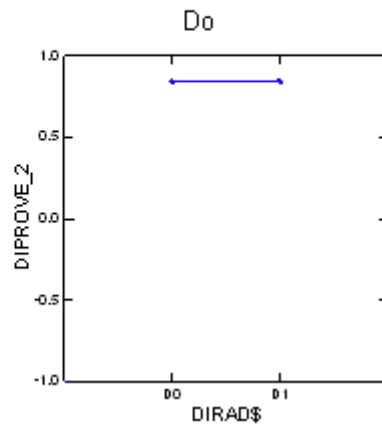
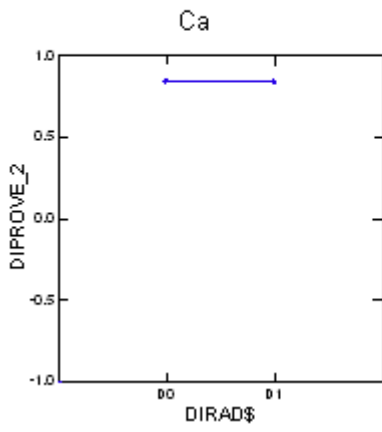
Least Squares Means



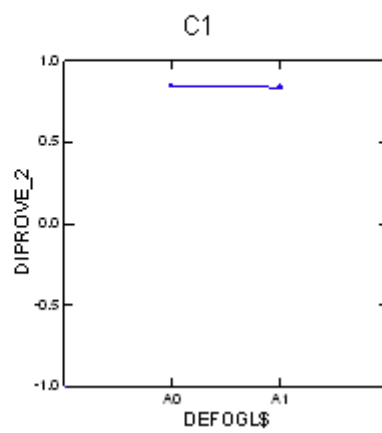
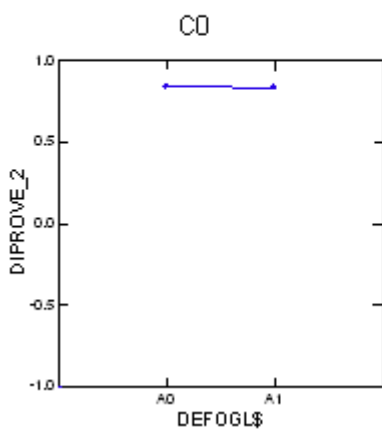
Least Squares Means



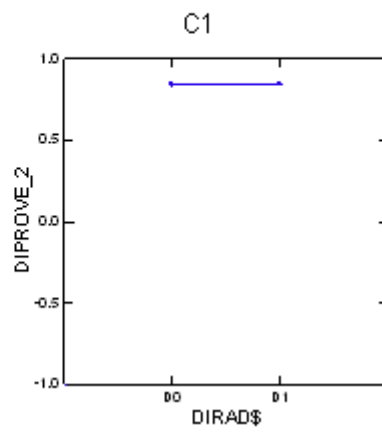
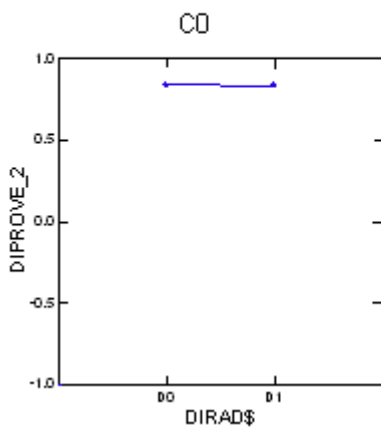
Least Squares Means



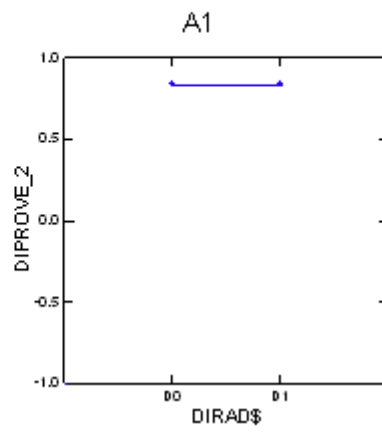
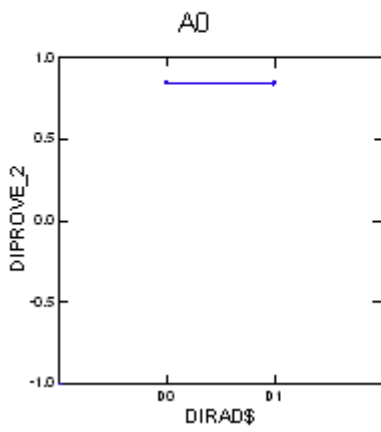
Least Squares Means



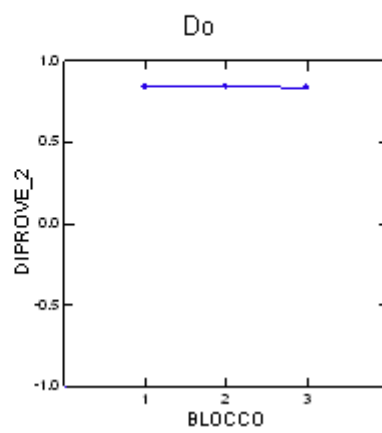
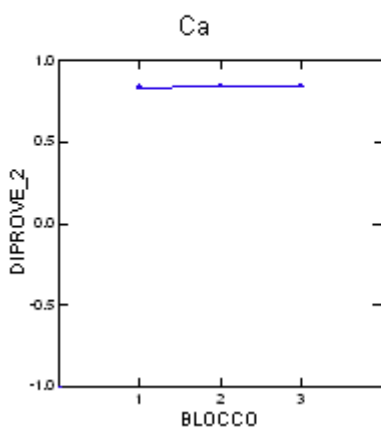
Least Squares Means



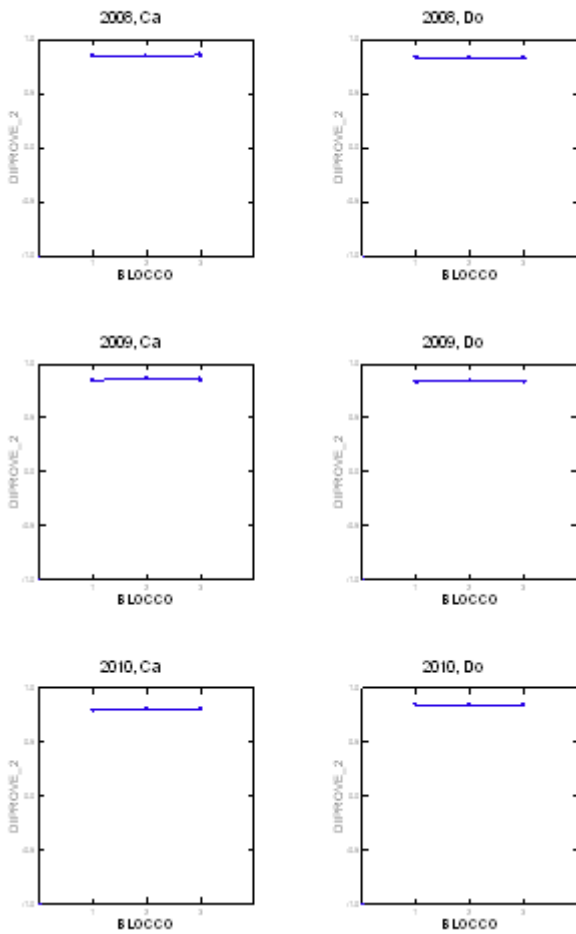
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 150 is an Outlier (Studentized Residual : -4.836)

Durbin-Watson D Statistic | 1.964
 First Order Autocorrelation | 0.017

Information Criteria

AIC | -977.420
 AIC (Corrected) | -955.585
 Schwarz's BIC | -876.446

▼ General Linear Model DIPROVE3 2008-2009-2010

Data for the following results were selected according to
 SELECT (VARIETA\$ = 'Cabernet')

Effects coding used for categorical variables in model.
 The categorical values encountered during processing are

Variables	Levels		
ANNO (3 levels)	2,008.000	2,009.000	2,010.000
LOCALITA\$ (2 levels)	Ca	Do	

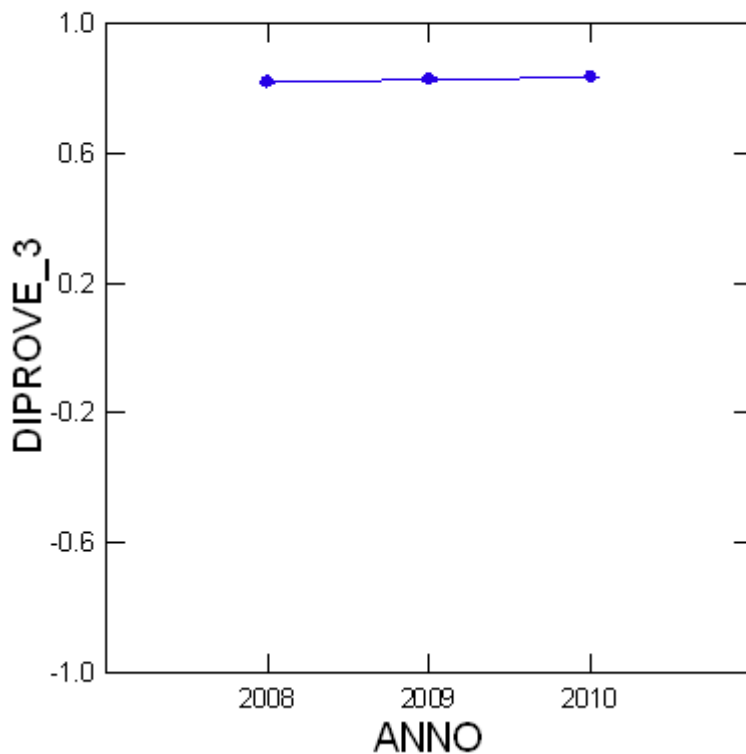
N_GEMME\$ (2 levels)	C0	C1	
DEFOGL\$ (2 levels)	A0	A1	
DIRAD\$ (2 levels)	D0	D1	
BLOCCO (3 levels)	1.000	2.000	3.000

Dependent Variable	DIPROVE_3
N	144
Multiple R	0.876
Squared Multiple R	0.768

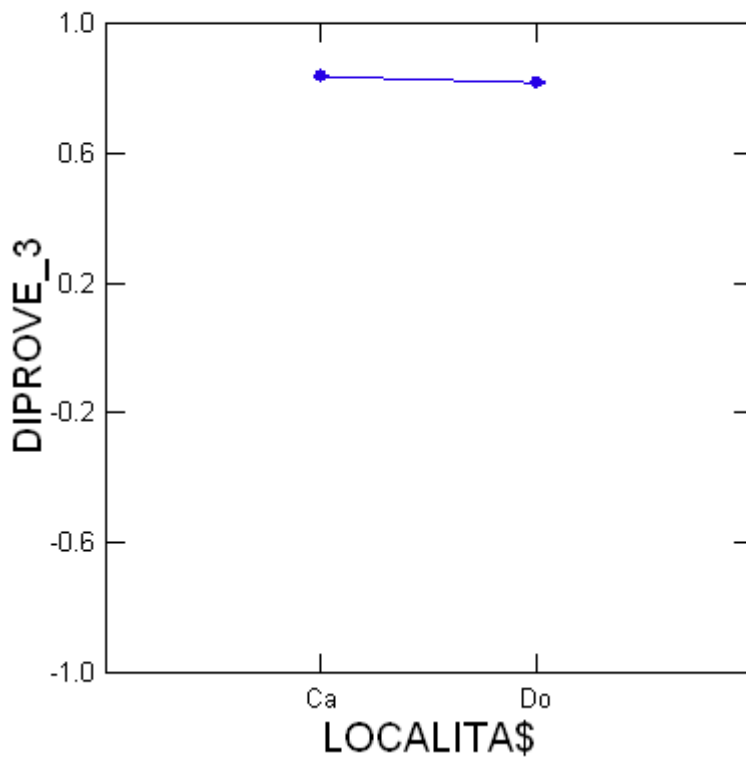
Analysis of Variance

Source	Type III SS	df	Mean Squares	F-ratio	p-value
ANNO	0.005	2	0.003	33.034	0.000
LOCALITA\$	0.010	1	0.010	127.882	0.000
N_GEMME\$	0.000	1	0.000	0.936	0.335
DEFOGL\$	0.000	1	0.000	6.206	0.014
DIRAD\$	0.000	1	0.000	0.240	0.625
LOCALITA\$*ANNO	0.004	2	0.002	24.782	0.000
N_GEMME\$*ANNO	0.000	2	0.000	0.750	0.475
DEFOGL\$*ANNO	0.001	2	0.000	3.279	0.041
DIRAD\$*ANNO	0.000	2	0.000	0.560	0.573
N_GEMME\$*LOCALITA\$	0.000	1	0.000	0.740	0.392
DEFOGL\$*LOCALITA\$	0.000	1	0.000	0.197	0.658
DIRAD\$*LOCALITA\$	0.000	1	0.000	0.005	0.943
DEFOGL\$*N_GEMME\$	0.000	1	0.000	2.463	0.119
DIRAD\$*N_GEMME\$	0.000	1	0.000	4.339	0.040
DIRAD\$*DEFOGL\$	0.000	1	0.000	0.219	0.640
BLOCCO(LOCALITA\$)	0.002	4	0.000	5.378	0.001
BLOCCO*ANNO(LOCALITA\$)	0.002	8	0.000	3.633	0.001
Error	0.008	111	0.000		

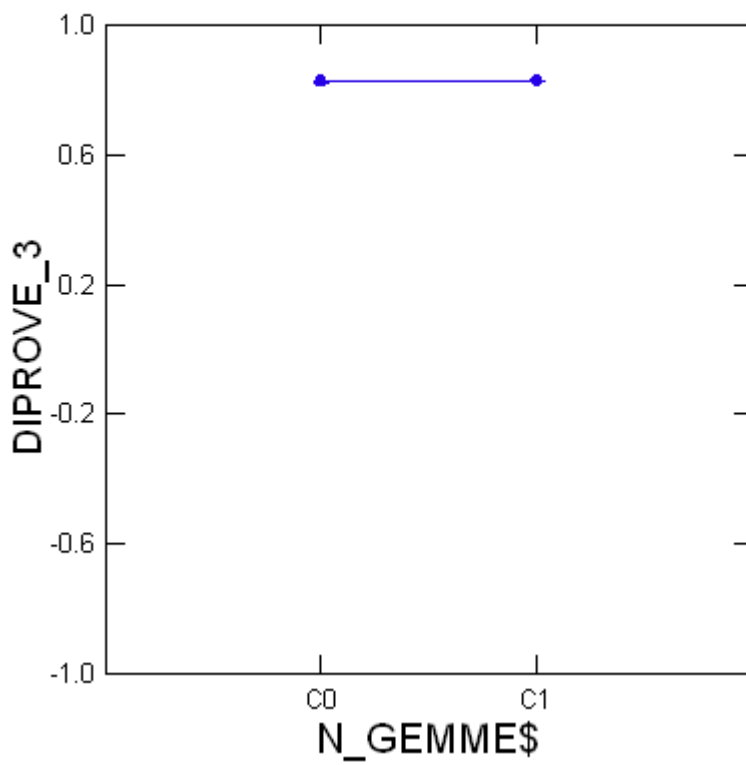
Least Squares Means



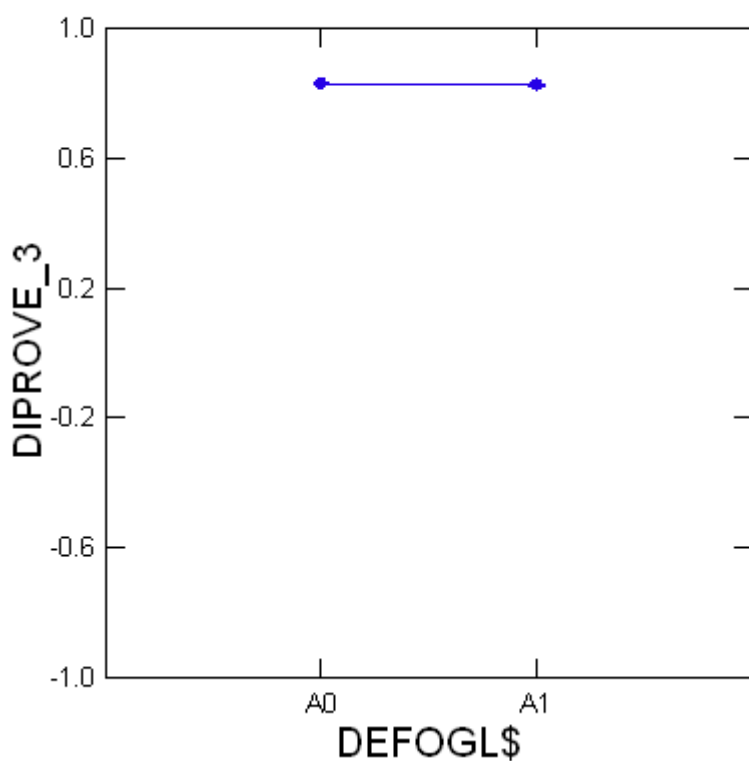
Least Squares Means



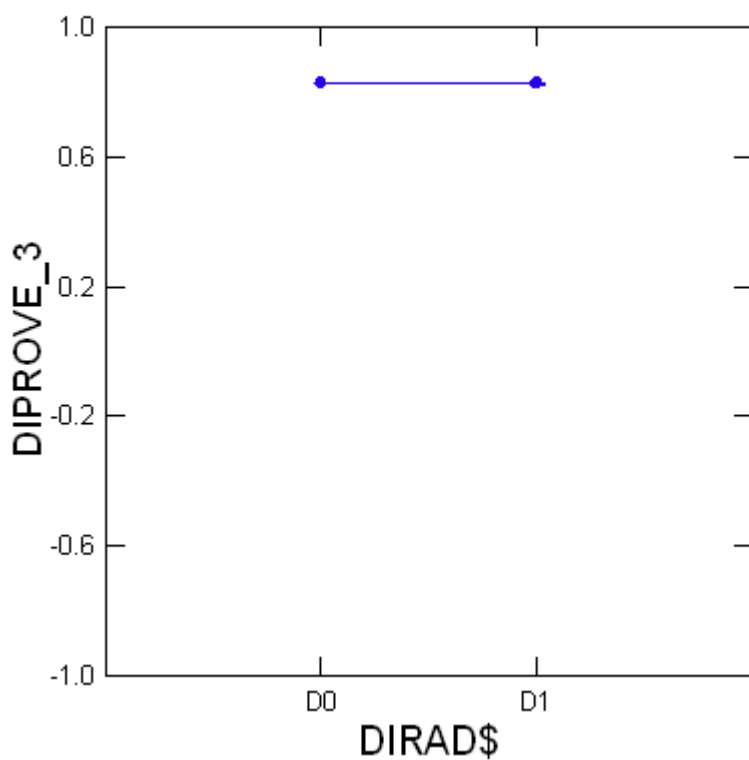
Least Squares Means



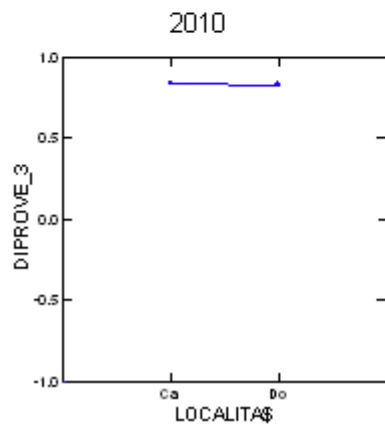
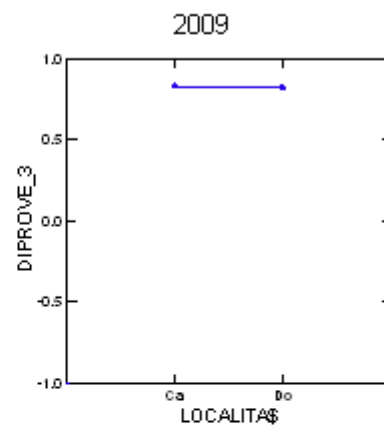
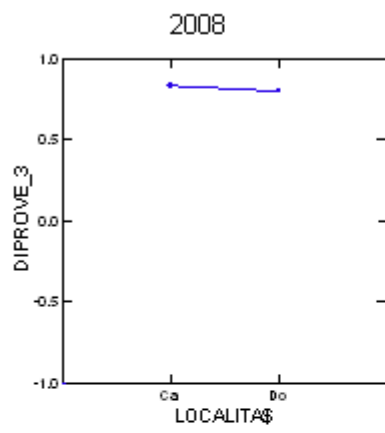
Least Squares Means



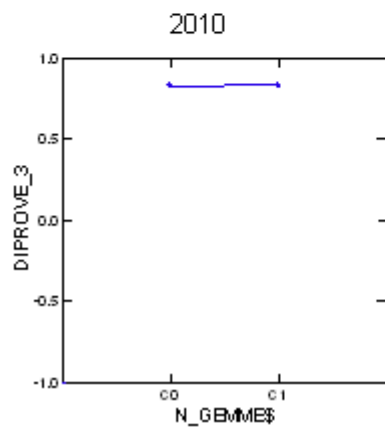
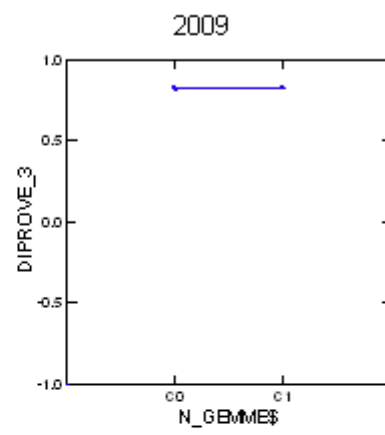
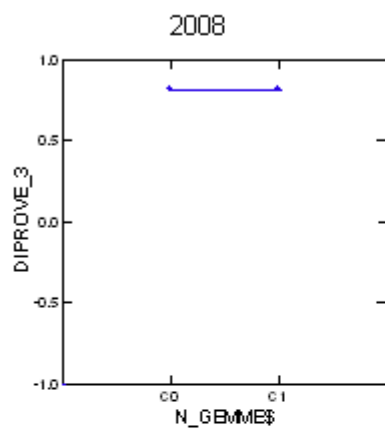
Least Squares Means



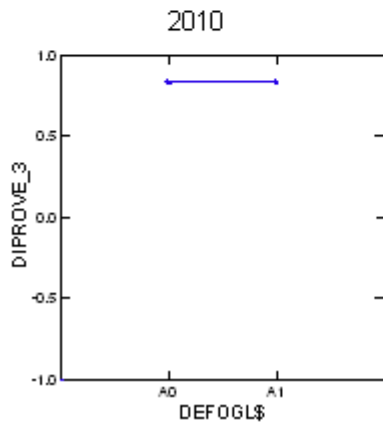
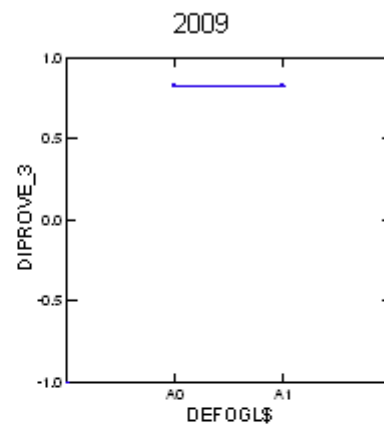
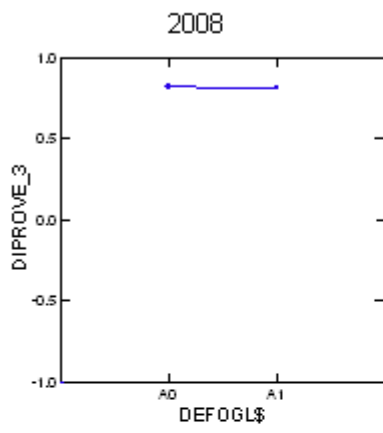
Least Squares Means



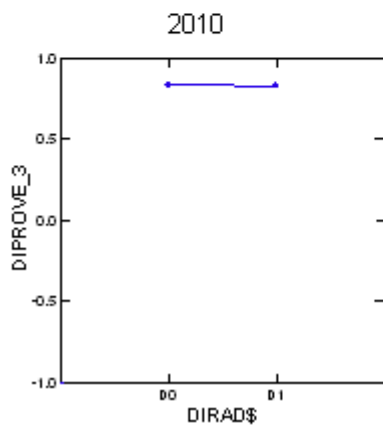
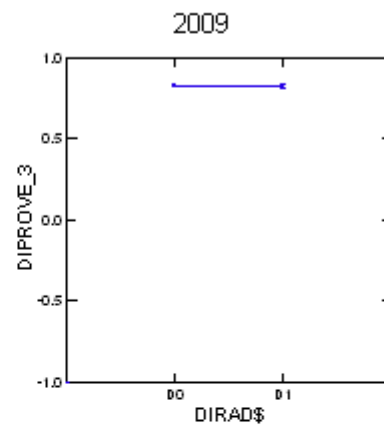
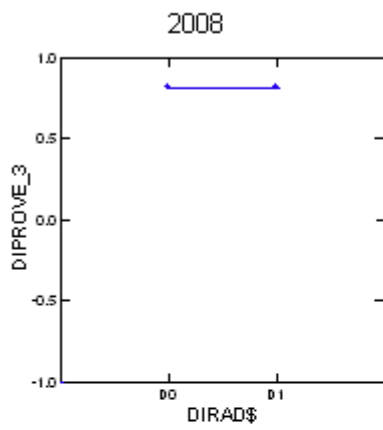
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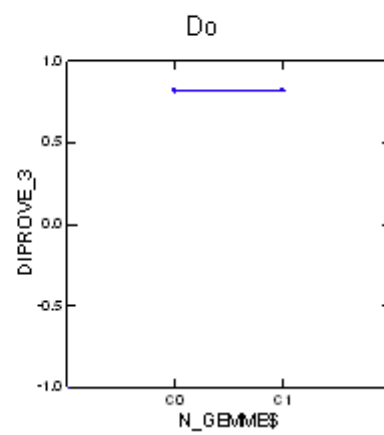
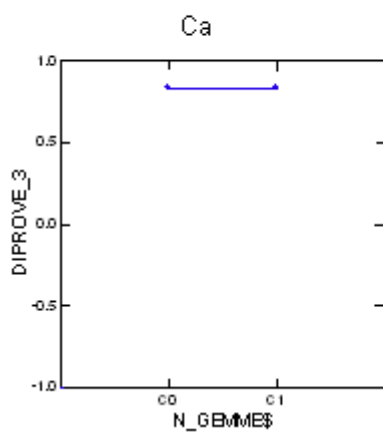
Least Squares Means



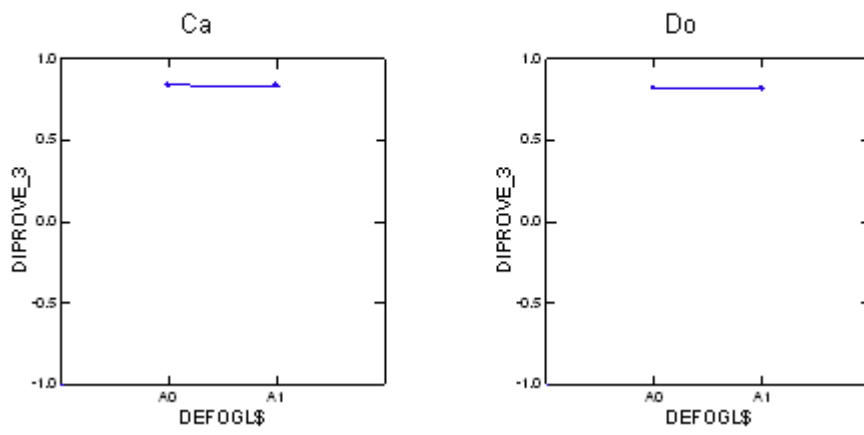
Least Squares Means



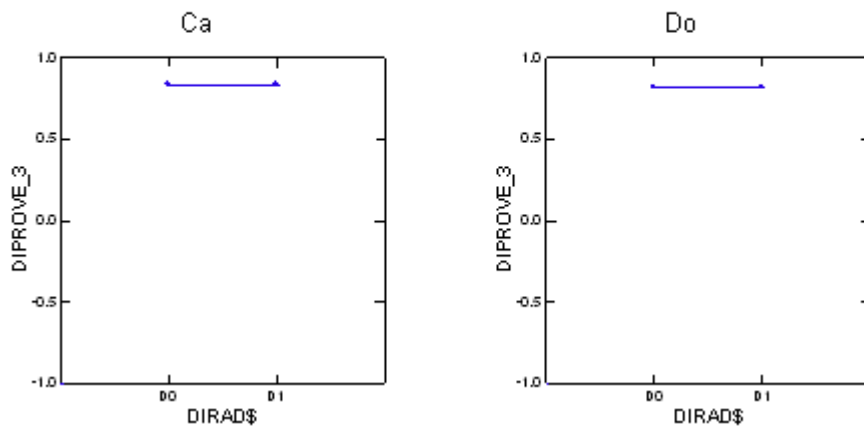
Least Squares Means



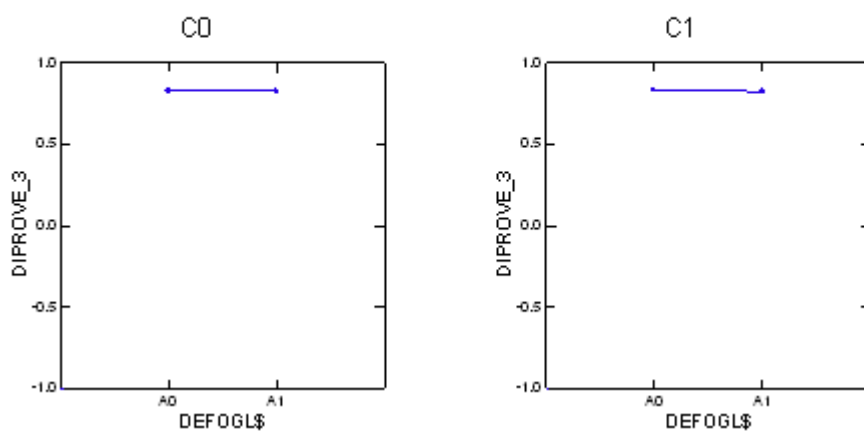
Least Squares Means



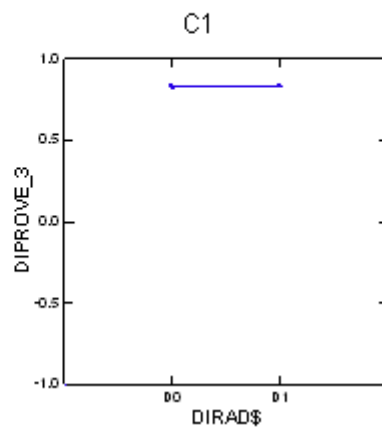
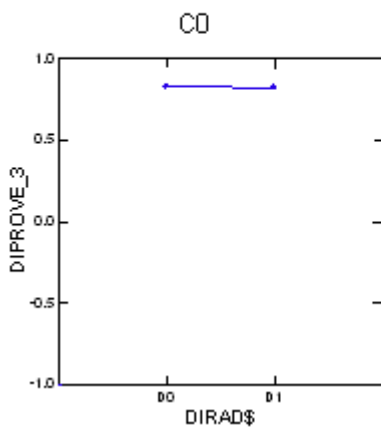
Least Squares Means



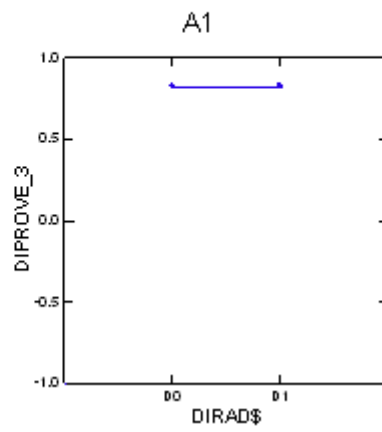
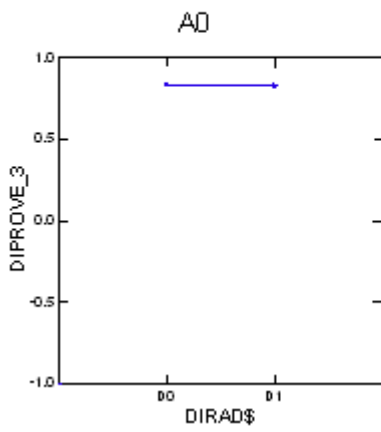
Least Squares Means



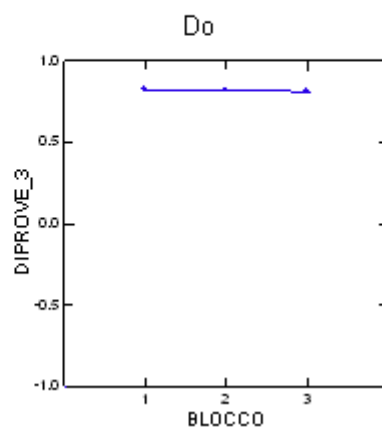
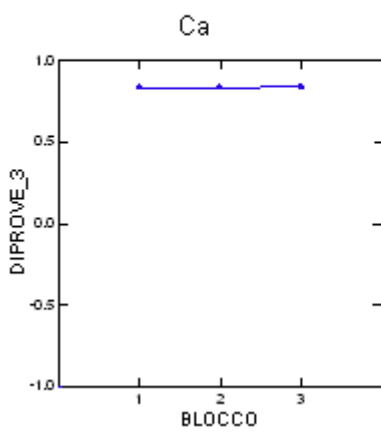
Least Squares Means



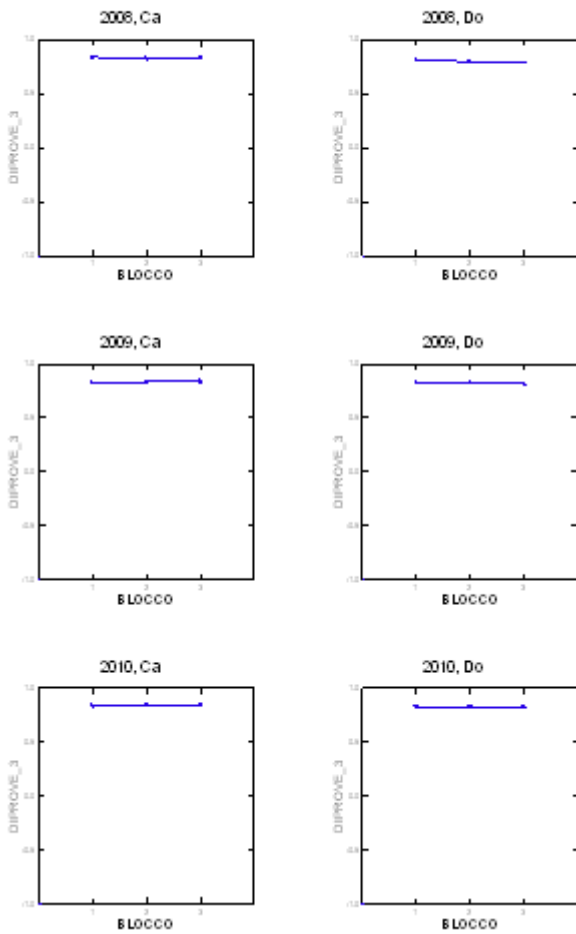
Least Squares Means



Least Squares Means



Least Squares Means



*** WARNING *** :

Case 329 is an Outlier (Studentized Residual : -3.896)

Durbin-Watson D Statistic | 1.805
First Order Autocorrelation | 0.093

Information Criteria

AIC | -925.840
AIC (Corrected) | -904.005
Schwarz's BIC | -824.866

Allegato 2.3 – Modelli lineari

BROLIO 2008 IBIMET

PvinacciaA

Analysis of Variance Table

Response: Br_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	2038219	2038219	31.8985	6.014e-05	***
IBIMET_3	1	7849	7849	0.1228	0.73119	
N_Gemme	1	5055	5055	0.0791	0.78262	
Defogl	1	150317	150317	2.3525	0.14737	
Dirad	1	26236	26236	0.4106	0.53202	
Vigoria	2	143063	71532	1.1195	0.35399	
N_Gemme:Defogl	1	51968	51968	0.8133	0.38240	
N_Gemme:Dirad	1	18287	18287	0.2862	0.60107	
Defogl:Dirad	1	240687	240687	3.7668	0.07269	.
N_Gemme:Vigoria	2	11669	5835	0.0913	0.91327	
Defogl:Vigoria	2	231843	115922	1.8142	0.19925	
Dirad:Vigoria	2	241749	120874	1.8917	0.18741	
N_Gemme:Defogl:Dirad	1	43162	43162	0.6755	0.42492	
Residuals	14	894558	63897			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.834

IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Br_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.00026477	0.00026477	9.6908	0.007633	**
IBIMET_3	1	0.00003575	0.00003575	1.3083	0.271880	
N_Gemme	1	0.00000003	0.00000003	0.0011	0.974480	
Defogl	1	0.00034380	0.00034380	12.5834	0.003218	**
Dirad	1	0.00002282	0.00002282	0.8353	0.376217	
Vigoria	2	0.00020993	0.00010497	3.8419	0.046768	*
N_Gemme:Defogl	1	0.00003774	0.00003774	1.3812	0.259506	
N_Gemme:Dirad	1	0.00001229	0.00001229	0.4500	0.513257	
Defogl:Dirad	1	0.00000865	0.00000865	0.3168	0.582453	
N_Gemme:Vigoria	2	0.00006097	0.00003049	1.1158	0.355116	
Defogl:Vigoria	2	0.00003478	0.00001739	0.6366	0.543751	
Dirad:Vigoria	2	0.00003689	0.00001844	0.6750	0.524967	
N_Gemme:Defogl:Dirad	1	0.00000288	0.00000288	0.1054	0.750269	
Residuals	14	0.00038250	0.00002732			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.64

IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Br_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	12.3283	12.3283	9.3080	0.008634	**
IBIMET_3	1	1.7029	1.7029	1.2857	0.275883	
N_Gemme	1	0.0123	0.0123	0.0093	0.924493	
Defogl	1	15.8643	15.8643	11.9776	0.003821	**
Dirad	1	0.9193	0.9193	0.6941	0.418765	
Vigoria	2	9.5810	4.7905	3.6169	0.054164	.
N_Gemme:Defogl	1	1.7139	1.7139	1.2940	0.274406	
N_Gemme:Dirad	1	0.5895	0.5895	0.4450	0.515546	
Defogl:Dirad	1	0.2802	0.2802	0.2115	0.652624	
N_Gemme:Vigoria	2	2.8178	1.4089	1.0637	0.371484	
Defogl:Vigoria	2	1.8706	0.9353	0.7062	0.510293	
Dirad:Vigoria	2	1.7576	0.8788	0.6635	0.530513	
N_Gemme:Defogl:Dirad	1	0.1468	0.1468	0.1108	0.744127	
Residuals	14	18.5429	1.3245			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.632

IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Br_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.036222	0.036222	8.3198	0.01201	*
IBIMET_3	1	0.011046	0.011046	2.5373	0.13351	
N_Gemme	1	0.000001	0.000001	0.0002	0.98983	
Defogl	1	0.004080	0.004080	0.9371	0.34945	
Dirad	1	0.000047	0.000047	0.0108	0.91852	
Vigoria	2	0.023246	0.011623	2.6698	0.10418	
N_Gemme:Defogl	1	0.000148	0.000148	0.0340	0.85635	
N_Gemme:Dirad	1	0.000244	0.000244	0.0560	0.81641	
Defogl:Dirad	1	0.003645	0.003645	0.8373	0.37566	
N_Gemme:Vigoria	2	0.002846	0.001423	0.3269	0.72653	
Defogl:Vigoria	2	0.005273	0.002636	0.6055	0.55947	
Dirad:Vigoria	2	0.003965	0.001982	0.4553	0.64332	
N_Gemme:Defogl:Dirad	1	0.004279	0.004279	0.9829	0.33831	
Residuals	14	0.060951	0.004354			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.611

IBIMET_3: n.s

AcTitola

Analysis of Variance Table

Response: Br_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	3.7138	3.7138	29.4853	8.869e-05	***
IBIMET_3	1	2.3019	2.3019	18.2758	0.0007699	***
N_Gemme	1	0.0461	0.0461	0.3663	0.5547011	
Defogl	1	0.5466	0.5466	4.3401	0.0560381	.
Dirad	1	0.1147	0.1147	0.9110	0.3560424	

Viguria	2	1.5717	0.7859	6.2393	0.0115514	*
N_Gemme:Defogl	1	0.1806	0.1806	1.4342	0.2509714	
N_Gemme:Dirad	1	0.0017	0.0017	0.0138	0.9080677	
Defogl:Dirad	1	0.0031	0.0031	0.0248	0.8770001	
N_Gemme:Viguria	2	0.0150	0.0075	0.0596	0.9424112	
Defogl:Viguria	2	0.8940	0.4470	3.5491	0.0566490	.
Dirad:Viguria	2	0.4373	0.2187	1.7361	0.2120596	
N_Gemme:Defogl:Dirad	1	0.0360	0.0360	0.2855	0.6015041	
Residuals	14	1.7634	0.1260			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.823
 IBIMET_3: 0.752

AcTartA

Analysis of Variance Table

Response: Br_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.05541	0.05541	0.4383	0.5187087	
IBIMET_3	1	2.92083	2.92083	23.1013	0.0002792	***
N_Gemme	1	0.27065	0.27065	2.1406	0.1655284	
Defogl	1	0.79581	0.79581	6.2942	0.0250365	*
Dirad	1	0.26727	0.26727	2.1139	0.1680146	
Viguria	2	2.20864	1.10432	8.7342	0.0034496	**
N_Gemme:Defogl	1	0.08257	0.08257	0.6531	0.4325416	
N_Gemme:Dirad	1	0.01915	0.01915	0.1515	0.7029827	
Defogl:Dirad	1	0.00879	0.00879	0.0695	0.7958598	
N_Gemme:Viguria	2	0.03643	0.01821	0.1441	0.8671072	
Defogl:Viguria	2	0.90252	0.45126	3.5691	0.0559020	.
Dirad:Viguria	2	0.70921	0.35460	2.8046	0.0945524	.
N_Gemme:Defogl:Dirad	1	0.02264	0.02264	0.1790	0.6786324	
Residuals	14	1.77010	0.12644			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: 0.789

AcMalicoA

Analysis of Variance Table

Response: Br_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	3.7514	3.7514	96.9208	1.133e-07	***
IBIMET_3	1	0.2291	0.2291	5.9189	0.02898	*
N_Gemme	1	0.0517	0.0517	1.3352	0.26722	
Defogl	1	0.2936	0.2936	7.5845	0.01552	*
Dirad	1	0.0213	0.0213	0.5508	0.47024	
Viguria	2	0.3861	0.1931	4.9882	0.02314	*
N_Gemme:Defogl	1	0.1204	0.1204	3.1096	0.09963	.
N_Gemme:Dirad	1	0.0220	0.0220	0.5671	0.46388	
Defogl:Dirad	1	0.0008	0.0008	0.0207	0.88767	
N_Gemme:Viguria	2	0.0097	0.0049	0.1254	0.88309	
Defogl:Viguria	2	0.0822	0.0411	1.0616	0.37216	
Dirad:Viguria	2	0.0578	0.0289	0.7463	0.49208	

N_Gemme:Defogl:Dirad 1 0.0204 0.0204 0.5270 0.47983
 Residuals 14 0.5419 0.0387

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.935

IBIMET_3: -0.545

KA

Analysis of Variance Table

Response: Br_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.011484	0.011484	2.4252	0.141711
IBIMET_3	1	0.067156	0.067156	14.1820	0.002087 **
N_Gemme	1	0.000324	0.000324	0.0684	0.797512
Defogl	1	0.000243	0.000243	0.0514	0.823993
Dirad	1	0.001726	0.001726	0.3646	0.555636
Vigoria	2	0.007716	0.003858	0.8147	0.462680
N_Gemme:Defogl	1	0.017336	0.017336	3.6610	0.076373 .
N_Gemme:Dirad	1	0.000043	0.000043	0.0091	0.925180
Defogl:Dirad	1	0.000019	0.000019	0.0040	0.950281
N_Gemme:Vigoria	2	0.004525	0.002263	0.4778	0.629896
Defogl:Vigoria	2	0.019085	0.009542	2.0152	0.170161
Dirad:Vigoria	2	0.006548	0.003274	0.6914	0.517171
N_Gemme:Defogl:Dirad	1	0.002145	0.002145	0.4529	0.511916
Residuals	14	0.066294	0.004735		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: 0.709

ApaA

Analysis of Variance Table

Response: Br_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	4721.5	4721.5	15.9821	0.001322 **
IBIMET_3	1	3056.6	3056.6	10.3466	0.006212 **
N_Gemme	1	0.7	0.7	0.0023	0.962237
Defogl	1	2858.7	2858.7	9.6766	0.007667 **
Dirad	1	9.4	9.4	0.0317	0.861333
Vigoria	2	1203.1	601.5	2.0362	0.167416
N_Gemme:Defogl	1	217.9	217.9	0.7376	0.404905
N_Gemme:Dirad	1	420.5	420.5	1.4234	0.252667
Defogl:Dirad	1	41.2	41.2	0.1394	0.714519
N_Gemme:Vigoria	2	667.1	333.6	1.1291	0.351079
Defogl:Vigoria	2	579.1	289.6	0.9802	0.399580
Dirad:Vigoria	2	370.7	185.3	0.6274	0.548356
N_Gemme:Defogl:Dirad	1	172.1	172.1	0.5827	0.457938
Residuals	14	4135.9	295.4		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.73

IBIMET_3: -0.652

Antoctot1A

Analysis of Variance Table

Response: Br_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	169781	169781	32.3539	5.602e-05 ***
IBIMET_3	1	1114	1114	0.2123	0.6521
N_Gemme	1	468	468	0.0893	0.7695
Defogl	1	17310	17310	3.2986	0.0908 .
Dirad	1	1	1	0.0001	0.9910
Vigoria	2	22321	11161	2.1268	0.1561
N_Gemme:Defogl	1	10571	10571	2.0144	0.1777
N_Gemme:Dirad	1	740	740	0.1410	0.7129
Defogl:Dirad	1	1778	1778	0.3388	0.5698
N_Gemme:Vigoria	2	9315	4658	0.8876	0.4336
Defogl:Vigoria	2	10079	5040	0.9604	0.4066
Dirad:Vigoria	2	4179	2089	0.3982	0.6789
N_Gemme:Defogl:Dirad	1	3905	3905	0.7441	0.4029
Residuals	14	73467	5248		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.835

IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Br_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	486552	486552	22.3242	0.0003255 ***
IBIMET_3	1	2121	2121	0.0973	0.7596571
N_Gemme	1	4864	4864	0.2232	0.6439237
Defogl	1	93683	93683	4.2984	0.0570828 .
Dirad	1	152	152	0.0070	0.9345278
Vigoria	2	65994	32997	1.5140	0.2539545
N_Gemme:Defogl	1	12088	12088	0.5546	0.4687471
N_Gemme:Dirad	1	35386	35386	1.6236	0.2233358
Defogl:Dirad	1	261	261	0.0120	0.9144467
N_Gemme:Vigoria	2	21008	10504	0.4820	0.6274504
Defogl:Vigoria	2	8313	4157	0.1907	0.8284789
Dirad:Vigoria	2	17946	8973	0.4117	0.6702914
N_Gemme:Defogl:Dirad	1	18456	18456	0.8468	0.3730374
Residuals	14	305127	21795		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.784

IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Br_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	381603	381603	33.1596	4.95e-05	***
IBIMET_3	1	26506	26506	2.3032	0.151360	
N_Gemme	1	3805	3805	0.3307	0.574387	
Defogl	1	166153	166153	14.4379	0.001952	**
Dirad	1	391	391	0.0340	0.856366	
Vigoria	2	136164	68082	5.9161	0.013733	*
N_Gemme:Defogl	1	10742	10742	0.9334	0.350366	
N_Gemme:Dirad	1	36268	36268	3.1515	0.097591	.
Defogl:Dirad	1	6119	6119	0.5317	0.477911	
N_Gemme:Vigoria	2	4706	2353	0.2045	0.817464	
Defogl:Vigoria	2	67506	33753	2.9330	0.086324	.
Dirad:Vigoria	2	27238	13619	1.1835	0.335066	
N_Gemme:Defogl:Dirad	1	7365	7365	0.6399	0.437093	
Residuals	14	161113	11508			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.839
IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Br_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	1090120	1090120	12.7131	0.003104	**
IBIMET_3	1	292928	292928	3.4162	0.085793	.
N_Gemme	1	3793	3793	0.0442	0.836446	
Defogl	1	849619	849619	9.9083	0.007123	**
Dirad	1	94	94	0.0011	0.974056	
Vigoria	2	370967	185484	2.1631	0.151838	
N_Gemme:Defogl	1	7	7	0.0001	0.992719	
N_Gemme:Dirad	1	405280	405280	4.7264	0.047346	*
Defogl:Dirad	1	4690	4690	0.0547	0.818477	
N_Gemme:Vigoria	2	28689	14345	0.1673	0.847623	
Defogl:Vigoria	2	6821	3410	0.0398	0.961117	
Dirad:Vigoria	2	16798	8399	0.0980	0.907309	
N_Gemme:Defogl:Dirad	1	36588	36588	0.4267	0.524198	
Residuals	14	1200471	85748			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.69
IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	7849966	7849966	256.1616	2.151e-10	***
IBIMET_3	1	711	711	0.0232	0.88111	
N_Gemme	1	74526	74526	2.4319	0.14120	
Defogl	1	1029635	1029635	33.5992	4.632e-05	***
Dirad	1	189832	189832	6.1946	0.02602	*
Vigoria	2	198955	99477	3.2462	0.06946	.

N_Gemme:Defogl	1	23	23	0.0008	0.97836
N_Gemme:Dirad	1	106966	106966	3.4905	0.08279 .
Defogl:Dirad	1	188388	188388	6.1475	0.02650 *
N_Gemme:Vigoria	2	36440	18220	0.5946	0.56516
Defogl:Vigoria	2	7010	3505	0.1144	0.89275
Dirad:Vigoria	2	11848	5924	0.1933	0.82639
N_Gemme:Defogl:Dirad	1	18001	18001	0.5874	0.45615
Residuals	14	429024	30645		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.974
IBIMET_3: n.s

DensMostoB

Analysis of Variance Table

Response: Br_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.00015104	0.00015104	5.8875	0.029344 *
IBIMET_3	1	0.00000753	0.00000753	0.2936	0.596467
N_Gemme	1	0.00000369	0.00000369	0.1437	0.710362
Defogl	1	0.00039824	0.00039824	15.5238	0.001481 **
Dirad	1	0.00001103	0.00001103	0.4300	0.522628
Vigoria	2	0.00014106	0.00007053	2.7493	0.098373 .
N_Gemme:Defogl	1	0.00000121	0.00000121	0.0472	0.831059
N_Gemme:Dirad	1	0.00013184	0.00013184	5.1393	0.039758 *
Defogl:Dirad	1	0.00000375	0.00000375	0.1464	0.707773
N_Gemme:Vigoria	2	0.00000084	0.00000042	0.0165	0.983695
Defogl:Vigoria	2	0.00005930	0.00002965	1.1557	0.343125
Dirad:Vigoria	2	0.00002302	0.00001151	0.4486	0.647365
N_Gemme:Defogl:Dirad	1	0.00001851	0.00001851	0.7215	0.409946
Residuals	14	0.00035915	0.00002565		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.544
IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Br_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	6.8384	6.8384	5.9648	0.028462 *
IBIMET_3	1	0.2479	0.2479	0.2162	0.649066
N_Gemme	1	0.1767	0.1767	0.1541	0.700539
Defogl	1	18.9585	18.9585	16.5367	0.001155 **
Dirad	1	0.7451	0.7451	0.6499	0.433614
Vigoria	2	5.9969	2.9985	2.6154	0.108368
N_Gemme:Defogl	1	0.0379	0.0379	0.0331	0.858330
N_Gemme:Dirad	1	6.0860	6.0860	5.3086	0.037066 *
Defogl:Dirad	1	0.1105	0.1105	0.0964	0.760785
N_Gemme:Vigoria	2	0.0525	0.0263	0.0229	0.977397
Defogl:Vigoria	2	2.7005	1.3503	1.1778	0.336695
Dirad:Vigoria	2	1.2057	0.6028	0.5258	0.602286
N_Gemme:Defogl:Dirad	1	0.7682	0.7682	0.6700	0.426750

Residuals 14 16.0502 1.1464

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.547

IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Br_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.001177	0.001177	0.4624	0.507599
IBIMET_3	1	0.036821	0.036821	14.4628	0.001939 **
N_Gemme	1	0.000689	0.000689	0.2708	0.610945
Defogl	1	0.003586	0.003586	1.4084	0.255077
Dirad	1	0.001225	0.001225	0.4811	0.499263
Vigoria	2	0.011333	0.005667	2.2258	0.144761
N_Gemme:Defogl	1	0.000314	0.000314	0.1233	0.730690
N_Gemme:Dirad	1	0.019332	0.019332	7.5934	0.015473 *
Defogl:Dirad	1	0.000075	0.000075	0.0296	0.865863
N_Gemme:Vigoria	2	0.001860	0.000930	0.3653	0.700403
Defogl:Vigoria	2	0.003415	0.001707	0.6707	0.527050
Dirad:Vigoria	2	0.000717	0.000358	0.1408	0.869890
N_Gemme:Defogl:Dirad	1	0.004751	0.004751	1.8661	0.193461
Residuals	14	0.035642	0.002546		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: -0.713

AcTitolB

Analysis of Variance Table

Response: Br_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1.7781	1.7781	9.4304	0.0082984 **
IBIMET_3	1	4.2513	4.2513	22.5467	0.0003114 ***
N_Gemme	1	0.0381	0.0381	0.2022	0.6598502
Defogl	1	0.6562	0.6562	3.4801	0.0832043 .
Dirad	1	0.0525	0.0525	0.2785	0.6059535
Vigoria	2	4.7978	2.3989	12.7227	0.0007095 ***
N_Gemme:Defogl	1	0.0578	0.0578	0.3066	0.5885323
N_Gemme:Dirad	1	0.0120	0.0120	0.0638	0.8042230
Defogl:Dirad	1	0.0079	0.0079	0.0416	0.8412451
N_Gemme:Vigoria	2	0.1180	0.0590	0.3129	0.7362832
Defogl:Vigoria	2	0.0652	0.0326	0.1730	0.8429278
Dirad:Vigoria	2	0.1678	0.0839	0.4448	0.6496793
N_Gemme:Defogl:Dirad	1	0.1071	0.1071	0.5682	0.4634562
Residuals	14	2.6398	0.1886		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.634

IBIMET_3: 0.785

AcTartB

Analysis of Variance Table

Response: Br_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.1714	0.1714	1.5956	0.2271666
IBIMET_3	1	3.8087	3.8087	35.4539	3.521e-05 ***
N_Gemme	1	0.0339	0.0339	0.3154	0.5832562
Defogl	1	3.0112	3.0112	28.0295	0.0001133 ***
Dirad	1	0.3538	0.3538	3.2934	0.0910389 .
Vigoria	2	2.9600	1.4800	13.7769	0.0004927 ***
N_Gemme:Defogl	1	0.0248	0.0248	0.2308	0.6383555
N_Gemme:Dirad	1	0.0556	0.0556	0.5175	0.4837466
Defogl:Dirad	1	0.0022	0.0022	0.0203	0.8886928
N_Gemme:Vigoria	2	0.0370	0.0185	0.1724	0.8434351
Defogl:Vigoria	2	0.2799	0.1400	1.3028	0.3027694
Dirad:Vigoria	2	0.0054	0.0027	0.0252	0.9751761
N_Gemme:Defogl:Dirad	1	0.0652	0.0652	0.6067	0.4490163
Residuals	14	1.5040	0.1074		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: 0.847

AcMalicoB

Analysis of Variance Table

Response: Br_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	2.33336	2.33336	13.1446	0.002755 **
IBIMET_3	1	0.00240	0.00240	0.0135	0.909116
N_Gemme	1	0.07396	0.07396	0.4166	0.529059
Defogl	1	0.92914	0.92914	5.2342	0.038222 *
Dirad	1	0.13296	0.13296	0.7490	0.401371
Vigoria	2	0.35176	0.17588	0.9908	0.395879
N_Gemme:Defogl	1	0.00059	0.00059	0.0033	0.954751
N_Gemme:Dirad	1	0.21698	0.21698	1.2223	0.287543
Defogl:Dirad	1	0.00029	0.00029	0.0016	0.968344
N_Gemme:Vigoria	2	0.08446	0.04223	0.2379	0.791416
Defogl:Vigoria	2	0.19849	0.09924	0.5591	0.583994
Dirad:Vigoria	2	0.24570	0.12285	0.6920	0.516883
N_Gemme:Defogl:Dirad	1	0.00723	0.00723	0.0408	0.842924
Residuals	14	2.48521	0.17751		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.696
IBIMET_3: n.s

KB

Analysis of Variance Table

Response: Br_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IBIMET_2	1	0.022871	0.022871	2.0129	0.17784
IBIMET_3	1	0.036960	0.036960	3.2529	0.09285 .
N_Gemme	1	0.012265	0.012265	1.0795	0.31642
Defogl	1	0.048765	0.048765	4.2918	0.05725 .
Dirad	1	0.000000	0.000000	0.0000	0.99959
Vigoria	2	0.042085	0.021043	1.8520	0.19338
N_Gemme:Defogl	1	0.000059	0.000059	0.0052	0.94375
N_Gemme:Dirad	1	0.036344	0.036344	3.1986	0.09535 .
Defogl:Dirad	1	0.001982	0.001982	0.1744	0.68256
N_Gemme:Vigoria	2	0.007861	0.003930	0.3459	0.71344
Defogl:Vigoria	2	0.025600	0.012800	1.1265	0.35184
Dirad:Vigoria	2	0.006114	0.003057	0.2691	0.76797
N_Gemme:Defogl:Dirad	1	0.009255	0.009255	0.8145	0.38205
Residuals	14	0.159071	0.011362		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

ApaB

Analysis of Variance Table

Response: Br_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	3257.5	3257.5	15.8828	0.001354	**
IBIMET_3	1	2188.2	2188.2	10.6693	0.005626	**
N_Gemme	1	25.2	25.2	0.1230	0.730979	
Defogl	1	5817.9	5817.9	28.3670	0.000107	***
Dirad	1	20.0	20.0	0.0973	0.759638	
Vigoria	2	1194.7	597.4	2.9126	0.087575	.
N_Gemme:Defogl	1	76.7	76.7	0.3738	0.550728	
N_Gemme:Dirad	1	300.2	300.2	1.4636	0.246387	
Defogl:Dirad	1	108.7	108.7	0.5301	0.478549	
N_Gemme:Vigoria	2	366.2	183.1	0.8927	0.431626	
Defogl:Vigoria	2	645.5	322.8	1.5737	0.241819	
Dirad:Vigoria	2	52.0	26.0	0.1268	0.881945	
N_Gemme:Defogl:Dirad	1	581.1	581.1	2.8332	0.114492	
Residuals	14	2871.3	205.1			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.729
IBIMET_3: -0.658

Antoctot1B

Analysis of Variance Table

Response: Br_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	39531	39531	33.4675	4.725e-05	***
IBIMET_3	1	1652	1652	1.3988	0.25662	
N_Gemme	1	4910	4910	4.1572	0.06080	.
Defogl	1	40942	40942	34.6616	3.953e-05	***
Dirad	1	23	23	0.0193	0.89156	
Vigoria	2	10270	5135	4.3472	0.03400	*
N_Gemme:Defogl	1	302	302	0.2554	0.62118	

N_Gemme:Dirad	1	1918	1918	1.6240	0.22328
Defogl:Dirad	1	7532	7532	6.3765	0.02426 *
N_Gemme:Viguria	2	559	280	0.2367	0.79229
Defogl:Viguria	2	14805	7402	6.2670	0.01138 *
Dirad:Viguria	2	2498	1249	1.0572	0.37358
N_Gemme:Defogl:Dirad	1	2044	2044	1.7305	0.20948
Residuals	14	16537	1181		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.84
IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	19391	19391	1.0711	0.31824
IBIMET_3	1	1420	1420	0.0785	0.78349
N_Gemme	1	3192	3192	0.1763	0.68094
Defogl	1	69061	69061	3.8148	0.07109 .
Dirad	1	51111	51111	2.8233	0.11507
Viguria	2	25341	12671	0.6999	0.51320
N_Gemme:Defogl	1	2253	2253	0.1245	0.72950
N_Gemme:Dirad	1	210	210	0.0116	0.91583
Defogl:Dirad	1	65163	65163	3.5995	0.07862 .
N_Gemme:Viguria	2	3181	1590	0.0878	0.91640
Defogl:Viguria	2	104135	52068	2.8761	0.08986 .
Dirad:Viguria	2	7685	3842	0.2122	0.81132
N_Gemme:Defogl:Dirad	1	20349	20349	1.1241	0.30699
Residuals	14	253448	18103		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	44082	44082	5.7763	0.030669 *
IBIMET_3	1	78049	78049	10.2272	0.006446 **
N_Gemme	1	34	34	0.0045	0.947669
Defogl	1	474224	474224	62.1403	1.625e-06 ***
Dirad	1	251	251	0.0329	0.858682
Viguria	2	74588	37294	4.8868	0.024560 *
N_Gemme:Defogl	1	1499	1499	0.1965	0.664344
N_Gemme:Dirad	1	13020	13020	1.7060	0.212553
Defogl:Dirad	1	18707	18707	2.4513	0.139749
N_Gemme:Viguria	2	194	97	0.0127	0.987362
Defogl:Viguria	2	40638	20319	2.6625	0.104728
Dirad:Viguria	2	674	337	0.0441	0.956950
N_Gemme:Defogl:Dirad	1	7989	7989	1.0468	0.323601
Residuals	14	106841	7632		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.54

IBIMET_3: 0.65

Poliftot2B

Analysis of Variance Table

Response: Br_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	35073	35073	0.3743	0.5505
IBIMET_3	1	170310	170310	1.8178	0.1990
N_Gemme	1	35430	35430	0.3782	0.5485
Defogl	1	1214044	1214044	12.9578	0.0029 **
Dirad	1	279441	279441	2.9825	0.1062
Vigoria	2	254783	127392	1.3597	0.2886
N_Gemme:Defogl	1	11112	11112	0.1186	0.7357
N_Gemme:Dirad	1	37	37	0.0004	0.9845
Defogl:Dirad	1	194091	194091	2.0716	0.1720
N_Gemme:Vigoria	2	28240	14120	0.1507	0.8615
Defogl:Vigoria	2	297998	148999	1.5903	0.2386
Dirad:Vigoria	2	130783	65392	0.6979	0.5141
N_Gemme:Defogl:Dirad	1	77666	77666	0.8289	0.3780
Residuals	14	1311693	93692		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

BROLIO 2008 DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Br_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1355733	1355733	26.4299	0.0001896 ***
DIPROVE_2	1	1333491	1333491	25.9963	0.0002042 ***
DIPROVE_3	1	9907	9907	0.1931	0.6675411
N_Gemme	1	2023	2023	0.0394	0.8456593
Defogl	1	20310	20310	0.3959	0.5400907
Dirad	1	84811	84811	1.6534	0.2209319
Vigoria	2	9695	4847	0.0945	0.9104456
N_Gemme:Defogl	1	104840	104840	2.0439	0.1764120
N_Gemme:Dirad	1	3814	3814	0.0744	0.7893818
Defogl:Dirad	1	52075	52075	1.0152	0.3320483
N_Gemme:Vigoria	2	15109	7554	0.1473	0.8644807
Defogl:Vigoria	2	217498	108749	2.1201	0.1596330
Dirad:Vigoria	2	184518	92259	1.7986	0.2043666
N_Gemme:Defogl:Dirad	1	43999	43999	0.8578	0.3712399
Residuals	13	666840	51295		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.819
 DIPROVE_2: 0.816
 DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Br_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.00065375	0.00065375	41.7983	2.116e-05	***
DIPROVE_2	1	0.00017584	0.00017584	11.2425	0.005191	**
DIPROVE_3	1	0.00008085	0.00008085	5.1691	0.040600	*
N_Gemme	1	0.00000176	0.00000176	0.1124	0.742749	
Defogl	1	0.00006379	0.00006379	4.0783	0.064547	.
Dirad	1	0.00000644	0.00000644	0.4118	0.532206	
Vigoria	2	0.00006383	0.00003191	2.0405	0.169548	
N_Gemme:Defogl	1	0.00006788	0.00006788	4.3402	0.057528	.
N_Gemme:Dirad	1	0.00000932	0.00000932	0.5959	0.453956	
Defogl:Dirad	1	0.00000343	0.00000343	0.2194	0.647291	
N_Gemme:Vigoria	2	0.00002782	0.00001391	0.8895	0.434459	
Defogl:Vigoria	2	0.00002960	0.00001480	0.9461	0.413428	
Dirad:Vigoria	2	0.00006597	0.00003298	2.1089	0.160984	
N_Gemme:Defogl:Dirad	1	0.00000020	0.00000020	0.0128	0.911494	
Residuals	13	0.00020333	0.00001564			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.873
 DIPROVE_2: -0.681
 DIPROVE_3: 0.533

BrixA

Analysis of Variance Table

Response: Br_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	30.1210	30.1210	39.3136	2.878e-05	***
DIPROVE_2	1	8.1992	8.1992	10.7015	0.006074	**
DIPROVE_3	1	3.4990	3.4990	4.5669	0.052177	.
N_Gemme	1	0.1238	0.1238	0.1615	0.694274	
Defogl	1	3.0174	3.0174	3.9382	0.068720	.
Dirad	1	0.2239	0.2239	0.2923	0.597899	
Vigoria	2	3.0806	1.5403	2.0104	0.173489	
N_Gemme:Defogl	1	3.0338	3.0338	3.9596	0.068062	.
N_Gemme:Dirad	1	0.4657	0.4657	0.6078	0.449588	
Defogl:Dirad	1	0.2469	0.2469	0.3222	0.579934	
N_Gemme:Vigoria	2	1.4373	0.7186	0.9380	0.416376	
Defogl:Vigoria	2	1.5678	0.7839	1.0231	0.386674	
Dirad:Vigoria	2	3.1343	1.5672	2.0455	0.168911	
N_Gemme:Defogl:Dirad	1	0.0166	0.0166	0.0216	0.885379	
Residuals	13	9.9603	0.7662			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.867
 DIPROVE_2: -0.672
 DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Br_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.023187	0.0231874	7.8742	0.01486	*
DIPROVE_2	1	0.008930	0.0089298	3.0325	0.10521	
DIPROVE_3	1	0.016943	0.0169427	5.7536	0.03217	*
N_Gemme	1	0.002981	0.0029808	1.0123	0.33272	
Defogl	1	0.007233	0.0072325	2.4561	0.14108	
Dirad	1	0.000001	0.0000007	0.0002	0.98800	
Vigoria	2	0.039427	0.0197137	6.6946	0.01003	*
N_Gemme:Defogl	1	0.001163	0.0011626	0.3948	0.54067	
N_Gemme:Dirad	1	0.002183	0.0021827	0.7412	0.40487	
Defogl:Dirad	1	0.000007	0.0000074	0.0025	0.96088	
N_Gemme:Vigoria	2	0.001218	0.0006088	0.2068	0.81584	
Defogl:Vigoria	2	0.004434	0.0022171	0.7529	0.49046	
Dirad:Vigoria	2	0.005614	0.0028068	0.9532	0.41089	
N_Gemme:Defogl:Dirad	1	0.004392	0.0043922	1.4916	0.24366	
Residuals	13	0.038281	0.0029447			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.614
DIPROVE_2: n.s
DIPROVE_3: -0.554

AcTitola

Analysis of Variance Table

Response: Br_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1.2039	1.2039	14.0861	0.0024130	**
DIPROVE_2	1	0.5037	0.5037	5.8938	0.0304613	*
DIPROVE_3	1	3.2676	3.2676	38.2341	3.306e-05	***
N_Gemme	1	0.2118	0.2118	2.4784	0.1394361	
Defogl	1	0.2781	0.2781	3.2541	0.0944577	.
Dirad	1	0.0939	0.0939	1.0984	0.3137217	
Vigoria	2	3.0330	1.5165	17.7444	0.0001923	***
N_Gemme:Defogl	1	0.2901	0.2901	3.3948	0.0883213	.
N_Gemme:Dirad	1	0.0180	0.0180	0.2100	0.6543099	
Defogl:Dirad	1	0.0631	0.0631	0.7383	0.4057850	
N_Gemme:Vigoria	2	0.0216	0.0108	0.1262	0.8824764	
Defogl:Vigoria	2	0.7776	0.3888	4.5492	0.0317892	*
Dirad:Vigoria	2	0.7382	0.3691	4.3185	0.0364624	*
N_Gemme:Defogl:Dirad	1	0.0147	0.0147	0.1715	0.6855637	
Residuals	13	1.1110	0.0855			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.721
DIPROVE_2: 0.559
DIPROVE_3: 0.864

AcTarta

Analysis of Variance Table

Response: Br_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.0336	0.03364	0.4243	0.526142	
DIPROVE_2	1	0.2529	0.25294	3.1908	0.097384	.
DIPROVE_3	1	2.7306	2.73058	34.4458	5.510e-05	***
N_Gemme	1	0.0358	0.03580	0.4516	0.513365	
Defogl	1	0.2303	0.23033	2.9056	0.112040	
Dirad	1	0.3685	0.36855	4.6492	0.050383	.
Vigoria	2	3.2583	1.62913	20.5512	9.435e-05	***
N_Gemme:Defogl	1	0.1150	0.11503	1.4511	0.249820	
N_Gemme:Dirad	1	0.0037	0.00366	0.0461	0.833294	
Defogl:Dirad	1	0.1685	0.16850	2.1256	0.168585	
N_Gemme:Vigoria	2	0.0540	0.02698	0.3404	0.717645	
Defogl:Vigoria	2	0.5913	0.29564	3.7295	0.052467	.
Dirad:Vigoria	2	1.1723	0.58613	7.3939	0.007171	**
N_Gemme:Defogl:Dirad	1	0.0247	0.02471	0.3117	0.586121	
Residuals	13	1.0305	0.07927			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: 0.852

AcMalicoA

Analysis of Variance Table

Response: Br_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1.21387	1.21387	56.2027	4.520e-06	***
DIPROVE_2	1	2.02344	2.02344	93.6863	2.623e-07	***
DIPROVE_3	1	0.01120	0.01120	0.5183	0.48429	
N_Gemme	1	0.02453	0.02453	1.1357	0.30596	
Defogl	1	0.07583	0.07583	3.5110	0.08361	.
Dirad	1	0.09812	0.09812	4.5429	0.05271	.
Vigoria	2	1.44586	0.72293	33.4721	7.456e-06	***
N_Gemme:Defogl	1	0.12737	0.12737	5.8974	0.03042	*
N_Gemme:Dirad	1	0.00168	0.00168	0.0779	0.78453	
Defogl:Dirad	1	0.01958	0.01958	0.9065	0.35841	
N_Gemme:Vigoria	2	0.00214	0.00107	0.0495	0.95185	
Defogl:Vigoria	2	0.13645	0.06823	3.1589	0.07619	.
Dirad:Vigoria	2	0.09438	0.04719	2.1850	0.15204	
N_Gemme:Defogl:Dirad	1	0.03310	0.03310	1.5325	0.23762	
Residuals	13	0.28077	0.02160			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.901
 DIPROVE_2: -0.937
 DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Br_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.007182	0.007182	1.1000	0.31338
DIPROVE_2	1	0.000154	0.000154	0.0236	0.88031
DIPROVE_3	1	0.044574	0.044574	6.8269	0.02148 *
N_Gemme	1	0.005931	0.005931	0.9083	0.35795
Defogl	1	0.001778	0.001778	0.2723	0.61058
Dirad	1	0.001246	0.001246	0.1908	0.66941
Vigoria	2	0.000379	0.000190	0.0290	0.97144
N_Gemme:Defogl	1	0.014194	0.014194	2.1740	0.16417
N_Gemme:Dirad	1	0.002905	0.002905	0.4449	0.51643
Defogl:Dirad	1	0.000652	0.000652	0.0998	0.75704
N_Gemme:Vigoria	2	0.007328	0.003664	0.5612	0.58376
Defogl:Vigoria	2	0.026199	0.013099	2.0063	0.17403
Dirad:Vigoria	2	0.004290	0.002145	0.3285	0.72578
N_Gemme:Defogl:Dirad	1	0.002954	0.002954	0.4524	0.51298
Residuals	13	0.084879	0.006529		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.587

ApaA

Analysis of Variance Table

Response: Br_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	4441.5	4441.5	18.1872	0.0009218 ***
DIPROVE_2	1	4704.7	4704.7	19.2650	0.0007321 ***
DIPROVE_3	1	1526.6	1526.6	6.2513	0.0265759 *
N_Gemme	1	301.7	301.7	1.2354	0.2864988
Defogl	1	575.5	575.5	2.3566	0.1487265
Dirad	1	165.0	165.0	0.6757	0.4258762
Vigoria	2	746.5	373.2	1.5283	0.2534420
N_Gemme:Defogl	1	446.3	446.3	1.8277	0.1994535
N_Gemme:Dirad	1	87.3	87.3	0.3575	0.5601873
Defogl:Dirad	1	188.9	188.9	0.7736	0.3950695
N_Gemme:Vigoria	2	923.8	461.9	1.8913	0.1901243
Defogl:Vigoria	2	596.3	298.1	1.2209	0.3266732
Dirad:Vigoria	2	278.3	139.1	0.5698	0.5791514
N_Gemme:Defogl:Dirad	1	297.3	297.3	1.2174	0.2898803
Residuals	13	3174.7	244.2		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.764
DIPROVE_2: 0.773
DIPROVE_3: -0.57

Antoctot1A

Analysis of Variance Table

Response: Br_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	157313	157313	41.8647	2.099e-05 ***

DIPROVE_2	1	60228	60228	16.0280	0.001502	**
DIPROVE_3	1	638	638	0.1699	0.686921	
N_Gemme	1	40	40	0.0107	0.919366	
Defogl	1	411	411	0.1093	0.746182	
Dirad	1	995	995	0.2649	0.615423	
Vigoria	2	8801	4400	1.1710	0.340708	
N_Gemme:Defogl	1	19261	19261	5.1258	0.041323	*
N_Gemme:Dirad	1	3596	3596	0.9569	0.345821	
Defogl:Dirad	1	207	207	0.0552	0.817915	
N_Gemme:Vigoria	2	816	408	0.1086	0.897867	
Defogl:Vigoria	2	10823	5412	1.4402	0.272298	
Dirad:Vigoria	2	11310	5655	1.5049	0.258302	
N_Gemme:Defogl:Dirad	1	1741	1741	0.4634	0.507987	
Residuals	13	48849	3758			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.874
DIPROVE_2: 0.743
DIPROVE_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Br_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	362685	362685	15.4021	0.001743	**
DIPROVE_2	1	228672	228672	9.7110	0.008187	**
DIPROVE_3	1	6253	6253	0.2655	0.614993	
N_Gemme	1	2236	2236	0.0950	0.762843	
Defogl	1	29805	29805	1.2657	0.280915	
Dirad	1	5711	5711	0.2425	0.630587	
Vigoria	2	4172	2086	0.0886	0.915772	
N_Gemme:Defogl	1	21525	21525	0.9141	0.356468	
N_Gemme:Dirad	1	11587	11587	0.4920	0.495383	
Defogl:Dirad	1	7255	7255	0.3081	0.588265	
N_Gemme:Vigoria	2	5741	2871	0.1219	0.886228	
Defogl:Vigoria	2	2854	1427	0.0606	0.941463	
Dirad:Vigoria	2	51855	25928	1.1011	0.361622	
N_Gemme:Defogl:Dirad	1	25479	25479	1.0820	0.317211	
Residuals	13	306122	23548			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.736
DIPROVE_2: 0.654
DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Br_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	410945	410945	45.4819	1.375e-05	***
DIPROVE_2	1	263649	263649	29.1797	0.0001208	***
DIPROVE_3	1	30142	30142	3.3360	0.0908251	.
N_Gemme	1	33	33	0.0036	0.9527500	

Defogl	1	36644	36644	4.0556	0.0652017	.
Dirad	1	2271	2271	0.2513	0.6245280	
Vigoria	2	19636	9818	1.0866	0.3661224	
N_Gemme:Defogl	1	25245	25245	2.7940	0.1185024	
N_Gemme:Dirad	1	23231	23231	2.5711	0.1328381	
Defogl:Dirad	1	285	285	0.0315	0.8617968	
N_Gemme:Vigoria	2	1480	740	0.0819	0.9218589	
Defogl:Vigoria	2	75530	37765	4.1797	0.0396552	*
Dirad:Vigoria	2	21773	10887	1.2049	0.3310980	
N_Gemme:Defogl:Dirad	1	7356	7356	0.8141	0.3833232	
Residuals	13	117460	9035			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.882
 DIPROVE_2: 0.832
 DIPROVE_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Br_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	793370	793370	7.7387	0.015561	*
DIPROVE_2	1	1081538	1081538	10.5496	0.006353	**
DIPROVE_3	1	58726	58726	0.5728	0.462634	
N_Gemme	1	13112	13112	0.1279	0.726362	
Defogl	1	493223	493223	4.8110	0.047064	*
Dirad	1	22958	22958	0.2239	0.643905	
Vigoria	2	194090	97045	0.9466	0.413247	
N_Gemme:Defogl	1	1344	1344	0.0131	0.910606	
N_Gemme:Dirad	1	60653	60653	0.5916	0.455541	
Defogl:Dirad	1	25397	25397	0.2477	0.626992	
N_Gemme:Vigoria	2	38442	19221	0.1875	0.831242	
Defogl:Vigoria	2	66257	33129	0.3231	0.729520	
Dirad:Vigoria	2	22909	11455	0.1117	0.895135	
N_Gemme:Defogl:Dirad	1	102093	102093	0.9958	0.336533	
Residuals	13	1332754	102520			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.611
 DIPROVE_2: -0.669
 DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	3572194	3572194	68.4015	1.551e-06	***
DIPROVE_2	1	4297571	4297571	82.2912	5.505e-07	***
DIPROVE_3	1	314542	314542	6.0229	0.028985	*
N_Gemme	1	121886	121886	2.3339	0.150543	
Defogl	1	671478	671478	12.8577	0.003321	**
Dirad	1	41495	41495	0.7946	0.388927	
Vigoria	2	286351	143175	2.7416	0.101529	

N_Gemme:Defogl	1	1670	1670	0.0320	0.860827
N_Gemme:Dirad	1	7651	7651	0.1465	0.708090
Defogl:Dirad	1	97232	97232	1.8618	0.195569
N_Gemme:Vigoria	2	3504	1752	0.0335	0.967094
Defogl:Vigoria	2	4966	2483	0.0475	0.953736
Dirad:Vigoria	2	34268	17134	0.3281	0.726092
N_Gemme:Defogl:Dirad	1	7606	7606	0.1456	0.708894
Residuals	13	678911	52224		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.917
DIPROVE_2: -0.929
DIPROVE_3: 0.563

DensMostoB

Analysis of Variance Table

Response: Br_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.00043241	0.00043241	19.3782	0.000715	***
DIPROVE_2	1	0.00006282	0.00006282	2.8152	0.117238	
DIPROVE_3	1	0.00005562	0.00005562	2.4924	0.138409	
N_Gemme	1	0.00000539	0.00000539	0.2417	0.631145	
Defogl	1	0.00015136	0.00015136	6.7831	0.021822	*
Dirad	1	0.00000285	0.00000285	0.1279	0.726337	
Vigoria	2	0.00004279	0.00002139	0.9588	0.408887	
N_Gemme:Defogl	1	0.00000595	0.00000595	0.2668	0.614132	
N_Gemme:Dirad	1	0.00013025	0.00013025	5.8369	0.031139	*
Defogl:Dirad	1	0.00000039	0.00000039	0.0175	0.896858	
N_Gemme:Vigoria	2	0.00001423	0.00000712	0.3190	0.732431	
Defogl:Vigoria	2	0.00005340	0.00002670	1.1966	0.333434	
Dirad:Vigoria	2	0.00004311	0.00002156	0.9661	0.406287	
N_Gemme:Defogl:Dirad	1	0.00001955	0.00001955	0.8762	0.366309	
Residuals	13	0.00029008	0.00002231			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.774
DIPROVE_2: n.s
DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Br_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	19.4522	19.4522	19.1152	0.0007555	***
DIPROVE_2	1	2.7366	2.7366	2.6892	0.1249931	
DIPROVE_3	1	2.2890	2.2890	2.2493	0.1575644	
N_Gemme	1	0.2261	0.2261	0.2222	0.6452060	
Defogl	1	7.7421	7.7421	7.6080	0.0162790	*
Dirad	1	0.2689	0.2689	0.2642	0.6158440	
Vigoria	2	1.8435	0.9218	0.9058	0.4282769	
N_Gemme:Defogl	1	0.2152	0.2152	0.2115	0.6532079	
N_Gemme:Dirad	1	5.8317	5.8317	5.7307	0.0324543	*
Defogl:Dirad	1	0.0057	0.0057	0.0056	0.9415584	

N_Gemme:Vigoria	2	0.6203	0.3102	0.3048	0.7424116
Defogl:Vigoria	2	2.5187	1.2593	1.2375	0.3221274
Dirad:Vigoria	2	2.1402	1.0701	1.0516	0.3773111
N_Gemme:Defogl:Dirad	1	0.8557	0.8557	0.8409	0.3758485
Residuals	13	13.2292	1.0176		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.771
DIPROVE_2: n.s
DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Br_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.0050864	0.0050864	2.4400	0.142279
DIPROVE_2	1	0.0014529	0.0014529	0.6970	0.418882
DIPROVE_3	1	0.0142532	0.0142532	6.8375	0.021394 *
N_Gemme	1	0.0014769	0.0014769	0.7085	0.415156
Defogl	1	0.0098253	0.0098253	4.7134	0.049034 *
Dirad	1	0.0016389	0.0016389	0.7862	0.391355
Vigoria	2	0.0298398	0.0149199	7.1573	0.008018 **
N_Gemme:Defogl	1	0.0005286	0.0005286	0.2536	0.622994
N_Gemme:Dirad	1	0.0106936	0.0106936	5.1299	0.041254 *
Defogl:Dirad	1	0.0003932	0.0003932	0.1886	0.671192
N_Gemme:Vigoria	2	0.0032565	0.0016282	0.7811	0.478253
Defogl:Vigoria	2	0.0071497	0.0035749	1.7149	0.218280
Dirad:Vigoria	2	0.0020526	0.0010263	0.4923	0.622144
N_Gemme:Defogl:Dirad	1	0.0061907	0.0061907	2.9698	0.108517
Residuals	13	0.0270993	0.0020846		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.587

AcTitolB

Analysis of Variance Table

Response: Br_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.0686	0.0686	0.2979	0.5944252
DIPROVE_2	1	0.0013	0.0013	0.0055	0.9422047
DIPROVE_3	1	4.8760	4.8760	21.1788	0.0004959 ***
N_Gemme	1	0.3935	0.3935	1.7090	0.2137628
Defogl	1	0.0133	0.0133	0.0577	0.8138557
Dirad	1	0.0835	0.0835	0.3629	0.5572801
Vigoria	2	5.4553	2.7277	11.8475	0.0011768 **
N_Gemme:Defogl	1	0.0898	0.0898	0.3899	0.5431306
N_Gemme:Dirad	1	0.0976	0.0976	0.4239	0.5263226
Defogl:Dirad	1	0.0005	0.0005	0.0021	0.9638201
N_Gemme:Vigoria	2	0.1918	0.0959	0.4165	0.6678716
Defogl:Vigoria	2	0.0973	0.0487	0.2113	0.8122226
Dirad:Vigoria	2	0.2913	0.1456	0.6326	0.5468209

N_Gemme:Defogl:Dirad 1 0.0969 0.0969 0.4209 0.5277758
 Residuals 13 2.9930 0.2302

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: 0.787

AcTartB

Analysis of Variance Table

Response: Br_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.1405	0.1405	1.0973	0.3139435	
DIPROVE_2	1	0.0753	0.0753	0.5881	0.4568418	
DIPROVE_3	1	4.0592	4.0592	31.6987	8.198e-05	***
N_Gemme	1	0.2425	0.2425	1.8940	0.1919992	
Defogl	1	0.9486	0.9486	7.4079	0.0174545	*
Dirad	1	0.5143	0.5143	4.0161	0.0663602	.
Vigoria	2	4.1682	2.0841	16.2748	0.0002887	***
N_Gemme:Defogl	1	0.0387	0.0387	0.3023	0.5917663	
N_Gemme:Dirad	1	0.0030	0.0030	0.0232	0.8812218	
Defogl:Dirad	1	0.0750	0.0750	0.5859	0.4576645	
N_Gemme:Vigoria	2	0.0917	0.0459	0.3582	0.7055994	
Defogl:Vigoria	2	0.1484	0.0742	0.5793	0.5741411	
Dirad:Vigoria	2	0.0731	0.0366	0.2855	0.7562103	
N_Gemme:Defogl:Dirad	1	0.0698	0.0698	0.5448	0.4735544	
Residuals	13	1.6647	0.1281			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: 0.842

AcMalicoB

Analysis of Variance Table

Response: Br_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.87235	0.87235	4.9388	0.04463	*
DIPROVE_2	1	0.38399	0.38399	2.1740	0.16416	
DIPROVE_3	1	0.08486	0.08486	0.4804	0.50042	
N_Gemme	1	0.01902	0.01902	0.1077	0.74800	
Defogl	1	0.70653	0.70653	4.0000	0.06684	.
Dirad	1	0.21960	0.21960	1.2433	0.28503	
Vigoria	2	1.35922	0.67961	3.8476	0.04869	*
N_Gemme:Defogl	1	0.00110	0.00110	0.0062	0.93843	
N_Gemme:Dirad	1	0.51652	0.51652	2.9243	0.11100	
Defogl:Dirad	1	0.03052	0.03052	0.1728	0.68442	
N_Gemme:Vigoria	2	0.06692	0.03346	0.1894	0.82967	
Defogl:Vigoria	2	0.24984	0.12492	0.7072	0.51102	
Dirad:Vigoria	2	0.22194	0.11097	0.6283	0.54896	
N_Gemme:Defogl:Dirad	1	0.03391	0.03391	0.1920	0.66846	
Residuals	13	2.29620	0.17663			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.525
DIPROVE_2: n.s
DIPROVE_3: n.s

KB

Analysis of Variance Table

Response: Br_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.006354	0.006354	0.5851	0.45800
DIPROVE_2	1	0.000190	0.000190	0.0175	0.89667
DIPROVE_3	1	0.113958	0.113958	10.4929	0.00646 **
N_Gemme	1	0.000236	0.000236	0.0218	0.88499
Defogl	1	0.008784	0.008784	0.8088	0.38482
Dirad	1	0.000091	0.000091	0.0083	0.92863
Vigoria	2	0.055173	0.027586	2.5401	0.11717
N_Gemme:Defogl	1	0.000144	0.000144	0.0133	0.91004
N_Gemme:Dirad	1	0.030744	0.030744	2.8308	0.11632
Defogl:Dirad	1	0.000180	0.000180	0.0166	0.89943
N_Gemme:Vigoria	2	0.004278	0.002139	0.1970	0.82363
Defogl:Vigoria	2	0.030183	0.015092	1.3896	0.28384
Dirad:Vigoria	2	0.006313	0.003157	0.2907	0.75250
N_Gemme:Defogl:Dirad	1	0.011415	0.011415	1.0510	0.32397
Residuals	13	0.141187	0.010861		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.668

ApaB

Analysis of Variance Table

Response: Br_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	6868.4	6868.4	48.4259	9.934e-06 ***
DIPROVE_2	1	1192.7	1192.7	8.4089	0.0124123 *
DIPROVE_3	1	2730.1	2730.1	19.2488	0.0007346 ***
N_Gemme	1	190.8	190.8	1.3454	0.2669462
Defogl	1	876.8	876.8	6.1819	0.0272826 *
Dirad	1	152.2	152.2	1.0734	0.3190741
Vigoria	2	1132.5	566.2	3.9922	0.0444942 *
N_Gemme:Defogl	1	0.2	0.2	0.0016	0.9685821
N_Gemme:Dirad	1	575.6	575.6	4.0585	0.0651159 .
Defogl:Dirad	1	205.0	205.0	1.4456	0.2506744
N_Gemme:Vigoria	2	720.8	360.4	2.5412	0.1170797
Defogl:Vigoria	2	695.0	347.5	2.4500	0.1250524
Dirad:Vigoria	2	74.1	37.0	0.2612	0.7740639
N_Gemme:Defogl:Dirad	1	246.9	246.9	1.7411	0.2097672
Residuals	13	1843.8	141.8		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.888

DIPROVE_2: 0.627
 DIPROVE_3: -0.773

Antoctot1B

Analysis of Variance Table

Response: Br_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	37368	37368	50.1641	8.260e-06	***
DIPROVE_2	1	32315	32315	43.3802	1.752e-05	***
DIPROVE_3	1	1882	1882	2.5271	0.1359210	
N_Gemme	1	3130	3130	4.2015	0.0611224	.
Defogl	1	23684	23684	31.7934	8.083e-05	***
Dirad	1	809	809	1.0857	0.3164184	
Vigoria	2	4873	2436	3.2705	0.0707093	.
N_Gemme:Defogl	1	225	225	0.3022	0.5918354	
N_Gemme:Dirad	1	426	426	0.5725	0.4627465	
Defogl:Dirad	1	4554	4554	6.1134	0.0280011	*
N_Gemme:Vigoria	2	618	309	0.4150	0.6687555	
Defogl:Vigoria	2	19860	9930	13.3300	0.0007101	***
Dirad:Vigoria	2	2471	1236	1.6586	0.2282634	
N_Gemme:Defogl:Dirad	1	1623	1623	2.1789	0.1637185	
Residuals	13	9684	745			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.891
 DIPROVE_2: -0.877
 DIPROVE_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	86994	86994	5.0025	0.04347	*
DIPROVE_2	1	12036	12036	0.6921	0.42047	
DIPROVE_3	1	9606	9606	0.5524	0.47057	
N_Gemme	1	4524	4524	0.2601	0.61857	
Defogl	1	17005	17005	0.9779	0.34078	
Dirad	1	62238	62238	3.5789	0.08100	.
Vigoria	2	13250	6625	0.3810	0.69059	
N_Gemme:Defogl	1	1627	1627	0.0935	0.76457	
N_Gemme:Dirad	1	399	399	0.0230	0.88189	
Defogl:Dirad	1	39791	39791	2.2881	0.15429	
N_Gemme:Vigoria	2	4489	2244	0.1291	0.88003	
Defogl:Vigoria	2	102136	51068	2.9366	0.08864	.
Dirad:Vigoria	2	21887	10943	0.6293	0.54845	
N_Gemme:Defogl:Dirad	1	23888	23888	1.3736	0.26222	
Residuals	13	226072	17390			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.527
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	141977	141977	24.9173	0.0002465	***
DIPROVE_2	1	63701	63701	11.1797	0.0052854	**
DIPROVE_3	1	131682	131682	23.1105	0.0003423	***
N_Gemme	1	12572	12572	2.2064	0.1612845	
Defogl	1	292603	292603	51.3526	7.303e-06	***
Dirad	1	4371	4371	0.7670	0.3970290	
Vigoria	2	55476	27738	4.8681	0.0264216	*
N_Gemme:Defogl	1	345	345	0.0606	0.8094445	
N_Gemme:Dirad	1	374	374	0.0656	0.8018362	
Defogl:Dirad	1	12836	12836	2.2527	0.1572701	
N_Gemme:Vigoria	2	1166	583	0.1023	0.9034551	
Defogl:Vigoria	2	51832	25916	4.5483	0.0318064	*
Dirad:Vigoria	2	5936	2968	0.5209	0.6058608	
N_Gemme:Defogl:Dirad	1	11845	11845	2.0788	0.1730134	
Residuals	13	74073	5698			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.811

DIPROVE_2: -0.68

DIPROVE_3: 0.8

Poliftot2B

Analysis of Variance Table

Response: Br_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	294747	294747	2.7981	0.11825	
DIPROVE_2	1	8870	8870	0.0842	0.77626	
DIPROVE_3	1	526517	526517	4.9984	0.04354	*
N_Gemme	1	102081	102081	0.9691	0.34288	
Defogl	1	555629	555629	5.2748	0.03890	*
Dirad	1	327346	327346	3.1076	0.10140	
Vigoria	2	187037	93519	0.8878	0.43510	
N_Gemme:Defogl	1	5150	5150	0.0489	0.82845	
N_Gemme:Dirad	1	73	73	0.0007	0.97933	
Defogl:Dirad	1	130742	130742	1.2412	0.28542	
N_Gemme:Vigoria	2	25432	12716	0.1207	0.88727	
Defogl:Vigoria	2	309746	154873	1.4703	0.26568	
Dirad:Vigoria	2	137085	68543	0.6507	0.53786	
N_Gemme:Defogl:Dirad	1	60867	60867	0.5778	0.46073	
Residuals	13	1369379	105337			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: 0.527

PvinacciaA

Analysis of Variance Table

Response: Br_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	934670	934670	16.3916	0.001380	**
IASMA_2	1	913720	913720	16.0242	0.001504	**
IASMA_3	1	168176	168176	2.9493	0.109624	
N_Gemme	1	6870	6870	0.1205	0.734071	
Defogl	1	78160	78160	1.3707	0.262705	
Dirad	1	60662	60662	1.0638	0.321149	
Vigoria	2	294087	147044	2.5787	0.113964	
N_Gemme:Defogl	1	59711	59711	1.0472	0.324827	
N_Gemme:Dirad	1	25095	25095	0.4401	0.518670	
Defogl:Dirad	1	278102	278102	4.8771	0.045783	*
N_Gemme:Vigoria	2	17949	8975	0.1574	0.855974	
Defogl:Vigoria	2	254963	127481	2.2357	0.146389	
Dirad:Vigoria	2	163254	81627	1.4315	0.274233	
N_Gemme:Defogl:Dirad	1	107966	107966	1.8934	0.192058	
Residuals	13	741278	57021			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.747

IASMA_2: 0.743

IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Br_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1.3540e-04	1.3540e-04	6.7149	0.022372	*
IASMA_2	1	2.7090e-04	2.7090e-04	13.4349	0.002852	**
IASMA_3	1	9.9390e-06	9.9390e-06	0.4929	0.495018	
N_Gemme	1	1.2200e-07	1.2200e-07	0.0061	0.939172	
Defogl	1	3.0847e-04	3.0847e-04	15.2981	0.001788	**
Dirad	1	3.3160e-06	3.3160e-06	0.1644	0.691697	
Vigoria	2	2.6104e-04	1.3052e-04	6.4729	0.011199	*
N_Gemme:Defogl	1	1.4353e-05	1.4353e-05	0.7118	0.414095	
N_Gemme:Dirad	1	1.9721e-05	1.9721e-05	0.9780	0.340734	
Defogl:Dirad	1	2.1657e-05	2.1657e-05	1.0740	0.318930	
N_Gemme:Vigoria	2	7.9485e-05	3.9743e-05	1.9710	0.178800	
Defogl:Vigoria	2	9.6400e-06	4.8200e-06	0.2390	0.790763	
Dirad:Vigoria	2	5.7625e-05	2.8813e-05	1.4289	0.274817	
N_Gemme:Defogl:Dirad	1	1.2000e-08	1.2000e-08	0.0006	0.980896	
Residuals	13	2.6213e-04	2.0164e-05			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.584

IASMA_2: -0.713

IASMA_3: n.s

BrixA

Analysis of Variance Table

Response: Br_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	6.1311	6.1311	6.5483	0.023786 *
IASMA_2	1	12.9284	12.9284	13.8081	0.002590 **
IASMA_3	1	0.3244	0.3244	0.3464	0.566228
N_Gemme	1	0.0261	0.0261	0.0279	0.869903
Defogl	1	14.5285	14.5285	15.5170	0.001696 **
Dirad	1	0.1008	0.1008	0.1077	0.748022
Vigoria	2	12.7517	6.3758	6.8097	0.009481 **
N_Gemme:Defogl	1	0.5133	0.5133	0.5482	0.472216
N_Gemme:Dirad	1	0.9839	0.9839	1.0508	0.324018
Defogl:Dirad	1	0.8782	0.8782	0.9380	0.350473
N_Gemme:Vigoria	2	3.7236	1.8618	1.9885	0.176421
Defogl:Vigoria	2	0.4173	0.2087	0.2229	0.803212
Dirad:Vigoria	2	2.6482	1.3241	1.4142	0.278155
N_Gemme:Defogl:Dirad	1	0.0001	0.0001	0.0001	0.993965
Residuals	13	12.1718	0.9363		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.579
IASMA_2: -0.718
IASMA_3: n.s

PhA

Analysis of Variance Table

Response: Br_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.000004	0.0000043	0.0013	0.971734
IASMA_2	1	0.014376	0.0143761	4.3407	0.057515 .
IASMA_3	1	0.007040	0.0070403	2.1258	0.168574
N_Gemme	1	0.000025	0.0000254	0.0077	0.931538
Defogl	1	0.000227	0.0002270	0.0685	0.797565
Dirad	1	0.000594	0.0005939	0.1793	0.678875
Vigoria	2	0.051760	0.0258798	7.8141	0.005908 **
N_Gemme:Defogl	1	0.000041	0.0000412	0.0124	0.912891
N_Gemme:Dirad	1	0.014324	0.0143240	4.3250	0.057909 .
Defogl:Dirad	1	0.008790	0.0087903	2.6541	0.127261
N_Gemme:Vigoria	2	0.005572	0.0027859	0.8412	0.453378
Defogl:Vigoria	2	0.002425	0.0012127	0.3662	0.700318
Dirad:Vigoria	2	0.005847	0.0029235	0.8827	0.437052
N_Gemme:Defogl:Dirad	1	0.001912	0.0019116	0.5772	0.460969
Residuals	13	0.043055	0.0033119		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Br_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.0381	0.0381	0.5984	0.4530319	
IASMA_2	1	1.0370	1.0370	16.2872	0.0014137	**
IASMA_3	1	1.5958	1.5958	25.0625	0.0002403	***
N_Gemme	1	0.0569	0.0569	0.8936	0.3617550	
Defogl	1	0.0317	0.0317	0.4986	0.4925784	
Dirad	1	0.0001	0.0001	0.0020	0.9645829	
Vigoria	2	6.3417	3.1708	49.7997	8.047e-07	***
N_Gemme:Defogl	1	0.0300	0.0300	0.4719	0.5041824	
N_Gemme:Dirad	1	0.5153	0.5153	8.0935	0.0137910	*
Defogl:Dirad	1	0.0279	0.0279	0.4377	0.5197866	
N_Gemme:Vigoria	2	0.1662	0.0831	1.3049	0.3044611	
Defogl:Vigoria	2	0.4356	0.2178	3.4207	0.0640358	.
Dirad:Vigoria	2	0.5167	0.2583	4.0572	0.0427429	*
N_Gemme:Defogl:Dirad	1	0.0054	0.0054	0.0850	0.7752927	
Residuals	13	0.8277	0.0637			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.746
IASMA_3: 0.811

AcTartA

Analysis of Variance Table

Response: Br_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	2.4538	2.45383	37.7050	3.542e-05	***
IASMA_2	1	0.1155	0.11552	1.7751	0.205641	
IASMA_3	1	0.6481	0.64807	9.9581	0.007589	**
N_Gemme	1	0.3649	0.36492	5.6073	0.034065	*
Defogl	1	0.0001	0.00013	0.0021	0.964402	
Dirad	1	0.0042	0.00425	0.0652	0.802408	
Vigoria	2	4.0890	2.04450	31.4152	1.051e-05	***
N_Gemme:Defogl	1	0.0011	0.00115	0.0176	0.896421	
N_Gemme:Dirad	1	0.3401	0.34011	5.2261	0.039671	*
Defogl:Dirad	1	0.0024	0.00240	0.0369	0.850626	
N_Gemme:Vigoria	2	0.0951	0.04756	0.7308	0.500284	
Defogl:Vigoria	2	0.3466	0.17330	2.6629	0.107334	
Dirad:Vigoria	2	0.7617	0.38084	5.8518	0.015406	*
N_Gemme:Defogl:Dirad	1	0.0012	0.00120	0.0184	0.894105	
Residuals	13	0.8460	0.06508			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.862
IASMA_2: n.s
IASMA_3: 0.659

AcMalicoA

Analysis of Variance Table

Response: Br_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	2.97386	2.97386	68.8972	1.49e-06	***
IASMA_2	1	0.65263	0.65263	15.1198	0.001867	**

IASMA_3	1	0.11634	0.11634	2.6953	0.124601
N_Gemme	1	0.08766	0.08766	2.0309	0.177689
Defogl	1	0.00614	0.00614	0.1422	0.712220
Dirad	1	0.00003	0.00003	0.0008	0.977856
Vigoria	2	0.89735	0.44868	10.3947	0.002012 **
N_Gemme:Defogl	1	0.10156	0.10156	2.3529	0.149024
N_Gemme:Dirad	1	0.01137	0.01137	0.2634	0.616373
Defogl:Dirad	1	0.00105	0.00105	0.0243	0.878559
N_Gemme:Vigoria	2	0.01357	0.00679	0.1572	0.856100
Defogl:Vigoria	2	0.09157	0.04578	1.0607	0.374352
Dirad:Vigoria	2	0.06743	0.03372	0.7811	0.478245
N_Gemme:Defogl:Dirad	1	0.00663	0.00663	0.1536	0.701479
Residuals	13	0.56113	0.04316		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.917
 IASMA_2: -0.733
 IASMA_3: n.s

KA

Analysis of Variance Table

Response: Br_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.003065	0.0030652	0.5037	0.4904
IASMA_2	1	0.002999	0.0029993	0.4929	0.4950
IASMA_3	1	0.021649	0.0216487	3.5577	0.0818 .
N_Gemme	1	0.000560	0.0005601	0.0920	0.7664
Defogl	1	0.005294	0.0052940	0.8700	0.3680
Dirad	1	0.007500	0.0075000	1.2325	0.2870
Vigoria	2	0.031801	0.0159005	2.6131	0.1112
N_Gemme:Defogl	1	0.011838	0.0118384	1.9455	0.1864
N_Gemme:Dirad	1	0.000253	0.0002532	0.0416	0.8415
Defogl:Dirad	1	0.002790	0.0027898	0.4585	0.5102
N_Gemme:Vigoria	2	0.006856	0.0034280	0.5634	0.5826
Defogl:Vigoria	2	0.011686	0.0058428	0.9602	0.4084
Dirad:Vigoria	2	0.011089	0.0055444	0.9112	0.4263
N_Gemme:Defogl:Dirad	1	0.008159	0.0081585	1.3408	0.2677
Residuals	13	0.079105	0.0060850		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

ApaA

Analysis of Variance Table

Response: Br_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	3809.4	3809.4	11.7790	0.004459 **
IASMA_2	1	3501.0	3501.0	10.8256	0.005857 **
IASMA_3	1	56.8	56.8	0.1758	0.681869
N_Gemme	1	1.6	1.6	0.0049	0.945218
Defogl	1	1336.5	1336.5	4.1325	0.063012 .

Dirad	1	39.4	39.4	0.1218	0.732697
Vigoria	2	3793.3	1896.7	5.8647	0.015302 *
N_Gemme:Defogl	1	18.7	18.7	0.0578	0.813805
N_Gemme:Dirad	1	110.9	110.9	0.3428	0.568235
Defogl:Dirad	1	49.1	49.1	0.1518	0.703107
N_Gemme:Vigoria	2	473.1	236.6	0.7315	0.499991
Defogl:Vigoria	2	121.1	60.6	0.1872	0.831434
Dirad:Vigoria	2	854.0	427.0	1.3203	0.300597
N_Gemme:Defogl:Dirad	1	85.3	85.3	0.2636	0.616257
Residuals	13	4204.3	323.4		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.689
IASMA_2: 0.674
IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Br_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	59918	59918	12.3081	0.003852	**
IASMA_2	1	76070	76070	15.6259	0.001652	**
IASMA_3	1	32100	32100	6.5938	0.023389	*
N_Gemme	1	315	315	0.0646	0.803281	
Defogl	1	14008	14008	2.8775	0.113625	
Dirad	1	1801	1801	0.3700	0.553494	
Vigoria	2	33917	16959	3.4836	0.061458	.
N_Gemme:Defogl	1	11017	11017	2.2631	0.156395	
N_Gemme:Dirad	1	2144	2144	0.4404	0.518503	
Defogl:Dirad	1	859	859	0.1765	0.681280	
N_Gemme:Vigoria	2	9716	4858	0.9979	0.395215	
Defogl:Vigoria	2	12536	6268	1.2875	0.308925	
Dirad:Vigoria	2	7070	3535	0.7261	0.502403	
N_Gemme:Defogl:Dirad	1	272	272	0.0559	0.816826	
Residuals	13	63287	4868			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.697
IASMA_2: -0.739
IASMA_3: 0.58

Poliftot1A

Analysis of Variance Table

Response: Br_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	175874	175874	8.2294	0.013176	*
IASMA_2	1	281626	281626	13.1777	0.003051	**
IASMA_3	1	36186	36186	1.6932	0.215772	
N_Gemme	1	5094	5094	0.2384	0.633525	
Defogl	1	83958	83958	3.9285	0.069021	.
Dirad	1	7858	7858	0.3677	0.554701	
Vigoria	2	101228	50614	2.3683	0.132733	
N_Gemme:Defogl	1	6386	6386	0.2988	0.593894	

N_Gemme:Dirad	1	4706	4706	0.2202	0.646657
Defogl:Dirad	1	1871	1871	0.0875	0.771995
N_Gemme:Vigoria	2	37334	18667	0.8734	0.440636
Defogl:Vigoria	2	2137	1069	0.0500	0.951413
Dirad:Vigoria	2	41585	20792	0.9729	0.403884
N_Gemme:Defogl:Dirad	1	8280	8280	0.3875	0.544407
Residuals	13	277829	21371		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.623
IASMA_2: -0.71
IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Br_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	283665	283665	32.4018	7.388e-05	***
IASMA_2	1	284694	284694	32.5194	7.262e-05	***
IASMA_3	1	18997	18997	2.1700	0.164525	
N_Gemme	1	3898	3898	0.4452	0.516289	
Defogl	1	111383	111383	12.7228	0.003443	**
Dirad	1	864	864	0.0987	0.758390	
Vigoria	2	47844	23922	2.7325	0.102179	
N_Gemme:Defogl	1	24404	24404	2.7876	0.118885	
N_Gemme:Dirad	1	300	300	0.0342	0.856037	
Defogl:Dirad	1	11374	11374	1.2992	0.274932	
N_Gemme:Vigoria	2	2515	1257	0.1436	0.867567	
Defogl:Vigoria	2	116866	58433	6.6745	0.010131	*
Dirad:Vigoria	2	14601	7300	0.8339	0.456311	
N_Gemme:Defogl:Dirad	1	465	465	0.0531	0.821280	
Residuals	13	113810	8755			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.845
IASMA_2: -0.845
IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Br_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	986588	986588	9.5580	0.008584	**
IASMA_2	1	985760	985760	9.5500	0.008606	**
IASMA_3	1	3380	3380	0.0327	0.859197	
N_Gemme	1	2453	2453	0.0238	0.879842	
Defogl	1	595744	595744	5.7715	0.031941	*
Dirad	1	5200	5200	0.0504	0.825904	
Vigoria	2	153673	76837	0.7444	0.494225	
N_Gemme:Defogl	1	1192	1192	0.0115	0.916065	
N_Gemme:Dirad	1	6665	6665	0.0646	0.803385	
Defogl:Dirad	1	27620	27620	0.2676	0.613649	
N_Gemme:Vigoria	2	34081	17040	0.1651	0.849571	

Defogl:Vigoria	2	83661	41830	0.4052	0.674949
Dirad:Vigoria	2	50752	25376	0.2458	0.785600
N_Gemme:Defogl:Dirad	1	28225	28225	0.2734	0.609830
Residuals	13	1341872	103221		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.651
IASMA_2: -0.651
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	4361743	4361743	88.7655	3.576e-07	***
IASMA_2	1	2123324	2123324	43.2116	1.787e-05	***
IASMA_3	1	1247892	1247892	25.3958	0.0002266	***
N_Gemme	1	120861	120861	2.4596	0.1408174	
Defogl	1	425230	425230	8.6538	0.0114522	*
Dirad	1	158439	158439	3.2244	0.0958179	.
Vigoria	2	729760	364880	7.4256	0.0070654	**
N_Gemme:Defogl	1	25684	25684	0.5227	0.4824985	
N_Gemme:Dirad	1	2070	2070	0.0421	0.8405430	
Defogl:Dirad	1	203984	203984	4.1513	0.0624924	.
N_Gemme:Vigoria	2	28917	14458	0.2942	0.7499295	
Defogl:Vigoria	2	1294	647	0.0132	0.9869373	
Dirad:Vigoria	2	18006	9003	0.1832	0.8346994	
N_Gemme:Defogl:Dirad	1	55331	55331	1.1260	0.3079373	
Residuals	13	638791	49138			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.934
IASMA_2: 0.877
IASMA_3: -0.813

DensMostoB

Analysis of Variance Table

Response: Br_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00012447	0.00012447	8.9082	0.0105453	*
IASMA_2	1	0.00025103	0.00025103	17.9666	0.0009673	***
IASMA_3	1	0.00005455	0.00005455	3.9043	0.0697791	.
N_Gemme	1	0.00000418	0.00000418	0.2990	0.5937694	
Defogl	1	0.00037227	0.00037227	26.6433	0.0001829	***
Dirad	1	0.00000036	0.00000036	0.0257	0.8750230	
Vigoria	2	0.00009881	0.00004941	3.5361	0.0593971	.
N_Gemme:Defogl	1	0.00000403	0.00000403	0.2887	0.6001029	
N_Gemme:Dirad	1	0.00002866	0.00002866	2.0510	0.1757110	
Defogl:Dirad	1	0.00000474	0.00000474	0.3394	0.5701673	
N_Gemme:Vigoria	2	0.00000239	0.00000120	0.0856	0.9184508	
Defogl:Vigoria	2	0.00011420	0.00005710	4.0866	0.0419767	*
Dirad:Vigoria	2	0.00005860	0.00002930	2.0971	0.1624246	
N_Gemme:Defogl:Dirad	1	0.00001028	0.00001028	0.7355	0.4066299	

Residuals 13 0.00018164 0.00001397

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.638
IASMA_2: -0.762
IASMA_3: n.s

BrixB

Analysis of Variance Table

Response: Br_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	5.5572	5.5572	8.8669	0.0106869	*
IASMA_2	1	11.1482	11.1482	17.7877	0.0010061	**
IASMA_3	1	2.6246	2.6246	4.1877	0.0614953	.
N_Gemme	1	0.1984	0.1984	0.3166	0.5832528	
Defogl	1	17.9101	17.9101	28.5769	0.0001330	***
Dirad	1	0.0007	0.0007	0.0012	0.9734217	
Vigoria	2	4.1544	2.0772	3.3143	0.0686833	.
N_Gemme:Defogl	1	0.1431	0.1431	0.2283	0.6407374	
N_Gemme:Dirad	1	1.3354	1.3354	2.1308	0.1681067	
Defogl:Dirad	1	0.1593	0.1593	0.2541	0.6226354	
N_Gemme:Vigoria	2	0.1292	0.0646	0.1031	0.9027959	
Defogl:Vigoria	2	5.1467	2.5734	4.1060	0.0414819	*
Dirad:Vigoria	2	2.8855	1.4427	2.3020	0.1393654	
N_Gemme:Defogl:Dirad	1	0.4347	0.4347	0.6936	0.4199853	
Residuals	13	8.1476	0.6267			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.637
IASMA_2: -0.76
IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Br_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.0063731	0.0063731	3.1560	0.099044	.
IASMA_2	1	0.0047268	0.0047268	2.3407	0.149993	
IASMA_3	1	0.0131613	0.0131613	6.5175	0.024058	*
N_Gemme	1	0.0011888	0.0011888	0.5887	0.456631	
Defogl	1	0.0206348	0.0206348	10.2184	0.007013	**
Dirad	1	0.0003734	0.0003734	0.1849	0.674250	
Vigoria	2	0.0264656	0.0132328	6.5529	0.010761	*
N_Gemme:Defogl	1	0.0006376	0.0006376	0.3157	0.583740	
N_Gemme:Dirad	1	0.0044504	0.0044504	2.2038	0.161506	
Defogl:Dirad	1	0.0002747	0.0002747	0.1360	0.718208	
N_Gemme:Vigoria	2	0.0023320	0.0011660	0.5774	0.575114	
Defogl:Vigoria	2	0.0079811	0.0039905	1.9761	0.178097	
Dirad:Vigoria	2	0.0032391	0.0016196	0.8020	0.469418	
N_Gemme:Defogl:Dirad	1	0.0028469	0.0028469	1.4098	0.256331	
Residuals	13	0.0262519	0.0020194			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.578

AcTitolB

Analysis of Variance Table

Response: Br_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.3242	0.3242	1.5509	0.2349838
IASMA_2	1	0.0459	0.0459	0.2196	0.6471267
IASMA_3	1	2.2170	2.2170	10.6053	0.0062487 **
N_Gemme	1	0.0226	0.0226	0.1081	0.7475016
Defogl	1	0.1123	0.1123	0.5372	0.4766098
Dirad	1	0.0018	0.0018	0.0084	0.9282017
Vigoria	2	8.3808	4.1904	20.0452	0.0001067 ***
N_Gemme:Defogl	1	0.0090	0.0090	0.0430	0.8388828
N_Gemme:Dirad	1	0.0883	0.0883	0.4222	0.5271572
Defogl:Dirad	1	0.0248	0.0248	0.1185	0.7362105
N_Gemme:Vigoria	2	0.4283	0.2141	1.0244	0.3862656
Defogl:Vigoria	2	0.1123	0.0561	0.2685	0.7686325
Dirad:Vigoria	2	0.1515	0.0758	0.3625	0.7027758
N_Gemme:Defogl:Dirad	1	0.1133	0.1133	0.5420	0.4746993
Residuals	13	2.7176	0.2090		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: -0.67

AcTartB

Analysis of Variance Table

Response: Br_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	2.5615	2.5615	21.6928	0.0004483 ***
IASMA_2	1	0.1093	0.10927	0.9254	0.3536252
IASMA_3	1	1.3917	1.39166	11.7856	0.0044509 **
N_Gemme	1	0.0608	0.06085	0.5153	0.4855515
Defogl	1	0.7206	0.72063	6.1028	0.0281145 *
Dirad	1	0.0291	0.02906	0.2461	0.6281318
Vigoria	2	5.5816	2.79082	23.6347	4.677e-05 ***
N_Gemme:Defogl	1	0.0088	0.00877	0.0742	0.7895518
N_Gemme:Dirad	1	0.0428	0.04279	0.3623	0.5575585
Defogl:Dirad	1	0.0059	0.00587	0.0497	0.8269985
N_Gemme:Vigoria	2	0.0876	0.04381	0.3710	0.6971202
Defogl:Vigoria	2	0.1028	0.05139	0.4352	0.6562182
Dirad:Vigoria	2	0.0589	0.02943	0.2492	0.7830460
N_Gemme:Defogl:Dirad	1	0.0168	0.01676	0.1419	0.7124305
Residuals	13	1.5351	0.11808		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.791
IASMA_2: n.s

IASMA_3: -0.69

AcMalicoB

Analysis of Variance Table

Response: Br_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	2.55037	2.55037	17.1312	0.001165	**
IASMA_2	1	0.35269	0.35269	2.3691	0.147739	
IASMA_3	1	0.23987	0.23987	1.6113	0.226570	
N_Gemme	1	0.01264	0.01264	0.0849	0.775392	
Defogl	1	0.09370	0.09370	0.6294	0.441824	
Dirad	1	0.00006	0.00006	0.0004	0.983654	
Vigoria	2	1.22289	0.61145	4.1072	0.041451	*
N_Gemme:Defogl	1	0.01224	0.01224	0.0822	0.778805	
N_Gemme:Dirad	1	0.13611	0.13611	0.9143	0.356424	
Defogl:Dirad	1	0.01498	0.01498	0.1006	0.756121	
N_Gemme:Vigoria	2	0.28716	0.14358	0.9644	0.406870	
Defogl:Vigoria	2	0.12226	0.06113	0.4106	0.671557	
Dirad:Vigoria	2	0.07147	0.03574	0.2401	0.789996	
N_Gemme:Defogl:Dirad	1	0.01073	0.01073	0.0721	0.792512	
Residuals	13	1.93534	0.14887			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.754

IASMA_2: n.s

IASMA_3: n.s

KB

Analysis of Variance Table

Response: Br_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.000056	0.000056	0.0072	0.933535	
IASMA_2	1	0.024599	0.024599	3.1949	0.097194	.
IASMA_3	1	0.012268	0.012268	1.5934	0.229023	
N_Gemme	1	0.009129	0.009129	1.1857	0.295974	
Defogl	1	0.050033	0.050033	6.4982	0.024231	*
Dirad	1	0.002287	0.002287	0.2971	0.594953	
Vigoria	2	0.117561	0.058780	7.6343	0.006414	**
N_Gemme:Defogl	1	0.000480	0.000480	0.0623	0.806775	
N_Gemme:Dirad	1	0.010491	0.010491	1.3626	0.264060	
Defogl:Dirad	1	0.004578	0.004578	0.5946	0.454425	
N_Gemme:Vigoria	2	0.012542	0.006271	0.8145	0.464237	
Defogl:Vigoria	2	0.044101	0.022051	2.8639	0.093211	.
Dirad:Vigoria	2	0.014684	0.007342	0.9536	0.410736	
N_Gemme:Defogl:Dirad	1	0.006328	0.006328	0.8219	0.381132	
Residuals	13	0.100094	0.007700			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Br_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	5680.2	5680.2	22.8206	0.0003614	***
IASMA_2	1	286.7	286.7	1.1518	0.3026894	
IASMA_3	1	0.1	0.1	0.0003	0.9855368	
N_Gemme	1	1.0	1.0	0.0041	0.9497266	
Defogl	1	2201.2	2201.2	8.8437	0.0107673	*
Dirad	1	4.4	4.4	0.0177	0.8960949	
Vigoria	2	2647.5	1323.8	5.3183	0.0205273	*
N_Gemme:Defogl	1	112.9	112.9	0.4535	0.5124950	
N_Gemme:Dirad	1	1332.1	1332.1	5.3516	0.0377122	*
Defogl:Dirad	1	0.2	0.2	0.0008	0.9777663	
N_Gemme:Vigoria	2	125.2	62.6	0.2516	0.7812714	
Defogl:Vigoria	2	1015.8	507.9	2.0405	0.1695499	
Dirad:Vigoria	2	264.8	132.4	0.5318	0.5997745	
N_Gemme:Defogl:Dirad	1	597.4	597.4	2.3999	0.1453355	
Residuals	13	3235.8	248.9			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.798

IASMA_2: n.s

IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Br_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	56645	56645	44.5608	1.527e-05	***
IASMA_2	1	4637	4637	3.6480	0.078450	.
IASMA_3	1	3286	3286	2.5846	0.131914	
N_Gemme	1	6532	6532	5.1386	0.041108	*
Defogl	1	18644	18644	14.6663	0.002086	**
Dirad	1	144	144	0.1133	0.741757	
Vigoria	2	7473	3736	2.9394	0.088473	.
N_Gemme:Defogl	1	0	0	0.0000	0.997819	
N_Gemme:Dirad	1	2401	2401	1.8890	0.192545	
Defogl:Dirad	1	7160	7160	5.6323	0.033731	*
N_Gemme:Vigoria	2	97	48	0.0381	0.962735	
Defogl:Vigoria	2	15186	7593	5.9731	0.014458	*
Dirad:Vigoria	2	2007	1003	0.7893	0.474784	
N_Gemme:Defogl:Dirad	1	2785	2785	2.1909	0.162653	
Residuals	13	16526	1271			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.88

IASMA_2: n.s

IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	65518	65518	3.9186	0.06933 .
IASMA_2	1	1515	1515	0.0906	0.76815
IASMA_3	1	3183	3183	0.1904	0.66976
N_Gemme	1	3578	3578	0.2140	0.65129
Defogl	1	45050	45050	2.6944	0.12466
Dirad	1	39496	39496	2.3622	0.14828
Vigoria	2	12408	6204	0.3710	0.69709
N_Gemme:Defogl	1	4675	4675	0.2796	0.60586
N_Gemme:Dirad	1	381	381	0.0228	0.88239
Defogl:Dirad	1	73486	73486	4.3951	0.05617 .
N_Gemme:Vigoria	2	1272	636	0.0381	0.96277
Defogl:Vigoria	2	129358	64679	3.8684	0.04806 *
Dirad:Vigoria	2	17964	8982	0.5372	0.59682
N_Gemme:Defogl:Dirad	1	10696	10696	0.6397	0.43819
Residuals	13	217359	16720		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	274970	274970	46.9085	1.172e-05 ***
IASMA_2	1	2389	2389	0.4076	0.5342719
IASMA_3	1	2799	2799	0.4774	0.5017402
N_Gemme	1	1310	1310	0.2235	0.6442121
Defogl	1	223389	223389	38.1091	3.360e-05 ***
Dirad	1	4230	4230	0.7216	0.4109757
Vigoria	2	177391	88695	15.1310	0.0004036 ***
N_Gemme:Defogl	1	2526	2526	0.4310	0.5229583
N_Gemme:Dirad	1	15793	15793	2.6943	0.1246677
Defogl:Dirad	1	30972	30972	5.2837	0.0387572 *
N_Gemme:Vigoria	2	1334	667	0.1138	0.8933417
Defogl:Vigoria	2	38182	19091	3.2568	0.0713568 .
Dirad:Vigoria	2	5663	2831	0.4830	0.6275544
N_Gemme:Defogl:Dirad	1	3637	3637	0.6205	0.4449923
Residuals	13	76204	5862		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.885
IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Br_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IASMA_1	1	320660	320660	3.7796	0.07385	.
IASMA_2	1	142502	142502	1.6797	0.21751	
IASMA_3	1	238089	238089	2.8063	0.11777	
N_Gemme	1	27606	27606	0.3254	0.57811	
Defogl	1	724083	724083	8.5347	0.01191	*
Dirad	1	134884	134884	1.5899	0.22951	
Vigoria	2	525172	262586	3.0951	0.07955	.
N_Gemme:Defogl	1	67061	67061	0.7904	0.39012	
N_Gemme:Dirad	1	1167	1167	0.0138	0.90843	
Defogl:Dirad	1	296004	296004	3.4890	0.08448	.
N_Gemme:Vigoria	2	15046	7523	0.0887	0.91569	
Defogl:Vigoria	2	310061	155031	1.8273	0.19983	
Dirad:Vigoria	2	58769	29385	0.3464	0.71359	
N_Gemme:Defogl:Dirad	1	76673	76673	0.9037	0.35912	
Residuals	13	1102922	84840			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

BROLIO 2009M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Br_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	4218841	4218841	66.9587	1.745e-06	***
IBIMET_2	1	421987	421987	6.6975	0.022515	*
IBIMET_3	1	36568	36568	0.5804	0.459760	
N_Gemme	1	57724	57724	0.9162	0.355948	
Defogl	1	58024	58024	0.9209	0.354742	
Dirad	1	1007003	1007003	15.9825	0.001518	**
Vigoria	2	570029	285014	4.5236	0.032273	*
N_Gemme:Defogl	1	17068	17068	0.2709	0.611488	
N_Gemme:Dirad	1	1145	1145	0.0182	0.894830	
Defogl:Dirad	1	50637	50637	0.8037	0.386298	
N_Gemme:Vigoria	2	78059	39030	0.6195	0.553393	
Defogl:Vigoria	2	395452	197726	3.1382	0.077263	.
Dirad:Vigoria	2	30717	15359	0.2438	0.787177	
N_Gemme:Defogl:Dirad	1	205039	205039	3.2542	0.094450	.
Residuals	13	819086	63007			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.915
IBIMET_2: -0.583
IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Br_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.00059329	0.00059329	48.8408	9.501e-06	***

IBIMET_2	1	0.00002103	0.00002103	1.7312	0.210992	
IBIMET_3	1	0.00000769	0.00000769	0.6328	0.440610	
N_Gemme	1	0.00000266	0.00000266	0.2186	0.647866	
Defogl	1	0.00001472	0.00001472	1.2121	0.290887	
Dirad	1	0.00003937	0.00003937	3.2409	0.095057	.
Vigoria	2	0.00038141	0.00019070	15.6991	0.000341	***
N_Gemme:Defogl	1	0.00008706	0.00008706	7.1673	0.019003	*
N_Gemme:Dirad	1	0.00001684	0.00001684	1.3866	0.260092	
Defogl:Dirad	1	0.00005413	0.00005413	4.4562	0.054711	.
N_Gemme:Vigoria	2	0.00002326	0.00001163	0.9575	0.409328	
Defogl:Vigoria	2	0.00006879	0.00003440	2.8316	0.095328	.
Dirad:Vigoria	2	0.00005510	0.00002755	2.2681	0.142904	
N_Gemme:Defogl:Dirad	1	0.00000054	0.00000054	0.0444	0.836308	
Residuals	13	0.00015792	0.00001215			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.889
 IBIMET_2: n.s
 IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Br_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	27.0709	27.0709	49.9378	8.459e-06	***
IBIMET_2	1	1.0168	1.0168	1.8756	0.194027	
IBIMET_3	1	0.3722	0.3722	0.6866	0.422289	
N_Gemme	1	0.1634	0.1634	0.3014	0.592330	
Defogl	1	0.7026	0.7026	1.2961	0.275478	
Dirad	1	1.7348	1.7348	3.2002	0.096941	.
Vigoria	2	16.6908	8.3454	15.3949	0.000373	***
N_Gemme:Defogl	1	4.1861	4.1861	7.7221	0.015650	*
N_Gemme:Dirad	1	0.7188	0.7188	1.3259	0.270274	
Defogl:Dirad	1	2.5069	2.5069	4.6245	0.050914	.
N_Gemme:Vigoria	2	1.1865	0.5932	1.0944	0.363702	
Defogl:Vigoria	2	3.2525	1.6263	3.0000	0.084867	.
Dirad:Vigoria	2	2.8222	1.4111	2.6030	0.112003	
N_Gemme:Defogl:Dirad	1	0.0215	0.0215	0.0397	0.845059	
Residuals	13	7.0472	0.5421			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.891
 IBIMET_2: n.s
 IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Br_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.050027	0.050027	14.9574	0.0019421	**
IBIMET_2	1	0.002690	0.002690	0.8042	0.3861508	
IBIMET_3	1	0.049976	0.049976	14.9422	0.0019494	**
N_Gemme	1	0.009468	0.009468	2.8309	0.1163140	

Defogl	1	0.003966	0.003966	1.1857	0.2959805	
Dirad	1	0.014669	0.014669	4.3859	0.0563964	.
Vigoria	2	0.113255	0.056628	16.9309	0.0002401	***
N_Gemme:Defogl	1	0.003420	0.003420	1.0226	0.3303666	
N_Gemme:Dirad	1	0.014275	0.014275	4.2682	0.0593623	.
Defogl:Dirad	1	0.003970	0.003970	1.1871	0.2957031	
N_Gemme:Vigoria	2	0.002008	0.001004	0.3002	0.7456643	
Defogl:Vigoria	2	0.021493	0.010747	3.2131	0.0734692	.
Dirad:Vigoria	2	0.017758	0.008879	2.6547	0.1079581	
N_Gemme:Defogl:Dirad	1	0.005755	0.005755	1.7206	0.2123093	
Residuals	13	0.043480	0.003345			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.731
 IBIMET_2: n.s
 IBIMET_3: -0.731

AcTitola

Analysis of Variance Table

Response: Br_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	1.2961	1.2961	9.7276	0.0081447	**
IBIMET_2	1	0.0002	0.0002	0.0016	0.9683742	
IBIMET_3	1	5.1868	5.1868	38.9298	3.023e-05	***
N_Gemme	1	0.0958	0.0958	0.7193	0.4117284	
Defogl	1	1.5180	1.5180	11.3932	0.0049721	**
Dirad	1	1.1318	1.1318	8.4946	0.0120663	*
Vigoria	2	4.2344	2.1172	15.8906	0.0003225	***
N_Gemme:Defogl	1	0.0339	0.0339	0.2547	0.6222637	
N_Gemme:Dirad	1	0.4809	0.4809	3.6092	0.0798725	.
Defogl:Dirad	1	0.0246	0.0246	0.1848	0.6743127	
N_Gemme:Vigoria	2	0.0142	0.0071	0.0533	0.9483025	
Defogl:Vigoria	2	1.9344	0.9672	7.2594	0.0076391	**
Dirad:Vigoria	2	0.4859	0.2430	1.8236	0.2004068	
N_Gemme:Defogl:Dirad	1	0.0717	0.0717	0.5384	0.4761108	
Residuals	13	1.7321	0.1332			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.654
 IBIMET_2: n.s
 IBIMET_3: 0.866

AcTartA

Analysis of Variance Table

Response: Br_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.0232	0.0232	0.1159	0.738971	
IBIMET_2	1	0.3566	0.3566	1.7840	0.204574	
IBIMET_3	1	6.0965	6.0965	30.4989	9.83e-05	***
N_Gemme	1	0.1339	0.1339	0.6698	0.427865	
Defogl	1	1.1186	1.1186	5.5960	0.034217	*
Dirad	1	0.7031	0.7031	3.5172	0.083369	.
Vigoria	2	4.6838	2.3419	11.7159	0.001233	**

N_Gemme:Defogl	1	0.0073	0.0073	0.0364	0.851712
N_Gemme:Dirad	1	0.3962	0.3962	1.9819	0.182655
Defogl:Dirad	1	0.0192	0.0192	0.0959	0.761721
N_Gemme:Vigoria	2	0.0260	0.0130	0.0650	0.937327
Defogl:Vigoria	2	1.6467	0.8234	4.1191	0.041151 *
Dirad:Vigoria	2	0.7209	0.3604	1.8031	0.203641
N_Gemme:Defogl:Dirad	1	0.0093	0.0093	0.0465	0.832528
Residuals	13	2.5986	0.1999		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: 0.837

AcMalicoA

Analysis of Variance Table

Response: Br_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.87003	0.87003	21.3124	0.000483 ***
IBIMET_2	1	0.00125	0.00125	0.0306	0.863743
IBIMET_3	1	0.23796	0.23796	5.8291	0.031235 *
N_Gemme	1	0.23152	0.23152	5.6714	0.033216 *
Defogl	1	0.00044	0.00044	0.0108	0.918885
Dirad	1	0.05076	0.05076	1.2435	0.284989
Vigoria	2	0.43677	0.21839	5.3496	0.020177 *
N_Gemme:Defogl	1	0.09446	0.09446	2.3139	0.152169
N_Gemme:Dirad	1	0.09726	0.09726	2.3825	0.146685
Defogl:Dirad	1	0.00346	0.00346	0.0847	0.775649
N_Gemme:Vigoria	2	0.03201	0.01600	0.3920	0.683405
Defogl:Vigoria	2	0.02390	0.01195	0.2927	0.751045
Dirad:Vigoria	2	0.09654	0.04827	1.1825	0.337423
N_Gemme:Defogl:Dirad	1	0.00138	0.00138	0.0338	0.856964
Residuals	13	0.53070	0.04082		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.788
IBIMET_2: n.s
IBIMET_3: 0.556

KA

Analysis of Variance Table

Response: Br_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.015855	0.015855	1.9345	0.18762
IBIMET_2	1	0.045098	0.045098	5.5025	0.03551 *
IBIMET_3	1	0.039735	0.039735	4.8482	0.04634 *
N_Gemme	1	0.037089	0.037089	4.5254	0.05311 .
Defogl	1	0.007070	0.007070	0.8626	0.36992
Dirad	1	0.000092	0.000092	0.0112	0.91738
Vigoria	2	0.065207	0.032604	3.9781	0.04489 *
N_Gemme:Defogl	1	0.001284	0.001284	0.1566	0.69871
N_Gemme:Dirad	1	0.026602	0.026602	3.2458	0.09484 .
Defogl:Dirad	1	0.000056	0.000056	0.0068	0.93544

N_Gemme:Vigoria	2	0.006680	0.003340	0.4075	0.67351
Defogl:Vigoria	2	0.001813	0.000906	0.1106	0.89614
Dirad:Vigoria	2	0.038685	0.019342	2.3600	0.13354
N_Gemme:Defogl:Dirad	1	0.000843	0.000843	0.1029	0.75352
Residuals	13	0.106546	0.008196		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: -0.545
 IBIMET_3: 0.521

ApaA

Analysis of Variance Table

Response: Br_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	2881.6	2881.59	9.7413	0.008110	**
IBIMET_2	1	641.2	641.21	2.1676	0.164736	
IBIMET_3	1	757.1	757.06	2.5593	0.133661	
N_Gemme	1	1118.9	1118.88	3.7824	0.073750	.
Defogl	1	2.1	2.13	0.0072	0.933617	
Dirad	1	1.4	1.44	0.0049	0.945370	
Vigoria	2	4962.1	2481.06	8.3873	0.004578	**
N_Gemme:Defogl	1	507.7	507.68	1.7162	0.212855	
N_Gemme:Dirad	1	265.5	265.51	0.8976	0.360711	
Defogl:Dirad	1	39.5	39.53	0.1336	0.720575	
N_Gemme:Vigoria	2	257.6	128.80	0.4354	0.656101	
Defogl:Vigoria	2	380.7	190.37	0.6435	0.541372	
Dirad:Vigoria	2	535.9	267.97	0.9059	0.428246	
N_Gemme:Defogl:Dirad	1	12.3	12.31	0.0416	0.841494	
Residuals	13	3845.5	295.81			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.654
 IBIMET_2: n.s
 IBIMET_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Br_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	267102	267102	107.7345	1.166e-07	***
IBIMET_2	1	41733	41733	16.8328	0.001247	**
IBIMET_3	1	4331	4331	1.7469	0.209052	
N_Gemme	1	1219	1219	0.4917	0.495533	
Defogl	1	222	222	0.0895	0.769504	
Dirad	1	34577	34577	13.9463	0.002500	**
Vigoria	2	47434	23717	9.5661	0.002789	**
N_Gemme:Defogl	1	5901	5901	2.3802	0.146868	
N_Gemme:Dirad	1	23497	23497	9.4776	0.008802	**
Defogl:Dirad	1	22	22	0.0088	0.926704	
N_Gemme:Vigoria	2	2615	1307	0.5273	0.602296	
Defogl:Vigoria	2	8773	4386	1.7692	0.209134	
Dirad:Vigoria	2	107	53	0.0215	0.978729	

N_Gemme:Defogl:Dirad 1 576 576 0.2325 0.637692
 Residuals 13 32230 2479

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.945
 IBIMET_2: -0.751
 IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Br_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	784431	784431	132.4038	3.452e-08	***
IBIMET_2	1	26426	26426	4.4605	0.05461	.
IBIMET_3	1	4133	4133	0.6976	0.41869	
N_Gemme	1	25953	25953	4.3806	0.05653	.
Defogl	1	3	3	0.0005	0.98229	
Dirad	1	10808	10808	1.8242	0.19985	
Vigoria	2	50586	25293	4.2692	0.03756	*
N_Gemme:Defogl	1	10124	10124	1.7089	0.21378	
N_Gemme:Dirad	1	2843	2843	0.4798	0.50070	
Defogl:Dirad	1	6435	6435	1.0861	0.31633	
N_Gemme:Vigoria	2	3146	1573	0.2655	0.77085	
Defogl:Vigoria	2	34462	17231	2.9084	0.09038	.
Dirad:Vigoria	2	2747	1374	0.2318	0.79628	
N_Gemme:Defogl:Dirad	1	11049	11049	1.8650	0.19521	
Residuals	13	77019	5925			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.954
 IBIMET_2: n.s
 IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Br_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	1084313	1084313	57.8000	3.888e-06	***
IBIMET_2	1	67599	67599	3.6034	0.08009	.
IBIMET_3	1	15813	15813	0.8429	0.37528	
N_Gemme	1	197	197	0.0105	0.92004	
Defogl	1	9398	9398	0.5010	0.49157	
Dirad	1	62021	62021	3.3061	0.09213	.
Vigoria	2	218247	109124	5.8169	0.01569	*
N_Gemme:Defogl	1	698	698	0.0372	0.85006	
N_Gemme:Dirad	1	129502	129502	6.9032	0.02089	*
Defogl:Dirad	1	1034	1034	0.0551	0.81808	
N_Gemme:Vigoria	2	44280	22140	1.1802	0.33808	
Defogl:Vigoria	2	39836	19918	1.0617	0.37402	
Dirad:Vigoria	2	9835	4918	0.2621	0.77337	
N_Gemme:Defogl:Dirad	1	43419	43419	2.3145	0.15212	
Residuals	13	243877	18760			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.904
IBIMET_2: n.s
IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Br_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	3031908	3031908	85.9137	4.309e-07	***
IBIMET_2	1	1834	1834	0.0520	0.82320	
IBIMET_3	1	6220	6220	0.1763	0.68147	
N_Gemme	1	66604	66604	1.8873	0.19273	
Defogl	1	18275	18275	0.5178	0.48450	
Dirad	1	4837	4837	0.1371	0.71718	
Vigoria	2	131373	65686	1.8613	0.19460	
N_Gemme:Defogl	1	24073	24073	0.6821	0.42375	
N_Gemme:Dirad	1	21550	21550	0.6107	0.44854	
Defogl:Dirad	1	31126	31126	0.8820	0.36478	
N_Gemme:Vigoria	2	2924	1462	0.0414	0.95955	
Defogl:Vigoria	2	332145	166073	4.7059	0.02901	*
Dirad:Vigoria	2	73798	36899	1.0456	0.37926	
N_Gemme:Defogl:Dirad	1	2185	2185	0.0619	0.80736	
Residuals	13	458772	35290			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.932
IBIMET_2: n.s
IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	5898969	5898969	55.4691	4.849e-06	***
IBIMET_2	1	946804	946804	8.9030	0.0105632	*
IBIMET_3	1	275682	275682	2.5923	0.1313882	
N_Gemme	1	133025	133025	1.2509	0.2836302	
Defogl	1	171511	171511	1.6128	0.2263664	
Dirad	1	2486691	2486691	23.3828	0.0003254	***
Vigoria	2	1350516	675258	6.3496	0.0119168	*
N_Gemme:Defogl	1	96104	96104	0.9037	0.3591346	
N_Gemme:Dirad	1	2834	2834	0.0266	0.8728456	
Defogl:Dirad	1	104376	104376	0.9815	0.3399198	
N_Gemme:Vigoria	2	422247	211124	1.9852	0.1768566	
Defogl:Vigoria	2	164831	82416	0.7750	0.4808767	
Dirad:Vigoria	2	2272836	1136418	10.6860	0.0018001	**
N_Gemme:Defogl:Dirad	1	75834	75834	0.7131	0.4136927	
Residuals	13	1382509	106347			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.9

IBIMET_2: -0.638
IBIMET_3: n.s

DensMostoB

Analysis of Variance Table

Response: Br_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.00031578	3.1578e-04	14.3669	0.002248	**
IBIMET_2	1	0.00014090	1.4090e-04	6.4103	0.025037	*
IBIMET_3	1	0.00004822	4.8223e-05	2.1940	0.162379	
N_Gemme	1	0.00000957	9.5660e-06	0.4352	0.520951	
Defogl	1	0.00000798	7.9770e-06	0.3629	0.557246	
Dirad	1	0.00003219	3.2186e-05	1.4644	0.247782	
Vigoria	2	0.00032790	1.6395e-04	7.4592	0.006956	**
N_Gemme:Defogl	1	0.00002166	2.1658e-05	0.9854	0.338995	
N_Gemme:Dirad	1	0.00006460	6.4601e-05	2.9391	0.110184	
Defogl:Dirad	1	0.00001282	1.2825e-05	0.5835	0.458590	
N_Gemme:Vigoria	2	0.00003855	1.9276e-05	0.8770	0.439263	
Defogl:Vigoria	2	0.00002070	1.0350e-05	0.4709	0.634689	
Dirad:Vigoria	2	0.00003298	1.6489e-05	0.7502	0.491663	
N_Gemme:Defogl:Dirad	1	0.00002150	2.1503e-05	0.9783	0.340665	
Residuals	13	0.00028574	2.1980e-05			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.725
IBIMET_2: -0.575
IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Br_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	14.7732	14.7732	14.4575	0.002198	**
IBIMET_2	1	6.5993	6.5993	6.4582	0.024594	*
IBIMET_3	1	2.5533	2.5533	2.4988	0.137950	
N_Gemme	1	0.4097	0.4097	0.4009	0.537592	
Defogl	1	0.3511	0.3511	0.3436	0.567814	
Dirad	1	1.4896	1.4896	1.4577	0.248799	
Vigoria	2	13.8907	6.9454	6.7969	0.009540	**
N_Gemme:Defogl	1	0.9534	0.9534	0.9330	0.351705	
N_Gemme:Dirad	1	3.0441	3.0441	2.9790	0.108021	
Defogl:Dirad	1	0.7262	0.7262	0.7107	0.414445	
N_Gemme:Vigoria	2	1.7753	0.8876	0.8687	0.442494	
Defogl:Vigoria	2	1.0274	0.5137	0.5027	0.616168	
Dirad:Vigoria	2	1.5338	0.7669	0.7505	0.491517	
N_Gemme:Defogl:Dirad	1	0.9671	0.9671	0.9464	0.348381	
Residuals	13	13.2839	1.0218			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.726
IBIMET_2: -0.576
IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Br_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.016425	0.016425	2.2929	0.15390
IBIMET_2	1	0.012166	0.012166	1.6983	0.21511
IBIMET_3	1	0.049873	0.049873	6.9621	0.02045 *
N_Gemme	1	0.002028	0.002028	0.2830	0.60369
Defogl	1	0.000525	0.000525	0.0733	0.79091
Dirad	1	0.051203	0.051203	7.1477	0.01914 *
Vigoria	2	0.090825	0.045412	6.3394	0.01198 *
N_Gemme:Defogl	1	0.010344	0.010344	1.4440	0.25093
N_Gemme:Dirad	1	0.015515	0.015515	2.1658	0.16490
Defogl:Dirad	1	0.000423	0.000423	0.0590	0.81185
N_Gemme:Vigoria	2	0.015924	0.007962	1.1115	0.35841
Defogl:Vigoria	2	0.009582	0.004791	0.6688	0.52910
Dirad:Vigoria	2	0.022256	0.011128	1.5534	0.24836
N_Gemme:Defogl:Dirad	1	0.000468	0.000468	0.0654	0.80222
Residuals	13	0.093126	0.007164		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: -0.591

AcTitolB

Analysis of Variance Table

Response: Br_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.0034	0.00338	0.0126	0.91250
IBIMET_2	1	0.1067	0.10670	0.3958	0.54018
IBIMET_3	1	1.6506	1.65064	6.1225	0.02790 *
N_Gemme	1	0.0737	0.07366	0.2732	0.60997
Defogl	1	0.3565	0.35652	1.3224	0.27088
Dirad	1	2.0736	2.07358	7.6912	0.01582 *
Vigoria	2	2.9920	1.49599	5.5488	0.01810 *
N_Gemme:Defogl	1	0.0105	0.01045	0.0388	0.84695
N_Gemme:Dirad	1	0.4776	0.47760	1.7715	0.20607
Defogl:Dirad	1	0.0222	0.02220	0.0824	0.77866
N_Gemme:Vigoria	2	0.3881	0.19404	0.7197	0.50530
Defogl:Vigoria	2	0.8904	0.44521	1.6513	0.22959
Dirad:Vigoria	2	0.1147	0.05736	0.2128	0.81110
N_Gemme:Defogl:Dirad	1	0.1413	0.14127	0.5240	0.48197
Residuals	13	3.5049	0.26960		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: -0.566

AcTartB

Analysis of Variance Table

Response: Br_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.2142	0.21421	0.8467	0.374260
IBIMET_2	1	0.3689	0.36887	1.4580	0.248762
IBIMET_3	1	2.6936	2.69363	10.6466	0.006173 **
N_Gemme	1	0.2077	0.20769	0.8209	0.381408
Defogl	1	0.5032	0.50324	1.9891	0.181915
Dirad	1	2.2267	2.22666	8.8009	0.010917 *
Viguria	2	3.4043	1.70213	6.7277	0.009869 **
N_Gemme:Defogl	1	0.0265	0.02653	0.1049	0.751232
N_Gemme:Dirad	1	0.5635	0.56349	2.2272	0.159463
Defogl:Dirad	1	0.1148	0.11478	0.4537	0.512400
N_Gemme:Viguria	2	0.1765	0.08827	0.3489	0.711866
Defogl:Viguria	2	0.5298	0.26491	1.0470	0.378779
Dirad:Viguria	2	0.1809	0.09047	0.3576	0.706034
N_Gemme:Defogl:Dirad	1	0.0532	0.05323	0.2104	0.654044
Residuals	13	3.2890	0.25300		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: n.s
 IBIMET_3: -0.671

AcMalicoB

Analysis of Variance Table

Response: Br_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1.89855	1.89855	129.7669	3.891e-08 ***
IBIMET_2	1	0.18420	0.18420	12.5902	0.003568 **
IBIMET_3	1	0.01226	0.01226	0.8381	0.376619
N_Gemme	1	0.00971	0.00971	0.6637	0.429936
Defogl	1	0.02181	0.02181	1.4907	0.243792
Dirad	1	0.00023	0.00023	0.0155	0.902920
Viguria	2	0.71182	0.35591	24.3265	4.036e-05 ***
N_Gemme:Defogl	1	0.00565	0.00565	0.3860	0.545145
N_Gemme:Dirad	1	0.00506	0.00506	0.3457	0.566649
Defogl:Dirad	1	0.02267	0.02267	1.5496	0.235173
N_Gemme:Viguria	2	0.00293	0.00146	0.1000	0.905533
Defogl:Viguria	2	0.11476	0.05738	3.9218	0.046484 *
Dirad:Viguria	2	0.18637	0.09319	6.3693	0.011798 *
N_Gemme:Defogl:Dirad	1	0.04733	0.04733	3.2348	0.095336 .
Residuals	13	0.19020	0.01463		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.953
 IBIMET_2: 0.701
 IBIMET_3: n.s

KB

Analysis of Variance Table

Response: Br_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.082332	0.082332	11.7272	0.004524	**
IBIMET_2	1	0.014054	0.014054	2.0019	0.180608	
IBIMET_3	1	0.035636	0.035636	5.0760	0.042174	*
N_Gemme	1	0.001024	0.001024	0.1458	0.708718	
Defogl	1	0.003667	0.003667	0.5223	0.482678	
Dirad	1	0.043004	0.043004	6.1255	0.027873	*
Vigoria	2	0.151190	0.075595	10.7676	0.001746	**
N_Gemme:Defogl	1	0.036705	0.036705	5.2282	0.039637	*
N_Gemme:Dirad	1	0.002209	0.002209	0.3146	0.584382	
Defogl:Dirad	1	0.010000	0.010000	1.4244	0.254006	
N_Gemme:Vigoria	2	0.006431	0.003216	0.4580	0.642363	
Defogl:Vigoria	2	0.018280	0.009140	1.3019	0.305245	
Dirad:Vigoria	2	0.143896	0.071948	10.2481	0.002129	**
N_Gemme:Defogl:Dirad	1	0.007072	0.007072	1.0073	0.333856	
Residuals	13	0.091268	0.007021			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.689
IBIMET_2: n.s
IBIMET_3: -0.53

ApaB

Analysis of Variance Table

Response: Br_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	1611.1	1611.13	6.4472	0.024695	*
IBIMET_2	1	908.3	908.32	3.6348	0.078930	.
IBIMET_3	1	229.7	229.66	0.9190	0.355221	
N_Gemme	1	434.2	434.24	1.7377	0.210191	
Defogl	1	4.6	4.56	0.0183	0.894561	
Dirad	1	0.0	0.04	0.0001	0.990644	
Vigoria	2	3416.8	1708.38	6.8363	0.009358	**
N_Gemme:Defogl	1	215.8	215.80	0.8636	0.369681	
N_Gemme:Dirad	1	382.0	382.01	1.5287	0.238184	
Defogl:Dirad	1	30.6	30.59	0.1224	0.732018	
N_Gemme:Vigoria	2	62.1	31.05	0.1242	0.884204	
Defogl:Vigoria	2	1335.1	667.56	2.6713	0.106691	
Dirad:Vigoria	2	3939.2	1969.61	7.8817	0.005730	**
N_Gemme:Defogl:Dirad	1	631.9	631.92	2.5287	0.135804	
Residuals	13	3248.7	249.90			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.576
IBIMET_2: n.s
IBIMET_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Br_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	64141	64141	28.3720	0.0001374	***
IBIMET_2	1	6703	6703	2.9648	0.1087871	

IBIMET_3	1	1035	1035	0.4578	0.5105255
N_Gemme	1	3356	3356	1.4846	0.2447143
Defogl	1	18022	18022	7.9719	0.0143698 *
Dirad	1	3771	3771	1.6682	0.2189899
Vigoria	2	16479	8240	3.6447	0.0553847 .
N_Gemme:Defogl	1	1885	1885	0.8337	0.3778413
N_Gemme:Dirad	1	3	3	0.0012	0.9724480
Defogl:Dirad	1	8086	8086	3.5766	0.0810898 .
N_Gemme:Vigoria	2	1416	708	0.3133	0.7364184
Defogl:Vigoria	2	807	403	0.1785	0.8385761
Dirad:Vigoria	2	14488	7244	3.2044	0.0738991 .
N_Gemme:Defogl:Dirad	1	6590	6590	2.9150	0.1115123
Residuals	13	29389	2261		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.828

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	294775	294775	71.7991	1.185e-06	***
IBIMET_2	1	638	638	0.1553	0.699879	
IBIMET_3	1	10606	10606	2.5833	0.131999	
N_Gemme	1	6848	6848	1.6679	0.219026	
Defogl	1	37498	37498	9.1335	0.009812	**
Dirad	1	8725	8725	2.1251	0.168636	
Vigoria	2	12919	6459	1.5734	0.244391	
N_Gemme:Defogl	1	40	40	0.0099	0.922436	
N_Gemme:Dirad	1	9006	9006	2.1937	0.162404	
Defogl:Dirad	1	4294	4294	1.0458	0.325122	
N_Gemme:Vigoria	2	2206	1103	0.2687	0.768522	
Defogl:Vigoria	2	10297	5149	1.2541	0.317688	
Dirad:Vigoria	2	57171	28586	6.9627	0.008802	**
N_Gemme:Defogl:Dirad	1	1515	1515	0.3690	0.554013	
Residuals	13	53372	4106			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.92

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	479894	479894	18.9602	0.0007807	***
IBIMET_2	1	27382	27382	1.0818	0.3172448	
IBIMET_3	1	9852	9852	0.3893	0.5434854	
N_Gemme	1	329	329	0.0130	0.9109719	
Defogl	1	117178	117178	4.6296	0.0508026	.

Dirad	1	270	270	0.0107	0.9192939
Vigoria	2	89019	44509	1.7585	0.2108964
N_Gemme:Defogl	1	380	380	0.0150	0.9043896
N_Gemme:Dirad	1	43	43	0.0017	0.9677413
Defogl:Dirad	1	23314	23314	0.9211	0.3546923
N_Gemme:Vigoria	2	5595	2798	0.1105	0.8961863
Defogl:Vigoria	2	68888	34444	1.3608	0.2906599
Dirad:Vigoria	2	8388	4194	0.1657	0.8490549
N_Gemme:Defogl:Dirad	1	2288	2288	0.0904	0.7684406
Residuals	13	329037	25311		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.77
 IBIMET_2: n.s
 IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Br_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	2158407	2158407	39.9683	2.650e-05	***
IBIMET_2	1	313	313	0.0058	0.94048	
IBIMET_3	1	58784	58784	1.0885	0.31581	
N_Gemme	1	1922	1922	0.0356	0.85327	
Defogl	1	255367	255367	4.7288	0.04872	*
Dirad	1	2982	2982	0.0552	0.81787	
Vigoria	2	89033	44517	0.8243	0.46020	
N_Gemme:Defogl	1	40023	40023	0.7411	0.40490	
N_Gemme:Dirad	1	41442	41442	0.7674	0.39692	
Defogl:Dirad	1	5483	5483	0.1015	0.75507	
N_Gemme:Vigoria	2	7839	3919	0.0726	0.93037	
Defogl:Vigoria	2	77878	38939	0.7211	0.50470	
Dirad:Vigoria	2	28349	14174	0.2625	0.77312	
N_Gemme:Defogl:Dirad	1	43744	43744	0.8100	0.38448	
Residuals	13	702039	54003			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.869
 IBIMET_2: n.s
 IBIMET_3: n.s

BROLIO 2009M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Br_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1037907	1037907	15.2503	0.001808	**
DIPROVE_2	1	3283012	3283012	48.2383	1.014e-05	***
DIPROVE_3	1	388008	388008	5.7011	0.032832	*
N_Gemme	1	423	423	0.0062	0.938396	
Defogl	1	346424	346424	5.0901	0.041930	*

Dirad	1	608657	608657	8.9432	0.010427	*
Vigoria	2	680734	340367	5.0011	0.024498	*
N_Gemme:Defogl	1	1834	1834	0.0270	0.872126	
N_Gemme:Dirad	1	5915	5915	0.0869	0.772801	
Defogl:Dirad	1	46604	46604	0.6848	0.422881	
N_Gemme:Vigoria	2	62603	31302	0.4599	0.641220	
Defogl:Vigoria	2	371623	185812	2.7302	0.102346	
Dirad:Vigoria	2	46670	23335	0.3429	0.715962	
N_Gemme:Defogl:Dirad	1	202209	202209	2.9711	0.108447	
Residuals	13	884757	68058			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.735
DIPROVE_2: 0.888
DIPROVE_3: 0.552

DensMostoA

Analysis of Variance Table

Response: Br_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.00049096	0.00049096	13.2144	0.003022	**
DIPROVE_2	1	0.00012770	0.00012770	3.4372	0.086567	.
DIPROVE_3	1	0.00001448	0.00001448	0.3897	0.543250	
N_Gemme	1	0.00000298	0.00000298	0.0803	0.781357	
Defogl	1	0.00000001	0.00000001	0.0003	0.986647	
Dirad	1	0.00003149	0.00003149	0.8476	0.374006	
Vigoria	2	0.00009149	0.00004574	1.2312	0.323845	
N_Gemme:Defogl	1	0.00002757	0.00002757	0.7421	0.404590	
N_Gemme:Dirad	1	0.00001327	0.00001327	0.3572	0.560306	
Defogl:Dirad	1	0.00005209	0.00005209	1.4020	0.257586	
N_Gemme:Vigoria	2	0.00000093	0.00000046	0.0125	0.987609	
Defogl:Vigoria	2	0.00001285	0.00000642	0.1729	0.843102	
Dirad:Vigoria	2	0.00016040	0.00008020	2.1587	0.155060	
N_Gemme:Defogl:Dirad	1	0.00001459	0.00001459	0.3926	0.541792	
Residuals	13	0.00048299	0.00003715			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.71
DIPROVE_2: n.s
DIPROVE_3: n.s

BrixA

Analysis of Variance Table

Response: Br_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	22.4849	22.4849	13.5568	0.002764	**
DIPROVE_2	1	5.8307	5.8307	3.5155	0.083437	.
DIPROVE_3	1	0.6119	0.6119	0.3689	0.554056	
N_Gemme	1	0.1761	0.1761	0.1062	0.749696	
Defogl	1	0.0000	0.0000	0.0000	0.999908	
Dirad	1	1.3802	1.3802	0.8321	0.378259	
Vigoria	2	3.8781	1.9391	1.1691	0.341264	
N_Gemme:Defogl	1	1.3755	1.3755	0.8293	0.379050	

N_Gemme:Dirad	1	0.6445	0.6445	0.3886	0.543827
Defogl:Dirad	1	2.4281	2.4281	1.4640	0.247841
N_Gemme:Vigoria	2	0.0743	0.0371	0.0224	0.977893
Defogl:Vigoria	2	0.5989	0.2995	0.1805	0.836871
Dirad:Vigoria	2	7.7825	3.8912	2.3461	0.134907
N_Gemme:Defogl:Dirad	1	0.6660	0.6660	0.4016	0.537273
Residuals	13	21.5614	1.6586		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.714
DIPROVE_2: n.s
DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Br_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.024659	0.024659	3.2946	0.09264 .
DIPROVE_2	1	0.021361	0.021361	2.8540	0.11497
DIPROVE_3	1	0.030918	0.030918	4.1310	0.06306 .
N_Gemme	1	0.006751	0.006751	0.9020	0.35956
Defogl	1	0.022122	0.022122	2.9558	0.10928
Dirad	1	0.021246	0.021246	2.8387	0.11586
Vigoria	2	0.099388	0.049694	6.6396	0.01031 *
N_Gemme:Defogl	1	0.000049	0.000049	0.0065	0.93705
N_Gemme:Dirad	1	0.000001	0.000001	0.0002	0.99036
Defogl:Dirad	1	0.004804	0.004804	0.6419	0.43744
N_Gemme:Vigoria	2	0.007728	0.003864	0.5163	0.60847
Defogl:Vigoria	2	0.004749	0.002374	0.3173	0.73363
Dirad:Vigoria	2	0.014860	0.007430	0.9927	0.39700
N_Gemme:Defogl:Dirad	1	0.000278	0.000278	0.0372	0.85004
Residuals	13	0.097298	0.007484		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTitola

Analysis of Variance Table

Response: Br_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.2915	0.29151	0.7478	0.40285
DIPROVE_2	1	0.5684	0.56841	1.4581	0.24874
DIPROVE_3	1	0.4302	0.43022	1.1036	0.31261
N_Gemme	1	0.1456	0.14563	0.3736	0.55159
Defogl	1	2.3824	2.38237	6.1114	0.02802 *
Dirad	1	1.6874	1.68740	4.3286	0.05782 .
Vigoria	2	5.5899	2.79496	7.1698	0.00797 **
N_Gemme:Defogl	1	0.0207	0.02070	0.0531	0.82135
N_Gemme:Dirad	1	0.0178	0.01777	0.0456	0.83423
Defogl:Dirad	1	0.0902	0.09022	0.2314	0.63845
N_Gemme:Vigoria	2	0.3491	0.17457	0.4478	0.64852

Defogl:Vigoria	2	0.9501	0.47503	1.2186	0.32730
Dirad:Vigoria	2	0.5914	0.29569	0.7585	0.48801
N_Gemme:Defogl:Dirad	1	0.0584	0.05843	0.1499	0.70491
Residuals	13	5.0677	0.38983		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTartA

Analysis of Variance Table

Response: Br_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.0097	0.0097	0.0310	0.863028
DIPROVE_2	1	0.1476	0.1476	0.4701	0.505004
DIPROVE_3	1	0.0091	0.0091	0.0289	0.867713
N_Gemme	1	0.0985	0.0985	0.3137	0.584946
Defogl	1	2.8487	2.8487	9.0745	0.009998 **
Dirad	1	1.4450	1.4450	4.6031	0.051379 .
Vigoria	2	7.6253	3.8126	12.1450	0.001060 **
N_Gemme:Defogl	1	0.0699	0.0699	0.2227	0.644795
N_Gemme:Dirad	1	0.0538	0.0538	0.1713	0.685747
Defogl:Dirad	1	0.0813	0.0813	0.2590	0.619361
N_Gemme:Vigoria	2	0.1479	0.0739	0.2355	0.793453
Defogl:Vigoria	2	0.7874	0.3937	1.2541	0.317667
Dirad:Vigoria	2	0.9647	0.4824	1.5365	0.251766
N_Gemme:Defogl:Dirad	1	0.1698	0.1698	0.5410	0.475067
Residuals	13	4.0810	0.3139		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Br_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.23619	0.23619	5.7016	0.032826 *
DIPROVE_2	1	0.21815	0.21815	5.2660	0.039035 *
DIPROVE_3	1	0.03166	0.03166	0.7643	0.397842
N_Gemme	1	0.19103	0.19103	4.6115	0.051195 .
Defogl	1	0.23779	0.23779	5.7403	0.032333 *
Dirad	1	0.06971	0.06971	1.6828	0.217109
Vigoria	2	0.73097	0.36549	8.8228	0.003795 **
N_Gemme:Defogl	1	0.24510	0.24510	5.9168	0.030192 *
N_Gemme:Dirad	1	0.03194	0.03194	0.7710	0.395854
Defogl:Dirad	1	0.00437	0.00437	0.1055	0.750535
N_Gemme:Vigoria	2	0.04359	0.02180	0.5261	0.602945
Defogl:Vigoria	2	0.01474	0.00737	0.1779	0.839060
Dirad:Vigoria	2	0.11428	0.05714	1.3793	0.286259
N_Gemme:Defogl:Dirad	1	0.00039	0.00039	0.0095	0.923909

Residuals 13 0.53853 0.04143

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.552

DIPROVE_2: 0.537

DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Br_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.003106	0.003106	0.3068	0.58906
DIPROVE_2	1	0.014305	0.014305	1.4130	0.25581
DIPROVE_3	1	0.066696	0.066696	6.5881	0.02344 *
N_Gemme	1	0.020962	0.020962	2.0706	0.17380
Defogl	1	0.007558	0.007558	0.7466	0.40322
Dirad	1	0.000825	0.000825	0.0815	0.77982
Vigoria	2	0.057105	0.028553	2.8204	0.09607 .
N_Gemme:Defogl	1	0.008207	0.008207	0.8107	0.38430
N_Gemme:Dirad	1	0.003023	0.003023	0.2986	0.59402
Defogl:Dirad	1	0.000503	0.000503	0.0496	0.82714
N_Gemme:Vigoria	2	0.006795	0.003398	0.3356	0.72091
Defogl:Vigoria	2	0.006645	0.003322	0.3282	0.72603
Dirad:Vigoria	2	0.063743	0.031871	3.1482	0.07674 .
N_Gemme:Defogl:Dirad	1	0.001574	0.001574	0.1554	0.69978
Residuals	13	0.131607	0.010124		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: -0.58

ApaA

Analysis of Variance Table

Response: Br_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	2487.3	2487.32	4.3054	0.05841 .
DIPROVE_2	1	1612.3	1612.29	2.7908	0.11870
DIPROVE_3	1	743.9	743.87	1.2876	0.27699
N_Gemme	1	455.6	455.65	0.7887	0.39063
Defogl	1	52.2	52.20	0.0904	0.76848
Dirad	1	67.9	67.92	0.1176	0.73717
Vigoria	2	497.8	248.91	0.4308	0.65891
N_Gemme:Defogl	1	44.9	44.88	0.0777	0.78483
N_Gemme:Dirad	1	5.4	5.45	0.0094	0.92412
Defogl:Dirad	1	12.5	12.51	0.0217	0.88527
N_Gemme:Vigoria	2	408.4	204.22	0.3535	0.70877
Defogl:Vigoria	2	782.9	391.45	0.6776	0.52491
Dirad:Vigoria	2	1513.4	756.71	1.3098	0.30323
N_Gemme:Defogl:Dirad	1	14.2	14.15	0.0245	0.87804
Residuals	13	7510.5	577.73		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Br_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	67572	67572	13.6436	0.002702	**
DIPROVE_2	1	192896	192896	38.9479	3.016e-05	***
DIPROVE_3	1	43491	43491	8.7813	0.010987	*
N_Gemme	1	8	8	0.0016	0.969045	
Defogl	1	2252	2252	0.4548	0.511885	
Dirad	1	19751	19751	3.9880	0.067200	.
Vigoria	2	40185	20093	4.0569	0.042751	*
N_Gemme:Defogl	1	8106	8106	1.6366	0.223153	
N_Gemme:Dirad	1	14706	14706	2.9693	0.108546	
Defogl:Dirad	1	18	18	0.0036	0.952947	
N_Gemme:Vigoria	2	4747	2374	0.4793	0.629750	
Defogl:Vigoria	2	8883	4442	0.8968	0.431674	
Dirad:Vigoria	2	1294	647	0.1306	0.878694	
N_Gemme:Defogl:Dirad	1	2045	2045	0.4128	0.531706	
Residuals	13	64385	4953			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.716
 DIPROVE_2: -0.866
 DIPROVE_3: -0.635

Poliftot1A

Analysis of Variance Table

Response: Br_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	214132	214132	19.2061	0.0007412	***
DIPROVE_2	1	364222	364222	32.6682	7.105e-05	***
DIPROVE_3	1	73255	73255	6.5704	0.0235918	*
N_Gemme	1	12191	12191	1.0935	0.3147601	
Defogl	1	24777	24777	2.2223	0.1598897	
Dirad	1	2875	2875	0.2579	0.6201014	
Vigoria	2	120717	60359	5.4137	0.0194815	*
N_Gemme:Defogl	1	6879	6879	0.6170	0.4462406	
N_Gemme:Dirad	1	1234	1234	0.1106	0.7447222	
Defogl:Dirad	1	8758	8758	0.7855	0.3915510	
N_Gemme:Vigoria	2	1071	535	0.0480	0.9532780	
Defogl:Vigoria	2	66067	33034	2.9629	0.0870534	.
Dirad:Vigoria	2	5921	2960	0.2655	0.7708592	
N_Gemme:Defogl:Dirad	1	3129	3129	0.2806	0.6052140	
Residuals	13	144939	11149			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.772
 DIPROVE_2: -0.846

DIPROVE_3: 0.579

Antoctot2A

Analysis of Variance Table

Response: Br_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	362772	362772	12.8945	0.003289	**
DIPROVE_2	1	652540	652540	23.1942	0.000337	***
DIPROVE_3	1	133888	133888	4.7590	0.048102	*
N_Gemme	1	2667	2667	0.0948	0.763025	
Defogl	1	44260	44260	1.5732	0.231828	
Dirad	1	22850	22850	0.8122	0.383870	
Vigoria	2	92575	46288	1.6453	0.230702	
N_Gemme:Defogl	1	858	858	0.0305	0.864080	
N_Gemme:Dirad	1	80932	80932	2.8767	0.113672	
Defogl:Dirad	1	3241	3241	0.1152	0.739727	
N_Gemme:Vigoria	2	50601	25300	0.8993	0.430728	
Defogl:Vigoria	2	72813	36407	1.2941	0.307238	
Dirad:Vigoria	2	24869	12434	0.4420	0.652078	
N_Gemme:Defogl:Dirad	1	59463	59463	2.1136	0.169709	
Residuals	13	365739	28134			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.706

DIPROVE_2: -0.801

DIPROVE_3: -0.518

Poliftot2A

Analysis of Variance Table

Response: Br_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1180074	1180074	22.2617	0.0004018	***
DIPROVE_2	1	939816	939816	17.7293	0.0010192	**
DIPROVE_3	1	181915	181915	3.4318	0.0867890	.
N_Gemme	1	30552	30552	0.5764	0.4612870	
Defogl	1	289972	289972	5.4702	0.0359656	*
Dirad	1	23571	23571	0.4447	0.5165412	
Vigoria	2	128551	64275	1.2125	0.3289707	
N_Gemme:Defogl	1	20210	20210	0.3813	0.5475972	
N_Gemme:Dirad	1	5934	5934	0.1119	0.7432764	
Defogl:Dirad	1	23715	23715	0.4474	0.5152847	
N_Gemme:Vigoria	2	4297	2148	0.0405	0.9604014	
Defogl:Vigoria	2	572965	286482	5.4044	0.0195813	*
Dirad:Vigoria	2	98547	49273	0.9295	0.4194603	
N_Gemme:Defogl:Dirad	1	18385	18385	0.3468	0.5660075	
Residuals	13	689120	53009			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.795

DIPROVE_2: -0.76

DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1301016	1301016	14.0439	0.0024389	**
DIPROVE_2	1	4663789	4663789	50.3437	8.107e-06	***
DIPROVE_3	1	451553	451553	4.8743	0.0458369	*
N_Gemme	1	16272	16272	0.1757	0.6819785	
Defogl	1	87800	87800	0.9478	0.3480558	
Dirad	1	1858483	1858483	20.0616	0.0006207	***
Vigoria	2	1650511	825255	8.9083	0.0036604	**
N_Gemme:Defogl	1	12537	12537	0.1353	0.7188823	
N_Gemme:Dirad	1	180909	180909	1.9528	0.1856745	
Defogl:Dirad	1	82947	82947	0.8954	0.3612802	
N_Gemme:Vigoria	2	261397	130699	1.4108	0.2789273	
Defogl:Vigoria	2	266816	133408	1.4401	0.2723164	
Dirad:Vigoria	2	3407357	1703679	18.3905	0.0001621	***
N_Gemme:Defogl:Dirad	1	339074	339074	3.6602	0.0780084	.
Residuals	13	1204307	92639			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.721
 DIPROVE_2: 0.891
 DIPROVE_3: -0.522

DensMostoB

Analysis of Variance Table

Response: Br_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.00025765	2.5765e-04	4.6602	0.05015	.
DIPROVE_2	1	0.00001176	1.1756e-05	0.2126	0.65234	
DIPROVE_3	1	0.00001842	1.8423e-05	0.3332	0.57363	
N_Gemme	1	0.00002495	2.4953e-05	0.4513	0.51347	
Defogl	1	0.00003984	3.9839e-05	0.7206	0.41131	
Dirad	1	0.00002118	2.1175e-05	0.3830	0.54670	
Vigoria	2	0.00006461	3.2307e-05	0.5843	0.57147	
N_Gemme:Defogl	1	0.00000095	9.5400e-07	0.0172	0.89752	
N_Gemme:Dirad	1	0.00001361	1.3612e-05	0.2462	0.62804	
Defogl:Dirad	1	0.00000337	3.3700e-06	0.0609	0.80886	
N_Gemme:Vigoria	2	0.00003785	1.8926e-05	0.3423	0.71633	
Defogl:Vigoria	2	0.00001277	6.3850e-06	0.1155	0.89183	
Dirad:Vigoria	2	0.00013466	6.7331e-05	1.2178	0.32750	
N_Gemme:Defogl:Dirad	1	0.00002071	2.0706e-05	0.3745	0.55111	
Residuals	13	0.00071874	5.5288e-05			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Br_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	11.906	11.9056	4.6604	0.05014 .
DIPROVE_2	1	0.501	0.5010	0.1961	0.66515
DIPROVE_3	1	0.811	0.8112	0.3176	0.58267
N_Gemme	1	1.078	1.0777	0.4219	0.52732
Defogl	1	1.784	1.7839	0.6983	0.41845
Dirad	1	1.032	1.0323	0.4041	0.53602
Vigoria	2	2.622	1.3109	0.5132	0.61024
N_Gemme:Defogl	1	0.035	0.0350	0.0137	0.90864
N_Gemme:Dirad	1	0.650	0.6505	0.2546	0.62228
Defogl:Dirad	1	0.234	0.2338	0.0915	0.76704
N_Gemme:Vigoria	2	1.675	0.8373	0.3277	0.72633
Defogl:Vigoria	2	0.557	0.2783	0.1089	0.89759
Dirad:Vigoria	2	6.343	3.1716	1.2415	0.32105
N_Gemme:Defogl:Dirad	1	0.941	0.9409	0.3683	0.55437
Residuals	13	33.210	2.5546		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Br_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.000861	0.000861	0.1250	0.729361
DIPROVE_2	1	0.002054	0.002054	0.2982	0.594238
DIPROVE_3	1	0.014559	0.014559	2.1141	0.169659
N_Gemme	1	0.005642	0.005642	0.8192	0.381872
Defogl	1	0.005236	0.005236	0.7603	0.399061
Dirad	1	0.053050	0.053050	7.7035	0.015751 *
Vigoria	2	0.145561	0.072780	10.5686	0.001882 **
N_Gemme:Defogl	1	0.001992	0.001992	0.2892	0.599808
N_Gemme:Dirad	1	0.006218	0.006218	0.9030	0.359324
Defogl:Dirad	1	0.000222	0.000222	0.0323	0.860242
N_Gemme:Vigoria	2	0.017031	0.008516	1.2366	0.322380
Defogl:Vigoria	2	0.003457	0.001729	0.2510	0.781706
Dirad:Vigoria	2	0.045189	0.022594	3.2810	0.070217 .
N_Gemme:Defogl:Dirad	1	0.000088	0.000088	0.0127	0.911838
Residuals	13	0.089524	0.006886		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTitolB

Analysis of Variance Table

Response: Br_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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DIPROVE_1	1	0.0073	0.00727	0.0218	0.88476
DIPROVE_2	1	0.0049	0.00494	0.0148	0.90495
DIPROVE_3	1	0.0309	0.03087	0.0927	0.76557
N_Gemme	1	0.1812	0.18120	0.5443	0.47377
Defogl	1	0.3271	0.32714	0.9827	0.33964
Dirad	1	2.3849	2.38492	7.1638	0.01903 *
Viguria	2	3.8515	1.92575	5.7846	0.01596 *
N_Gemme:Defogl	1	0.0440	0.04399	0.1321	0.72208
N_Gemme:Dirad	1	0.2537	0.25372	0.7621	0.39850
Defogl:Dirad	1	0.0257	0.02567	0.0771	0.78561
N_Gemme:Viguria	2	0.5347	0.26735	0.8031	0.46897
Defogl:Viguria	2	0.3544	0.17722	0.5323	0.59951
Dirad:Viguria	2	0.1944	0.09718	0.2919	0.75161
N_Gemme:Defogl:Dirad	1	0.2835	0.28351	0.8516	0.37291
Residuals	13	4.3279	0.33291		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTartB

Analysis of Variance Table

Response: Br_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.1000	0.10003	0.2812	0.60483
DIPROVE_2	1	0.1158	0.11584	0.3257	0.57793
DIPROVE_3	1	0.2102	0.21024	0.5911	0.45573
N_Gemme	1	0.3962	0.39619	1.1139	0.31045
Defogl	1	0.4585	0.45851	1.2891	0.27671
Dirad	1	2.6764	2.67645	7.5251	0.01675 *
Viguria	2	4.7508	2.37540	6.6787	0.01011 *
N_Gemme:Defogl	1	0.1051	0.10505	0.2954	0.59600
N_Gemme:Dirad	1	0.2964	0.29639	0.8333	0.37793
Defogl:Dirad	1	0.1057	0.10571	0.2972	0.59486
N_Gemme:Viguria	2	0.2498	0.12491	0.3512	0.71033
Defogl:Viguria	2	0.1223	0.06114	0.1719	0.84395
Dirad:Viguria	2	0.2046	0.10229	0.2876	0.75472
N_Gemme:Defogl:Dirad	1	0.1373	0.13731	0.3861	0.54511
Residuals	13	4.6237	0.35567		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Br_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.47682	0.47682	17.3454	0.0011102 **
DIPROVE_2	1	0.68724	0.68724	24.9998	0.0002430 ***
DIPROVE_3	1	0.35352	0.35352	12.8598	0.0033192 **

N_Gemme	1	0.01146	0.01146	0.4170	0.5296578	
Defogl	1	0.05199	0.05199	1.8914	0.1922813	
Dirad	1	0.00380	0.00380	0.1382	0.7161080	
Vigoria	2	1.15875	0.57938	21.0760	8.327e-05	***
N_Gemme:Defogl	1	0.01209	0.01209	0.4397	0.5188549	
N_Gemme:Dirad	1	0.02348	0.02348	0.8541	0.3722304	
Defogl:Dirad	1	0.02267	0.02267	0.8248	0.3803085	
N_Gemme:Vigoria	2	0.00086	0.00043	0.0156	0.9845618	
Defogl:Vigoria	2	0.06418	0.03209	1.1673	0.3417881	
Dirad:Vigoria	2	0.16694	0.08347	3.0364	0.0827849	.
N_Gemme:Defogl:Dirad	1	0.02236	0.02236	0.8136	0.3834785	
Residuals	13	0.35737	0.02749			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.756
DIPROVE_2: 0.811
DIPROVE_3: -0.705

KB

Analysis of Variance Table

Response: Br_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.005966	0.005966	0.9244	0.3538687	
DIPROVE_2	1	0.011906	0.011906	1.8447	0.1975108	
DIPROVE_3	1	0.027302	0.027302	4.2300	0.0603636	.
N_Gemme	1	0.002460	0.002460	0.3811	0.5476820	
Defogl	1	0.003492	0.003492	0.5410	0.4750618	
Dirad	1	0.043429	0.043429	6.7286	0.0222599	*
Vigoria	2	0.261756	0.130878	20.2773	0.0001008	***
N_Gemme:Defogl	1	0.011824	0.011824	1.8319	0.1989689	
N_Gemme:Dirad	1	0.000019	0.000019	0.0030	0.9570455	
Defogl:Dirad	1	0.008388	0.008388	1.2995	0.2748646	
N_Gemme:Vigoria	2	0.007990	0.003995	0.6189	0.5536550	
Defogl:Vigoria	2	0.026663	0.013332	2.0655	0.1663577	
Dirad:Vigoria	2	0.151665	0.075832	11.7489	0.0012187	**
N_Gemme:Defogl:Dirad	1	0.000000	0.000000	0.0000	0.9998245	
Residuals	13	0.083907	0.006454			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Br_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	3204.5	3204.5	11.8998	0.004311	**
DIPROVE_2	1	601.8	601.8	2.2346	0.158822	
DIPROVE_3	1	818.3	818.3	3.0386	0.104892	
N_Gemme	1	201.7	201.7	0.7491	0.402447	
Defogl	1	193.8	193.8	0.7197	0.411592	
Dirad	1	106.5	106.5	0.3957	0.540237	

Vigoria	2	895.5	447.7	1.6627	0.227521
N_Gemme:Defogl	1	162.4	162.4	0.6030	0.451352
N_Gemme:Dirad	1	674.8	674.8	2.5057	0.137448
Defogl:Dirad	1	11.3	11.3	0.0419	0.841050
N_Gemme:Vigoria	2	90.8	45.4	0.1687	0.846616
Defogl:Vigoria	2	459.9	230.0	0.8540	0.448271
Dirad:Vigoria	2	5348.5	2674.2	9.9306	0.002411 **
N_Gemme:Defogl:Dirad	1	179.5	179.5	0.6666	0.428936
Residuals	13	3500.8	269.3		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.691
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Br_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	35040	35040	14.8483	0.001995	**
DIPROVE_2	1	38161	38161	16.1710	0.001453	**
DIPROVE_3	1	4870	4870	2.0638	0.174462	
N_Gemme	1	3465	3465	1.4682	0.247191	
Defogl	1	1052	1052	0.4459	0.515987	
Dirad	1	1547	1547	0.6554	0.432743	
Vigoria	2	889	445	0.1884	0.830500	
N_Gemme:Defogl	1	11853	11853	5.0229	0.043104	*
N_Gemme:Dirad	1	907	907	0.3846	0.545893	
Defogl:Dirad	1	7552	7552	3.2003	0.096936	.
N_Gemme:Vigoria	2	1283	642	0.2719	0.766151	
Defogl:Vigoria	2	4110	2055	0.8709	0.441621	
Dirad:Vigoria	2	18842	9421	3.9923	0.044492	*
N_Gemme:Defogl:Dirad	1	15921	15921	6.7466	0.022114	*
Residuals	13	30678	2360			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.73
 DIPROVE_2: -0.745
 DIPROVE_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	166213	166213	39.6483	2.759e-05	***
DIPROVE_2	1	91721	91721	21.8788	0.0004325	***
DIPROVE_3	1	15205	15205	3.6269	0.0792169	.
N_Gemme	1	7000	7000	1.6697	0.2187979	
Defogl	1	10244	10244	2.4437	0.1420065	
Dirad	1	5285	5285	1.2608	0.2818167	
Vigoria	2	12304	6152	1.4675	0.2662833	
N_Gemme:Defogl	1	7422	7422	1.7705	0.2061890	
N_Gemme:Dirad	1	26795	26795	6.3917	0.0252117	*

Defogl:Dirad	1	4000	4000	0.9541	0.3465055	
N_Gemme:Vigoria	2	1617	808	0.1928	0.8269255	
Defogl:Vigoria	2	30750	15375	3.6676	0.0545796	.
Dirad:Vigoria	2	61193	30596	7.2984	0.0074999	**
N_Gemme:Defogl:Dirad	1	15662	15662	3.7359	0.0753354	.
Residuals	13	54499	4192			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.868

DIPROVE_2: -0.792

DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	314856	314856	10.2571	0.006932	**
DIPROVE_2	1	204563	204563	6.6641	0.022792	*
DIPROVE_3	1	52708	52708	1.7171	0.212748	
N_Gemme	1	2011	2011	0.0655	0.801983	
Defogl	1	2692	2692	0.0877	0.771787	
Dirad	1	1258	1258	0.0410	0.842692	
Vigoria	2	2488	1244	0.0405	0.960399	
N_Gemme:Defogl	1	36740	36740	1.1969	0.293801	
N_Gemme:Dirad	1	6235	6235	0.2031	0.659648	
Defogl:Dirad	1	24173	24173	0.7875	0.390984	
N_Gemme:Vigoria	2	7327	3663	0.1193	0.888467	
Defogl:Vigoria	2	49858	24929	0.8121	0.465217	
Dirad:Vigoria	2	47698	23849	0.7769	0.480037	
N_Gemme:Defogl:Dirad	1	10197	10197	0.3322	0.574225	
Residuals	13	399053	30696			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.664

DIPROVE_2: -0.582

DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Br_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1489938	1489938	23.6218	0.0003114	***
DIPROVE_2	1	478528	478528	7.5867	0.0163995	*
DIPROVE_3	1	169414	169414	2.6859	0.1252015	
N_Gemme	1	59	59	0.0009	0.9761294	
Defogl	1	38057	38057	0.6034	0.4512022	
Dirad	1	11610	11610	0.1841	0.6749234	
Vigoria	2	127669	63835	1.0120	0.3903974	
N_Gemme:Defogl	1	1598	1598	0.0253	0.8759804	
N_Gemme:Dirad	1	144209	144209	2.2863	0.1544415	
Defogl:Dirad	1	8689	8689	0.1378	0.7164994	
N_Gemme:Vigoria	2	12677	6339	0.1005	0.9050870	
Defogl:Vigoria	2	68266	34133	0.5412	0.5946365	

Dirad:Vigoria	2	132226	66113	1.0482	0.3784113
N_Gemme:Defogl:Dirad	1	10695	10695	0.1696	0.6872076
Residuals	13	819970	63075		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.803
 DIPROVE_2: -0.607
 DIPROVE_3: n.s

BROLIO 2009M IASMA

PvinacciaA

Analysis of Variance Table

Response: Br_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	3557332	3557332	46.5467	1.22e-05	***
IASMA_2	1	217583	217583	2.8470	0.115376	
IASMA_3	1	243049	243049	3.1802	0.097885	.
N_Gemme	1	50952	50952	0.6667	0.428918	
Defogl	1	73512	73512	0.9619	0.344614	
Dirad	1	1165285	1165285	15.2474	0.001810	**
Vigoria	2	982488	491244	6.4278	0.011456	*
N_Gemme:Defogl	1	8159	8159	0.1068	0.749064	
N_Gemme:Dirad	1	1266	1266	0.0166	0.899546	
Defogl:Dirad	1	22746	22746	0.2976	0.594610	
N_Gemme:Vigoria	2	28180	14090	0.1844	0.833769	
Defogl:Vigoria	2	335152	167576	2.1927	0.151160	
Dirad:Vigoria	2	74092	37046	0.4847	0.626558	
N_Gemme:Defogl:Dirad	1	214057	214057	2.8009	0.118090	
Residuals	13	993526	76425			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.884
 IASMA_2: n.s
 IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Br_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00052993	0.00052993	19.4278	0.0007076	***
IASMA_2	1	0.00000158	0.00000158	0.0578	0.8136915	
IASMA_3	1	0.00004304	0.00004304	1.5777	0.2311951	
N_Gemme	1	0.00000017	0.00000017	0.0064	0.9375731	
Defogl	1	0.00003754	0.00003754	1.3761	0.2618135	
Dirad	1	0.00006212	0.00006212	2.2773	0.1551966	
Vigoria	2	0.00004697	0.00002348	0.8610	0.4455181	
N_Gemme:Defogl	1	0.00007263	0.00007263	2.6628	0.1266970	
N_Gemme:Dirad	1	0.00000212	0.00000212	0.0777	0.7847609	
Defogl:Dirad	1	0.00002096	0.00002096	0.7683	0.3966576	
N_Gemme:Vigoria	2	0.00002008	0.00001004	0.3681	0.6990577	
Defogl:Vigoria	2	0.00022660	0.00011330	4.1536	0.0402908	*
Dirad:Vigoria	2	0.00010440	0.00005220	1.9137	0.1868682	

N_Gemme:Defogl:Dirad 1 0.00000107 0.00000107 0.0394 0.8458210
 Residuals 13 0.00035460 0.00002728

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.774
 IASMA_2: n.s
 IASMA_3: n.s

BrixA

Analysis of Variance Table

Response: Br_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	23.8296	23.8296	19.5648	0.0006876	***
IASMA_2	1	0.0900	0.0900	0.0739	0.7900349	
IASMA_3	1	2.0468	2.0468	1.6805	0.2174008	
N_Gemme	1	0.0027	0.0027	0.0022	0.9632958	
Defogl	1	1.8125	1.8125	1.4881	0.2441780	
Dirad	1	2.7736	2.7736	2.2772	0.1552063	
Vigoria	2	1.9123	0.9561	0.7850	0.4765836	
N_Gemme:Defogl	1	3.4721	3.4721	2.8507	0.1151643	
N_Gemme:Dirad	1	0.1138	0.1138	0.0934	0.7646947	
Defogl:Dirad	1	0.9837	0.9837	0.8076	0.3851666	
N_Gemme:Vigoria	2	1.0446	0.5223	0.4288	0.6601598	
Defogl:Vigoria	2	10.3789	5.1895	4.2607	0.0377551	*
Dirad:Vigoria	2	5.1506	2.5753	2.1144	0.1603180	
N_Gemme:Defogl:Dirad	1	0.0482	0.0482	0.0396	0.8453858	
Residuals	13	15.8338	1.2180			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.775
 IASMA_2: n.s
 IASMA_3: n.s

PhA

Analysis of Variance Table

Response: Br_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.034053	0.034053	3.7178	0.07596	.
IASMA_2	1	0.038123	0.038123	4.1621	0.06219	.
IASMA_3	1	0.001255	0.001255	0.1370	0.71728	
N_Gemme	1	0.000789	0.000789	0.0862	0.77373	
Defogl	1	0.005688	0.005688	0.6209	0.44483	
Dirad	1	0.011335	0.011335	1.2375	0.28611	
Vigoria	2	0.065091	0.032545	3.5532	0.05874	.
N_Gemme:Defogl	1	0.000934	0.000934	0.1019	0.75458	
N_Gemme:Dirad	1	0.003790	0.003790	0.4138	0.53125	
Defogl:Dirad	1	0.000223	0.000223	0.0244	0.87836	
N_Gemme:Vigoria	2	0.000615	0.000307	0.0336	0.96708	
Defogl:Vigoria	2	0.060088	0.030044	3.2801	0.07026	.
Dirad:Vigoria	2	0.006344	0.003172	0.3463	0.71362	
N_Gemme:Defogl:Dirad	1	0.008811	0.008811	0.9620	0.34459	
Residuals	13	0.119074	0.009160			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Br_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.2389	0.23891	0.5660	0.46525
IASMA_2	1	2.7492	2.74917	6.5131	0.02410 *
IASMA_3	1	0.0000	0.00003	0.0001	0.99353
N_Gemme	1	0.0306	0.03057	0.0724	0.79206
Defogl	1	0.7370	0.73698	1.7460	0.20917
Dirad	1	1.0037	1.00369	2.3779	0.14705
Vigoria	2	3.9875	1.99375	4.7234	0.02872 *
N_Gemme:Defogl	1	0.0006	0.00060	0.0014	0.97057
N_Gemme:Dirad	1	0.0907	0.09066	0.2148	0.65071
Defogl:Dirad	1	0.0369	0.03693	0.0875	0.77206
N_Gemme:Vigoria	2	0.0285	0.01427	0.0338	0.96684
Defogl:Vigoria	2	3.4608	1.73038	4.0994	0.04165 *
Dirad:Vigoria	2	0.2738	0.13690	0.3243	0.72870
N_Gemme:Defogl:Dirad	1	0.1154	0.11544	0.2735	0.60979
Residuals	13	5.4873	0.42210		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.578
IASMA_3: n.s

AcTarta

Analysis of Variance Table

Response: Br_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.5708	0.5708	1.4618	0.248178
IASMA_2	1	3.7112	3.7112	9.5049	0.008728 **
IASMA_3	1	0.2428	0.2428	0.6219	0.444490
N_Gemme	1	0.0042	0.0042	0.0108	0.918807
Defogl	1	0.5327	0.5327	1.3644	0.263753
Dirad	1	0.5412	0.5412	1.3860	0.260182
Vigoria	2	4.2646	2.1323	5.4611	0.018985 *
N_Gemme:Defogl	1	0.0112	0.0112	0.0287	0.868074
N_Gemme:Dirad	1	0.0818	0.0818	0.2096	0.654651
Defogl:Dirad	1	0.0425	0.0425	0.1088	0.746771
N_Gemme:Vigoria	2	0.0056	0.0028	0.0072	0.992867
Defogl:Vigoria	2	2.9777	1.4888	3.8131	0.049763 *
Dirad:Vigoria	2	0.4583	0.2291	0.5868	0.570156
N_Gemme:Defogl:Dirad	1	0.0191	0.0191	0.0490	0.828323
Residuals	13	5.0759	0.3905		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: 0.65
IASMA_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Br_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.62503	0.62503	12.0735	0.004109	**
IASMA_2	1	0.16380	0.16380	3.1641	0.098652	.
IASMA_3	1	0.01094	0.01094	0.2114	0.653261	
N_Gemme	1	0.11517	0.11517	2.2247	0.159686	
Defogl	1	0.20554	0.20554	3.9704	0.067732	.
Dirad	1	0.04128	0.04128	0.7974	0.388105	
Vigoria	2	0.52922	0.26461	5.1114	0.023025	*
N_Gemme:Defogl	1	0.12998	0.12998	2.5107	0.137087	
N_Gemme:Dirad	1	0.07001	0.07001	1.3524	0.265754	
Defogl:Dirad	1	0.01882	0.01882	0.3634	0.556971	
N_Gemme:Vigoria	2	0.04755	0.02377	0.4592	0.641629	
Defogl:Vigoria	2	0.02014	0.01007	0.1945	0.825584	
Dirad:Vigoria	2	0.05220	0.02610	0.5042	0.615356	
N_Gemme:Defogl:Dirad	1	0.00576	0.00576	0.1112	0.744104	
Residuals	13	0.67300	0.05177			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.694
IASMA_2: n.s
IASMA_3: n.s

KA

Analysis of Variance Table

Response: Br_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.036473	0.036473	2.8284	0.1165	
IASMA_2	1	0.003238	0.003238	0.2511	0.6247	
IASMA_3	1	0.000562	0.000562	0.0436	0.8378	
N_Gemme	1	0.016689	0.016689	1.2942	0.2758	
Defogl	1	0.038172	0.038172	2.9602	0.1090	
Dirad	1	0.000446	0.000446	0.0346	0.8553	
Vigoria	2	0.065374	0.032687	2.5348	0.1176	
N_Gemme:Defogl	1	0.000141	0.000141	0.0109	0.9184	
N_Gemme:Dirad	1	0.007822	0.007822	0.6066	0.4500	
Defogl:Dirad	1	0.001546	0.001546	0.1199	0.7347	
N_Gemme:Vigoria	2	0.006439	0.003220	0.2497	0.7827	
Defogl:Vigoria	2	0.012503	0.006252	0.4848	0.6265	
Dirad:Vigoria	2	0.035597	0.017799	1.3803	0.2860	
N_Gemme:Defogl:Dirad	1	0.000012	0.000012	0.0009	0.9764	
Residuals	13	0.167637	0.012895			

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

ApaA

Analysis of Variance Table

Response: Br_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	1947.9	1947.87	3.2514	0.09458 .
IASMA_2	1	0.9	0.85	0.0014	0.97047
IASMA_3	1	1353.8	1353.82	2.2598	0.15667
N_Gemme	1	686.5	686.47	1.1459	0.30389
Defogl	1	212.1	212.06	0.3540	0.56208
Dirad	1	33.6	33.59	0.0561	0.81650
Vigoria	2	491.7	245.84	0.4104	0.67171
N_Gemme:Defogl	1	299.6	299.63	0.5001	0.49192
N_Gemme:Dirad	1	1.9	1.89	0.0031	0.95610
Defogl:Dirad	1	358.9	358.95	0.5992	0.45275
N_Gemme:Vigoria	2	304.9	152.43	0.2544	0.77913
Defogl:Vigoria	2	1357.9	678.94	1.1333	0.35181
Dirad:Vigoria	2	1281.9	640.93	1.0699	0.37142
N_Gemme:Defogl:Dirad	1	89.8	89.80	0.1499	0.70489
Residuals	13	7788.1	599.08		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Br_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	289638	289638	60.7590	2.969e-06 ***
IASMA_2	1	8115	8115	1.7024	0.21460
IASMA_3	1	7242	7242	1.5191	0.23958
N_Gemme	1	2504	2504	0.5253	0.48144
Defogl	1	1225	1225	0.2570	0.62070
Dirad	1	42970	42970	9.0140	0.01019 *
Vigoria	2	14028	7014	1.4714	0.26544
N_Gemme:Defogl	1	5233	5233	1.0978	0.31384
N_Gemme:Dirad	1	17768	17768	3.7273	0.07563 .
Defogl:Dirad	1	1240	1240	0.2600	0.61863
N_Gemme:Vigoria	2	4389	2195	0.4604	0.64096
Defogl:Vigoria	2	11602	5801	1.2169	0.32777
Dirad:Vigoria	2	1957	978	0.2052	0.81705
N_Gemme:Defogl:Dirad	1	458	458	0.0960	0.76157
Residuals	13	61971	4767		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.908
 IASMA_2: n.s
 IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Br_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	725696	725696	68.3749	1.554e-06	***
IASMA_2	1	10269	10269	0.9675	0.34326	
IASMA_3	1	995	995	0.0938	0.76431	
N_Gemme	1	24414	24414	2.3003	0.15328	
Defogl	1	3408	3408	0.3211	0.58058	
Dirad	1	17008	17008	1.6025	0.22776	
Vigoria	2	79836	39918	3.7610	0.05143	.
N_Gemme:Defogl	1	9725	9725	0.9163	0.35591	
N_Gemme:Dirad	1	1404	1404	0.1323	0.72194	
Defogl:Dirad	1	2077	2077	0.1957	0.66545	
N_Gemme:Vigoria	2	5193	2596	0.2446	0.78651	
Defogl:Vigoria	2	20091	10045	0.9465	0.41329	
Dirad:Vigoria	2	5217	2609	0.2458	0.78564	
N_Gemme:Defogl:Dirad	1	6857	6857	0.6460	0.43599	
Residuals	13	137975	10613			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.917

IASMA_2: n.s

IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Br_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1171159	1171159	38.2158	3.314e-05	***
IASMA_2	1	6624	6624	0.2162	0.6497	
IASMA_3	1	7758	7758	0.2532	0.6233	
N_Gemme	1	1463	1463	0.0478	0.8304	
Defogl	1	2234	2234	0.0729	0.7914	
Dirad	1	74004	74004	2.4148	0.1442	
Vigoria	2	33055	16528	0.5393	0.5957	
N_Gemme:Defogl	1	227	227	0.0074	0.9327	
N_Gemme:Dirad	1	83365	83365	2.7202	0.1230	
Defogl:Dirad	1	13022	13022	0.4249	0.5259	
N_Gemme:Vigoria	2	51879	25939	0.8464	0.4513	
Defogl:Vigoria	2	56670	28335	0.9246	0.4213	
Dirad:Vigoria	2	32510	16255	0.5304	0.6006	
N_Gemme:Defogl:Dirad	1	37698	37698	1.2301	0.2875	
Residuals	13	398397	30646			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.864

IASMA_2: n.s

IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Br_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	2746094	2746094	52.9987	6.181e-06	***

IASMA_2	1	1277	1277	0.0247	0.8776
IASMA_3	1	18894	18894	0.3647	0.5563
N_Gemme	1	54693	54693	1.0556	0.3230
Defogl	1	154778	154778	2.9872	0.1076
Dirad	1	4235	4235	0.0817	0.7795
Vigoria	2	151412	75706	1.4611	0.2677
N_Gemme:Defogl	1	16150	16150	0.3117	0.5861
N_Gemme:Dirad	1	2584	2584	0.0499	0.8268
Defogl:Dirad	1	5794	5794	0.1118	0.7434
N_Gemme:Vigoria	2	14400	7200	0.1390	0.8715
Defogl:Vigoria	2	235755	117878	2.2750	0.1422
Dirad:Vigoria	2	122693	61347	1.1840	0.3370
N_Gemme:Defogl:Dirad	1	5277	5277	0.1018	0.7547
Residuals	13	673587	51814		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.896
IASMA_2: n.s
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Br_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	5033905	5033905	38.6505	3.133e-05	***
IASMA_2	1	626260	626260	4.8084	0.0471147	*
IASMA_3	1	223054	223054	1.7126	0.2133096	
N_Gemme	1	100570	100570	0.7722	0.3954949	
Defogl	1	73	73	0.0006	0.9814355	
Dirad	1	3226760	3226760	24.7752	0.0002528	***
Vigoria	2	769645	384822	2.9547	0.0875453	.
N_Gemme:Defogl	1	2329	2329	0.0179	0.8956612	
N_Gemme:Dirad	1	208105	208105	1.5978	0.2284057	
Defogl:Dirad	1	61326	61326	0.4709	0.5046457	
N_Gemme:Vigoria	2	255980	127990	0.9827	0.4004545	
Defogl:Vigoria	2	481189	240595	1.8473	0.1967388	
Dirad:Vigoria	2	3032263	1516131	11.6409	0.0012666	**
N_Gemme:Defogl:Dirad	1	70169	70169	0.5388	0.4759775	
Residuals	13	1693142	130242			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.865
IASMA_2: -0.52
IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Br_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00031299	3.1299e-04	9.6325	0.008388	**
IASMA_2	1	0.00001235	1.2349e-05	0.3800	0.548229	
IASMA_3	1	0.00016336	1.6336e-04	5.0274	0.043023	*
N_Gemme	1	0.00002954	2.9540e-05	0.9091	0.357745	

Defogl	1	0.00002766	2.7664e-05	0.8514	0.372968
Dirad	1	0.00005652	5.6520e-05	1.7395	0.209969
Vigoria	2	0.00008793	4.3967e-05	1.3531	0.292527
N_Gemme:Defogl	1	0.00002680	2.6802e-05	0.8248	0.380295
N_Gemme:Dirad	1	0.00000232	2.3190e-06	0.0714	0.793548
Defogl:Dirad	1	0.00000101	1.0080e-06	0.0310	0.862883
N_Gemme:Vigoria	2	0.00004375	2.1875e-05	0.6732	0.526976
Defogl:Vigoria	2	0.00007473	3.7367e-05	1.1500	0.346845
Dirad:Vigoria	2	0.00011428	5.7142e-05	1.7586	0.210888
N_Gemme:Defogl:Dirad	1	0.00000542	5.4230e-06	0.1669	0.689539
Residuals	13	0.00042241	3.2493e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.652
IASMA_2: n.s
IASMA_3: 0.528

BrixB

Analysis of Variance Table

Response: Br_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	14.3821	14.3821	9.5804	0.008525	**
IASMA_2	1	0.6558	0.6558	0.4369	0.520178	
IASMA_3	1	7.6328	7.6328	5.0845	0.042027	*
N_Gemme	1	1.2809	1.2809	0.8532	0.372468	
Defogl	1	1.2756	1.2756	0.8497	0.373417	
Dirad	1	2.5994	2.5994	1.7315	0.210949	
Vigoria	2	3.4396	1.7198	1.1456	0.348146	
N_Gemme:Defogl	1	1.1569	1.1569	0.7707	0.395949	
N_Gemme:Dirad	1	0.1173	0.1173	0.0781	0.784244	
Defogl:Dirad	1	0.0905	0.0905	0.0603	0.809924	
N_Gemme:Vigoria	2	2.0025	1.0012	0.6670	0.529978	
Defogl:Vigoria	2	3.5281	1.7640	1.1751	0.339545	
Dirad:Vigoria	2	5.4696	2.7348	1.8217	0.200700	
N_Gemme:Defogl:Dirad	1	0.2314	0.2314	0.1541	0.700989	
Residuals	13	19.5157	1.5012			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.651
IASMA_2: n.s
IASMA_3: 0.53

PhB

Analysis of Variance Table

Response: Br_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.027160	0.027160	4.2992	0.058562	.
IASMA_2	1	0.056144	0.056144	8.8871	0.010617	*
IASMA_3	1	0.000662	0.000662	0.1048	0.751278	
N_Gemme	1	0.000081	0.000081	0.0128	0.911536	
Defogl	1	0.002208	0.002208	0.3496	0.564487	
Dirad	1	0.035770	0.035770	5.6621	0.033337	*
Vigoria	2	0.099118	0.049559	7.8448	0.005827	**

N_Gemme:Defogl	1	0.006566	0.006566	1.0393	0.326584
N_Gemme:Dirad	1	0.004026	0.004026	0.6373	0.439046
Defogl:Dirad	1	0.000097	0.000097	0.0154	0.903285
N_Gemme:Vigoria	2	0.012766	0.006383	1.0104	0.390969
Defogl:Vigoria	2	0.012451	0.006225	0.9854	0.399517
Dirad:Vigoria	2	0.050079	0.025039	3.9636	0.045292 *
N_Gemme:Defogl:Dirad	1	0.001428	0.001428	0.2261	0.642330
Residuals	13	0.082127	0.006317		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.637
IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Br_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.1910	0.19103	0.7290	0.408661
IASMA_2	1	2.5024	2.50242	9.5498	0.008606 **
IASMA_3	1	0.1901	0.19010	0.7255	0.409768
N_Gemme	1	0.0289	0.02894	0.1104	0.744934
Defogl	1	0.0011	0.00107	0.0041	0.949931
Dirad	1	1.4256	1.42563	5.4405	0.036394 *
Vigoria	2	2.5250	1.26249	4.8180	0.027192 *
N_Gemme:Defogl	1	0.0505	0.05054	0.1929	0.667737
N_Gemme:Dirad	1	0.3147	0.31472	1.2011	0.292998
Defogl:Dirad	1	0.0021	0.00206	0.0079	0.930666
N_Gemme:Vigoria	2	0.2933	0.14664	0.5596	0.584610
Defogl:Vigoria	2	1.2956	0.64779	2.4721	0.123059
Dirad:Vigoria	2	0.4863	0.24315	0.9279	0.420055
N_Gemme:Defogl:Dirad	1	0.0929	0.09290	0.3545	0.561791
Residuals	13	3.4065	0.26204		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.651
IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Br_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.6868	0.6868	2.6403	0.128169
IASMA_2	1	4.0644	4.0644	15.6251	0.001653 **
IASMA_3	1	0.0319	0.0319	0.1225	0.731893
N_Gemme	1	0.0029	0.0029	0.0113	0.916995
Defogl	1	0.0075	0.0075	0.0289	0.867662
Dirad	1	1.4708	1.4708	5.6543	0.033440 *
Vigoria	2	2.8472	1.4236	5.4728	0.018865 *
N_Gemme:Defogl	1	0.1093	0.1093	0.4203	0.528094
N_Gemme:Dirad	1	0.3436	0.3436	1.3210	0.271127
Defogl:Dirad	1	0.0331	0.0331	0.1271	0.727136

N_Gemme:Vigoria	2	0.1296	0.0648	0.2491	0.783163
Defogl:Vigoria	2	0.8849	0.4425	1.7011	0.220691
Dirad:Vigoria	2	0.5355	0.2678	1.0294	0.384602
N_Gemme:Defogl:Dirad	1	0.0240	0.0240	0.0921	0.766323
Residuals	13	3.3815	0.2601		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.739
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Br_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1.73227	1.73227	63.5129	2.332e-06	***
IASMA_2	1	0.59465	0.59465	21.8028	0.0004389	***
IASMA_3	1	0.01091	0.01091	0.4000	0.5380770	
N_Gemme	1	0.00705	0.00705	0.2586	0.6195791	
Defogl	1	0.01756	0.01756	0.6437	0.4368006	
Dirad	1	0.03836	0.03836	1.4065	0.2568576	
Vigoria	2	0.38515	0.19258	7.0607	0.0083967	**
N_Gemme:Defogl	1	0.00201	0.00201	0.0735	0.7905105	
N_Gemme:Dirad	1	0.02423	0.02423	0.8883	0.3631326	
Defogl:Dirad	1	0.02803	0.02803	1.0276	0.3292296	
N_Gemme:Vigoria	2	0.02539	0.01270	0.4655	0.6379008	
Defogl:Vigoria	2	0.05101	0.02551	0.9352	0.4173957	
Dirad:Vigoria	2	0.12022	0.06011	2.2040	0.1498925	
N_Gemme:Defogl:Dirad	1	0.02213	0.02213	0.8113	0.3841276	
Residuals	13	0.35457	0.02727			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.911
IASMA_2: 0.791
IASMA_3: n.s

KB

Analysis of Variance Table

Response: Br_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.080390	0.080390	7.5670	0.01651	*
IASMA_2	1	0.071591	0.071591	6.7388	0.02218	*
IASMA_3	1	0.002231	0.002231	0.2100	0.65436	
N_Gemme	1	0.001322	0.001322	0.1245	0.72990	
Defogl	1	0.009714	0.009714	0.9144	0.35640	
Dirad	1	0.020927	0.020927	1.9698	0.18390	
Vigoria	2	0.140080	0.070040	6.5928	0.01055	*
N_Gemme:Defogl	1	0.020995	0.020995	1.9762	0.18324	
N_Gemme:Dirad	1	0.000170	0.000170	0.0160	0.90118	
Defogl:Dirad	1	0.009549	0.009549	0.8989	0.36037	
N_Gemme:Vigoria	2	0.005097	0.002549	0.2399	0.79012	
Defogl:Vigoria	2	0.009561	0.004780	0.4500	0.64722	
Dirad:Vigoria	2	0.132538	0.066269	6.2378	0.01261	*

N_Gemme:Defogl:Dirad 1 0.004492 0.004492 0.4229 0.52683
 Residuals 13 0.138109 0.010624

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.607
 IASMA_2: 0.584
 IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Br_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	792.3	792.28	2.2126	0.160735
IASMA_2	1	0.2	0.24	0.0007	0.979649
IASMA_3	1	158.3	158.25	0.4420	0.517800
N_Gemme	1	163.6	163.57	0.4568	0.510971
Defogl	1	1356.0	1355.99	3.7869	0.073599 .
Dirad	1	0.4	0.39	0.0011	0.974255
Vigoria	2	958.2	479.11	1.3380	0.296203
N_Gemme:Defogl	1	172.0	171.98	0.4803	0.500486
N_Gemme:Dirad	1	425.8	425.78	1.1891	0.295311
Defogl:Dirad	1	118.0	118.02	0.3296	0.575699
N_Gemme:Vigoria	2	19.1	9.55	0.0267	0.973739
Defogl:Vigoria	2	1265.4	632.68	1.7669	0.209512
Dirad:Vigoria	2	5674.2	2837.10	7.9233	0.005624 **
N_Gemme:Defogl:Dirad	1	691.8	691.83	1.9321	0.187871
Residuals	13	4654.9	358.07		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Br_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	64854	64854	16.7188	0.001280 **
IASMA_2	1	10114	10114	2.6073	0.130370
IASMA_3	1	78	78	0.0200	0.889595
N_Gemme	1	1818	1818	0.4687	0.505622
Defogl	1	3073	3073	0.7921	0.389634
Dirad	1	6290	6290	1.6214	0.225192
Vigoria	2	3047	1524	0.3928	0.682909
N_Gemme:Defogl	1	4970	4970	1.2811	0.278134
N_Gemme:Dirad	1	1143	1143	0.2946	0.596481
Defogl:Dirad	1	5928	5928	1.5282	0.238260
N_Gemme:Vigoria	2	2624	1312	0.3383	0.719104
Defogl:Vigoria	2	191	96	0.0246	0.975698
Dirad:Vigoria	2	15484	7742	1.9959	0.175424
N_Gemme:Defogl:Dirad	1	6130	6130	1.5803	0.230833
Residuals	13	50428	3879		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.75
IASMA_2: n.s
IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Br_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	242221	242221	32.8478	6.922e-05 ***
IASMA_2	1	11193	11193	1.5179	0.23976
IASMA_3	1	2407	2407	0.3264	0.57753
N_Gemme	1	2003	2003	0.2716	0.61102
Defogl	1	23953	23953	3.2482	0.09472 .
Dirad	1	15506	15506	2.1028	0.17072
Vigoria	2	15787	7893	1.0704	0.37123
N_Gemme:Defogl	1	1711	1711	0.2320	0.63803
N_Gemme:Dirad	1	27232	27232	3.6929	0.07684 .
Defogl:Dirad	1	2034	2034	0.2758	0.60830
N_Gemme:Vigoria	2	4268	2134	0.2894	0.75343
Defogl:Vigoria	2	15893	7947	1.0776	0.36895
Dirad:Vigoria	2	47406	23703	3.2144	0.07341 .
N_Gemme:Defogl:Dirad	1	2433	2433	0.3300	0.57546
Residuals	13	95863	7374		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.846
IASMA_2: n.s
IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Br_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	442862	442862	11.5857	0.004708 **
IASMA_2	1	68040	68040	1.7800	0.205052
IASMA_3	1	11318	11318	0.2961	0.595555
N_Gemme	1	2104	2104	0.0550	0.818175
Defogl	1	7661	7661	0.2004	0.661749
Dirad	1	7588	7588	0.1985	0.663262
Vigoria	2	51259	25629	0.6705	0.528286
N_Gemme:Defogl	1	3575	3575	0.0935	0.764596
N_Gemme:Dirad	1	5821	5821	0.1523	0.702693
Defogl:Dirad	1	10653	10653	0.2787	0.606450
N_Gemme:Vigoria	2	2149	1075	0.0281	0.972335
Defogl:Vigoria	2	34363	17181	0.4495	0.647511
Dirad:Vigoria	2	17534	8767	0.2294	0.798195
N_Gemme:Defogl:Dirad	1	8	8	0.0002	0.988978
Residuals	13	496924	38225		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.686

IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Br_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1676023	1676023	23.4604	0.0003208	***
IASMA_2	1	83061	83061	1.1627	0.3005149	
IASMA_3	1	136536	136536	1.9112	0.1901218	
N_Gemme	1	36517	36517	0.5111	0.4872833	
Defogl	1	92093	92093	1.2891	0.2767164	
Dirad	1	4853	4853	0.0679	0.7984531	
Vigoria	2	280523	140261	1.9633	0.1798536	
N_Gemme:Defogl	1	21583	21583	0.3021	0.5918813	
N_Gemme:Dirad	1	129731	129731	1.8159	0.2008162	
Defogl:Dirad	1	100	100	0.0014	0.9707291	
N_Gemme:Vigoria	2	1886	943	0.0132	0.9869017	
Defogl:Vigoria	2	21165	10582	0.1481	0.8637544	
Dirad:Vigoria	2	35251	17625	0.2467	0.7849405	
N_Gemme:Defogl:Dirad	1	65557	65557	0.9177	0.3555701	
Residuals	13	928727	71441			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.802
IASMA_2: n.s
IASMA_3: n.s

CACCIAGRANDE 2008M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Ca_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	1013525	1013525	23.1147	0.0029770	**
IBIMET_3	1	120197	120197	2.7413	0.1488675	
N_Gemme	1	21903	21903	0.4995	0.5062147	
Defogl	1	1472499	1472499	33.5822	0.0011566	**
Dirad	1	1731523	1731523	39.4896	0.0007556	***
Vigoria	2	14260	7130	0.1626	0.8535439	
N_Gemme:Defogl	1	31370	31370	0.7154	0.4300778	
N_Gemme:Dirad	1	485571	485571	11.0741	0.0158505	*
Defogl:Dirad	1	321157	321157	7.3244	0.0352763	*
N_Gemme:Vigoria	2	307262	153631	3.5038	0.0981455	.
Defogl:Vigoria	2	13001	6500	0.1482	0.8652796	
Dirad:Vigoria	2	47492	23746	0.5416	0.6078258	
N_Gemme:Defogl:Dirad	1	21228	21228	0.4841	0.5125939	
Residuals	6	263085	43848			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.891
IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Ca_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	6.3707e-05	6.3707e-05	30.4282	0.0014922	**
IBIMET_3	1	2.6777e-05	2.6777e-05	12.7894	0.0116961	*
N_Gemme	1	1.1562e-05	1.1562e-05	5.5222	0.0570627	.
Defogl	1	1.2094e-04	1.2094e-04	57.7639	0.0002701	***
Dirad	1	1.0180e-04	1.0180e-04	48.6241	0.0004324	***
Viguria	2	1.5838e-05	7.9190e-06	3.7823	0.0865437	.
N_Gemme:Defogl	1	1.7730e-06	1.7730e-06	0.8470	0.3929091	
N_Gemme:Dirad	1	1.9018e-05	1.9018e-05	9.0837	0.0235812	*
Defogl:Dirad	1	1.6700e-07	1.6700e-07	0.0796	0.7872647	
N_Gemme:Viguria	2	1.3333e-05	6.6660e-06	3.1841	0.1141672	
Defogl:Viguria	2	5.3970e-06	2.6980e-06	1.2888	0.3422578	
Dirad:Viguria	2	6.6490e-06	3.3250e-06	1.5879	0.2795880	
N_Gemme:Defogl:Dirad	1	2.1200e-07	2.1200e-07	0.1014	0.7609694	
Residuals	6	1.2562e-05	2.0940e-06			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.914

IBIMET_3: -0.825

BrixA

Analysis of Variance Table

Response: Ca_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	3.0559	3.0559	27.0170	0.0020192	**
IBIMET_3	1	1.1491	1.1491	10.1588	0.0189011	*
N_Gemme	1	0.6440	0.6440	5.6935	0.0543135	.
Defogl	1	6.2159	6.2159	54.9540	0.0003098	***
Dirad	1	4.9070	4.9070	43.3822	0.0005881	***
Viguria	2	0.8308	0.4154	3.6726	0.0908842	.
N_Gemme:Defogl	1	0.0828	0.0828	0.7322	0.4250220	
N_Gemme:Dirad	1	0.9022	0.9022	7.9759	0.0301885	*
Defogl:Dirad	1	0.0422	0.0422	0.3731	0.5637148	
N_Gemme:Viguria	2	0.5778	0.2889	2.5543	0.1575707	
Defogl:Viguria	2	0.2998	0.1499	1.3253	0.3336776	
Dirad:Viguria	2	0.3796	0.1898	1.6781	0.2637239	
N_Gemme:Defogl:Dirad	1	0.0062	0.0062	0.0547	0.8228429	
Residuals	6	0.6787	0.1131			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.905

IBIMET_3: -0.793

PhA

Analysis of Variance Table

Response: Ca_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IBIMET_2	1	0.0117501	0.0117501	6.7492	0.04078	*
IBIMET_3	1	0.0000721	0.0000721	0.0414	0.84550	
N_Gemme	1	0.0024899	0.0024899	1.4302	0.27685	
Defogl	1	0.0216561	0.0216561	12.4390	0.01241	*
Dirad	1	0.0149147	0.0149147	8.5669	0.02639	*
Vigoria	2	0.0056562	0.0028281	1.6244	0.27302	
N_Gemme:Defogl	1	0.0004147	0.0004147	0.2382	0.64284	
N_Gemme:Dirad	1	0.0137654	0.0137654	7.9067	0.03068	*
Defogl:Dirad	1	0.0007927	0.0007927	0.4553	0.52494	
N_Gemme:Vigoria	2	0.0029568	0.0014784	0.8492	0.47344	
Defogl:Vigoria	2	0.0095953	0.0047977	2.7557	0.14160	
Dirad:Vigoria	2	0.0014902	0.0007451	0.4280	0.67026	
N_Gemme:Defogl:Dirad	1	0.0048959	0.0048959	2.8121	0.14457	
Residuals	6	0.0104458	0.0017410			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.728
IBIMET_3: n.s

AcTitola

Analysis of Variance Table

Response: Ca_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.15160	0.15160	3.2235	0.122730
IBIMET_3	1	0.75320	0.75320	16.0152	0.007104 **
N_Gemme	1	0.01939	0.01939	0.4123	0.544550
Defogl	1	0.33659	0.33659	7.1568	0.036770 *
Dirad	1	0.27648	0.27648	5.8787	0.051539 .
Vigoria	2	0.37449	0.18724	3.9814	0.079349 .
N_Gemme:Defogl	1	0.01853	0.01853	0.3941	0.553285
N_Gemme:Dirad	1	0.04660	0.04660	0.9908	0.357991
Defogl:Dirad	1	0.02683	0.02683	0.5704	0.478692
N_Gemme:Vigoria	2	0.02995	0.01497	0.3184	0.738895
Defogl:Vigoria	2	0.00119	0.00059	0.0126	0.987472
Dirad:Vigoria	2	0.00156	0.00078	0.0166	0.983576
N_Gemme:Defogl:Dirad	1	0.00633	0.00633	0.1345	0.726408
Residuals	6	0.28218	0.04703		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: 0.853

AcTartA

Analysis of Variance Table

Response: Ca_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.00108	0.00108	0.0269	0.875178
IBIMET_3	1	1.06698	1.06698	26.4646	0.002127 **
N_Gemme	1	0.01159	0.01159	0.2874	0.611175
Defogl	1	0.08135	0.08135	2.0178	0.205283
Dirad	1	0.20270	0.20270	5.0277	0.066143 .
Vigoria	2	0.95450	0.47725	11.8373	0.008266 **
N_Gemme:Defogl	1	0.10052	0.10052	2.4933	0.165409

N_Gemme:Dirad	1	0.00273	0.00273	0.0677	0.803377
Defogl:Dirad	1	0.00734	0.00734	0.1821	0.684433
N_Gemme:Vigoria	2	0.00702	0.00351	0.0871	0.917732
Defogl:Vigoria	2	0.00647	0.00324	0.0803	0.923851
Dirad:Vigoria	2	0.02654	0.01327	0.3291	0.731754
N_Gemme:Defogl:Dirad	1	0.00872	0.00872	0.2163	0.658305
Residuals	6	0.24190	0.04032		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: 0.903

AcMalicoA

Analysis of Variance Table

Response: Ca_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	1.28316	1.28316	184.5748	9.87e-06	***
IBIMET_3	1	0.05908	0.05908	8.4976	0.026803	*
N_Gemme	1	0.00200	0.00200	0.2873	0.611232	
Defogl	1	0.16106	0.16106	23.1679	0.002960	**
Dirad	1	0.08174	0.08174	11.7574	0.013990	*
Vigoria	2	0.26744	0.13372	19.2348	0.002456	**
N_Gemme:Defogl	1	0.00210	0.00210	0.3027	0.602069	
N_Gemme:Dirad	1	0.03918	0.03918	5.6362	0.055213	.
Defogl:Dirad	1	0.02045	0.02045	2.9414	0.137161	
N_Gemme:Vigoria	2	0.00849	0.00425	0.6109	0.573462	
Defogl:Vigoria	2	0.01028	0.00514	0.7392	0.516437	
Dirad:Vigoria	2	0.01027	0.00514	0.7389	0.516588	
N_Gemme:Defogl:Dirad	1	0.00593	0.00593	0.8525	0.391471	
Residuals	6	0.04171	0.00695			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.984
IBIMET_3: 0.766

KA

Analysis of Variance Table

Response: Ca_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.061770	0.061770	17.4583	0.005824	**
IBIMET_3	1	0.002593	0.002593	0.7328	0.424838	
N_Gemme	1	0.003440	0.003440	0.9723	0.362175	
Defogl	1	0.011641	0.011641	3.2902	0.119627	
Dirad	1	0.053092	0.053092	15.0055	0.008231	**
Vigoria	2	0.031079	0.015539	4.3919	0.066848	.
N_Gemme:Defogl	1	0.005080	0.005080	1.4358	0.276004	
N_Gemme:Dirad	1	0.004362	0.004362	1.2327	0.309384	
Defogl:Dirad	1	0.001749	0.001749	0.4945	0.508295	
N_Gemme:Vigoria	2	0.017199	0.008599	2.4304	0.168600	
Defogl:Vigoria	2	0.014166	0.007083	2.0019	0.215749	
Dirad:Vigoria	2	0.006346	0.003173	0.8968	0.456276	
N_Gemme:Defogl:Dirad	1	0.011428	0.011428	3.2300	0.122423	
Residuals	6	0.021229	0.003538			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.863

IBIMET_3: n.s

ApaA

Analysis of Variance Table

Response: Ca_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	1923.47	1923.47	12.0033	0.01339	*
IBIMET_3	1	1309.70	1309.70	8.1731	0.02884	*
N_Gemme	1	178.14	178.14	1.1117	0.33232	
Defogl	1	693.70	693.70	4.3290	0.08266	.
Dirad	1	17.89	17.89	0.1117	0.74964	
Vigoria	2	229.63	114.82	0.7165	0.52597	
N_Gemme:Defogl	1	75.19	75.19	0.4692	0.51892	
N_Gemme:Dirad	1	9.76	9.76	0.0609	0.81328	
Defogl:Dirad	1	6.50	6.50	0.0406	0.84706	
N_Gemme:Vigoria	2	124.79	62.40	0.3894	0.69344	
Defogl:Vigoria	2	97.12	48.56	0.3030	0.74924	
Dirad:Vigoria	2	9.62	4.81	0.0300	0.97057	
N_Gemme:Defogl:Dirad	1	81.51	81.51	0.5086	0.50251	
Residuals	6	961.47	160.25			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.817

IBIMET_3: 0.759

Antoctot1A

Analysis of Variance Table

Response: Ca_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	90042	90042	23.0851	0.002986	**
IBIMET_3	1	27511	27511	7.0533	0.037735	*
N_Gemme	1	3511	3511	0.9001	0.379381	
Defogl	1	4499	4499	1.1534	0.324123	
Dirad	1	15549	15549	3.9865	0.092861	.
Vigoria	2	52913	26456	6.7829	0.028838	*
N_Gemme:Defogl	1	196	196	0.0501	0.830242	
N_Gemme:Dirad	1	3253	3253	0.8341	0.396307	
Defogl:Dirad	1	14817	14817	3.7988	0.099189	.
N_Gemme:Vigoria	2	11103	5551	1.4233	0.311987	
Defogl:Vigoria	2	26914	13457	3.4501	0.100617	
Dirad:Vigoria	2	5255	2627	0.6736	0.544607	
N_Gemme:Defogl:Dirad	1	8311	8311	2.1309	0.194650	
Residuals	6	23403	3900			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.891

IBIMET_3: -0.735

Poliftot1A

Analysis of Variance Table

Response: Ca_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	175033	175033	9.8094	0.02027	*
IBIMET_3	1	56598	56598	3.1720	0.12520	
N_Gemme	1	13724	13724	0.7691	0.41422	
Defogl	1	141287	141287	7.9182	0.03060	*
Dirad	1	97	97	0.0054	0.94359	
Vigoria	2	321500	160750	9.0089	0.01559	*
N_Gemme:Defogl	1	2354	2354	0.1319	0.72888	
N_Gemme:Dirad	1	8325	8325	0.4666	0.52006	
Defogl:Dirad	1	79266	79266	4.4423	0.07963	.
N_Gemme:Vigoria	2	40282	20141	1.1288	0.38362	
Defogl:Vigoria	2	50911	25455	1.4266	0.31128	
Dirad:Vigoria	2	30281	15140	0.8485	0.47368	
N_Gemme:Defogl:Dirad	1	24806	24806	1.3902	0.28300	
Residuals	6	107060	17843			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.788

IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	72815	72815	26.7460	0.002071	**
IBIMET_3	1	5731	5731	2.1052	0.196994	
N_Gemme	1	4968	4968	1.8249	0.225442	
Defogl	1	29961	29961	11.0050	0.016057	*
Dirad	1	36141	36141	13.2752	0.010790	*
Vigoria	2	6911	3456	1.2693	0.346974	
N_Gemme:Defogl	1	2381	2381	0.8745	0.385810	
N_Gemme:Dirad	1	3771	3771	1.3850	0.283814	
Defogl:Dirad	1	20493	20493	7.5274	0.033575	*
N_Gemme:Vigoria	2	13885	6943	2.5501	0.157927	
Defogl:Vigoria	2	2314	1157	0.4250	0.672048	
Dirad:Vigoria	2	2762	1381	0.5072	0.625868	
N_Gemme:Defogl:Dirad	1	78	78	0.0287	0.871106	
Residuals	6	16335	2722			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.904

IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Ca_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	52397	52397	7.1442	0.036885	*

IBIMET_3	1	3126	3126	0.4262	0.538070
N_Gemme	1	17325	17325	2.3622	0.175215
Defogl	1	6140	6140	0.8372	0.395492
Dirad	1	6	6	0.0007	0.979041
Vigoria	2	128504	64252	8.7606	0.016599 *
N_Gemme:Defogl	1	13	13	0.0018	0.967293
N_Gemme:Dirad	1	783	783	0.1068	0.754941
Defogl:Dirad	1	123394	123394	16.8244	0.006345 **
N_Gemme:Vigoria	2	48352	24176	3.2964	0.108167
Defogl:Vigoria	2	264	132	0.0180	0.982232
Dirad:Vigoria	2	38814	19407	2.6461	0.150011
N_Gemme:Defogl:Dirad	1	57	57	0.0078	0.932424
Residuals	6	44005	7334		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.737

IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	478154	478154	3.0282	0.13248
IBIMET_3	1	27219	27219	0.1724	0.69244
N_Gemme	1	87383	87383	0.5534	0.48503
Defogl	1	358079	358079	2.2678	0.18280
Dirad	1	1176023	1176023	7.4479	0.03423 *
Vigoria	2	226990	113495	0.7188	0.52500
N_Gemme:Defogl	1	29653	29653	0.1878	0.67991
N_Gemme:Dirad	1	762	762	0.0048	0.94688
Defogl:Dirad	1	87086	87086	0.5515	0.48574
N_Gemme:Vigoria	2	161517	80758	0.5115	0.62360
Defogl:Vigoria	2	169684	84842	0.5373	0.61002
Dirad:Vigoria	2	260708	130354	0.8255	0.48226
N_Gemme:Defogl:Dirad	1	12052	12052	0.0763	0.79161
Residuals	6	947402	157900		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1.1062e-04	1.1062e-04	4.6645	0.07410 .
IBIMET_3	1	1.5261e-05	1.5261e-05	0.6435	0.45303
N_Gemme	1	1.8477e-05	1.8477e-05	0.7791	0.41137
Defogl	1	6.1144e-05	6.1144e-05	2.5783	0.15946
Dirad	1	1.5260e-04	1.5260e-04	6.4351	0.04427 *
Vigoria	2	2.9860e-06	1.4930e-06	0.0629	0.93960
N_Gemme:Defogl	1	2.0840e-06	2.0840e-06	0.0879	0.77688
N_Gemme:Dirad	1	4.9657e-05	4.9657e-05	2.0939	0.19803

Defogl:Dirad	1	4.3500e-07	4.3500e-07	0.0183	0.89671
N_Gemme:Vigoria	2	1.1408e-05	5.7040e-06	0.2405	0.79345
Defogl:Vigoria	2	1.5022e-05	7.5110e-06	0.3167	0.74000
Dirad:Vigoria	2	7.9400e-07	3.9700e-07	0.0167	0.98345
N_Gemme:Defogl:Dirad	1	8.4000e-08	8.4000e-08	0.0036	0.95441
Residuals	6	1.4229e-04	2.3715e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	5.5252	5.5252	4.7604	0.07188 .
IBIMET_3	1	0.7480	0.7480	0.6445	0.45272
N_Gemme	1	0.8009	0.8009	0.6901	0.43794
Defogl	1	3.0925	3.0925	2.6644	0.15374
Dirad	1	7.4476	7.4476	6.4167	0.04449 *
Vigoria	2	0.2109	0.1054	0.0908	0.91439
N_Gemme:Defogl	1	0.0812	0.0812	0.0700	0.80019
N_Gemme:Dirad	1	2.4504	2.4504	2.1112	0.19644
Defogl:Dirad	1	0.0029	0.0029	0.0025	0.96179
N_Gemme:Vigoria	2	0.5546	0.2773	0.2389	0.79464
Defogl:Vigoria	2	0.8160	0.4080	0.3515	0.71721
Dirad:Vigoria	2	0.0644	0.0322	0.0277	0.97277
N_Gemme:Defogl:Dirad	1	0.0155	0.0155	0.0134	0.91173
Residuals	6	6.9640	1.1607		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.002705	0.002705	0.6632	0.44653
IBIMET_3	1	0.000019	0.000019	0.0047	0.94734
N_Gemme	1	0.005703	0.005703	1.3982	0.28175
Defogl	1	0.029096	0.029096	7.1341	0.03698 *
Dirad	1	0.034765	0.034765	8.5240	0.02665 *
Vigoria	2	0.011849	0.005925	1.4526	0.30586
N_Gemme:Defogl	1	0.002303	0.002303	0.5646	0.48085
N_Gemme:Dirad	1	0.009632	0.009632	2.3617	0.17526
Defogl:Dirad	1	0.000343	0.000343	0.0841	0.78155
N_Gemme:Vigoria	2	0.001652	0.000826	0.2025	0.82204
Defogl:Vigoria	2	0.034371	0.017186	4.2137	0.07193 .
Dirad:Vigoria	2	0.003242	0.001621	0.3975	0.68848
N_Gemme:Defogl:Dirad	1	0.006872	0.006872	1.6848	0.24194
Residuals	6	0.024471	0.004079		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

AcTitolB

Analysis of Variance Table

Response: Ca_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.34371	0.34371	42.1575	0.0006349	***
IBIMET_3	1	0.17650	0.17650	21.6479	0.0034937	**
N_Gemme	1	0.12276	0.12276	15.0571	0.0081674	**
Defogl	1	0.04417	0.04417	5.4170	0.0588454	.
Dirad	1	0.02143	0.02143	2.6284	0.1560925	
Vigoria	2	0.00129	0.00065	0.0792	0.9247657	
N_Gemme:Defogl	1	0.13612	0.13612	16.6953	0.0064583	**
N_Gemme:Dirad	1	0.01726	0.01726	2.1172	0.1958943	
Defogl:Dirad	1	0.02663	0.02663	3.2666	0.1207127	
N_Gemme:Vigoria	2	0.04122	0.02061	2.5279	0.1598379	
Defogl:Vigoria	2	0.03031	0.01515	1.8587	0.2354027	
Dirad:Vigoria	2	0.06173	0.03087	3.7858	0.0864100	.
N_Gemme:Defogl:Dirad	1	0.00254	0.00254	0.3121	0.5966181	
Residuals	6	0.04892	0.00815			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.936

IBIMET_3: 0.885

AcTartB

Analysis of Variance Table

Response: Ca_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.110182	0.110182	25.6540	0.0022994	**
IBIMET_3	1	0.301445	0.301445	70.1867	0.0001574	***
N_Gemme	1	0.002367	0.002367	0.5510	0.4859277	
Defogl	1	0.002552	0.002552	0.5941	0.4700784	
Dirad	1	0.000729	0.000729	0.1697	0.6946765	
Vigoria	2	0.011685	0.005842	1.3603	0.3256915	
N_Gemme:Defogl	1	0.259065	0.259065	60.3191	0.0002397	***
N_Gemme:Dirad	1	0.002179	0.002179	0.5074	0.5030078	
Defogl:Dirad	1	0.031338	0.031338	7.2965	0.0355190	*
N_Gemme:Vigoria	2	0.030683	0.015342	3.5720	0.0951185	.
Defogl:Vigoria	2	0.100038	0.050019	11.6462	0.0085940	**
Dirad:Vigoria	2	0.151650	0.075825	17.6546	0.0030642	**
N_Gemme:Defogl:Dirad	1	0.006863	0.006863	1.5980	0.2530743	
Residuals	6	0.025769	0.004295			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.9

IBIMET_3: 0.96

AcMalicoB

Analysis of Variance Table

Response: Ca_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1.89622	1.89622	271.0364	3.201e-06 ***
IBIMET_3	1	0.00264	0.00264	0.3777	0.56142
N_Gemme	1	0.03857	0.03857	5.5134	0.05721 .
Defogl	1	0.01149	0.01149	1.6422	0.24732
Dirad	1	0.00048	0.00048	0.0684	0.80237
Vigoria	2	0.05553	0.02776	3.9685	0.07979 .
N_Gemme:Defogl	1	0.02416	0.02416	3.4529	0.11251
N_Gemme:Dirad	1	0.01929	0.01929	2.7565	0.14793
Defogl:Dirad	1	0.00362	0.00362	0.5178	0.49884
N_Gemme:Vigoria	2	0.07221	0.03611	5.1607	0.04968 *
Defogl:Vigoria	2	0.00126	0.00063	0.0900	0.91518
Dirad:Vigoria	2	0.00318	0.00159	0.2274	0.80319
N_Gemme:Defogl:Dirad	1	0.04809	0.04809	6.8732	0.03950 *
Residuals	6	0.04198	0.00700		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.989

IBIMET_3: n.s

KB

Analysis of Variance Table

Response: Ca_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.163440	0.163440	7.9208	0.03058 *
IBIMET_3	1	0.003645	0.003645	0.1766	0.68891
N_Gemme	1	0.008746	0.008746	0.4238	0.53914
Defogl	1	0.132697	0.132697	6.4310	0.04432 *
Dirad	1	0.172757	0.172757	8.3724	0.02757 *
Vigoria	2	0.098304	0.049152	2.3821	0.17319
N_Gemme:Defogl	1	0.002213	0.002213	0.1073	0.75440
N_Gemme:Dirad	1	0.037867	0.037867	1.8352	0.22430
Defogl:Dirad	1	0.010852	0.010852	0.5259	0.49562
N_Gemme:Vigoria	2	0.011508	0.005754	0.2789	0.76594
Defogl:Vigoria	2	0.147586	0.073793	3.5763	0.09494 .
Dirad:Vigoria	2	0.000297	0.000149	0.0072	0.99283
N_Gemme:Defogl:Dirad	1	0.043068	0.043068	2.0872	0.19866
Residuals	6	0.123805	0.020634		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.754

IBIMET_3: n.s

ApaB

Analysis of Variance Table

Response: Ca_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1202.55	1202.55	27.4154	0.001946 **
IBIMET_3	1	775.32	775.32	17.6756	0.005659 **

N_Gemme	1	127.25	127.25	2.9010	0.139416
Defogl	1	274.18	274.18	6.2508	0.046518 *
Dirad	1	280.38	280.38	6.3921	0.044784 *
Vigoria	2	327.10	163.55	3.7286	0.088632 .
N_Gemme:Defogl	1	164.07	164.07	3.7405	0.101275
N_Gemme:Dirad	1	138.30	138.30	3.1528	0.126138
Defogl:Dirad	1	0.13	0.13	0.0029	0.958548
N_Gemme:Vigoria	2	34.56	17.28	0.3939	0.690647
Defogl:Vigoria	2	116.92	58.46	1.3327	0.331959
Dirad:Vigoria	2	50.88	25.44	0.5799	0.588493
N_Gemme:Defogl:Dirad	1	1.40	1.40	0.0320	0.863998
Residuals	6	263.18	43.86		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.906

IBIMET_3: 0.864

Antoctot1B

Analysis of Variance Table

Response: Ca_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	8976	8975.9	1.4478	0.27419
IBIMET_3	1	1085	1085.0	0.1750	0.69026
N_Gemme	1	133	132.9	0.0214	0.88840
Defogl	1	1527	1527.3	0.2464	0.63731
Dirad	1	80	80.3	0.0130	0.91309
Vigoria	2	57471	28735.6	4.6352	0.06066 .
N_Gemme:Defogl	1	425	424.8	0.0685	0.80225
N_Gemme:Dirad	1	2516	2516.3	0.4059	0.54759
Defogl:Dirad	1	37	37.1	0.0060	0.94082
N_Gemme:Vigoria	2	8752	4376.2	0.7059	0.53050
Defogl:Vigoria	2	578	288.8	0.0466	0.95482
Dirad:Vigoria	2	2802	1401.0	0.2260	0.80422
N_Gemme:Defogl:Dirad	1	182	181.7	0.0293	0.86971
Residuals	6	37197	6199.5		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	26217	26217	2.9489	0.13674
IBIMET_3	1	1868	1868	0.2101	0.66279
N_Gemme	1	151	151	0.0170	0.90043
Defogl	1	3474	3474	0.3908	0.55491
Dirad	1	101840	101840	11.4552	0.01477 *
Vigoria	2	167734	83867	9.4335	0.01405 *
N_Gemme:Defogl	1	28584	28584	3.2152	0.12312
N_Gemme:Dirad	1	2484	2484	0.2794	0.61605
Defogl:Dirad	1	9516	9516	1.0704	0.34074

N_Gemme:Vigoria	2	3244	1622	0.1825	0.83768
Defogl:Vigoria	2	8983	4492	0.5052	0.62693
Dirad:Vigoria	2	10328	5164	0.5809	0.58803
N_Gemme:Defogl:Dirad	1	22832	22832	2.5682	0.16015
Residuals	6	53342	8890		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	9177	9177	1.9114	0.21607
IBIMET_3	1	3707	3707	0.7721	0.41338
N_Gemme	1	979	979	0.2038	0.66751
Defogl	1	25486	25486	5.3085	0.06077 .
Dirad	1	17648	17648	3.6759	0.10366
Vigoria	2	90317	45158	9.4061	0.01414 *
N_Gemme:Defogl	1	42	42	0.0087	0.92882
N_Gemme:Dirad	1	2054	2054	0.4279	0.53729
Defogl:Dirad	1	696	696	0.1450	0.71646
N_Gemme:Vigoria	2	9231	4615	0.9614	0.43434
Defogl:Vigoria	2	4410	2205	0.4592	0.65226
Dirad:Vigoria	2	4935	2467	0.5139	0.62228
N_Gemme:Defogl:Dirad	1	1301	1301	0.2710	0.62132
Residuals	6	28806	4801		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	15895	15895	0.5241	0.4964
IBIMET_3	1	2725	2725	0.0898	0.7745
N_Gemme	1	1920	1920	0.0633	0.8098
Defogl	1	46019	46019	1.5172	0.2641
Dirad	1	106483	106483	3.5107	0.1101
Vigoria	2	79268	39634	1.3067	0.3380
N_Gemme:Defogl	1	37315	37315	1.2302	0.3098
N_Gemme:Dirad	1	137	137	0.0045	0.9486
Defogl:Dirad	1	10910	10910	0.3597	0.5706
N_Gemme:Vigoria	2	7055	3527	0.1163	0.8922
Defogl:Vigoria	2	22839	11420	0.3765	0.7014
Dirad:Vigoria	2	25571	12786	0.4215	0.6741
N_Gemme:Defogl:Dirad	1	55419	55419	1.8271	0.2252
Residuals	6	181988	30331		

IBIMET_2: n.s

IBIMET_3: n.s

CACCIAGRANDE 2008M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Ca_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	563986	563986	8.7635	0.031507 *
DIPROVE_2	1	764732	764732	11.8828	0.018294 *
DIPROVE_3	1	97605	97605	1.5166	0.272887
N_Gemme	1	20188	20188	0.3137	0.599585
Defogl	1	893413	893413	13.8824	0.013629 *
Dirad	1	1598645	1598645	24.8407	0.004161 **
Vigoria	2	360613	180306	2.8017	0.152690
N_Gemme:Defogl	1	42879	42879	0.6663	0.451474
N_Gemme:Dirad	1	624993	624993	9.7115	0.026361 *
Defogl:Dirad	1	146461	146461	2.2758	0.191780
N_Gemme:Vigoria	2	361954	180977	2.8121	0.151942
Defogl:Vigoria	2	36039	18020	0.2800	0.766900
Dirad:Vigoria	2	30645	15322	0.2381	0.796584
N_Gemme:Defogl:Dirad	1	142	142	0.0022	0.964353
Residuals	5	321780	64356		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.798

DIPROVE_2: -0.839

DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Ca_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	3.0097e-05	3.0097e-05	5.0350	0.074849 .
DIPROVE_2	1	1.1536e-05	1.1536e-05	1.9299	0.223462
DIPROVE_3	1	9.0895e-05	9.0895e-05	15.2060	0.011415 *
N_Gemme	1	5.0300e-07	5.0300e-07	0.0842	0.783382
Defogl	1	1.1773e-04	1.1773e-04	19.6946	0.006778 **
Dirad	1	4.4118e-05	4.4118e-05	7.3806	0.041938 *
Vigoria	2	2.5550e-05	1.2775e-05	2.1372	0.213410
N_Gemme:Defogl	1	7.5600e-07	7.5600e-07	0.1265	0.736642
N_Gemme:Dirad	1	2.4494e-05	2.4494e-05	4.0977	0.098829 .
Defogl:Dirad	1	1.5620e-06	1.5620e-06	0.2613	0.630980
N_Gemme:Vigoria	2	5.8680e-06	2.9340e-06	0.4908	0.638802
Defogl:Vigoria	2	7.5820e-06	3.7910e-06	0.6342	0.568220
Dirad:Vigoria	2	7.4250e-06	3.7120e-06	0.6210	0.574247
N_Gemme:Defogl:Dirad	1	1.7410e-06	1.7410e-06	0.2913	0.612551
Residuals	5	2.9888e-05	5.9780e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: 0.867

BrixA

Analysis of Variance Table

Response: Ca_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1.7973	1.7973	6.3504	0.053175 .
DIPROVE_2	1	0.5195	0.5195	1.8355	0.233473
DIPROVE_3	1	4.6045	4.6045	16.2696	0.009986 **
N_Gemme	1	0.0304	0.0304	0.1074	0.756367
Defogl	1	5.3714	5.3714	18.9793	0.007314 **
Dirad	1	2.3915	2.3915	8.4499	0.033521 *
Vigoria	2	1.0649	0.5325	1.8814	0.245932
N_Gemme:Defogl	1	0.0348	0.0348	0.1231	0.739958
N_Gemme:Dirad	1	1.2463	1.2463	4.4038	0.089922 .
Defogl:Dirad	1	0.0548	0.0548	0.1936	0.678306
N_Gemme:Vigoria	2	0.2508	0.1254	0.4430	0.665073
Defogl:Vigoria	2	0.5592	0.2796	0.9880	0.434919
Dirad:Vigoria	2	0.3415	0.1707	0.6033	0.582497
N_Gemme:Defogl:Dirad	1	0.0899	0.0899	0.3175	0.597442
Residuals	5	1.4151	0.2830		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: 0.875

PhA

Analysis of Variance Table

Response: Ca_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.0118944	0.0118944	6.0313	0.05752 .
DIPROVE_2	1	0.0028306	0.0028306	1.4353	0.28459
DIPROVE_3	1	0.0004009	0.0004009	0.2033	0.67097
N_Gemme	1	0.0005003	0.0005003	0.2537	0.63588
Defogl	1	0.0183588	0.0183588	9.3091	0.02839 *
Dirad	1	0.0105926	0.0105926	5.3712	0.06826 .
Vigoria	2	0.0144041	0.0072021	3.6519	0.10527
N_Gemme:Defogl	1	0.0000231	0.0000231	0.0117	0.91797
N_Gemme:Dirad	1	0.0126120	0.0126120	6.3951	0.05260 .
Defogl:Dirad	1	0.0013836	0.0013836	0.7016	0.44045
N_Gemme:Vigoria	2	0.0015036	0.0007518	0.3812	0.70131
Defogl:Vigoria	2	0.0117691	0.0058846	2.9839	0.14032
Dirad:Vigoria	2	0.0000185	0.0000093	0.0047	0.99532
N_Gemme:Defogl:Dirad	1	0.0047435	0.0047435	2.4052	0.18162
Residuals	5	0.0098606	0.0019721		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

AcTitola

Analysis of Variance Table

Response: Ca_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.12647	0.12647	2.4312	0.17968
DIPROVE_2	1	0.00097	0.00097	0.0187	0.89659
DIPROVE_3	1	0.70978	0.70978	13.6441	0.01409 *
N_Gemme	1	0.03468	0.03468	0.6667	0.45134
Defogl	1	0.14467	0.14467	2.7810	0.15627
Dirad	1	0.08088	0.08088	1.5548	0.26765
Vigoria	2	0.52002	0.26001	4.9981	0.06419 .
N_Gemme:Defogl	1	0.00033	0.00033	0.0063	0.93974
N_Gemme:Dirad	1	0.06316	0.06316	1.2141	0.32071
Defogl:Dirad	1	0.01730	0.01730	0.3326	0.58911
N_Gemme:Vigoria	2	0.05857	0.02928	0.5629	0.60187
Defogl:Vigoria	2	0.09688	0.04844	0.9312	0.45315
Dirad:Vigoria	2	0.20915	0.10457	2.0102	0.22875
N_Gemme:Defogl:Dirad	1	0.00191	0.00191	0.0367	0.85552
Residuals	5	0.26011	0.05202		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: -0.855

AcTartA

Analysis of Variance Table

Response: Ca_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.00116	0.00116	0.0295	0.87034
DIPROVE_2	1	0.21844	0.21844	5.5726	0.06470 .
DIPROVE_3	1	0.89486	0.89486	22.8286	0.00498 **
N_Gemme	1	0.02918	0.02918	0.7444	0.42768
Defogl	1	0.23234	0.23234	5.9273	0.05904 .
Dirad	1	0.00721	0.00721	0.1839	0.68592
Vigoria	2	0.97823	0.48911	12.4777	0.01138 *
N_Gemme:Defogl	1	0.08305	0.08305	2.1188	0.20527
N_Gemme:Dirad	1	0.01060	0.01060	0.2705	0.62522
Defogl:Dirad	1	0.01435	0.01435	0.3660	0.57160
N_Gemme:Vigoria	2	0.00468	0.00234	0.0597	0.94274
Defogl:Vigoria	2	0.00138	0.00069	0.0176	0.98258
Dirad:Vigoria	2	0.04683	0.02341	0.5973	0.58533
N_Gemme:Defogl:Dirad	1	0.00116	0.00116	0.0296	0.87011
Residuals	5	0.19600	0.03920		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: -0.906

AcMalicoA

Analysis of Variance Table

Response: Ca_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.11518	0.11518	4.8284	0.079357 .
DIPROVE_2	1	0.47897	0.47897	20.0786	0.006513 **
DIPROVE_3	1	0.21144	0.21144	8.8638	0.030899 *
N_Gemme	1	0.01764	0.01764	0.7395	0.429105
Defogl	1	0.00058	0.00058	0.0245	0.881798
Dirad	1	0.07104	0.07104	2.9780	0.144993
Vigoria	2	0.18571	0.09286	3.8926	0.095645 .
N_Gemme:Defogl	1	0.03313	0.03313	1.3890	0.291600
N_Gemme:Dirad	1	0.14625	0.14625	6.1310	0.056110 .
Defogl:Dirad	1	0.11917	0.11917	4.9955	0.075683 .
N_Gemme:Vigoria	2	0.06139	0.03069	1.2867	0.354151
Defogl:Vigoria	2	0.09053	0.04527	1.8976	0.243676
Dirad:Vigoria	2	0.26742	0.13371	5.6052	0.052838 .
N_Gemme:Defogl:Dirad	1	0.07515	0.07515	3.1502	0.136091
Residuals	5	0.11927	0.02385		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.895
DIPROVE_3: -0.8

KA

Analysis of Variance Table

Response: Ca_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.000008	0.000008	0.0029	0.958835
DIPROVE_2	1	0.022870	0.022870	8.0481	0.036378 *
DIPROVE_3	1	0.002099	0.002099	0.7385	0.429406
N_Gemme	1	0.003564	0.003564	1.2542	0.313645
Defogl	1	0.069011	0.069011	24.2849	0.004367 **
Dirad	1	0.027527	0.027527	9.6868	0.026480 *
Vigoria	2	0.018600	0.009300	3.2726	0.123430
N_Gemme:Defogl	1	0.006768	0.006768	2.3816	0.183423
N_Gemme:Dirad	1	0.000896	0.000896	0.3154	0.598616
Defogl:Dirad	1	0.007886	0.007886	2.7750	0.156626
N_Gemme:Vigoria	2	0.016139	0.008070	2.8397	0.149990
Defogl:Vigoria	2	0.008067	0.004034	1.4195	0.324923
Dirad:Vigoria	2	0.029589	0.014794	5.2061	0.059946 .
N_Gemme:Defogl:Dirad	1	0.017943	0.017943	6.3142	0.053643 .
Residuals	5	0.014209	0.002842		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.785
DIPROVE_3: n.s

ApaA

Analysis of Variance Table

Response: Ca_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	61.95	61.95	0.2416	0.64390
DIPROVE_2	1	324.47	324.47	1.2651	0.31176

DIPROVE_3	1	1136.63	1136.63	4.4317	0.08917	.
N_Gemme	1	590.36	590.36	2.3018	0.18967	
Defogl	1	366.24	366.24	1.4280	0.28568	
Dirad	1	99.75	99.75	0.3889	0.56021	
Vigoria	2	1082.45	541.23	2.1102	0.21654	
N_Gemme:Defogl	1	208.37	208.37	0.8124	0.40872	
N_Gemme:Dirad	1	11.66	11.66	0.0455	0.83959	
Defogl:Dirad	1	79.77	79.77	0.3110	0.60109	
N_Gemme:Vigoria	2	164.39	82.20	0.3205	0.73968	
Defogl:Vigoria	2	93.00	46.50	0.1813	0.83943	
Dirad:Vigoria	2	170.29	85.14	0.3320	0.73220	
N_Gemme:Defogl:Dirad	1	46.77	46.77	0.1823	0.68714	
Residuals	5	1282.39	256.48			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Ca_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	10785	10785	2.3772	0.183756	
DIPROVE_2	1	151408	151408	33.3728	0.002186	**
DIPROVE_3	1	1124	1124	0.2478	0.639754	
N_Gemme	1	51	51	0.0113	0.919471	
Defogl	1	1071	1071	0.2360	0.647638	
Dirad	1	22562	22562	4.9730	0.076162	.
Vigoria	2	35758	17879	3.9408	0.093864	.
N_Gemme:Defogl	1	4384	4384	0.9663	0.370740	
N_Gemme:Dirad	1	72	72	0.0159	0.904611	
Defogl:Dirad	1	6596	6596	1.4540	0.281833	
N_Gemme:Vigoria	2	15864	7932	1.7484	0.265640	
Defogl:Vigoria	2	4683	2342	0.5161	0.625503	
Dirad:Vigoria	2	2909	1455	0.3206	0.739591	
N_Gemme:Defogl:Dirad	1	7323	7323	1.6141	0.259824	
Residuals	5	22684	4537			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.933
DIPROVE_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Ca_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	77483	77483	5.4359	0.067086	.
DIPROVE_2	1	445505	445505	31.2545	0.002527	**
DIPROVE_3	1	3015	3015	0.2116	0.664862	
N_Gemme	1	3902	3902	0.2737	0.623183	
Defogl	1	128710	128710	9.0297	0.029929	*

Dirad	1	9680	9680	0.6791	0.447418
Vigoria	2	136386	68193	4.7841	0.069009 .
N_Gemme:Defogl	1	67	67	0.0047	0.948138
N_Gemme:Dirad	1	713	713	0.0500	0.831909
Defogl:Dirad	1	27054	27054	1.8980	0.226772
N_Gemme:Vigoria	2	37664	18832	1.3212	0.346229
Defogl:Vigoria	2	18740	9370	0.6574	0.557875
Dirad:Vigoria	2	55641	27821	1.9518	0.236330
N_Gemme:Defogl:Dirad	1	35695	35695	2.5042	0.174391
Residuals	5	71270	14254		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: -0.928
 DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	4855	4855	1.1629	0.330126
DIPROVE_2	1	69925	69925	16.7476	0.009425 **
DIPROVE_3	1	5372	5372	1.2866	0.308104
N_Gemme	1	11021	11021	2.6397	0.165152
Defogl	1	15570	15570	3.7292	0.111330
Dirad	1	34759	34759	8.3251	0.034374 *
Vigoria	2	15925	7963	1.9071	0.242360
N_Gemme:Defogl	1	4363	4363	1.0450	0.353550
N_Gemme:Dirad	1	2210	2210	0.5293	0.499543
Defogl:Dirad	1	11011	11011	2.6373	0.165311
N_Gemme:Vigoria	2	18062	9031	2.1630	0.210468
Defogl:Vigoria	2	1231	616	0.1474	0.866538
Dirad:Vigoria	2	3329	1665	0.3987	0.690789
N_Gemme:Defogl:Dirad	1	36	36	0.0085	0.929927
Residuals	5	20876	4175		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: -0.878
 DIPROVE_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Ca_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	127	127	0.0277	0.874357
DIPROVE_2	1	141572	141572	30.8737	0.002596 **
DIPROVE_3	1	6785	6785	1.4797	0.278106
N_Gemme	1	3913	3913	0.8533	0.398020
Defogl	1	65559	65559	14.2969	0.012875 *
Dirad	1	5983	5983	1.3048	0.305063
Vigoria	2	16623	8311	1.8125	0.255870
N_Gemme:Defogl	1	2387	2387	0.5206	0.502898

N_Gemme:Dirad	1	80	80	0.0174	0.900081
Defogl:Dirad	1	46419	46419	10.1230	0.024493 *
N_Gemme:Vigoria	2	41195	20597	4.4918	0.076449 .
Defogl:Vigoria	2	13565	6782	1.4791	0.312886
Dirad:Vigoria	2	91565	45783	9.9841	0.017945 *
N_Gemme:Defogl:Dirad	1	4480	4480	0.9770	0.368330
Residuals	5	22928	4586		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.928
DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	247925	247925	1.3894	0.29154
DIPROVE_2	1	24	24	0.0001	0.99126
DIPROVE_3	1	323539	323539	1.8131	0.23595
N_Gemme	1	6470	6470	0.0363	0.85647
Defogl	1	90236	90236	0.5057	0.50880
Dirad	1	1284112	1284112	7.1963	0.04368 *
Vigoria	2	280349	140174	0.7855	0.50505
N_Gemme:Defogl	1	56636	56636	0.3174	0.59751
N_Gemme:Dirad	1	226105	226105	1.2671	0.31142
Defogl:Dirad	1	124392	124392	0.6971	0.44181
N_Gemme:Vigoria	2	226063	113031	0.6334	0.56859
Defogl:Vigoria	2	127867	63933	0.3583	0.71546
Dirad:Vigoria	2	133571	66785	0.3743	0.70556
N_Gemme:Defogl:Dirad	1	3217	3217	0.0180	0.89842
Residuals	5	892206	178441		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	3.1520e-06	3.1520e-06	0.1204	0.74272
DIPROVE_2	1	8.1395e-05	8.1395e-05	3.1091	0.13814
DIPROVE_3	1	8.2690e-05	8.2690e-05	3.1586	0.13567
N_Gemme	1	6.7480e-06	6.7480e-06	0.2578	0.63325
Defogl	1	1.2746e-04	1.2746e-04	4.8686	0.07845 .
Dirad	1	5.8625e-05	5.8625e-05	2.2394	0.19479
Vigoria	2	5.6710e-06	2.8350e-06	0.1083	0.89940
N_Gemme:Defogl	1	3.5000e-06	3.5000e-06	0.1337	0.72958
N_Gemme:Dirad	1	3.0256e-05	3.0256e-05	1.1557	0.33147
Defogl:Dirad	1	3.0000e-09	3.0000e-09	0.0001	0.99172
N_Gemme:Vigoria	2	1.5001e-05	7.5010e-06	0.2865	0.76243

Defogl:Vigoria	2	1.3827e-05	6.9130e-06	0.2641	0.77799
Dirad:Vigoria	2	2.3139e-05	1.1570e-05	0.4419	0.66568
N_Gemme:Defogl:Dirad	1	4.9900e-07	4.9900e-07	0.0191	0.89555
Residuals	5	1.3090e-04	2.6179e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.1722	0.1722	0.1324	0.73082
DIPROVE_2	1	4.0614	4.0614	3.1224	0.13747
DIPROVE_3	1	3.9458	3.9458	3.0335	0.14203
N_Gemme	1	0.2666	0.2666	0.2049	0.66973
Defogl	1	6.1836	6.1836	4.7539	0.08108 .
Dirad	1	2.8812	2.8812	2.2151	0.19684
Vigoria	2	0.2466	0.1233	0.0948	0.91117
N_Gemme:Defogl	1	0.2125	0.2125	0.1634	0.70278
N_Gemme:Dirad	1	1.5270	1.5270	1.1740	0.32804
Defogl:Dirad	1	0.0198	0.0198	0.0152	0.90656
N_Gemme:Vigoria	2	0.7696	0.3848	0.2958	0.75609
Defogl:Vigoria	2	0.7682	0.3841	0.2953	0.75644
Dirad:Vigoria	2	1.1768	0.5884	0.4524	0.65981
N_Gemme:Defogl:Dirad	1	0.0390	0.0390	0.0300	0.86925
Residuals	5	6.5036	1.3007		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.005509	0.005509	1.3661	0.29517
DIPROVE_2	1	0.001608	0.001608	0.3987	0.55551
DIPROVE_3	1	0.001390	0.001390	0.3447	0.58263
N_Gemme	1	0.001732	0.001732	0.4295	0.54120
Defogl	1	0.056505	0.056505	14.0125	0.01339 *
Dirad	1	0.015507	0.015507	3.8455	0.10715
Vigoria	2	0.013319	0.006660	1.6515	0.28141
N_Gemme:Defogl	1	0.000029	0.000029	0.0071	0.93615
N_Gemme:Dirad	1	0.007402	0.007402	1.8357	0.23346
Defogl:Dirad	1	0.000797	0.000797	0.1976	0.67526
N_Gemme:Vigoria	2	0.001744	0.000872	0.2162	0.81271
Defogl:Vigoria	2	0.030981	0.015491	3.8415	0.09758 .
Dirad:Vigoria	2	0.003186	0.001593	0.3950	0.69297
N_Gemme:Defogl:Dirad	1	0.007155	0.007155	1.7743	0.24034

Residuals 5 0.020162 0.004032

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

AcTitolB

Analysis of Variance Table

Response: Ca_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.022065	0.022065	1.6542	0.254727
DIPROVE_2	1	0.129015	0.129015	9.6721	0.026551 *
DIPROVE_3	1	0.306363	0.306363	22.9675	0.004916 **
N_Gemme	1	0.055683	0.055683	4.1745	0.096481 .
Defogl	1	0.032496	0.032496	2.4361	0.179318
Dirad	1	0.000531	0.000531	0.0398	0.849767
Vigoria	2	0.085947	0.042973	3.2216	0.126197
N_Gemme:Defogl	1	0.081552	0.081552	6.1138	0.056349 .
N_Gemme:Dirad	1	0.025513	0.025513	1.9127	0.225240
Defogl:Dirad	1	0.000013	0.000013	0.0010	0.976483
N_Gemme:Vigoria	2	0.039377	0.019689	1.4760	0.313493
Defogl:Vigoria	2	0.121012	0.060506	4.5361	0.075254 .
Dirad:Vigoria	2	0.100504	0.050252	3.7673	0.100496
N_Gemme:Defogl:Dirad	1	0.007818	0.007818	0.5861	0.478496
Residuals	5	0.066695	0.013339		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: 0.812

DIPROVE_3: -0.906

AcTartB

Analysis of Variance Table

Response: Ca_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.00022	0.000217	0.0159	0.904597
DIPROVE_2	1	0.01322	0.013218	0.9691	0.370108
DIPROVE_3	1	0.04446	0.044457	3.2596	0.130839
N_Gemme	1	0.00309	0.003092	0.2267	0.654046
Defogl	1	0.01296	0.012964	0.9505	0.374363
Dirad	1	0.02273	0.022735	1.6669	0.253140
Vigoria	2	0.42248	0.211240	15.4880	0.007201 **
N_Gemme:Defogl	1	0.22402	0.224016	16.4248	0.009799 **
N_Gemme:Dirad	1	0.02812	0.028117	2.0615	0.210540
Defogl:Dirad	1	0.00455	0.004546	0.3333	0.588749
N_Gemme:Vigoria	2	0.07705	0.038525	2.8246	0.151053
Defogl:Vigoria	2	0.03102	0.015509	1.1371	0.391709
Dirad:Vigoria	2	0.06794	0.033970	2.4907	0.177604
N_Gemme:Defogl:Dirad	1	0.01650	0.016504	1.2101	0.321434
Residuals	5	0.06819	0.013639		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Ca_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.01754	0.01754	4.5774	0.0853738	.
DIPROVE_2	1	0.90096	0.90096	235.1728	2.139e-05	***
DIPROVE_3	1	0.21508	0.21508	56.1425	0.0006692	***
N_Gemme	1	0.00983	0.00983	2.5654	0.1701254	
Defogl	1	0.01715	0.01715	4.4769	0.0879643	.
Dirad	1	0.00090	0.00090	0.2353	0.6481445	
Vigoria	2	0.66075	0.33037	86.2361	0.0001332	***
N_Gemme:Defogl	1	0.05428	0.05428	14.1683	0.0131023	*
N_Gemme:Dirad	1	0.05135	0.05135	13.4042	0.0145769	*
Defogl:Dirad	1	0.04196	0.04196	10.9528	0.0212522	*
N_Gemme:Vigoria	2	0.12163	0.06081	15.8739	0.0068289	**
Defogl:Vigoria	2	0.01660	0.00830	2.1664	0.2100896	
Dirad:Vigoria	2	0.08133	0.04067	10.6151	0.0158642	*
N_Gemme:Defogl:Dirad	1	0.01020	0.01020	2.6614	0.1637410	
Residuals	5	0.01916	0.00383			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.99
DIPROVE_3: -0.958

KB

Analysis of Variance Table

Response: Ca_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.01284	0.01284	0.5305	0.49905	
DIPROVE_2	1	0.08738	0.08738	3.6108	0.11583	
DIPROVE_3	1	0.00173	0.00173	0.0717	0.79960	
N_Gemme	1	0.00145	0.00145	0.0599	0.81640	
Defogl	1	0.35630	0.35630	14.7237	0.01216	*
Dirad	1	0.07689	0.07689	3.1774	0.13476	
Vigoria	2	0.09650	0.04825	1.9939	0.23083	
N_Gemme:Defogl	1	0.00045	0.00045	0.0188	0.89631	
N_Gemme:Dirad	1	0.02038	0.02038	0.8423	0.40086	
Defogl:Dirad	1	0.01234	0.01234	0.5099	0.50710	
N_Gemme:Vigoria	2	0.01565	0.00782	0.3233	0.73784	
Defogl:Vigoria	2	0.08676	0.04338	1.7927	0.25884	
Dirad:Vigoria	2	0.03483	0.01741	0.7196	0.53132	
N_Gemme:Defogl:Dirad	1	0.03229	0.03229	1.3344	0.30023	
Residuals	5	0.12099	0.02420			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s

DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Ca_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	12.55	12.55	0.1654	0.70109
DIPROVE_2	1	195.67	195.67	2.5790	0.16920
DIPROVE_3	1	1271.64	1271.64	16.7606	0.00941 **
N_Gemme	1	7.51	7.51	0.0990	0.76572
Defogl	1	477.10	477.10	6.2883	0.05398 .
Dirad	1	7.31	7.31	0.0963	0.76880
Vigoria	2	899.16	449.58	5.9256	0.04796 *
N_Gemme:Defogl	1	227.63	227.63	3.0003	0.14380
N_Gemme:Dirad	1	72.04	72.04	0.9496	0.37459
Defogl:Dirad	1	2.67	2.67	0.0351	0.85867
N_Gemme:Vigoria	2	60.39	30.20	0.3980	0.69120
Defogl:Vigoria	2	46.21	23.10	0.3045	0.75024
Dirad:Vigoria	2	70.09	35.05	0.4619	0.65451
N_Gemme:Defogl:Dirad	1	26.89	26.89	0.3544	0.57756
Residuals	5	379.35	75.87		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: -0.878

Antoctot1B

Analysis of Variance Table

Response: Ca_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	5886	5885.8	0.9399	0.3768
DIPROVE_2	1	1939	1938.6	0.3096	0.6019
DIPROVE_3	1	2515	2515.1	0.4016	0.5541
N_Gemme	1	477	477.0	0.0762	0.7936
Defogl	1	3192	3191.8	0.5097	0.5072
Dirad	1	2263	2263.0	0.3614	0.5739
Vigoria	2	38015	19007.6	3.0354	0.1371
N_Gemme:Defogl	1	102	102.1	0.0163	0.9034
N_Gemme:Dirad	1	1553	1553.3	0.2481	0.6396
Defogl:Dirad	1	12289	12289.0	1.9625	0.2202
N_Gemme:Vigoria	2	11093	5546.4	0.8857	0.4685
Defogl:Vigoria	2	974	487.2	0.0778	0.9262
Dirad:Vigoria	2	9309	4654.6	0.7433	0.5216
N_Gemme:Defogl:Dirad	1	845	844.6	0.1349	0.7285
Residuals	5	31310	6261.9		

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	8305	8305	0.7779	0.41816
DIPROVE_2	1	42749	42749	4.0044	0.10180
DIPROVE_3	1	161	161	0.0151	0.90710
N_Gemme	1	153	153	0.0144	0.90925
Defogl	1	4362	4362	0.4086	0.55079
Dirad	1	85396	85396	7.9992	0.03675 *
Vigoria	2	142390	71195	6.6690	0.03882 *
N_Gemme:Defogl	1	10675	10675	1.0000	0.36322
N_Gemme:Dirad	1	5904	5904	0.5531	0.49051
Defogl:Dirad	1	3643	3643	0.3412	0.58449
N_Gemme:Vigoria	2	6368	3184	0.2983	0.75445
Defogl:Vigoria	2	2807	1403	0.1315	0.87975
Dirad:Vigoria	2	44006	22003	2.0611	0.22242
N_Gemme:Defogl:Dirad	1	30303	30303	2.8385	0.15285
Residuals	5	53378	10676		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	37348	37348	3.5284	0.1191
DIPROVE_2	1	10091	10091	0.9534	0.3737
DIPROVE_3	1	2598	2598	0.2454	0.6413
N_Gemme	1	235	235	0.0222	0.8874
Defogl	1	312	312	0.0295	0.8703
Dirad	1	22815	22815	2.1555	0.2020
Vigoria	2	20276	10138	0.9578	0.4445
N_Gemme:Defogl	1	23	23	0.0022	0.9645
N_Gemme:Dirad	1	1018	1018	0.0962	0.7689
Defogl:Dirad	1	15101	15101	1.4267	0.2859
N_Gemme:Vigoria	2	17230	8615	0.8139	0.4943
Defogl:Vigoria	2	6744	3372	0.3186	0.7409
Dirad:Vigoria	2	11918	5959	0.5630	0.6018
N_Gemme:Defogl:Dirad	1	153	153	0.0145	0.9089
Residuals	5	52924	10585		

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	43292	43292	2.0018	0.21626

DIPROVE_2	1	6993	6993	0.3234	0.59418
DIPROVE_3	1	818	818	0.0378	0.85349
N_Gemme	1	24	24	0.0011	0.97449
Defogl	1	5255	5255	0.2430	0.64295
Dirad	1	118302	118302	5.4702	0.06647 .
Vigoria	2	53488	26744	1.2366	0.36614
N_Gemme:Defogl	1	30423	30423	1.4068	0.28888
N_Gemme:Dirad	1	24679	24679	1.1412	0.33425
Defogl:Dirad	1	2529	2529	0.1170	0.74626
N_Gemme:Vigoria	2	30290	15145	0.7003	0.53935
Defogl:Vigoria	2	4269	2134	0.0987	0.90775
Dirad:Vigoria	2	84037	42019	1.9429	0.23751
N_Gemme:Defogl:Dirad	1	81011	81011	3.7459	0.11072
Residuals	5	108133	21627		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

CACCIAGRANDE 2008M IASMA

PvinacciaA

Analysis of Variance Table

Response: Ca_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	252544	252544	5.0340	0.074870 .
IASMA_2	1	711882	711882	14.1901	0.013064 *
IASMA_3	1	291059	291059	5.8017	0.060964 .
N_Gemme	1	1074	1074	0.0214	0.889393
Defogl	1	1063690	1063690	21.2027	0.005816 **
Dirad	1	2043732	2043732	40.7380	0.001398 **
Vigoria	2	479719	239859	4.7812	0.069080 .
N_Gemme:Defogl	1	1268	1268	0.0253	0.879896
N_Gemme:Dirad	1	337341	337341	6.7243	0.048653 *
Defogl:Dirad	1	136481	136481	2.7205	0.159981
N_Gemme:Vigoria	2	181450	90725	1.8084	0.256479
Defogl:Vigoria	2	15789	7894	0.1574	0.858467
Dirad:Vigoria	2	78991	39496	0.7873	0.504384
N_Gemme:Defogl:Dirad	1	18215	18215	0.3631	0.573080
Residuals	5	250838	50168		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.86
IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Ca_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	2.5057e-05	2.5057e-05	4.2310	0.09480 .
IASMA_2	1	7.0180e-06	7.0180e-06	1.1850	0.32601

IASMA_3	1	6.5753e-05	6.5753e-05	11.1025	0.02073	*
N_Gemme	1	1.0710e-06	1.0710e-06	0.1809	0.68830	
Defogl	1	8.9005e-05	8.9005e-05	15.0286	0.01168	*
Dirad	1	6.5594e-05	6.5594e-05	11.0756	0.02082	*
Vigoria	2	8.2267e-05	4.1134e-05	6.9455	0.03604	*
N_Gemme:Defogl	1	2.0700e-07	2.0700e-07	0.0350	0.85901	
N_Gemme:Dirad	1	1.2254e-05	1.2254e-05	2.0691	0.20983	
Defogl:Dirad	1	6.4400e-07	6.4400e-07	0.1088	0.75493	
N_Gemme:Vigoria	2	8.0200e-06	4.0100e-06	0.6771	0.54925	
Defogl:Vigoria	2	8.1910e-06	4.0960e-06	0.6916	0.54305	
Dirad:Vigoria	2	4.6410e-06	2.3200e-06	0.3918	0.69491	
N_Gemme:Defogl:Dirad	1	4.0400e-07	4.0400e-07	0.0682	0.80434	
Residuals	5	2.9612e-05	5.9220e-06			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.83

BrixA

Analysis of Variance Table

Response: Ca_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	1.3799	1.3799	4.2409	0.09451 .
IASMA_2	1	0.4412	0.4412	1.3561	0.29675
IASMA_3	1	3.2130	3.2130	9.8749	0.02560 *
N_Gemme	1	0.0792	0.0792	0.2435	0.64259
Defogl	1	4.5393	4.5393	13.9511	0.01350 *
Dirad	1	3.3158	3.3158	10.1906	0.02420 *
Vigoria	2	3.5640	1.7820	5.4767	0.05499 .
N_Gemme:Defogl	1	0.0192	0.0192	0.0590	0.81769
N_Gemme:Dirad	1	0.5014	0.5014	1.5410	0.26952
Defogl:Dirad	1	0.0215	0.0215	0.0660	0.80748
N_Gemme:Vigoria	2	0.3226	0.1613	0.4958	0.63616
Defogl:Vigoria	2	0.4536	0.2268	0.6971	0.54071
Dirad:Vigoria	2	0.2728	0.1364	0.4192	0.67872
N_Gemme:Defogl:Dirad	1	0.0214	0.0214	0.0656	0.80802
Residuals	5	1.6269	0.3254		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.815

PhA

Analysis of Variance Table

Response: Ca_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.0005410	0.0005410	0.3100	0.60169
IASMA_2	1	0.0011989	0.0011989	0.6870	0.44495
IASMA_3	1	0.0040202	0.0040202	2.3035	0.18954
N_Gemme	1	0.0018221	0.0018221	1.0440	0.35376
Defogl	1	0.0193934	0.0193934	11.1119	0.02070 *

Dirad	1	0.0140573	0.0140573	8.0545	0.03633	*
Vigoria	2	0.0261578	0.0130789	7.4939	0.03130	*
N_Gemme:Defogl	1	0.0044368	0.0044368	2.5421	0.17173	
N_Gemme:Dirad	1	0.0058394	0.0058394	3.3458	0.12690	
Defogl:Dirad	1	0.0000452	0.0000452	0.0259	0.87851	
N_Gemme:Vigoria	2	0.0048833	0.0024417	1.3990	0.32921	
Defogl:Vigoria	2	0.0037548	0.0018774	1.0757	0.40874	
Dirad:Vigoria	2	0.0009237	0.0004618	0.2646	0.77761	
N_Gemme:Defogl:Dirad	1	0.0050956	0.0050956	2.9196	0.14821	
Residuals	5	0.0087264	0.0017453			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Ca_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.30658	0.30658	3.6456	0.11448
IASMA_2	1	0.00118	0.00118	0.0141	0.91021
IASMA_3	1	0.23295	0.23295	2.7701	0.15692
N_Gemme	1	0.00066	0.00066	0.0078	0.93309
Defogl	1	0.18915	0.18915	2.2492	0.19397
Dirad	1	0.09835	0.09835	1.1695	0.32889
Vigoria	2	0.86837	0.43418	5.1630	0.06079
N_Gemme:Defogl	1	0.01281	0.01281	0.1523	0.71241
N_Gemme:Dirad	1	0.00926	0.00926	0.1102	0.75343
Defogl:Dirad	1	0.00761	0.00761	0.0905	0.77562
N_Gemme:Vigoria	2	0.05344	0.02672	0.3177	0.74150
Defogl:Vigoria	2	0.06199	0.03100	0.3686	0.70905
Dirad:Vigoria	2	0.04661	0.02331	0.2771	0.76888
N_Gemme:Defogl:Dirad	1	0.01545	0.01545	0.1837	0.68604
Residuals	5	0.42048	0.08410		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

AcTartA

Analysis of Variance Table

Response: Ca_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.21477	0.21477	9.7111	0.0263634	*
IASMA_2	1	0.26942	0.26942	12.1823	0.0174646	*
IASMA_3	1	0.01324	0.01324	0.5986	0.4740755	
N_Gemme	1	0.00415	0.00415	0.1874	0.6831095	
Defogl	1	0.05358	0.05358	2.4227	0.1803171	
Dirad	1	0.07589	0.07589	3.4315	0.1231650	
Vigoria	2	1.83653	0.91826	41.5202	0.0007686	***
N_Gemme:Defogl	1	0.08249	0.08249	3.7298	0.1113092	

N_Gemme:Dirad	1	0.00917	0.00917	0.4146	0.5480067
Defogl:Dirad	1	0.01600	0.01600	0.7233	0.4338919
N_Gemme:Vigoria	2	0.02567	0.01284	0.5804	0.5933903
Defogl:Vigoria	2	0.00278	0.00139	0.0628	0.9398801
Dirad:Vigoria	2	0.00279	0.00139	0.0630	0.9396869
N_Gemme:Defogl:Dirad	1	0.00242	0.00242	0.1095	0.7541625
Residuals	5	0.11058	0.02212		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.812
IASMA_2: 0.842
IASMA_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Ca_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.12298	0.12298	2.7018	0.16116
IASMA_2	1	0.45371	0.45371	9.9673	0.02518 *
IASMA_3	1	0.34889	0.34889	7.6646	0.03944 *
N_Gemme	1	0.00928	0.00928	0.2038	0.67057
Defogl	1	0.05129	0.05129	1.1267	0.33704
Dirad	1	0.01922	0.01922	0.4222	0.54451
Vigoria	2	0.42537	0.21269	4.6724	0.07173 .
N_Gemme:Defogl	1	0.01866	0.01866	0.4100	0.55017
N_Gemme:Dirad	1	0.00457	0.00457	0.1005	0.76404
Defogl:Dirad	1	0.04226	0.04226	0.9285	0.37951
N_Gemme:Vigoria	2	0.04568	0.02284	0.5018	0.63299
Defogl:Vigoria	2	0.07951	0.03975	0.8733	0.47283
Dirad:Vigoria	2	0.14352	0.07176	1.5764	0.29454
N_Gemme:Defogl:Dirad	1	0.00037	0.00037	0.0081	0.93170
Residuals	5	0.22760	0.04552		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.816
IASMA_3: -0.778

KA

Analysis of Variance Table

Response: Ca_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.000311	0.000311	0.1392	0.724382
IASMA_2	1	0.017546	0.017546	7.8545	0.037877 *
IASMA_3	1	0.000919	0.000919	0.4116	0.549417
N_Gemme	1	0.003215	0.003215	1.4391	0.284026
Defogl	1	0.017616	0.017616	7.8857	0.037630 *
Dirad	1	0.070513	0.070513	31.5648	0.002472 **
Vigoria	2	0.026547	0.013274	5.9419	0.047725 *
N_Gemme:Defogl	1	0.017910	0.017910	8.0171	0.036612 *
N_Gemme:Dirad	1	0.005037	0.005037	2.2546	0.193522
Defogl:Dirad	1	0.000016	0.000016	0.0073	0.935281
N_Gemme:Vigoria	2	0.032171	0.016085	7.2005	0.033718 *

Defogl:Vigoria	2	0.017813	0.008907	3.9870	0.092203	.
Dirad:Vigoria	2	0.017163	0.008582	3.8415	0.097582	.
N_Gemme:Defogl:Dirad	1	0.007229	0.007229	3.2358	0.131954	
Residuals	5	0.011170	0.002234			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.782
IASMA_3: n.s

ApaA

Analysis of Variance Table

Response: Ca_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	905.64	905.64	4.2383	0.09459	.
IASMA_2	1	10.23	10.23	0.0479	0.83542	
IASMA_3	1	4.77	4.77	0.0223	0.88705	
N_Gemme	1	251.33	251.33	1.1762	0.32764	
Defogl	1	387.65	387.65	1.8142	0.23584	
Dirad	1	0.41	0.41	0.0019	0.96683	
Vigoria	2	2096.91	1048.45	4.9066	0.06619	.
N_Gemme:Defogl	1	226.52	226.52	1.0601	0.35041	
N_Gemme:Dirad	1	21.52	21.52	0.1007	0.76379	
Defogl:Dirad	1	38.34	38.34	0.1794	0.68946	
N_Gemme:Vigoria	2	134.84	67.42	0.3155	0.74294	
Defogl:Vigoria	2	122.73	61.36	0.2872	0.76197	
Dirad:Vigoria	2	235.32	117.66	0.5506	0.60796	
N_Gemme:Defogl:Dirad	1	213.85	213.85	1.0008	0.36304	
Residuals	5	1068.41	213.68			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Ca_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	28410	28410	6.7586	0.04827	*
IASMA_2	1	47056	47056	11.1943	0.02042	*
IASMA_3	1	2515	2515	0.5983	0.47418	
N_Gemme	1	3405	3405	0.8099	0.40939	
Defogl	1	3679	3679	0.8751	0.39249	
Dirad	1	8304	8304	1.9756	0.21885	
Vigoria	2	94698	47349	11.2640	0.01406	*
N_Gemme:Defogl	1	2983	2983	0.7097	0.43798	
N_Gemme:Dirad	1	6953	6953	1.6540	0.25474	
Defogl:Dirad	1	19756	19756	4.7000	0.08236	.
N_Gemme:Vigoria	2	15426	7713	1.8349	0.25258	
Defogl:Vigoria	2	14224	7112	1.6920	0.27467	
Dirad:Vigoria	2	18525	9263	2.2035	0.20597	
N_Gemme:Defogl:Dirad	1	323	323	0.0769	0.79266	

Residuals 5 21018 4204

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.758

IASMA_2: 0.831

IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Ca_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	198808	198808	7.5275	0.04062	*
IASMA_2	1	212515	212515	8.0465	0.03639	*
IASMA_3	1	4078	4078	0.1544	0.71054	
N_Gemme	1	3093	3093	0.1171	0.74612	
Defogl	1	123303	123303	4.6686	0.08311	.
Dirad	1	14475	14475	0.5480	0.49239	
Vigoria	2	177989	88995	3.3696	0.11839	
N_Gemme:Defogl	1	603	603	0.0228	0.88585	
N_Gemme:Dirad	1	483	483	0.0183	0.89769	
Defogl:Dirad	1	81934	81934	3.1023	0.13849	
N_Gemme:Vigoria	2	51789	25895	0.9804	0.43728	
Defogl:Vigoria	2	36953	18476	0.6996	0.53966	
Dirad:Vigoria	2	13093	6547	0.2479	0.78951	
N_Gemme:Defogl:Dirad	1	354	354	0.0134	0.91238	
Residuals	5	132055	26411			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.775

IASMA_2: -0.785

IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	2164	2164	1.4952	0.275903	
IASMA_2	1	33424	33424	23.0906	0.004861	**
IASMA_3	1	1125	1125	0.7770	0.418399	
N_Gemme	1	7109	7109	4.9110	0.077511	.
Defogl	1	21562	21562	14.8958	0.011886	*
Dirad	1	40994	40994	28.3202	0.003136	**
Vigoria	2	49965	24983	17.2587	0.005694	**
N_Gemme:Defogl	1	1124	1124	0.7763	0.418617	
N_Gemme:Dirad	1	5100	5100	3.5230	0.119340	
Defogl:Dirad	1	13439	13439	9.2841	0.028522	*
N_Gemme:Vigoria	2	14949	7474	5.1635	0.060784	.
Defogl:Vigoria	2	3336	1668	1.1523	0.387645	
Dirad:Vigoria	2	14673	7337	5.0684	0.062710	.
N_Gemme:Defogl:Dirad	1	2345	2345	1.6202	0.259035	
Residuals	5	7238	1448			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: 0.907
 IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Ca_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	35458	35458	2.2979	0.18999
IASMA_2	1	158497	158497	10.2715	0.02386 *
IASMA_3	1	819	819	0.0531	0.82692
N_Gemme	1	2092	2092	0.1356	0.72779
Defogl	1	7523	7523	0.4876	0.51615
Dirad	1	7376	7376	0.4780	0.52011
Vigoria	2	15597	7799	0.5054	0.63110
N_Gemme:Defogl	1	6437	6437	0.4172	0.54683
N_Gemme:Dirad	1	10709	10709	0.6940	0.44277
Defogl:Dirad	1	73362	73362	4.7543	0.08107 .
N_Gemme:Vigoria	2	40125	20062	1.3002	0.35103
Defogl:Vigoria	2	6641	3320	0.2152	0.81349
Dirad:Vigoria	2	17200	8600	0.5573	0.60463
N_Gemme:Defogl:Dirad	1	4190	4190	0.2715	0.62456
Residuals	5	77154	15431		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: -0.82
 IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	88144	88144	0.3790	0.56506
IASMA_2	1	14980	14980	0.0644	0.80975
IASMA_3	1	279505	279505	1.2019	0.32291
N_Gemme	1	10276	10276	0.0442	0.84180
Defogl	1	261053	261053	1.1226	0.33784
Dirad	1	1063131	1063131	4.5716	0.08552 .
Vigoria	2	150310	75155	0.3232	0.73791
N_Gemme:Defogl	1	8070	8070	0.0347	0.85954
N_Gemme:Dirad	1	7746	7746	0.0333	0.86236
Defogl:Dirad	1	11237	11237	0.0483	0.83471
N_Gemme:Vigoria	2	513901	256950	1.1049	0.40051
Defogl:Vigoria	2	176732	88366	0.3800	0.70206
Dirad:Vigoria	2	194845	97422	0.4189	0.67888
N_Gemme:Defogl:Dirad	1	80023	80023	0.3441	0.58295
Residuals	5	1162757	232551		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s

IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	7.4310e-06	7.4310e-06	0.5482	0.49232
IASMA_2	1	7.6567e-05	7.6567e-05	5.6488	0.06342 .
IASMA_3	1	5.7484e-05	5.7484e-05	4.2410	0.09451 .
N_Gemme	1	1.7069e-05	1.7069e-05	1.2593	0.31277
Defogl	1	5.0562e-05	5.0562e-05	3.7303	0.11129
Dirad	1	5.7062e-05	5.7062e-05	4.2098	0.09543 .
Vigoria	2	1.5798e-04	7.8992e-05	5.8277	0.04938 *
N_Gemme:Defogl	1	7.1900e-07	7.1900e-07	0.0530	0.82703
N_Gemme:Dirad	1	2.6138e-05	2.6138e-05	1.9284	0.22362
Defogl:Dirad	1	4.2200e-07	4.2200e-07	0.0312	0.86680
N_Gemme:Vigoria	2	2.9336e-05	1.4668e-05	1.0821	0.40691
Defogl:Vigoria	2	1.3165e-05	6.5830e-06	0.4856	0.64159
Dirad:Vigoria	2	8.2700e-06	4.1350e-06	0.3051	0.74987
N_Gemme:Defogl:Dirad	1	1.2879e-05	1.2879e-05	0.9502	0.37445
Residuals	5	6.7772e-05	1.3554e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.4021	0.4021	0.5802	0.48061
IASMA_2	1	3.9029	3.9029	5.6321	0.06370 .
IASMA_3	1	2.6032	2.6032	3.7565	0.11033
N_Gemme	1	0.7862	0.7862	1.1346	0.33551
Defogl	1	2.5840	2.5840	3.7288	0.11135
Dirad	1	2.7489	2.7489	3.9667	0.10303
Vigoria	2	7.6522	3.8261	5.5212	0.05423 .
N_Gemme:Defogl	1	0.0210	0.0210	0.0303	0.86870
N_Gemme:Dirad	1	1.3977	1.3977	2.0169	0.21479
Defogl:Dirad	1	0.0008	0.0008	0.0012	0.97352
N_Gemme:Vigoria	2	1.4165	0.7082	1.0220	0.42449
Defogl:Vigoria	2	0.6930	0.3465	0.5000	0.63392
Dirad:Vigoria	2	0.3893	0.1946	0.2809	0.76630
N_Gemme:Defogl:Dirad	1	0.7114	0.7114	1.0266	0.35746
Residuals	5	3.4649	0.6930		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.0000900	0.0000900	0.0572	0.820470
IASMA_2	1	0.0020778	0.0020778	1.3207	0.302458
IASMA_3	1	0.0041532	0.0041532	2.6399	0.165141
N_Gemme	1	0.0051821	0.0051821	3.2938	0.129256
Defogl	1	0.0273458	0.0273458	17.3816	0.008746 **
Dirad	1	0.0295552	0.0295552	18.7859	0.007469 **
Vigoria	2	0.0300488	0.0150244	9.5498	0.019606 *
N_Gemme:Defogl	1	0.0058820	0.0058820	3.7387	0.110981
N_Gemme:Dirad	1	0.0023919	0.0023919	1.5204	0.272371
Defogl:Dirad	1	0.0009944	0.0009944	0.6320	0.462657
N_Gemme:Vigoria	2	0.0065564	0.0032782	2.0837	0.219692
Defogl:Vigoria	2	0.0284721	0.0142361	9.0487	0.021803 *
Dirad:Vigoria	2	0.0118076	0.0059038	3.7526	0.101089
N_Gemme:Defogl:Dirad	1	0.0046005	0.0046005	2.9242	0.147954
Residuals	5	0.0078663	0.0015733		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Ca_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.04782	0.047825	3.1557	0.13582
IASMA_2	1	0.05358	0.053582	3.5356	0.11883
IASMA_3	1	0.12208	0.122077	8.0553	0.03632 *
N_Gemme	1	0.09080	0.090799	5.9913	0.05810 .
Defogl	1	0.01567	0.015666	1.0337	0.35594
Dirad	1	0.00007	0.000073	0.0048	0.94721
Vigoria	2	0.37629	0.188145	12.4147	0.01150 *
N_Gemme:Defogl	1	0.13618	0.136184	8.9861	0.03018 *
N_Gemme:Dirad	1	0.00044	0.000443	0.0292	0.87096
Defogl:Dirad	1	0.00004	0.000042	0.0027	0.96027
N_Gemme:Vigoria	2	0.07270	0.036350	2.3985	0.18607
Defogl:Vigoria	2	0.00535	0.002677	0.1766	0.84309
Dirad:Vigoria	2	0.07044	0.035221	2.3241	0.19334
N_Gemme:Defogl:Dirad	1	0.00733	0.007331	0.4837	0.51774
Residuals	5	0.07577	0.015155		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: -0.786

AcTartB

Analysis of Variance Table

Response: Ca_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.008343	0.008343	0.7515	0.425641
IASMA_2	1	0.155954	0.155954	14.0469	0.013322 *
IASMA_3	1	0.043779	0.043779	3.9432	0.103811
N_Gemme	1	0.000235	0.000235	0.0211	0.890114
Defogl	1	0.000354	0.000354	0.0318	0.865375
Dirad	1	0.069054	0.069054	6.2197	0.054896 .
Vigoria	2	0.181915	0.090958	8.1926	0.026433 *
N_Gemme:Defogl	1	0.203955	0.203955	18.3704	0.007818 **
N_Gemme:Dirad	1	0.006816	0.006816	0.6139	0.468789
Defogl:Dirad	1	0.018264	0.018264	1.6450	0.255874
N_Gemme:Vigoria	2	0.030753	0.015376	1.3850	0.332188
Defogl:Vigoria	2	0.115948	0.057974	5.2218	0.059643 .
Dirad:Vigoria	2	0.072243	0.036122	3.2535	0.124458
N_Gemme:Defogl:Dirad	1	0.073422	0.073422	6.6132	0.049938 *
Residuals	5	0.055512	0.011102		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.859
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Ca_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.11818	0.11818	48.0064	0.000961 ***
IASMA_2	1	0.68304	0.68304	277.4613	1.425e-05 ***
IASMA_3	1	0.00901	0.00901	3.6602	0.113922
N_Gemme	1	0.04273	0.04273	17.3561	0.008772 **
Defogl	1	0.00172	0.00172	0.6980	0.441533
Dirad	1	0.00071	0.00071	0.2871	0.615031
Vigoria	2	1.01855	0.50927	206.8745	1.558e-05 ***
N_Gemme:Defogl	1	0.02271	0.02271	9.2234	0.028849 *
N_Gemme:Dirad	1	0.00595	0.00595	2.4164	0.180786
Defogl:Dirad	1	0.02467	0.02467	10.0197	0.024943 *
N_Gemme:Vigoria	2	0.12939	0.06470	26.2802	0.002224 **
Defogl:Vigoria	2	0.09399	0.04700	19.0904	0.004562 **
Dirad:Vigoria	2	0.02930	0.01465	5.9515	0.047590 *
N_Gemme:Defogl:Dirad	1	0.02647	0.02647	10.7513	0.021981 *
Residuals	5	0.01231	0.00246		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.952
IASMA_2: 0.991
IASMA_3: n.s

KB

Analysis of Variance Table

Response: Ca_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IASMA_1	1	0.003877	0.003877	0.6841	0.445837
IASMA_2	1	0.004844	0.004844	0.8548	0.397621
IASMA_3	1	0.000127	0.000127	0.0223	0.887017
N_Gemme	1	0.011845	0.011845	2.0902	0.207879
Defogl	1	0.156221	0.156221	27.5679	0.003325 **
Dirad	1	0.173751	0.173751	30.6613	0.002636 **
Vigoria	2	0.245982	0.122991	21.7039	0.003429 **
N_Gemme:Defogl	1	0.013266	0.013266	2.3410	0.186559
N_Gemme:Dirad	1	0.023493	0.023493	4.1457	0.097353 .
Defogl:Dirad	1	0.001149	0.001149	0.2027	0.671388
N_Gemme:Vigoria	2	0.043328	0.021664	3.8230	0.098297 .
Defogl:Vigoria	2	0.151812	0.075906	13.3949	0.009811 **
Dirad:Vigoria	2	0.038429	0.019214	3.3907	0.117337
N_Gemme:Defogl:Dirad	1	0.060328	0.060328	10.6459	0.022376 *
Residuals	5	0.028334	0.005667		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Ca_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	477.93	477.93	31.2503	0.0025276 **
IASMA_2	1	39.06	39.06	2.5542	0.1708922
IASMA_3	1	88.35	88.35	5.7772	0.0613493 .
N_Gemme	1	84.89	84.89	5.5510	0.0650695 .
Defogl	1	126.18	126.18	8.2504	0.0348988 *
Dirad	1	151.15	151.15	9.8831	0.0255585 *
Vigoria	2	1875.57	937.78	61.3191	0.0003037 ***
N_Gemme:Defogl	1	161.24	161.24	10.5428	0.0227723 *
N_Gemme:Dirad	1	189.53	189.53	12.3928	0.0169130 *
Defogl:Dirad	1	16.01	16.01	1.0468	0.3531669
N_Gemme:Vigoria	2	54.11	27.06	1.7691	0.2624212
Defogl:Vigoria	2	215.08	107.54	7.0319	0.0352292 *
Dirad:Vigoria	2	20.91	10.46	0.6836	0.5464437
N_Gemme:Defogl:Dirad	1	179.74	179.74	11.7530	0.0186707 *
Residuals	5	76.47	15.29		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.928
IASMA_2: n.s
IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Ca_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	7814	7813.6	0.7054	0.4393
IASMA_2	1	8312	8311.8	0.7504	0.4259
IASMA_3	1	1633	1632.6	0.1474	0.7168

N_Gemme	1	44	43.8	0.0040	0.9523
Defogl	1	2024	2023.6	0.1827	0.6868
Dirad	1	1140	1140.4	0.1030	0.7613
Vigoria	2	18195	9097.7	0.8214	0.4915
N_Gemme:Defogl	1	2081	2080.9	0.1879	0.6828
N_Gemme:Dirad	1	666	665.8	0.0601	0.8161
Defogl:Dirad	1	2260	2259.7	0.2040	0.6704
N_Gemme:Vigoria	2	15883	7941.3	0.7170	0.5324
Defogl:Vigoria	2	367	183.4	0.0166	0.9836
Dirad:Vigoria	2	4431	2215.4	0.2000	0.8250
N_Gemme:Defogl:Dirad	1	1533	1533.1	0.1384	0.7251
Residuals	5	55380	11076.1		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	14757	14757	1.5891	0.26308
IASMA_2	1	39720	39720	4.2772	0.09346 .
IASMA_3	1	14010	14010	1.5086	0.27401
N_Gemme	1	4610	4610	0.4964	0.51253
Defogl	1	5264	5264	0.5669	0.48542
Dirad	1	132183	132183	14.2340	0.01299 *
Vigoria	2	75291	37645	4.0538	0.08987 .
N_Gemme:Defogl	1	22544	22544	2.4276	0.17995
N_Gemme:Dirad	1	1322	1322	0.1423	0.72145
Defogl:Dirad	1	231	231	0.0249	0.88086
N_Gemme:Vigoria	2	9120	4560	0.4910	0.63870
Defogl:Vigoria	2	8548	4274	0.4602	0.65545
Dirad:Vigoria	2	34221	17110	1.8425	0.25147
N_Gemme:Defogl:Dirad	1	32348	32348	3.4833	0.12098
Residuals	5	46432	9286		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	6266	6265.8	0.5939	0.4758
IASMA_2	1	1292	1291.8	0.1224	0.7407
IASMA_3	1	25007	25006.9	2.3701	0.1843
N_Gemme	1	2	2.5	0.0002	0.9884
Defogl	1	29018	29018.0	2.7502	0.1581
Dirad	1	2441	2441.1	0.2314	0.6508
Vigoria	2	26794	13396.8	1.2697	0.3582
N_Gemme:Defogl	1	2262	2261.7	0.2144	0.6628

N_Gemme:Dirad	1	288	287.7	0.0273	0.8753
Defogl:Dirad	1	2411	2411.1	0.2285	0.6528
N_Gemme:Vigoria	2	24994	12497.0	1.1844	0.3792
Defogl:Vigoria	2	3198	1599.2	0.1516	0.8632
Dirad:Vigoria	2	9946	4972.9	0.4713	0.6493
N_Gemme:Defogl:Dirad	1	12113	12113.3	1.1481	0.3329
Residuals	5	52755	10551.1		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	2615	2615	0.0765	0.7932
IASMA_2	1	898	898	0.0263	0.8776
IASMA_3	1	6158	6158	0.1801	0.6889
N_Gemme	1	10757	10757	0.3146	0.5990
Defogl	1	54070	54070	1.5816	0.2641
Dirad	1	139457	139457	4.0792	0.0994 .
Vigoria	2	3626	1813	0.0530	0.9489
N_Gemme:Defogl	1	23933	23933	0.7000	0.4409
N_Gemme:Dirad	1	175	175	0.0051	0.9457
Defogl:Dirad	1	1931	1931	0.0565	0.8216
N_Gemme:Vigoria	2	45619	22809	0.6672	0.5536
Defogl:Vigoria	2	17709	8855	0.2590	0.7816
Dirad:Vigoria	2	48034	24017	0.7025	0.5384
N_Gemme:Defogl:Dirad	1	67624	67624	1.9780	0.2186
Residuals	5	170937	34187		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

CACCIAGRANDE 2009M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Ca_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1644092	1644092	70.0495	4.241e-06 ***
IBIMET_2	1	679978	679978	28.9717	0.0002225 ***
IBIMET_3	1	4672	4672	0.1991	0.6641256
N_Gemme	1	278130	278130	11.8502	0.0055012 **
Defogl	1	83825	83825	3.5715	0.0854047 .
Vigoria	2	59513	29757	1.2678	0.3195373
N_Gemme:Defogl	1	133	133	0.0057	0.9413801
N_Gemme:Vigoria	2	335315	167658	7.1433	0.0102743 *
Defogl:Vigoria	2	199057	99529	4.2406	0.0431296 *
Residuals	11	258175	23470		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.93
IBIMET_2: -0.851
IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Ca_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	8.091e-04	8.091e-04	149.0682	9.736e-08	***
IBIMET_2	1	2.658e-05	2.658e-05	4.8978	0.04895	*
IBIMET_3	1	8.770e-06	8.770e-06	1.6159	0.22989	
N_Gemme	1	1.428e-05	1.428e-05	2.6309	0.13309	
Defogl	1	2.391e-05	2.391e-05	4.4049	0.05973	.
Vigoria	2	1.301e-05	6.500e-06	1.1981	0.33827	
N_Gemme:Defogl	1	1.800e-06	1.800e-06	0.3313	0.57650	
N_Gemme:Vigoria	2	6.770e-06	3.380e-06	0.6235	0.55398	
Defogl:Vigoria	2	3.140e-06	1.570e-06	0.2892	0.75438	
Residuals	11	5.970e-05	5.430e-06			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.965
IBIMET_2: -0.555
IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Ca_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	38.784	38.784	146.8856	1.050e-07	***
IBIMET_2	1	1.349	1.349	5.1077	0.04509	*
IBIMET_3	1	0.401	0.401	1.5182	0.24358	
N_Gemme	1	0.869	0.869	3.2914	0.09698	.
Defogl	1	1.212	1.212	4.5919	0.05533	.
Vigoria	2	0.490	0.245	0.9270	0.42457	
N_Gemme:Defogl	1	0.119	0.119	0.4495	0.51640	
N_Gemme:Vigoria	2	0.393	0.197	0.7449	0.49729	
Defogl:Vigoria	2	0.134	0.067	0.2538	0.78029	
Residuals	11	2.904	0.264			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.965
IBIMET_2: -0.563
IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Ca_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IBIMET_1	1	0.133854	0.133854	65.7040	5.764e-06	***
IBIMET_2	1	0.003580	0.003580	1.7575	0.211813	
IBIMET_3	1	0.001367	0.001367	0.6712	0.430022	
N_Gemme	1	0.027705	0.027705	13.5992	0.003577	**
Defogl	1	0.000065	0.000065	0.0317	0.861904	
Vigoria	2	0.003717	0.001858	0.9122	0.430000	
N_Gemme:Defogl	1	0.004988	0.004988	2.4486	0.145922	
N_Gemme:Vigoria	2	0.001340	0.000670	0.3290	0.726518	
Defogl:Vigoria	2	0.008358	0.004179	2.0513	0.174930	
Residuals	11	0.022409	0.002037			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.926
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTitola

Analysis of Variance Table

Response: Ca_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.59692	0.59692	22.5671	0.0005992	***
IBIMET_2	1	0.05574	0.05574	2.1071	0.1745297	
IBIMET_3	1	0.10262	0.10262	3.8795	0.0745767	.
N_Gemme	1	0.00087	0.00087	0.0328	0.8596008	
Defogl	1	0.06004	0.06004	2.2699	0.1600720	
Vigoria	2	0.64566	0.32283	12.2048	0.0016125	**
N_Gemme:Defogl	1	0.24682	0.24682	9.3312	0.0109551	*
N_Gemme:Vigoria	2	0.08494	0.04247	1.6056	0.2444591	
Defogl:Vigoria	2	0.00200	0.00100	0.0378	0.9630179	
Residuals	11	0.29096	0.02645			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.82
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTarta

Analysis of Variance Table

Response: Ca_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.20027	0.200268	5.0137	0.04677	*
IBIMET_2	1	0.00643	0.006430	0.1610	0.69594	
IBIMET_3	1	0.01374	0.013740	0.3440	0.56938	
N_Gemme	1	0.01052	0.010517	0.2633	0.61801	
Defogl	1	0.00021	0.000205	0.0051	0.94415	
Vigoria	2	0.04275	0.021374	0.5351	0.60012	
N_Gemme:Defogl	1	0.20222	0.202223	5.0627	0.04589	*
N_Gemme:Vigoria	2	0.00979	0.004897	0.1226	0.88582	
Defogl:Vigoria	2	0.02739	0.013693	0.3428	0.71710	
Residuals	11	0.43939	0.039944			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.56
 IBIMET_2: n.s
 IBIMET_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Ca_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1.21325	1.21325	18.1167	0.001351 **
IBIMET_2	1	0.08253	0.08253	1.2324	0.290621
IBIMET_3	1	0.20844	0.20844	3.1125	0.105405
N_Gemme	1	0.00459	0.00459	0.0685	0.798314
Defogl	1	0.14959	0.14959	2.2337	0.163150
Vigoria	2	0.12573	0.06287	0.9387	0.420332
N_Gemme:Defogl	1	0.05211	0.05211	0.7781	0.396594
N_Gemme:Vigoria	2	0.07242	0.03621	0.5407	0.597049
Defogl:Vigoria	2	0.00860	0.00430	0.0642	0.938187
Residuals	11	0.73665	0.06697		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.789
 IBIMET_2: n.s
 IBIMET_3: n.s

KA

Analysis of Variance Table

Response: Ca_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.286219	0.286219	36.9551	7.971e-05 ***
IBIMET_2	1	0.000091	0.000091	0.0117	0.915856
IBIMET_3	1	0.000083	0.000083	0.0107	0.919580
N_Gemme	1	0.117285	0.117285	15.1433	0.002513 **
Defogl	1	0.002078	0.002078	0.2683	0.614742
Vigoria	2	0.083829	0.041914	5.4117	0.023098 *
N_Gemme:Defogl	1	0.005628	0.005628	0.7267	0.412140
N_Gemme:Vigoria	2	0.007372	0.003686	0.4759	0.633532
Defogl:Vigoria	2	0.006166	0.003083	0.3980	0.680932
Residuals	11	0.085196	0.007745		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.878
 IBIMET_2: n.s
 IBIMET_3: n.s

ApaA

Analysis of Variance Table

Response: Ca_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	547.78	547.78	2.6426	0.13231
IBIMET_2	1	777.22	777.22	3.7495	0.07893 .

IBIMET_3	1	50.63	50.63	0.2443	0.63087	
N_Gemme	1	1104.00	1104.00	5.3260	0.04146	*
Defogl	1	281.24	281.24	1.3568	0.26873	
Vigoria	2	35.51	17.76	0.0857	0.91851	
N_Gemme:Defogl	1	92.05	92.05	0.4441	0.51889	
N_Gemme:Vigoria	2	42.19	21.10	0.1018	0.90408	
Defogl:Vigoria	2	598.24	299.12	1.4430	0.27763	
Residuals	11	2280.13	207.28			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: n.s
 IBIMET_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Ca_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	68145	68145	49.2935	2.208e-05	***
IBIMET_2	1	7025	7025	5.0816	0.04555	*
IBIMET_3	1	294	294	0.2128	0.65360	
N_Gemme	1	275	275	0.1991	0.66407	
Defogl	1	29	29	0.0210	0.88733	
Vigoria	2	7287	3643	2.6355	0.11611	
N_Gemme:Defogl	1	888	888	0.6424	0.43983	
N_Gemme:Vigoria	2	9733	4867	3.5204	0.06580	.
Defogl:Vigoria	2	957	479	0.3462	0.71483	
Residuals	11	15207	1382			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.904
 IBIMET_2: 0.562
 IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Ca_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	125.5	125.5	0.1200	0.73557	
IBIMET_2	1	7634.3	7634.3	7.2999	0.02059	*
IBIMET_3	1	483.0	483.0	0.4619	0.51080	
N_Gemme	1	189.9	189.9	0.1816	0.67824	
Defogl	1	2567.1	2567.1	2.4546	0.14548	
Vigoria	2	1893.2	946.6	0.9052	0.43260	
N_Gemme:Defogl	1	108.7	108.7	0.1039	0.75324	
N_Gemme:Vigoria	2	5880.6	2940.3	2.8115	0.10322	
Defogl:Vigoria	2	2461.8	1230.9	1.1770	0.34419	
Residuals	11	11504.0	1045.8			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: 0.632

IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	346345	346345	43.6439	3.831e-05	***
IBIMET_2	1	18954	18954	2.3884	0.15051	
IBIMET_3	1	761	761	0.0958	0.76266	
N_Gemme	1	997	997	0.1256	0.72969	
Defogl	1	414	414	0.0522	0.82345	
Vigoria	2	50796	25398	3.2005	0.08026	.
N_Gemme:Defogl	1	2525	2525	0.3182	0.58401	
N_Gemme:Vigoria	2	48267	24133	3.0411	0.08885	.
Defogl:Vigoria	2	299	150	0.0189	0.98136	
Residuals	11	87293	7936			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.894

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Ca_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	184	184.3	0.0189	0.8932	
IBIMET_2	1	5274	5274.3	0.5406	0.4776	
IBIMET_3	1	204	203.6	0.0209	0.8878	
N_Gemme	1	5059	5058.8	0.5185	0.4865	
Defogl	1	28359	28359.2	2.9065	0.1163	
Vigoria	2	24985	12492.4	1.2803	0.3163	
N_Gemme:Defogl	1	5053	5053.2	0.5179	0.4867	
N_Gemme:Vigoria	2	28776	14387.9	1.4746	0.2708	
Defogl:Vigoria	2	4493	2246.4	0.2302	0.7981	
Residuals	11	107328	9757.1			

IBIMET_1: n.s

IBIMET_2: n.s

IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	659959	659959	7.8383	0.017284	*
IBIMET_2	1	542652	542652	6.4450	0.027543	*
IBIMET_3	1	9888	9888	0.1174	0.738288	
N_Gemme	1	930086	930086	11.0466	0.006787	**
Defogl	1	14313	14313	0.1700	0.688039	
Vigoria	2	162955	81477	0.9677	0.410082	

N_Gemme:Defogl	1	50	50	0.0006	0.981087
N_Gemme:Vigoria	2	49234	24617	0.2924	0.752116
Defogl:Vigoria	2	186128	93064	1.1053	0.365242
Residuals	11	926166	84197		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.645
 IBIMET_2: -0.608
 IBIMET_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.00080527	0.00080527	54.9185	1.342e-05	***
IBIMET_2	1	0.00000511	0.00000511	0.3482	0.56707	
IBIMET_3	1	0.00003082	0.00003082	2.1017	0.17504	
N_Gemme	1	0.00005279	0.00005279	3.6001	0.08432	.
Defogl	1	0.00000020	0.00000020	0.0139	0.90843	
Vigoria	2	0.00002236	0.00001118	0.7626	0.48962	
N_Gemme:Defogl	1	0.00004979	0.00004979	3.3956	0.09246	.
N_Gemme:Vigoria	2	0.00000636	0.00000318	0.2170	0.80826	
Defogl:Vigoria	2	0.00001891	0.00000945	0.6448	0.54351	
Residuals	11	0.00016129	0.00001466			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.913
 IBIMET_2: n.s
 IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	39.093	39.093	55.3699	1.292e-05	***
IBIMET_2	1	0.250	0.250	0.3543	0.56375	
IBIMET_3	1	1.427	1.427	2.0213	0.18283	
N_Gemme	1	2.433	2.433	3.4463	0.09036	.
Defogl	1	0.012	0.012	0.0172	0.89797	
Vigoria	2	1.033	0.516	0.7313	0.50329	
N_Gemme:Defogl	1	2.619	2.619	3.7092	0.08034	.
N_Gemme:Vigoria	2	0.395	0.198	0.2801	0.76096	
Defogl:Vigoria	2	0.849	0.424	0.6010	0.56530	
Residuals	11	7.766	0.706			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.913
 IBIMET_2: n.s
 IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.180918	0.180918	30.1390	0.0001891	***
IBIMET_2	1	0.015197	0.015197	2.5317	0.1398898	
IBIMET_3	1	0.001991	0.001991	0.3317	0.5762415	
N_Gemme	1	0.034221	0.034221	5.7009	0.0360109	*
Defogl	1	0.005654	0.005654	0.9419	0.3526451	
Vigoria	2	0.008916	0.004458	0.7427	0.4982665	
N_Gemme:Defogl	1	0.000482	0.000482	0.0802	0.7822227	
N_Gemme:Vigoria	2	0.000362	0.000181	0.0301	0.9704172	
Defogl:Vigoria	2	0.003295	0.001647	0.2744	0.7650629	
Residuals	11	0.066031	0.006003			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.856
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTitolB

Analysis of Variance Table

Response: Ca_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.65697	0.65697	23.6357	0.0005013	***
IBIMET_2	1	0.00190	0.00190	0.0682	0.7988184	
IBIMET_3	1	0.07358	0.07358	2.6472	0.1320094	
N_Gemme	1	0.13398	0.13398	4.8201	0.0504836	.
Defogl	1	0.01256	0.01256	0.4520	0.5152758	
Vigoria	2	0.04158	0.02079	0.7480	0.4959438	
N_Gemme:Defogl	1	0.10504	0.10504	3.7792	0.0779074	.
N_Gemme:Vigoria	2	0.02481	0.01241	0.4463	0.6510807	
Defogl:Vigoria	2	0.01716	0.00858	0.3087	0.7405409	
Residuals	11	0.30575	0.02780			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.826
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTartB

Analysis of Variance Table

Response: Ca_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.005023	0.005023	0.2063	0.65852	
IBIMET_2	1	0.142445	0.142445	5.8498	0.03409	*
IBIMET_3	1	0.085119	0.085119	3.4956	0.08837	.
N_Gemme	1	0.047527	0.047527	1.9518	0.18994	
Defogl	1	0.023888	0.023888	0.9810	0.34323	
Vigoria	2	0.049570	0.024785	1.0179	0.39303	
N_Gemme:Defogl	1	0.076268	0.076268	3.1321	0.10444	
N_Gemme:Vigoria	2	0.020941	0.010471	0.4300	0.66099	

Defogl:Vigoria 2 0.018046 0.009023 0.3705 0.69866
 Residuals 11 0.267855 0.024350

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: -0.589
 IBIMET_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Ca_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.45677	0.45677	11.3564	0.006253	**
IBIMET_2	1	0.16121	0.16121	4.0082	0.070557	.
IBIMET_3	1	0.25340	0.25340	6.3003	0.028985	*
N_Gemme	1	0.00000	0.00000	0.0000	0.997308	
Defogl	1	0.00580	0.00580	0.1441	0.711439	
Vigoria	2	0.06754	0.03377	0.8397	0.457753	
N_Gemme:Defogl	1	0.05830	0.05830	1.4496	0.253865	
N_Gemme:Vigoria	2	0.00038	0.00019	0.0047	0.995297	
Defogl:Vigoria	2	0.05348	0.02674	0.6648	0.533853	
Residuals	11	0.44243	0.04022			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.713
 IBIMET_2: n.s
 IBIMET_3: 0.603

KB

Analysis of Variance Table

Response: Ca_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.35880	0.35880	31.7394	0.0001524	***
IBIMET_2	1	0.04377	0.04377	3.8716	0.0748336	.
IBIMET_3	1	0.14054	0.14054	12.4326	0.0047482	**
N_Gemme	1	0.14938	0.14938	13.2139	0.0039216	**
Defogl	1	0.02881	0.02881	2.5489	0.1386750	
Vigoria	2	0.00763	0.00381	0.3375	0.7207113	
N_Gemme:Defogl	1	0.03334	0.03334	2.9489	0.1139229	
N_Gemme:Vigoria	2	0.00370	0.00185	0.1637	0.8510542	
Defogl:Vigoria	2	0.01907	0.00954	0.8436	0.4561742	
Residuals	11	0.12435	0.01130			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.862
 IBIMET_2: n.s
 IBIMET_3: 0.728

ApaB

Analysis of Variance Table

Response: Ca_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	479.44	479.44	5.9122	0.0333214	*
IBIMET_2	1	628.72	628.72	7.7530	0.0177631	*
IBIMET_3	1	470.27	470.27	5.7990	0.0347299	*
N_Gemme	1	1635.53	1635.53	20.1682	0.0009148	***
Defogl	1	775.92	775.92	9.5681	0.0102270	*
Vigoria	2	471.93	235.97	2.9098	0.0967569	.
N_Gemme:Defogl	1	5.19	5.19	0.0639	0.8050337	
N_Gemme:Vigoria	2	188.86	94.43	1.1644	0.3477787	
Defogl:Vigoria	2	370.84	185.42	2.2865	0.1477840	
Residuals	11	892.04	81.09			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.591
IBIMET_2: 0.643
IBIMET_3: -0.588

Antoctot1B

Analysis of Variance Table

Response: Ca_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	34716	34716	9.4032	0.01073	*
IBIMET_2	1	989	989	0.2680	0.61493	
IBIMET_3	1	12202	12202	3.3051	0.09637	.
N_Gemme	1	3918	3918	1.0614	0.32503	
Defogl	1	3807	3807	1.0312	0.33169	
Vigoria	2	1073	536	0.1453	0.86640	
N_Gemme:Defogl	1	3338	3338	0.9042	0.36209	
N_Gemme:Vigoria	2	5075	2537	0.6873	0.52330	
Defogl:Vigoria	2	2433	1217	0.3295	0.72613	
Residuals	11	40611	3692			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.679
IBIMET_2: n.s
IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	7282.7	7282.7	3.7237	0.07983	.
IBIMET_2	1	9778.7	9778.7	4.9999	0.04703	*
IBIMET_3	1	8529.6	8529.6	4.3612	0.06081	.
N_Gemme	1	2201.5	2201.5	1.1256	0.31145	
Defogl	1	7626.3	7626.3	3.8994	0.07394	.
Vigoria	2	22087.5	11043.7	5.6467	0.02054	*
N_Gemme:Defogl	1	815.7	815.7	0.4171	0.53165	
N_Gemme:Vigoria	2	731.8	365.9	0.1871	0.83196	
Defogl:Vigoria	2	641.9	321.0	0.1641	0.85069	
Residuals	11	21513.5	1955.8			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.559
IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	102151	102151	6.3621	0.02836 *
IBIMET_2	1	182	182	0.0113	0.91718
IBIMET_3	1	44512	44512	2.7723	0.12410
N_Gemme	1	26336	26336	1.6403	0.22663
Defogl	1	7300	7300	0.4547	0.51404
Vigoria	2	1561	780	0.0486	0.95277
N_Gemme:Defogl	1	8949	8949	0.5574	0.47098
N_Gemme:Vigoria	2	10972	5486	0.3417	0.71787
Defogl:Vigoria	2	16100	8050	0.5013	0.61891
Residuals	11	176619	16056		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.605
IBIMET_2: n.s
IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	290954	290954	53.3339	1.537e-05 ***
IBIMET_2	1	5713	5713	1.0472	0.32814
IBIMET_3	1	207	207	0.0380	0.84909
N_Gemme	1	1313	1313	0.2406	0.63341
Defogl	1	39079	39079	7.1635	0.02154 *
Vigoria	2	15911	7955	1.4583	0.27430
N_Gemme:Defogl	1	20411	20411	3.7415	0.07921 .
N_Gemme:Vigoria	2	5799	2900	0.5315	0.60206
Defogl:Vigoria	2	1824	912	0.1672	0.84812
Residuals	11	60009	5455		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.911
IBIMET_2: n.s
IBIMET_3: n.s

CACCIAGRANDE 2009M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Ca_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	2014601	2014601	49.6425	2.138e-05	***
DIPROVE_2	1	13945	13945	0.3436	0.56958	
DIPROVE_3	1	30063	30063	0.7408	0.40777	
N_Gemme	1	304247	304247	7.4971	0.01930	*
Defogl	1	16504	16504	0.4067	0.53671	
Vigoria	2	284135	142067	3.5007	0.06660	.
N_Gemme:Defogl	1	15745	15745	0.3880	0.54606	
N_Gemme:Vigoria	2	241964	120982	2.9812	0.09236	.
Defogl:Vigoria	2	175283	87642	2.1596	0.16176	
Residuals	11	446403	40582			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.905

DIPROVE_2: n.s

DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Ca_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.00032350	0.00032350	69.5621	4.386e-06	***
DIPROVE_2	1	0.00000002	0.00000002	0.0038	0.9518485	
DIPROVE_3	1	0.00008648	0.00008648	18.5956	0.0012305	**
N_Gemme	1	0.00000047	0.00000047	0.1005	0.7571585	
Defogl	1	0.00013487	0.00013487	29.0004	0.0002216	***
Vigoria	2	0.00036649	0.00018324	39.4033	9.649e-06	***
N_Gemme:Defogl	1	0.00000003	0.00000003	0.0063	0.9380957	
N_Gemme:Vigoria	2	0.00000264	0.00000132	0.2843	0.7578763	
Defogl:Vigoria	2	0.00000141	0.00000071	0.1518	0.8609312	
Residuals	11	0.00005116	0.00000465			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.929

DIPROVE_2: n.s

DIPROVE_3: 0.793

BrixA

Analysis of Variance Table

Response: Ca_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	15.2992	15.2992	67.2018	5.175e-06	***
DIPROVE_2	1	0.0000	0.0000	0.0000	0.9968561	
DIPROVE_3	1	4.1521	4.1521	18.2383	0.0013193	**
N_Gemme	1	0.0602	0.0602	0.2643	0.6173613	
Defogl	1	6.4666	6.4666	28.4044	0.0002412	***
Vigoria	2	17.9329	8.9664	39.3850	9.670e-06	***
N_Gemme:Defogl	1	0.0031	0.0031	0.0138	0.9086263	
N_Gemme:Vigoria	2	0.1545	0.0773	0.3394	0.7194202	
Defogl:Vigoria	2	0.0817	0.0409	0.1795	0.8380572	
Residuals	11	2.5043	0.2277			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.927
DIPROVE_2: n.s
DIPROVE_3: 0.79

PhA

Analysis of Variance Table

Response: Ca_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.042446	0.042446	27.0696	0.0002932	***
DIPROVE_2	1	0.000277	0.000277	0.1765	0.6824951	
DIPROVE_3	1	0.024239	0.024239	15.4582	0.0023453	**
N_Gemme	1	0.017512	0.017512	11.1679	0.0065719	**
Defogl	1	0.040955	0.040955	26.1186	0.0003384	***
Vigoria	2	0.054618	0.027309	17.4161	0.0003901	***
N_Gemme:Defogl	1	0.006892	0.006892	4.3954	0.0599611	.
N_Gemme:Vigoria	2	0.000179	0.000090	0.0572	0.9447250	
Defogl:Vigoria	2	0.003016	0.001508	0.9618	0.4121398	
Residuals	11	0.017248	0.001568			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.843
DIPROVE_2: n.s
DIPROVE_3: 0.764

AcTitola

Analysis of Variance Table

Response: Ca_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.43914	0.43914	8.8729	0.01255	*
DIPROVE_2	1	0.46799	0.46799	9.4557	0.01056	*
DIPROVE_3	1	0.03018	0.03018	0.6098	0.45134	
N_Gemme	1	0.00296	0.00296	0.0597	0.81144	
Defogl	1	0.16699	0.16699	3.3741	0.09337	.
Vigoria	2	0.29354	0.14677	2.9655	0.09331	.
N_Gemme:Defogl	1	0.06335	0.06335	1.2801	0.28195	
N_Gemme:Vigoria	2	0.01030	0.00515	0.1041	0.90204	
Defogl:Vigoria	2	0.06769	0.03384	0.6838	0.52490	
Residuals	11	0.54442	0.04949			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.668
DIPROVE_2: -0.68
DIPROVE_3: n.s

AcTarta

Analysis of Variance Table

Response: Ca_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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DIPROVE_1	1	0.19400	0.193999	4.9012	0.04889	*
DIPROVE_2	1	0.01876	0.018757	0.4739	0.50547	
DIPROVE_3	1	0.01585	0.015852	0.4005	0.53977	
N_Gemme	1	0.00430	0.004297	0.1086	0.74799	
Defogl	1	0.01817	0.018166	0.4590	0.51211	
Vigoria	2	0.16802	0.084012	2.1225	0.16613	
N_Gemme:Defogl	1	0.08641	0.086409	2.1831	0.16759	
N_Gemme:Vigoria	2	0.00838	0.004188	0.1058	0.90049	
Defogl:Vigoria	2	0.00341	0.001707	0.0431	0.95794	
Residuals	11	0.43540	0.039582			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.555
DIPROVE_2: n.s
DIPROVE_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Ca_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.50365	0.50365	7.5727	0.01883	*
DIPROVE_2	1	0.23948	0.23948	3.6007	0.08430	.
DIPROVE_3	1	0.19628	0.19628	2.9512	0.11379	
N_Gemme	1	0.00570	0.00570	0.0857	0.77518	
Defogl	1	0.18413	0.18413	2.7686	0.12433	
Vigoria	2	0.59791	0.29895	4.4950	0.03743	*
N_Gemme:Defogl	1	0.04714	0.04714	0.7088	0.41776	
N_Gemme:Vigoria	2	0.02994	0.01497	0.2251	0.80206	
Defogl:Vigoria	2	0.11809	0.05904	0.8878	0.43911	
Residuals	11	0.73159	0.06651			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.639
DIPROVE_2: n.s
DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Ca_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.128806	0.128806	17.6129	0.001494	**
DIPROVE_2	1	0.040602	0.040602	5.5520	0.038064	*
DIPROVE_3	1	0.031837	0.031837	4.3534	0.061011	.
N_Gemme	1	0.092987	0.092987	12.7150	0.004427	**
Defogl	1	0.062263	0.062263	8.5138	0.013991	*
Vigoria	2	0.118535	0.059268	8.1043	0.006867	**
N_Gemme:Defogl	1	0.014206	0.014206	1.9425	0.190917	
N_Gemme:Vigoria	2	0.014588	0.007294	0.9974	0.399885	
Defogl:Vigoria	2	0.009677	0.004839	0.6617	0.535378	
Residuals	11	0.080445	0.007313			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.785
 DIPROVE_2: -0.579
 DIPROVE_3: n.s

ApaA

Analysis of Variance Table

Response: Ca_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1520.15	1520.15	8.1645	0.01559	*
DIPROVE_2	1	396.26	396.26	2.1283	0.17256	
DIPROVE_3	1	124.27	124.27	0.6674	0.43129	
N_Gemme	1	864.29	864.29	4.6420	0.05422	.
Defogl	1	356.04	356.04	1.9123	0.19414	
Vigoria	2	241.93	120.96	0.6497	0.54114	
N_Gemme:Defogl	1	129.65	129.65	0.6963	0.42177	
N_Gemme:Vigoria	2	42.77	21.38	0.1149	0.89255	
Defogl:Vigoria	2	85.56	42.78	0.2298	0.79844	
Residuals	11	2048.08	186.19			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.653
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Ca_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	14606	14605.6	10.7695	0.007312	**
DIPROVE_2	1	3021	3021.1	2.2276	0.163676	
DIPROVE_3	1	8313	8313.2	6.1298	0.030800	*
N_Gemme	1	196	196.2	0.1447	0.710896	
Defogl	1	19013	19013.0	14.0193	0.003242	**
Vigoria	2	41283	20641.3	15.2200	0.000679	***
N_Gemme:Defogl	1	0	0.1	0.0001	0.992573	
N_Gemme:Vigoria	2	6764	3382.1	2.4938	0.127896	
Defogl:Vigoria	2	1726	863.2	0.6365	0.547556	
Residuals	11	14918	1356.2			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.703
 DIPROVE_2: n.s
 DIPROVE_3: -0.598

Poliftot1A

Analysis of Variance Table

Response: Ca_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	3180.5	3180.5	2.7435	0.12587
DIPROVE_2	1	259.4	259.4	0.2238	0.64544

DIPROVE_3	1	235.0	235.0	0.2027	0.66127
N_Gemme	1	751.6	751.6	0.6483	0.43777
Defogl	1	7480.1	7480.1	6.4524	0.02747 *
Vigoria	2	1867.6	933.8	0.8055	0.47157
N_Gemme:Defogl	1	44.8	44.8	0.0386	0.84780
N_Gemme:Vigoria	2	3633.6	1816.8	1.5672	0.25185
Defogl:Vigoria	2	2643.5	1321.7	1.1401	0.35483
Residuals	11	12752.1	1159.3		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	95915	95915	11.6906	0.005732	**
DIPROVE_2	1	15850	15850	1.9318	0.192044	
DIPROVE_3	1	40232	40232	4.9037	0.048839	*
N_Gemme	1	836	836	0.1019	0.755493	
Defogl	1	98459	98459	12.0006	0.005294	**
Vigoria	2	183183	91591	11.1635	0.002250	**
N_Gemme:Defogl	1	248	248	0.0302	0.865246	
N_Gemme:Vigoria	2	30032	15016	1.8302	0.205994	
Defogl:Vigoria	2	1646	823	0.1003	0.905375	
Residuals	11	90250	8205			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.718
DIPROVE_2: n.s
DIPROVE_3: -0.555

Poliftot2A

Analysis of Variance Table

Response: Ca_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	7671	7671	0.7106	0.4172	
DIPROVE_2	1	972	972	0.0900	0.7698	
DIPROVE_3	1	1921	1921	0.1780	0.6812	
N_Gemme	1	5240	5240	0.4854	0.5004	
Defogl	1	33884	33884	3.1388	0.1041	
Vigoria	2	18845	9423	0.8728	0.4448	
N_Gemme:Defogl	1	2998	2998	0.2778	0.6086	
N_Gemme:Vigoria	2	13280	6640	0.6151	0.5582	
Defogl:Vigoria	2	6156	3078	0.2851	0.7573	
Residuals	11	118747	10795			

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	986158	986158	11.6714	0.005760	**
DIPROVE_2	1	12350	12350	0.1462	0.709507	
DIPROVE_3	1	6956	6956	0.0823	0.779498	
N_Gemme	1	1022817	1022817	12.1052	0.005155	**
Defogl	1	1006	1006	0.0119	0.915083	
Vigoria	2	319718	159859	1.8920	0.196706	
N_Gemme:Defogl	1	6530	6530	0.0773	0.786175	
N_Gemme:Vigoria	2	29091	14546	0.1722	0.844076	
Defogl:Vigoria	2	167370	83685	0.9904	0.402245	
Residuals	11	929432	84494			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.718

DIPROVE_2: n.s

DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	2.8986e-04	2.8986e-04	16.9794	0.001699	**
DIPROVE_2	1	6.3048e-05	6.3048e-05	3.6932	0.080908	.
DIPROVE_3	1	3.9055e-05	3.9055e-05	2.2878	0.158583	
N_Gemme	1	2.1920e-05	2.1920e-05	1.2841	0.281239	
Defogl	1	1.8584e-04	1.8584e-04	10.8863	0.007085	**
Vigoria	2	3.0764e-04	1.5382e-04	9.0106	0.004816	**
N_Gemme:Defogl	1	2.0125e-05	2.0125e-05	1.1789	0.300799	
N_Gemme:Vigoria	2	6.6870e-06	3.3440e-06	0.1959	0.824931	
Defogl:Vigoria	2	3.0953e-05	1.5476e-05	0.9066	0.432063	
Residuals	11	1.8778e-04	1.7071e-05			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.779

DIPROVE_2: n.s

DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	14.1455	14.1455	16.9129	0.001722	**
DIPROVE_2	1	3.0628	3.0628	3.6620	0.082035	.
DIPROVE_3	1	1.8630	1.8630	2.2275	0.163690	
N_Gemme	1	0.9810	0.9810	1.1729	0.301980	
Defogl	1	8.8399	8.8399	10.5692	0.007721	**
Vigoria	2	14.8083	7.4041	8.8526	0.005115	**

N_Gemme:Defogl	1	1.1480	1.1480	1.3726	0.266116
N_Gemme:Vigoria	2	0.4445	0.2223	0.2658	0.771410
Defogl:Vigoria	2	1.3851	0.6926	0.8280	0.462395
Residuals	11	9.2001	0.8364		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.778
 DIPROVE_2: n.s
 DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.031388	0.031388	5.5690	0.03782	*
DIPROVE_2	1	0.018848	0.018848	3.3441	0.09466	.
DIPROVE_3	1	0.030966	0.030966	5.4942	0.03890	*
N_Gemme	1	0.019942	0.019942	3.5382	0.08669	.
Defogl	1	0.066716	0.066716	11.8371	0.00552	**
Vigoria	2	0.079167	0.039583	7.0231	0.01083	*
N_Gemme:Defogl	1	0.000677	0.000677	0.1201	0.73547	
N_Gemme:Vigoria	2	0.001478	0.000739	0.1311	0.87848	
Defogl:Vigoria	2	0.005888	0.002944	0.5223	0.60715	
Residuals	11	0.061998	0.005636			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.58
 DIPROVE_2: n.s
 DIPROVE_3: -0.577

AcTitolB

Analysis of Variance Table

Response: Ca_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.21324	0.213242	10.1919	0.008569	**
DIPROVE_2	1	0.27739	0.277395	13.2581	0.003880	**
DIPROVE_3	1	0.00823	0.008233	0.3935	0.543258	
N_Gemme	1	0.10440	0.104401	4.9899	0.047211	*
Defogl	1	0.03863	0.038629	1.8463	0.201427	
Vigoria	2	0.33555	0.167773	8.0187	0.007110	**
N_Gemme:Defogl	1	0.10154	0.101540	4.8531	0.049826	*
N_Gemme:Vigoria	2	0.04706	0.023530	1.1246	0.359427	
Defogl:Vigoria	2	0.01714	0.008569	0.4096	0.673667	
Residuals	11	0.23015	0.020923			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.693
 DIPROVE_2: -0.739
 DIPROVE_3: n.s

AcTartB

Analysis of Variance Table

Response: Ca_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.000058	0.000058	0.0028	0.958509
DIPROVE_2	1	0.031541	0.031541	1.5432	0.239979
DIPROVE_3	1	0.000077	0.000077	0.0038	0.952045
N_Gemme	1	0.016017	0.016017	0.7837	0.394964
Defogl	1	0.012707	0.012707	0.6217	0.447081
Vigoria	2	0.302718	0.151359	7.4053	0.009178 **
N_Gemme:Defogl	1	0.023596	0.023596	1.1544	0.305627
N_Gemme:Vigoria	2	0.057775	0.028887	1.4133	0.284258
Defogl:Vigoria	2	0.067363	0.033682	1.6479	0.236600
Residuals	11	0.224831	0.020439		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Ca_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.16145	0.161454	4.1945	0.06519 .
DIPROVE_2	1	0.03136	0.031359	0.8147	0.38607
DIPROVE_3	1	0.07260	0.072599	1.8861	0.19699
N_Gemme	1	0.01464	0.014643	0.3804	0.54993
Defogl	1	0.16565	0.165651	4.3035	0.06229 .
Vigoria	2	0.50947	0.254735	6.6179	0.01298 *
N_Gemme:Defogl	1	0.02754	0.027542	0.7155	0.41563
N_Gemme:Vigoria	2	0.02065	0.010327	0.2683	0.76954
Defogl:Vigoria	2	0.07253	0.036267	0.9422	0.41909
Residuals	11	0.42341	0.038492		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

KB

Analysis of Variance Table

Response: Ca_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.016634	0.016634	0.9521	0.35016
DIPROVE_2	1	0.118123	0.118123	6.7612	0.02468 *
DIPROVE_3	1	0.023999	0.023999	1.3737	0.26594
N_Gemme	1	0.098163	0.098163	5.6187	0.03713 *
Defogl	1	0.195629	0.195629	11.1975	0.00652 **
Vigoria	2	0.209529	0.104764	5.9966	0.01733 *
N_Gemme:Defogl	1	0.008403	0.008403	0.4810	0.50235
N_Gemme:Vigoria	2	0.006543	0.003272	0.1873	0.83182

Defogl:Vigoria 2 0.040190 0.020095 1.1502 0.35188
 Residuals 11 0.192178 0.017471

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: 0.617
 DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Ca_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	51.85	51.85	0.5039	0.492584
DIPROVE_2	1	64.86	64.86	0.6302	0.444062
DIPROVE_3	1	887.11	887.11	8.6202	0.013544 *
N_Gemme	1	1429.15	1429.15	13.8873	0.003343 **
Defogl	1	540.44	540.44	5.2515	0.042656 *
Vigoria	2	572.77	286.39	2.7829	0.105197
N_Gemme:Defogl	1	200.35	200.35	1.9469	0.190454
N_Gemme:Vigoria	2	172.61	86.30	0.8386	0.458167
Defogl:Vigoria	2	867.59	433.79	4.2153	0.043752 *
Residuals	11	1132.02	102.91		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: -0.663

Antoctot1B

Analysis of Variance Table

Response: Ca_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	3025	3024.5	0.8369	0.37991
DIPROVE_2	1	16097	16097.3	4.4540	0.05853 .
DIPROVE_3	1	2746	2745.7	0.7597	0.40204
N_Gemme	1	2426	2425.5	0.6711	0.43005
Defogl	1	10512	10512.3	2.9087	0.11615
Vigoria	2	18736	9368.0	2.5920	0.11958
N_Gemme:Defogl	1	3044	3044.2	0.8423	0.37843
N_Gemme:Vigoria	2	7590	3794.9	1.0500	0.38252
Defogl:Vigoria	2	4233	2116.3	0.5856	0.57325
Residuals	11	39755	3614.1		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	26819	26819.0	8.1699	0.01556 *
DIPROVE_2	1	5903	5903.2	1.7983	0.20695
DIPROVE_3	1	826	826.2	0.2517	0.62578
N_Gemme	1	2297	2296.8	0.6997	0.42069
Defogl	1	878	878.3	0.2676	0.61521
Vigoria	2	268	133.8	0.0408	0.96021
N_Gemme:Defogl	1	2698	2697.9	0.8219	0.38406
N_Gemme:Vigoria	2	3161	1580.3	0.4814	0.63034
Defogl:Vigoria	2	2250	1125.2	0.3428	0.71712
Residuals	11	36109	3282.7		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.653

DIPROVE_2: n.s

DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	10852	10852	0.7716	0.39850
DIPROVE_2	1	49058	49058	3.4882	0.08866 .
DIPROVE_3	1	13508	13508	0.9605	0.34813
N_Gemme	1	21868	21868	1.5549	0.23831
Defogl	1	17856	17856	1.2696	0.28382
Vigoria	2	51625	25812	1.8354	0.20520
N_Gemme:Defogl	1	16984	16984	1.2077	0.29526
N_Gemme:Vigoria	2	28335	14168	1.0074	0.39652
Defogl:Vigoria	2	29895	14947	1.0628	0.37844
Residuals	11	154701	14064		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	208553	208553	22.9824	0.0005587 ***
DIPROVE_2	1	117	117	0.0129	0.9116269
DIPROVE_3	1	3214	3214	0.3542	0.5638134
N_Gemme	1	20	20	0.0022	0.9630481
Defogl	1	7044	7044	0.7762	0.3971405
Vigoria	2	94406	47203	5.2017	0.0257036 *
N_Gemme:Defogl	1	11748	11748	1.2946	0.2793707
N_Gemme:Vigoria	2	13660	6830	0.7527	0.4938955
Defogl:Vigoria	2	2638	1319	0.1454	0.8663319
Residuals	11	99819	9074		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.822

DIPROVE_2: n.s

DIPROVE_3: n.s

CACCIAGRANDE 2009M IASMA

PvinacciaA

Analysis of Variance Table

Response: Ca_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	245907	245907	7.2876	0.0206735	*
IASMA_2	1	517169	517169	15.3265	0.0024139	**
IASMA_3	1	336393	336393	9.9691	0.0091212	**
N_Gemme	1	701330	701330	20.7842	0.0008180	***
Defogl	1	865948	865948	25.6627	0.0003629	***
Vigoria	2	90287	45144	1.3378	0.3019516	
N_Gemme:Defogl	1	19	19	0.0006	0.9815498	
N_Gemme:Vigoria	2	234961	117481	3.4816	0.0673836	.
Defogl:Vigoria	2	179700	89850	2.6627	0.1139984	
Residuals	11	371178	33743			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.631

IASMA_2: 0.763

IASMA_3: 0.69

DensMostoA

Analysis of Variance Table

Response: Ca_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00040211	0.00040211	57.5268	1.080e-05	***
IASMA_2	1	0.00027336	0.00027336	39.1075	6.229e-05	***
IASMA_3	1	0.00006835	0.00006835	9.7787	0.009628	**
N_Gemme	1	0.00001974	0.00001974	2.8241	0.121000	
Defogl	1	0.00000934	0.00000934	1.3356	0.272294	
Vigoria	2	0.00009339	0.00004670	6.6805	0.012614	*
N_Gemme:Defogl	1	0.00000541	0.00000541	0.7742	0.397729	
N_Gemme:Vigoria	2	0.00001767	0.00000884	1.2641	0.320513	
Defogl:Vigoria	2	0.00000079	0.00000040	0.0568	0.945013	
Residuals	11	0.00007689	0.00000699			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.916

IASMA_2: -0.883

IASMA_3: 0.686

BrixA

Analysis of Variance Table

Response: Ca_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	19.4095	19.4095	57.8766	1.050e-05	***
IASMA_2	1	13.4896	13.4896	40.2241	5.504e-05	***
IASMA_3	1	2.9967	2.9967	8.9357	0.01231	*
N_Gemme	1	1.1068	1.1068	3.3004	0.09658	.
Defogl	1	0.4355	0.4355	1.2987	0.27866	
Vigoria	2	4.2767	2.1383	6.3763	0.01450	*
N_Gemme:Defogl	1	0.3056	0.3056	0.9113	0.36029	
N_Gemme:Vigoria	2	0.9067	0.4534	1.3519	0.29857	
Defogl:Vigoria	2	0.0386	0.0193	0.0576	0.94433	
Residuals	11	3.6890	0.3354			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.917

IASMA_2: -0.886

IASMA_3: 0.669

PhA

Analysis of Variance Table

Response: Ca_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.045524	0.045524	19.2655	0.0010823	**
IASMA_2	1	0.058596	0.058596	24.7975	0.0004156	***
IASMA_3	1	0.011093	0.011093	4.6945	0.0530843	.
N_Gemme	1	0.029572	0.029572	12.5145	0.0046524	**
Defogl	1	0.000065	0.000065	0.0273	0.8717044	
Vigoria	2	0.024966	0.012483	5.2828	0.0246585	*
N_Gemme:Defogl	1	0.002707	0.002707	1.1456	0.3074009	
N_Gemme:Vigoria	2	0.002410	0.001205	0.5100	0.6140259	
Defogl:Vigoria	2	0.006457	0.003229	1.3663	0.2951304	
Residuals	11	0.025993	0.002363			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.798

IASMA_2: 0.832

IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Ca_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00119	0.00119	0.0265	0.873587	
IASMA_2	1	0.69817	0.69817	15.5385	0.002305	**
IASMA_3	1	0.55932	0.55932	12.4483	0.004730	**
N_Gemme	1	0.01652	0.01652	0.3676	0.556626	
Defogl	1	0.02346	0.02346	0.5222	0.484974	
Vigoria	2	0.10229	0.05114	1.1383	0.355384	
N_Gemme:Defogl	1	0.09081	0.09081	2.0211	0.182857	
N_Gemme:Vigoria	2	0.00137	0.00069	0.0153	0.984883	
Defogl:Vigoria	2	0.09918	0.04959	1.1037	0.365731	
Residuals	11	0.49425	0.04493			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.765
IASMA_3: -0.729

AcTartA

Analysis of Variance Table

Response: Ca_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.12568	0.125681	3.6921	0.08095 .
IASMA_2	1	0.08249	0.082486	2.4232	0.14784
IASMA_3	1	0.11750	0.117503	3.4519	0.09013 .
N_Gemme	1	0.03864	0.038640	1.1351	0.30951
Defogl	1	0.00335	0.003349	0.0984	0.75963
Vigoria	2	0.03620	0.018101	0.5318	0.60194
N_Gemme:Defogl	1	0.10117	0.101170	2.9720	0.11267
N_Gemme:Vigoria	2	0.03561	0.017804	0.5230	0.60676
Defogl:Vigoria	2	0.03761	0.018807	0.5525	0.59069
Residuals	11	0.37444	0.034040		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Ca_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	1.40569	1.40569	20.2857	0.0008953 ***
IASMA_2	1	0.10187	0.10187	1.4701	0.2507296
IASMA_3	1	0.02134	0.02134	0.3079	0.5900595
N_Gemme	1	0.00711	0.00711	0.1026	0.7547399
Defogl	1	0.01309	0.01309	0.1889	0.6722507
Vigoria	2	0.08365	0.04183	0.6036	0.5639945
N_Gemme:Defogl	1	0.08720	0.08720	1.2585	0.2858342
N_Gemme:Vigoria	2	0.00879	0.00440	0.0634	0.9388732
Defogl:Vigoria	2	0.16293	0.08147	1.1757	0.3445679
Residuals	11	0.76224	0.06929		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.805
IASMA_2: n.s
IASMA_3: n.s

KA

Analysis of Variance Table

Response: Ca_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IASMA_1	1	0.041636	0.041636	6.3028	0.0289591	*
IASMA_2	1	0.191933	0.191933	29.0547	0.0002199	***
IASMA_3	1	0.048299	0.048299	7.3115	0.0205105	*
N_Gemme	1	0.137420	0.137420	20.8025	0.0008153	***
Defogl	1	0.000669	0.000669	0.1012	0.7562987	
Vigoria	2	0.062813	0.031406	4.7543	0.0325103	*
N_Gemme:Defogl	1	0.005093	0.005093	0.7710	0.3986921	
N_Gemme:Vigoria	2	0.010589	0.005294	0.8015	0.4732205	
Defogl:Vigoria	2	0.022829	0.011414	1.7279	0.2225496	
Residuals	11	0.072665	0.006606			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.604
 IASMA_2: 0.852
 IASMA_3: -0.632

ApaA

Analysis of Variance Table

Response: Ca_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	38.13	38.13	0.1787	0.68064	
IASMA_2	1	199.37	199.37	0.9343	0.35453	
IASMA_3	1	228.99	228.99	1.0731	0.32249	
N_Gemme	1	546.73	546.73	2.5620	0.13776	
Defogl	1	1789.12	1789.12	8.3839	0.01456	*
Vigoria	2	67.97	33.98	0.1593	0.85471	
N_Gemme:Defogl	1	59.99	59.99	0.2811	0.60652	
N_Gemme:Vigoria	2	66.32	33.16	0.1554	0.85792	
Defogl:Vigoria	2	464.96	232.48	1.0894	0.37011	
Residuals	11	2347.39	213.40			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Ca_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	17403	17403	21.2175	0.0007572	***
IASMA_2	1	53313	53313	64.9965	6.069e-06	***
IASMA_3	1	6827	6827	8.3226	0.0148407	*
N_Gemme	1	74	74	0.0900	0.7697989	
Defogl	1	0	0	0.0000	0.9989919	
Vigoria	2	10784	5392	6.5735	0.0132406	*
N_Gemme:Defogl	1	1116	1116	1.3600	0.2682034	
N_Gemme:Vigoria	2	8197	4098	4.9966	0.0285907	*
Defogl:Vigoria	2	3106	1553	1.8932	0.1965240	
Residuals	11	9023	820			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.812
 IASMA_2: 0.925
 IASMA_3: -0.656

Poliftot1A

Analysis of Variance Table

Response: Ca_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	223.1	223.1	0.2185	0.64930
IASMA_2	1	2419.4	2419.4	2.3699	0.15195
IASMA_3	1	513.0	513.0	0.5025	0.49317
N_Gemme	1	1199.4	1199.4	1.1748	0.30160
Defogl	1	6540.9	6540.9	6.4069	0.02791 *
Vigoria	2	715.2	357.6	0.3503	0.71206
N_Gemme:Defogl	1	191.9	191.9	0.1880	0.67299
N_Gemme:Vigoria	2	5045.4	2522.7	2.4710	0.12992
Defogl:Vigoria	2	4769.8	2384.9	2.3361	0.14271
Residuals	11	11230.0	1020.9		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Ca_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	71479	71479	13.4799	0.00368 **
IASMA_2	1	266871	266871	50.3281	2.008e-05 ***
IASMA_3	1	45410	45410	8.5636	0.01378 *
N_Gemme	1	961	961	0.1811	0.67860
Defogl	1	1969	1969	0.3714	0.55463
Vigoria	2	61467	30733	5.7959	0.01910 *
N_Gemme:Defogl	1	4514	4514	0.8513	0.37598
N_Gemme:Vigoria	2	38084	19042	3.5910	0.06304 .
Defogl:Vigoria	2	7567	3784	0.7135	0.51124
Residuals	11	58329	5303		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.742
 IASMA_2: 0.906
 IASMA_3: -0.662

Poliftot2A

Analysis of Variance Table

Response: Ca_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	20009	20009.3	1.7854	0.2085
IASMA_2	1	5513	5513.3	0.4919	0.4976

IASMA_3	1	10876	10875.5	0.9704	0.3458
N_Gemme	1	6045	6044.6	0.5393	0.4781
Defogl	1	7914	7913.6	0.7061	0.4186
Vigoria	2	11950	5975.0	0.5331	0.6012
N_Gemme:Defogl	1	1045	1045.2	0.0933	0.7658
N_Gemme:Vigoria	2	21726	10863.2	0.9693	0.4095
Defogl:Vigoria	2	1358	678.8	0.0606	0.9415
Residuals	11	123279	11207.2		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Ca_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	77	77	0.0021	0.964172	
IASMA_2	1	385080	385080	10.5212	0.007823	**
IASMA_3	1	43382	43382	1.1853	0.299562	
N_Gemme	1	1353474	1353474	36.9797	7.948e-05	***
Defogl	1	650560	650560	17.7746	0.001446	**
Vigoria	2	288503	144252	3.9413	0.051207	.
N_Gemme:Defogl	1	48882	48882	1.3355	0.272306	
N_Gemme:Vigoria	2	43274	21637	0.5912	0.570353	
Defogl:Vigoria	2	265593	132796	3.6283	0.061639	.
Residuals	11	402605	36600			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.699
IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Ca_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00016609	0.00016609	14.1422	0.003151	**
IASMA_2	1	0.00057782	0.00057782	49.2002	2.228e-05	***
IASMA_3	1	0.00002895	0.00002895	2.4653	0.144683	
N_Gemme	1	0.00005112	0.00005112	4.3525	0.061035	.
Defogl	1	0.00001120	0.00001120	0.9533	0.349873	
Vigoria	2	0.00008899	0.00004449	3.7885	0.056015	.
N_Gemme:Defogl	1	0.00005853	0.00005853	4.9841	0.047318	*
N_Gemme:Vigoria	2	0.00001158	0.00000579	0.4928	0.623755	
Defogl:Vigoria	2	0.00002945	0.00001472	1.2536	0.323252	
Residuals	11	0.00012919	0.00001174			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.75
IASMA_2: -0.904
IASMA_3: n.s

BrixB

Analysis of Variance Table

Response: Ca_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	8.2188	8.2188	14.0006	0.003256	**
IASMA_2	1	27.5401	27.5401	46.9142	2.766e-05	***
IASMA_3	1	1.5538	1.5538	2.6469	0.132031	
N_Gemme	1	2.3999	2.3999	4.0882	0.068191	.
Defogl	1	0.5254	0.5254	0.8950	0.364444	
Vigoria	2	4.1259	2.0629	3.5142	0.066053	.
N_Gemme:Defogl	1	3.0178	3.0178	5.1407	0.044516	*
N_Gemme:Vigoria	2	0.6938	0.3469	0.5910	0.570462	
Defogl:Vigoria	2	1.3454	0.6727	1.1460	0.353126	
Residuals	11	6.4573	0.5870			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.748

IASMA_2: -0.9

IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Ca_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.029689	0.029689	5.4913	0.0389408	*
IASMA_2	1	0.141052	0.141052	26.0889	0.0003399	***
IASMA_3	1	0.008761	0.008761	1.6205	0.2292715	
N_Gemme	1	0.026441	0.026441	4.8906	0.0490928	*
Defogl	1	0.003311	0.003311	0.6124	0.4504183	
Vigoria	2	0.039813	0.019907	3.6819	0.0596845	.
N_Gemme:Defogl	1	0.003725	0.003725	0.6890	0.4241478	
N_Gemme:Vigoria	2	0.000681	0.000340	0.0630	0.9393170	
Defogl:Vigoria	2	0.004121	0.002060	0.3811	0.6918074	
Residuals	11	0.059472	0.005407			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.577

IASMA_2: -0.839

IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Ca_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.05364	0.05364	2.0741	0.1776640	
IASMA_2	1	0.64413	0.64413	24.9062	0.0004085	***
IASMA_3	1	0.07503	0.07503	2.9010	0.1165735	
N_Gemme	1	0.15573	0.15573	6.0215	0.0320246	*
Defogl	1	0.00290	0.00290	0.1122	0.7439646	
Vigoria	2	0.03119	0.01560	0.6030	0.5642769	

N_Gemme:Defogl	1	0.07622	0.07622	2.9473	0.1140107
N_Gemme:Vigoria	2	0.01296	0.00648	0.2506	0.7826466
Defogl:Vigoria	2	0.03705	0.01852	0.7162	0.5100240
Residuals	11	0.28448	0.02586		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.833
IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Ca_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.190733	0.190733	8.0801	0.01601 *
IASMA_2	1	0.000030	0.000030	0.0013	0.97229
IASMA_3	1	0.000896	0.000896	0.0380	0.84905
N_Gemme	1	0.025023	0.025023	1.0601	0.32531
Defogl	1	0.079203	0.079203	3.3553	0.09418 .
Vigoria	2	0.045605	0.022803	0.9660	0.41068
N_Gemme:Defogl	1	0.029465	0.029465	1.2482	0.28770
N_Gemme:Vigoria	2	0.065307	0.032654	1.3833	0.29114
Defogl:Vigoria	2	0.040764	0.020382	0.8634	0.44842
Residuals	11	0.259657	0.023605		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.651
IASMA_2: n.s
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Ca_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.50015	0.50015	13.5472	0.003622 **
IASMA_2	1	0.00888	0.00888	0.2404	0.633543
IASMA_3	1	0.27618	0.27618	7.4808	0.019401 *
N_Gemme	1	0.00206	0.00206	0.0558	0.817591
Defogl	1	0.03874	0.03874	1.0492	0.327680
Vigoria	2	0.19786	0.09893	2.6796	0.112712
N_Gemme:Defogl	1	0.06391	0.06391	1.7311	0.215030
N_Gemme:Vigoria	2	0.00058	0.00029	0.0078	0.992213
Defogl:Vigoria	2	0.00485	0.00243	0.0657	0.936788
Residuals	11	0.40611	0.03692		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.743
IASMA_2: n.s
IASMA_3: 0.636

KB

Analysis of Variance Table

Response: Ca_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.01946	0.01946	2.2998	0.15759	
IASMA_2	1	0.57103	0.57103	67.4846	5.072e-06	***
IASMA_3	1	0.00605	0.00605	0.7144	0.41598	
N_Gemme	1	0.07796	0.07796	9.2139	0.01134	*
Defogl	1	0.02846	0.02846	3.3639	0.09381	.
Vigoria	2	0.05348	0.02674	3.1604	0.08233	.
N_Gemme:Defogl	1	0.03298	0.03298	3.8972	0.07401	.
N_Gemme:Vigoria	2	0.00739	0.00369	0.4365	0.65703	
Defogl:Vigoria	2	0.01950	0.00975	1.1524	0.35126	
Residuals	11	0.09308	0.00846			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.927
IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Ca_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	72.83	72.83	0.5534	0.472531	
IASMA_2	1	116.82	116.82	0.8877	0.366345	
IASMA_3	1	44.58	44.58	0.3387	0.572299	
N_Gemme	1	1287.68	1287.68	9.7841	0.009613	**
Defogl	1	1034.93	1034.93	7.8637	0.017145	*
Vigoria	2	1012.91	506.45	3.8482	0.054075	.
N_Gemme:Defogl	1	161.94	161.94	1.2305	0.290974	
N_Gemme:Vigoria	2	247.84	123.92	0.9416	0.419315	
Defogl:Vigoria	2	491.50	245.75	1.8673	0.200356	
Residuals	11	1447.70	131.61			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Ca_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	427	427	0.1168	0.738977	
IASMA_2	1	46849	46849	12.8235	0.004311	**
IASMA_3	1	2755	2755	0.7540	0.403769	
N_Gemme	1	2740	2740	0.7500	0.404957	
Defogl	1	850	850	0.2326	0.639031	
Vigoria	2	2497	1248	0.3417	0.717860	
N_Gemme:Defogl	1	2454	2454	0.6716	0.429902	
N_Gemme:Vigoria	2	4927	2464	0.6744	0.529341	

Defogl:Vigoria	2	4479	2240	0.6130	0.559233
Residuals	11	40187	3653		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.734
IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Ca_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	4594.1	4594.1	1.6591	0.22416
IASMA_2	1	1852.7	1852.7	0.6691	0.43073
IASMA_3	1	6.6	6.6	0.0024	0.96182
N_Gemme	1	6472.8	6472.8	2.3376	0.15452
Defogl	1	21584.0	21584.0	7.7947	0.01753 *
Vigoria	2	2760.1	1380.0	0.4984	0.62059
N_Gemme:Defogl	1	8526.5	8526.5	3.0792	0.10707
N_Gemme:Vigoria	2	2735.7	1367.9	0.4940	0.62310
Defogl:Vigoria	2	2217.2	1108.6	0.4004	0.67946
Residuals	11	30459.4	2769.0		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Ca_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	8	8	0.0006	0.981131
IASMA_2	1	142450	142450	10.2499	0.008431 **
IASMA_3	1	183	183	0.0132	0.910768
N_Gemme	1	18877	18877	1.3583	0.268479
Defogl	1	100	100	0.0072	0.933824
Vigoria	2	21765	10882	0.7830	0.480910
N_Gemme:Defogl	1	15496	15496	1.1150	0.313641
N_Gemme:Vigoria	2	19323	9662	0.6952	0.519630
Defogl:Vigoria	2	23605	11802	0.8492	0.453969
Residuals	11	152874	13898		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.695
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Ca_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	149195	149195	18.4873	0.001257	**
IASMA_2	1	59772	59772	7.4066	0.019878	*
IASMA_3	1	19032	19032	2.3583	0.152863	
N_Gemme	1	7994	7994	0.9906	0.340986	
Defogl	1	48715	48715	6.0365	0.031852	*
Vigoria	2	22092	11046	1.3688	0.294555	
N_Gemme:Defogl	1	21027	21027	2.6055	0.134789	
N_Gemme:Vigoria	2	19322	9661	1.1971	0.338535	
Defogl:Vigoria	2	5300	2650	0.3284	0.726901	
Residuals	11	88771	8070			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.792

IASMA_2: -0.634

IASMA_3: n.s

CORTIGLIANO 2008 IBIMET

PvinacciaA

Analysis of Variance Table

Response: Co_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	6955	6955	0.0750	0.787374	
IBIMET_3	1	191	191	0.0021	0.964334	
N_Gemme	1	317117	317117	3.4176	0.080999	.
Defogl	1	536825	536825	5.7855	0.027130	*
Dirad	1	2451779	2451779	26.4233	6.859e-05	***
Vigoria	1	183613	183613	1.9788	0.176545	
N_Gemme:Defogl	1	40514	40514	0.4366	0.517126	
N_Gemme:Dirad	1	450604	450604	4.8562	0.040808	*
Defogl:Dirad	1	31981	31981	0.3447	0.564442	
N_Gemme:Vigoria	1	142374	142374	1.5344	0.231365	
Defogl:Vigoria	1	129093	129093	1.3913	0.253548	
Dirad:Vigoria	1	1007922	1007922	10.8626	0.004018	**
N_Gemme:Defogl:Dirad	1	31983	31983	0.3447	0.564426	
Residuals	18	1670196	92789			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	1.3101e-04	1.3101e-04	11.2487	0.003534	**
IBIMET_3	1	3.4107e-05	3.4107e-05	2.9286	0.104201	
N_Gemme	1	1.6830e-06	1.6830e-06	0.1445	0.708266	
Defogl	1	4.1993e-05	4.1993e-05	3.6057	0.073732	.
Dirad	1	4.3190e-05	4.3190e-05	3.7085	0.070082	.

Vigoria	1	1.0829e-05	1.0829e-05	0.9299	0.347674
N_Gemme:Defogl	1	5.2890e-06	5.2890e-06	0.4541	0.508947
N_Gemme:Dirad	1	5.2622e-05	5.2622e-05	4.5184	0.047631 *
Defogl:Dirad	1	3.3300e-07	3.3300e-07	0.0286	0.867573
N_Gemme:Vigoria	1	8.6800e-07	8.6800e-07	0.0745	0.787976
Defogl:Vigoria	1	2.9220e-06	2.9220e-06	0.2509	0.622529
Dirad:Vigoria	1	3.9778e-05	3.9778e-05	3.4155	0.081084 .
N_Gemme:Defogl:Dirad	1	2.3650e-06	2.3650e-06	0.2031	0.657643
Residuals	18	2.0963e-04	1.1646e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.62

IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Co_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	7.5761	7.5761	13.7341	0.001617	**
IBIMET_3	1	1.1010	1.1010	1.9959	0.174779	
N_Gemme	1	0.0406	0.0406	0.0736	0.789260	
Defogl	1	1.7305	1.7305	3.1371	0.093458	.
Dirad	1	2.0200	2.0200	3.6619	0.071710	.
Vigoria	1	0.3040	0.3040	0.5510	0.467481	
N_Gemme:Defogl	1	0.3123	0.3123	0.5661	0.461557	
N_Gemme:Dirad	1	2.6727	2.6727	4.8452	0.041013	*
Defogl:Dirad	1	0.0269	0.0269	0.0488	0.827570	
N_Gemme:Vigoria	1	0.0449	0.0449	0.0813	0.778765	
Defogl:Vigoria	1	0.2205	0.2205	0.3997	0.535176	
Dirad:Vigoria	1	2.0799	2.0799	3.7704	0.067985	.
N_Gemme:Defogl:Dirad	1	0.1311	0.1311	0.2376	0.631836	
Residuals	18	9.9293	0.5516			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.658

IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Co_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.052608	0.052608	16.0008	0.0008396	***
IBIMET_3	1	0.015072	0.015072	4.5842	0.0462048	*
N_Gemme	1	0.000341	0.000341	0.1038	0.7509962	
Defogl	1	0.002726	0.002726	0.8292	0.3745350	
Dirad	1	0.009378	0.009378	2.8524	0.1084825	
Vigoria	1	0.013442	0.013442	4.0884	0.0582992	.
N_Gemme:Defogl	1	0.012370	0.012370	3.7625	0.0682497	.
N_Gemme:Dirad	1	0.000058	0.000058	0.0175	0.8962238	
Defogl:Dirad	1	0.000471	0.000471	0.1434	0.7093865	
N_Gemme:Vigoria	1	0.000362	0.000362	0.1100	0.7439452	
Defogl:Vigoria	1	0.007483	0.007483	2.2761	0.1487386	
Dirad:Vigoria	1	0.002324	0.002324	0.7068	0.4115426	

N_Gemme:Defogl:Dirad 1 0.003421 0.003421 1.0404 0.3212427
 Residuals 18 0.059181 0.003288

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.686
 IBIMET_3: -0.451

AcTitola

Analysis of Variance Table

Response: Co_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.15174	0.15174	1.3004	0.26908
IBIMET_3	1	0.37651	0.37651	3.2268	0.08924
N_Gemme	1	0.11251	0.11251	0.9642	0.33914
Defogl	1	0.03418	0.03418	0.2930	0.59496
Dirad	1	0.03188	0.03188	0.2732	0.60754
Vigoria	1	0.19049	0.19049	1.6326	0.21758
N_Gemme:Defogl	1	0.01683	0.01683	0.1443	0.70851
N_Gemme:Dirad	1	0.25302	0.25302	2.1684	0.15814
Defogl:Dirad	1	0.00030	0.00030	0.0025	0.96040
N_Gemme:Vigoria	1	0.04744	0.04744	0.4065	0.53175
Defogl:Vigoria	1	0.00317	0.00317	0.0272	0.87093
Dirad:Vigoria	1	0.01876	0.01876	0.1607	0.69320
N_Gemme:Defogl:Dirad	1	0.03471	0.03471	0.2974	0.59219
Residuals	18	2.10027	0.11668		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: n.s

AcTarta

Analysis of Variance Table

Response: Co_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1.34712	1.34712	8.8086	0.00824 **
IBIMET_3	1	0.88040	0.88040	5.7568	0.02746 *
N_Gemme	1	0.00656	0.00656	0.0429	0.83826
Defogl	1	0.02875	0.02875	0.1880	0.66974
Dirad	1	0.03427	0.03427	0.2241	0.64165
Vigoria	1	0.23242	0.23242	1.5197	0.23352
N_Gemme:Defogl	1	0.01455	0.01455	0.0951	0.76129
N_Gemme:Dirad	1	0.11775	0.11775	0.7699	0.39180
Defogl:Dirad	1	0.00874	0.00874	0.0572	0.81373
N_Gemme:Vigoria	1	0.07549	0.07549	0.4936	0.49130
Defogl:Vigoria	1	0.39035	0.39035	2.5525	0.12753
Dirad:Vigoria	1	0.01015	0.01015	0.0664	0.79965
N_Gemme:Defogl:Dirad	1	0.16175	0.16175	1.0577	0.31737
Residuals	18	2.75279	0.15293		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.573
 IBIMET_3: -0.492

AcMalicoA

Analysis of Variance Table

Response: Co_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.01530	0.01530	0.5383	0.4726	
IBIMET_3	1	1.18852	1.18852	41.8153	4.398e-06	***
N_Gemme	1	0.02117	0.02117	0.7449	0.3995	
Defogl	1	0.03553	0.03553	1.2500	0.2782	
Dirad	1	0.00102	0.00102	0.0359	0.8518	
Vigoria	1	0.74473	0.74473	26.2015	7.185e-05	***
N_Gemme:Defogl	1	0.00275	0.00275	0.0966	0.7595	
N_Gemme:Dirad	1	0.05686	0.05686	2.0006	0.1743	
Defogl:Dirad	1	0.00216	0.00216	0.0759	0.7861	
N_Gemme:Vigoria	1	0.00583	0.00583	0.2052	0.6560	
Defogl:Vigoria	1	0.02143	0.02143	0.7539	0.3967	
Dirad:Vigoria	1	0.00859	0.00859	0.3022	0.5893	
N_Gemme:Defogl:Dirad	1	0.00008	0.00008	0.0027	0.9595	
Residuals	18	0.51162	0.02842			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: 0.836

KA

Analysis of Variance Table

Response: Co_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.13599	0.13599	6.9936	0.0164779	*
IBIMET_3	1	0.50024	0.50024	25.7254	7.946e-05	***
N_Gemme	1	0.00060	0.00060	0.0307	0.8628093	
Defogl	1	0.00740	0.00740	0.3807	0.5449388	
Dirad	1	0.00840	0.00840	0.4319	0.5193729	
Vigoria	1	0.45103	0.45103	23.1947	0.0001385	***
N_Gemme:Defogl	1	0.00021	0.00021	0.0107	0.9187103	
N_Gemme:Dirad	1	0.00109	0.00109	0.0562	0.8152518	
Defogl:Dirad	1	0.01528	0.01528	0.7860	0.3869942	
N_Gemme:Vigoria	1	0.00620	0.00620	0.3188	0.5792986	
Defogl:Vigoria	1	0.02840	0.02840	1.4606	0.2424680	
Dirad:Vigoria	1	0.00275	0.00275	0.1413	0.7113664	
N_Gemme:Defogl:Dirad	1	0.00844	0.00844	0.4339	0.5184162	
Residuals	18	0.35002	0.01945			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.529
IBIMET_3: 0.767

ApaA

Analysis of Variance Table

Response: Co_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	7787.2	7787.2	11.1886	0.003605	**
IBIMET_3	1	19985.6	19985.6	28.7151	4.301e-05	***
N_Gemme	1	163.4	163.4	0.2347	0.633893	
Defogl	1	5215.3	5215.3	7.4932	0.013532	*
Dirad	1	288.6	288.6	0.4147	0.527730	
Vigoria	1	7382.9	7382.9	10.6077	0.004378	**
N_Gemme:Defogl	1	5.8	5.8	0.0083	0.928469	
N_Gemme:Dirad	1	187.8	187.8	0.2698	0.609785	
Defogl:Dirad	1	331.3	331.3	0.4760	0.499023	
N_Gemme:Vigoria	1	79.5	79.5	0.1143	0.739217	
Defogl:Vigoria	1	3445.2	3445.2	4.9500	0.039117	*
Dirad:Vigoria	1	141.2	141.2	0.2029	0.657796	
N_Gemme:Defogl:Dirad	1	246.7	246.7	0.3544	0.559038	
Residuals	18	12528.0	696.0			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.619
IBIMET_3: -0.784

Antoctot1A

Analysis of Variance Table

Response: Co_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	48	48.0	0.0187	0.89265	
IBIMET_3	1	14	14.4	0.0056	0.94109	
N_Gemme	1	2	1.7	0.0007	0.97991	
Defogl	1	3779	3778.6	1.4750	0.24026	
Dirad	1	15406	15405.5	6.0135	0.02463	*
Vigoria	1	4042	4041.9	1.5777	0.22514	
N_Gemme:Defogl	1	1324	1323.7	0.5167	0.48148	
N_Gemme:Dirad	1	238	238.4	0.0931	0.76383	
Defogl:Dirad	1	494	494.4	0.1930	0.66566	
N_Gemme:Vigoria	1	2586	2585.6	1.0093	0.32839	
Defogl:Vigoria	1	5006	5006.4	1.9542	0.17912	
Dirad:Vigoria	1	10714	10714.1	4.1822	0.05575	.
N_Gemme:Defogl:Dirad	1	682	682.4	0.2664	0.61204	
Residuals	18	46113	2561.8			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Co_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	25	25	0.0027	0.95928	
IBIMET_3	1	7252	7252	0.7884	0.38630	
N_Gemme	1	1271	1271	0.1381	0.71447	
Defogl	1	64480	64480	7.0095	0.01637	*
Dirad	1	584	584	0.0635	0.80392	
Vigoria	1	55	55	0.0060	0.93922	

N_Gemme:Defogl	1	13747	13747	1.4944	0.23729
N_Gemme:Dirad	1	9390	9390	1.0207	0.32574
Defogl:Dirad	1	21246	21246	2.3097	0.14594
N_Gemme:Vigoria	1	27569	27569	2.9969	0.10053
Defogl:Vigoria	1	8428	8428	0.9162	0.35115
Dirad:Vigoria	1	1438	1438	0.1563	0.69726
N_Gemme:Defogl:Dirad	1	6	6	0.0006	0.98068
Residuals	18	165580	9199		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	7972	7972	1.1829	0.29112
IBIMET_3	1	18139	18139	2.6914	0.11825
N_Gemme	1	17284	17284	2.5645	0.12669
Defogl	1	9323	9323	1.3832	0.25487
Dirad	1	5665	5665	0.8406	0.37135
Vigoria	1	1089	1089	0.1616	0.69245
N_Gemme:Defogl	1	9312	9312	1.3816	0.25514
N_Gemme:Dirad	1	21	21	0.0030	0.95662
Defogl:Dirad	1	756	756	0.1121	0.74161
N_Gemme:Vigoria	1	2278	2278	0.3379	0.56823
Defogl:Vigoria	1	37020	37020	5.4928	0.03077 *
Dirad:Vigoria	1	14052	14052	2.0849	0.16594
N_Gemme:Defogl:Dirad	1	8175	8175	1.2129	0.28526
Residuals	18	121315	6740		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Co_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	49260	49260	1.1092	0.30620
IBIMET_3	1	7095	7095	0.1598	0.69408
N_Gemme	1	58177	58177	1.3099	0.26740
Defogl	1	1148	1148	0.0258	0.87407
Dirad	1	170938	170938	3.8489	0.06543 .
Vigoria	1	85922	85922	1.9347	0.18121
N_Gemme:Defogl	1	44923	44923	1.0115	0.32787
N_Gemme:Dirad	1	55953	55953	1.2599	0.27643
Defogl:Dirad	1	46449	46449	1.0459	0.32001
N_Gemme:Vigoria	1	4475	4475	0.1008	0.75458
Defogl:Vigoria	1	124472	124472	2.8027	0.11140
Dirad:Vigoria	1	62503	62503	1.4073	0.25092
N_Gemme:Defogl:Dirad	1	5511	5511	0.1241	0.72873

Residuals 18 799417 44412

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Co_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	155272	155272	0.3906	0.53985
IBIMET_3	1	180716	180716	0.4546	0.50874
N_Gemme	1	434948	434948	1.0940	0.30942
Defogl	1	1461394	1461394	3.6759	0.07122 .
Dirad	1	1002044	1002044	2.5205	0.12979
Vigoria	1	685305	685305	1.7238	0.20570
N_Gemme:Defogl	1	2539	2539	0.0064	0.93718
N_Gemme:Dirad	1	310970	310970	0.7822	0.38813
Defogl:Dirad	1	25956	25956	0.0653	0.80122
N_Gemme:Vigoria	1	945350	945350	2.3779	0.14046
Defogl:Vigoria	1	684554	684554	1.7219	0.20594
Dirad:Vigoria	1	106062	106062	0.2668	0.61179
N_Gemme:Defogl:Dirad	1	193992	193992	0.4880	0.49377
Residuals	18	7156136	397563		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

DensMostoB

Analysis of Variance Table

Response: Co_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.00034123	0.00034123	16.9185	0.0006527 ***
IBIMET_3	1	0.00005488	0.00005488	2.7211	0.1163732
N_Gemme	1	0.00000919	0.00000919	0.4555	0.5083202
Defogl	1	0.00026594	0.00026594	13.1859	0.0019097 **
Dirad	1	0.00004213	0.00004213	2.0890	0.1655466
Vigoria	1	0.00000933	0.00000933	0.4628	0.5049736
N_Gemme:Defogl	1	0.00000045	0.00000045	0.0221	0.8835375
N_Gemme:Dirad	1	0.00000033	0.00000033	0.0162	0.9002463
Defogl:Dirad	1	0.00011172	0.00011172	5.5393	0.0301600 *
N_Gemme:Vigoria	1	0.00000248	0.00000248	0.1232	0.7296775
Defogl:Vigoria	1	0.00000640	0.00000640	0.3171	0.5802770
Dirad:Vigoria	1	0.00013829	0.00013829	6.8567	0.0174071 *
N_Gemme:Defogl:Dirad	1	0.00000056	0.00000056	0.0279	0.8691849
Residuals	18	0.00036304	0.00002017		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.696

IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Co_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	17.8840	17.8840	17.8230	0.0005129	***
IBIMET_3	1	2.0879	2.0879	2.0807	0.1663433	
N_Gemme	1	0.5107	0.5107	0.5090	0.4847183	
Defogl	1	12.9972	12.9972	12.9529	0.0020517	**
Dirad	1	2.2723	2.2723	2.2646	0.1497080	
Vigoria	1	0.3296	0.3296	0.3285	0.5736570	
N_Gemme:Defogl	1	0.0324	0.0324	0.0323	0.8593402	
N_Gemme:Dirad	1	0.0253	0.0253	0.0252	0.8755928	
Defogl:Dirad	1	5.6269	5.6269	5.6077	0.0292805	*
N_Gemme:Vigoria	1	0.0482	0.0482	0.0481	0.8289064	
Defogl:Vigoria	1	0.3542	0.3542	0.3530	0.5598226	
Dirad:Vigoria	1	6.3184	6.3184	6.2969	0.0218782	*
N_Gemme:Defogl:Dirad	1	0.0335	0.0335	0.0333	0.8571601	
Residuals	18	18.0616	1.0034			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.705

IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Co_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.018284	0.0182840	1.7807	0.1987
IBIMET_3	1	0.001040	0.0010402	0.1013	0.7539
N_Gemme	1	0.006101	0.0061009	0.5942	0.4508
Defogl	1	0.006844	0.0068444	0.6666	0.4249
Dirad	1	0.000157	0.0001568	0.0153	0.9030
Vigoria	1	0.003285	0.0032849	0.3199	0.5786
N_Gemme:Defogl	1	0.006031	0.0060310	0.5874	0.4534
N_Gemme:Dirad	1	0.000249	0.0002488	0.0242	0.8780
Defogl:Dirad	1	0.003614	0.0036143	0.3520	0.5604
N_Gemme:Vigoria	1	0.002533	0.0025328	0.2467	0.6254
Defogl:Vigoria	1	0.016945	0.0169450	1.6503	0.2152
Dirad:Vigoria	1	0.003942	0.0039419	0.3839	0.5433
N_Gemme:Defogl:Dirad	1	0.002907	0.0029065	0.2831	0.6012
Residuals	18	0.184825	0.0102681		

IBIMET_2: n.s

IBIMET_3: n.s

AcTitolB

Analysis of Variance Table

Response: Co_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.42795	0.42795	2.5283	0.1292
IBIMET_3	1	0.04768	0.04768	0.2817	0.6021

N_Gemme	1	0.00147	0.00147	0.0087	0.9268
Defogl	1	0.10489	0.10489	0.6197	0.4414
Dirad	1	0.00787	0.00787	0.0465	0.8317
Vigoria	1	0.02501	0.02501	0.1478	0.7052
N_Gemme:Defogl	1	0.04230	0.04230	0.2499	0.6232
N_Gemme:Dirad	1	0.05504	0.05504	0.3252	0.5756
Defogl:Dirad	1	0.02754	0.02754	0.1627	0.6914
N_Gemme:Vigoria	1	0.21511	0.21511	1.2708	0.2744
Defogl:Vigoria	1	0.23972	0.23972	1.4162	0.2495
Dirad:Vigoria	1	0.00332	0.00332	0.0196	0.8901
N_Gemme:Defogl:Dirad	1	0.10978	0.10978	0.6485	0.4311
Residuals	18	3.04683	0.16927		

IBIMET_2: n.s
IBIMET_3: n.s

AcTartB

Analysis of Variance Table

Response: Co_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.57987	0.57987	8.0811	0.0108 *
IBIMET_3	1	0.01117	0.01117	0.1557	0.6978
N_Gemme	1	0.16565	0.16565	2.3085	0.1460
Defogl	1	0.00180	0.00180	0.0251	0.8759
Dirad	1	0.06200	0.06200	0.8640	0.3649
Vigoria	1	0.01270	0.01270	0.1769	0.6790
N_Gemme:Defogl	1	0.00545	0.00545	0.0759	0.7860
N_Gemme:Dirad	1	0.10775	0.10775	1.5016	0.2362
Defogl:Dirad	1	0.10631	0.10631	1.4816	0.2392
N_Gemme:Vigoria	1	0.09092	0.09092	1.2671	0.2751
Defogl:Vigoria	1	0.00054	0.00054	0.0075	0.9317
Dirad:Vigoria	1	0.01028	0.01028	0.1433	0.7094
N_Gemme:Defogl:Dirad	1	0.03738	0.03738	0.5210	0.4797
Residuals	18	1.29161	0.07176		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.557
IBIMET_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Co_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.89083	0.89083	12.0208	0.002750 **
IBIMET_3	1	0.33563	0.33563	4.5290	0.047397 *
N_Gemme	1	0.01553	0.01553	0.2096	0.652577
Defogl	1	0.35910	0.35910	4.8457	0.041002 *
Dirad	1	0.02564	0.02564	0.3459	0.563739
Vigoria	1	0.13343	0.13343	1.8005	0.196330
N_Gemme:Defogl	1	0.00851	0.00851	0.1148	0.738691
N_Gemme:Dirad	1	0.06359	0.06359	0.8581	0.366522
Defogl:Dirad	1	0.10082	0.10082	1.3604	0.258690
N_Gemme:Vigoria	1	0.01414	0.01414	0.1909	0.667391
Defogl:Vigoria	1	0.00780	0.00780	0.1052	0.749360

```

Dirad:Vigoria      1 0.14184 0.14184  1.9139 0.183447
N_Gemme:Defogl:Dirad 1 0.04289 0.04289  0.5788 0.456632
Residuals         18 1.33393 0.07411
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.633
IBIMET_3: -0.448

```

KB

Analysis of Variance Table

Response: Co_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.48275	0.48275	16.0468	0.000829	***
IBIMET_3	1	0.01929	0.01929	0.6411	0.433748	
N_Gemme	1	0.00871	0.00871	0.2894	0.597217	
Defogl	1	0.00147	0.00147	0.0488	0.827628	
Dirad	1	0.00076	0.00076	0.0251	0.875847	
Vigoria	1	0.05746	0.05746	1.9099	0.183882	
N_Gemme:Defogl	1	0.01989	0.01989	0.6612	0.426772	
N_Gemme:Dirad	1	0.00005	0.00005	0.0015	0.969117	
Defogl:Dirad	1	0.00105	0.00105	0.0348	0.854185	
N_Gemme:Vigoria	1	0.01955	0.01955	0.6498	0.430718	
Defogl:Vigoria	1	0.03372	0.03372	1.1207	0.303767	
Dirad:Vigoria	1	0.00924	0.00924	0.3071	0.586273	
N_Gemme:Defogl:Dirad	1	0.02032	0.02032	0.6754	0.421939	
Residuals	18	0.54151	0.03008			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

IBIMET_2: 0.687
IBIMET_3: n.s

```

ApaB

Analysis of Variance Table

Response: Co_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	4074.8	4074.8	8.1238	0.0106264	*
IBIMET_3	1	23917.3	23917.3	47.6829	1.863e-06	***
N_Gemme	1	877.2	877.2	1.7489	0.2025690	
Defogl	1	8498.9	8498.9	16.9439	0.0006483	***
Dirad	1	172.5	172.5	0.3439	0.5648962	
Vigoria	1	7316.1	7316.1	14.5858	0.0012567	**
N_Gemme:Defogl	1	832.1	832.1	1.6589	0.2140653	
N_Gemme:Dirad	1	762.2	762.2	1.5195	0.2335512	
Defogl:Dirad	1	580.4	580.4	1.1571	0.2962778	
N_Gemme:Vigoria	1	26.5	26.5	0.0528	0.8208859	
Defogl:Vigoria	1	4236.4	4236.4	8.4459	0.0094188	**
Dirad:Vigoria	1	1180.1	1180.1	2.3527	0.1424579	
N_Gemme:Defogl:Dirad	1	394.8	394.8	0.7870	0.3866982	
Residuals	18	9028.6	501.6			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

IBIMET_2: 0.558

```

IBIMET_3: -0.852

Antoctot1B

Analysis of Variance Table

Response: Co_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	34970	34970	24.2908	0.0001084	***
IBIMET_3	1	3244	3244	2.2532	0.1506807	
N_Gemme	1	455	455	0.3158	0.5810671	
Defogl	1	9392	9392	6.5240	0.0199233	*
Dirad	1	7485	7485	5.1990	0.0350054	*
Vigoria	1	10680	10680	7.4182	0.0139337	*
N_Gemme:Defogl	1	9	9	0.0064	0.9373194	
N_Gemme:Dirad	1	533	533	0.3704	0.5504034	
Defogl:Dirad	1	177	177	0.1232	0.7296811	
N_Gemme:Vigoria	1	2318	2318	1.6099	0.2206708	
Defogl:Vigoria	1	352	352	0.2444	0.6269920	
Dirad:Vigoria	1	3919	3919	2.7224	0.1162914	
N_Gemme:Defogl:Dirad	1	5	5	0.0037	0.9520364	
Residuals	18	25914	1440			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.758

IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Co_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	384071	384071	16.2413	0.0007855	***
IBIMET_3	1	163964	163964	6.9336	0.0168781	*
N_Gemme	1	25002	25002	1.0573	0.3174597	
Defogl	1	20854	20854	0.8819	0.3601207	
Dirad	1	79730	79730	3.3716	0.0829048	.
Vigoria	1	117976	117976	4.9889	0.0384408	*
N_Gemme:Defogl	1	1794	1794	0.0759	0.7861271	
N_Gemme:Dirad	1	88	88	0.0037	0.9518982	
Defogl:Dirad	1	8853	8853	0.3744	0.5482909	
N_Gemme:Vigoria	1	34109	34109	1.4424	0.2453200	
Defogl:Vigoria	1	14764	14764	0.6243	0.4397294	
Dirad:Vigoria	1	26460	26460	1.1189	0.3041404	
N_Gemme:Defogl:Dirad	1	12986	12986	0.5492	0.4682196	
Residuals	18	425661	23648			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.689

IBIMET_3: 0.527

Antoctot2B

Analysis of Variance Table

Response: Co_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1953	1953.0	0.5212	0.47961
IBIMET_3	1	9054	9054.4	2.4163	0.13749
N_Gemme	1	800	799.8	0.2134	0.64962
Defogl	1	23710	23710.3	6.3274	0.02160 *
Dirad	1	4885	4884.7	1.3035	0.26853
Vigoria	1	2291	2290.6	0.6113	0.44448
N_Gemme:Defogl	1	31	30.7	0.0082	0.92890
N_Gemme:Dirad	1	943	943.4	0.2518	0.62192
Defogl:Dirad	1	71	70.7	0.0189	0.89224
N_Gemme:Vigoria	1	102	101.7	0.0271	0.87097
Defogl:Vigoria	1	2378	2377.8	0.6345	0.43608
Dirad:Vigoria	1	551	550.9	0.1470	0.70590
N_Gemme:Defogl:Dirad	1	760	760.2	0.2029	0.65780
Residuals	18	67451	3747.3		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	142193	142193	1.6076	0.22098
IBIMET_3	1	84798	84798	0.9587	0.34049
N_Gemme	1	9843	9843	0.1113	0.74254
Defogl	1	11213	11213	0.1268	0.72594
Dirad	1	46115	46115	0.5214	0.47953
Vigoria	1	209141	209141	2.3645	0.14152
N_Gemme:Defogl	1	39	39	0.0004	0.98342
N_Gemme:Dirad	1	10799	10799	0.1221	0.73083
Defogl:Dirad	1	37347	37347	0.4222	0.52403
N_Gemme:Vigoria	1	322218	322218	3.6430	0.07238 .
Defogl:Vigoria	1	150443	150443	1.7009	0.20860
Dirad:Vigoria	1	10349	10349	0.1170	0.73626
N_Gemme:Defogl:Dirad	1	18152	18152	0.2052	0.65595
Residuals	18	1592091	88449		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

CORTIGLIANO 2008 DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Co_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	4113	4113	0.0420	0.8400407
DIPROVE_2	1	17030	17030	0.1739	0.6818529
DIPROVE_3	1	10649	10649	0.1088	0.7455899
N_Gemme	1	285883	285883	2.9199	0.1056869

Defogl	1	1013826	1013826	10.3547	0.0050495	**
Dirad	1	2225601	2225601	22.7311	0.0001786	***
Vigoria	1	842	842	0.0086	0.9272047	
N_Gemme:Defogl	1	30491	30491	0.3114	0.5840853	
N_Gemme:Dirad	1	381147	381147	3.8928	0.0649788	.
Defogl:Dirad	1	14893	14893	0.1521	0.7013693	
N_Gemme:Vigoria	1	183972	183972	1.8790	0.1882765	
Defogl:Vigoria	1	4850	4850	0.0495	0.8265252	
Dirad:Vigoria	1	1107937	1107937	11.3159	0.0036845	**
N_Gemme:Defogl:Dirad	1	55445	55445	0.5663	0.4620333	
Residuals	17	1664466	97910			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1.0677e-04	1.0677e-04	8.5194	0.009573	**
DIPROVE_2	1	6.9250e-05	6.9250e-05	5.5255	0.031070	*
DIPROVE_3	1	5.7900e-07	5.7900e-07	0.0462	0.832341	
N_Gemme	1	2.5284e-05	2.5284e-05	2.0175	0.173581	
Defogl	1	1.8000e-07	1.8000e-07	0.0144	0.905945	
Dirad	1	3.1591e-05	3.1591e-05	2.5207	0.130785	
Vigoria	1	3.2505e-05	3.2505e-05	2.5936	0.125709	
N_Gemme:Defogl	1	1.9880e-06	1.9880e-06	0.1587	0.695348	
N_Gemme:Dirad	1	2.9016e-05	2.9016e-05	2.3152	0.146499	
Defogl:Dirad	1	9.3990e-06	9.3990e-06	0.7499	0.398560	
N_Gemme:Vigoria	1	5.9530e-06	5.9530e-06	0.4750	0.500006	
Defogl:Vigoria	1	1.1159e-05	1.1159e-05	0.8904	0.358588	
Dirad:Vigoria	1	3.4030e-06	3.4030e-06	0.2715	0.609027	
N_Gemme:Defogl:Dirad	1	3.6482e-05	3.6482e-05	2.9109	0.106183	
Residuals	17	2.1306e-04	1.2533e-05			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.578
 DIPROVE_2: 0.495
 DIPROVE_3: n.s

BrixA

Analysis of Variance Table

Response: Co_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	5.9240	5.9240	9.7569	0.006184	**
DIPROVE_2	1	3.3855	3.3855	5.5760	0.030407	*
DIPROVE_3	1	0.0147	0.0147	0.0243	0.878073	
N_Gemme	1	0.9086	0.9086	1.4964	0.237910	
Defogl	1	0.0003	0.0003	0.0005	0.983078	
Dirad	1	1.5835	1.5835	2.6080	0.124731	
Vigoria	1	1.2171	1.2171	2.0046	0.174880	

N_Gemme:Defogl	1	0.1351	0.1351	0.2226	0.643099
N_Gemme:Dirad	1	1.5089	1.5089	2.4851	0.133352
Defogl:Dirad	1	0.4202	0.4202	0.6921	0.416998
N_Gemme:Vigoria	1	0.3055	0.3055	0.5031	0.487754
Defogl:Vigoria	1	0.4129	0.4129	0.6800	0.420996
Dirad:Vigoria	1	0.1895	0.1895	0.3122	0.583646
N_Gemme:Defogl:Dirad	1	1.8619	1.8619	3.0666	0.097937 .
Residuals	17	10.3218	0.6072		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.604
DIPROVE_2: 0.497
DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Co_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.019481	0.019481	6.8112	0.0183040	*
DIPROVE_2	1	0.001454	0.001454	0.5085	0.4854690	
DIPROVE_3	1	0.059469	0.059469	20.7920	0.0002779	***
N_Gemme	1	0.004084	0.004084	1.4277	0.2485435	
Defogl	1	0.000432	0.000432	0.1511	0.7022745	
Dirad	1	0.007016	0.007016	2.4530	0.1357213	
Vigoria	1	0.006719	0.006719	2.3491	0.1437553	
N_Gemme:Defogl	1	0.012977	0.012977	4.5371	0.0480727	*
N_Gemme:Dirad	1	0.000103	0.000103	0.0359	0.8519994	
Defogl:Dirad	1	0.000163	0.000163	0.0570	0.8141741	
N_Gemme:Vigoria	1	0.001226	0.001226	0.4286	0.5214596	
Defogl:Vigoria	1	0.014815	0.014815	5.1796	0.0360811	*
Dirad:Vigoria	1	0.000591	0.000591	0.2066	0.6551592	
N_Gemme:Defogl:Dirad	1	0.002086	0.002086	0.7294	0.4049407	
Residuals	17	0.048623	0.002860			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.535
DIPROVE_2: n.s
DIPROVE_3: 0.742

AcTitola

Analysis of Variance Table

Response: Co_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.01989	0.01989	0.1744	0.68148	
DIPROVE_2	1	0.01282	0.01282	0.1124	0.74157	
DIPROVE_3	1	0.63937	0.63937	5.6044	0.03004	*
N_Gemme	1	0.35902	0.35902	3.1470	0.09398	.
Defogl	1	0.01697	0.01697	0.1488	0.70448	
Dirad	1	0.02494	0.02494	0.2186	0.64602	
Vigoria	1	0.08110	0.08110	0.7109	0.41085	
N_Gemme:Defogl	1	0.03305	0.03305	0.2897	0.59739	
N_Gemme:Dirad	1	0.18210	0.18210	1.5962	0.22349	
Defogl:Dirad	1	0.00503	0.00503	0.0441	0.83622	

N_Gemme:Vigoria	1	0.00362	0.00362	0.0317	0.86077
Defogl:Vigoria	1	0.03132	0.03132	0.2745	0.60709
Dirad:Vigoria	1	0.01228	0.01228	0.1076	0.74690
N_Gemme:Defogl:Dirad	1	0.01084	0.01084	0.0950	0.76160
Residuals	17	1.93943	0.11408		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.498

AcTartA

Analysis of Variance Table

Response: Co_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.52319	0.52319	3.2574	0.088848 .
DIPROVE_2	1	0.00073	0.00073	0.0046	0.946973
DIPROVE_3	1	2.07035	2.07035	12.8899	0.002256 **
N_Gemme	1	0.05486	0.05486	0.3415	0.566621
Defogl	1	0.00831	0.00831	0.0517	0.822794
Dirad	1	0.00016	0.00016	0.0010	0.975264
Vigoria	1	0.12954	0.12954	0.8065	0.381690
N_Gemme:Defogl	1	0.00929	0.00929	0.0578	0.812855
N_Gemme:Dirad	1	0.07039	0.07039	0.4382	0.516843
Defogl:Dirad	1	0.04703	0.04703	0.2928	0.595437
N_Gemme:Vigoria	1	0.01510	0.01510	0.0940	0.762868
Defogl:Vigoria	1	0.21717	0.21717	1.3521	0.260976
Dirad:Vigoria	1	0.04155	0.04155	0.2587	0.617574
N_Gemme:Defogl:Dirad	1	0.14294	0.14294	0.8899	0.358711
Residuals	17	2.73050	0.16062		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.657

AcMalicoA

Analysis of Variance Table

Response: Co_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.02603	0.02603	0.7120	0.4104836
DIPROVE_2	1	0.00006	0.00006	0.0017	0.9677345
DIPROVE_3	1	0.87530	0.87530	23.9446	0.0001371 ***
N_Gemme	1	0.26092	0.26092	7.1377	0.0160992 *
Defogl	1	0.00275	0.00275	0.0751	0.7873151
Dirad	1	0.03119	0.03119	0.8532	0.3685649
Vigoria	1	0.42809	0.42809	11.7108	0.0032486 **
N_Gemme:Defogl	1	0.02567	0.02567	0.7021	0.4136955
N_Gemme:Dirad	1	0.01592	0.01592	0.4355	0.5181460
Defogl:Dirad	1	0.05109	0.05109	1.3976	0.2533956
N_Gemme:Vigoria	1	0.01286	0.01286	0.3517	0.5609595
Defogl:Vigoria	1	0.19869	0.19869	5.4352	0.0322961 *
Dirad:Vigoria	1	0.04432	0.04432	1.2125	0.2861791

```

N_Gemme:Defogl:Dirad  1 0.02126 0.02126  0.5815 0.4561650
Residuals              17 0.62144 0.03656
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.765

```

KA

Analysis of Variance Table

Response: Co_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.14082	0.14082	6.5984	0.0199260	*
DIPROVE_2	1	0.01587	0.01587	0.7437	0.4004682	
DIPROVE_3	1	0.54146	0.54146	25.3711	0.0001014	***
N_Gemme	1	0.03104	0.03104	1.4547	0.2442975	
Defogl	1	0.02677	0.02677	1.2544	0.2782814	
Dirad	1	0.05881	0.05881	2.7555	0.1152522	
Vigoria	1	0.20064	0.20064	9.4015	0.0069928	**
N_Gemme:Defogl	1	0.00287	0.00287	0.1345	0.7183650	
N_Gemme:Dirad	1	0.00833	0.00833	0.3902	0.5404673	
Defogl:Dirad	1	0.03796	0.03796	1.7786	0.1999126	
N_Gemme:Vigoria	1	0.00427	0.00427	0.2003	0.6601515	
Defogl:Vigoria	1	0.08085	0.08085	3.7886	0.0683228	.
Dirad:Vigoria	1	0.00147	0.00147	0.0690	0.7959716	
N_Gemme:Defogl:Dirad	1	0.00209	0.00209	0.0981	0.7579402	
Residuals	17	0.36280	0.02134			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

DIPROVE_1: 0.529
DIPROVE_2: n.s
DIPROVE_3: -0.774

```

ApaA

Analysis of Variance Table

Response: Co_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	3229.7	3229.7	4.4033	0.05112	.
DIPROVE_2	1	1170.2	1170.2	1.5955	0.22360	
DIPROVE_3	1	24040.8	24040.8	32.7768	2.477e-05	***
N_Gemme	1	3453.8	3453.8	4.7089	0.04447	*
Defogl	1	1801.4	1801.4	2.4560	0.13550	
Dirad	1	2727.0	2727.0	3.7179	0.07071	.
Vigoria	1	5664.1	5664.1	7.7223	0.01286	*
N_Gemme:Defogl	1	4.7	4.7	0.0064	0.93730	
N_Gemme:Dirad	1	28.8	28.8	0.0392	0.84534	
Defogl:Dirad	1	85.8	85.8	0.1169	0.73656	
N_Gemme:Vigoria	1	268.5	268.5	0.3660	0.55316	
Defogl:Vigoria	1	2460.4	2460.4	3.3544	0.08461	.
Dirad:Vigoria	1	218.7	218.7	0.2981	0.59214	
N_Gemme:Defogl:Dirad	1	165.8	165.8	0.2260	0.64055	
Residuals	17	12469.0	733.5			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.811

Antoctot1A

Analysis of Variance Table

Response: Co_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	3536	3535.8	1.6650	0.21421
DIPROVE_2	1	16226	16226.3	7.6408	0.01327 *
DIPROVE_3	1	1240	1240.0	0.5839	0.45526
N_Gemme	1	447	446.9	0.2104	0.65225
Defogl	1	540	540.4	0.2545	0.62041
Dirad	1	15932	15931.9	7.5022	0.01399 *
Vigoria	1	3349	3349.0	1.5770	0.22618
N_Gemme:Defogl	1	1645	1644.7	0.7745	0.39111
N_Gemme:Dirad	1	40	39.7	0.0187	0.89287
Defogl:Dirad	1	1506	1505.7	0.7090	0.41145
N_Gemme:Vigoria	1	14	14.4	0.0068	0.93540
Defogl:Vigoria	1	19	18.9	0.0089	0.92600
Dirad:Vigoria	1	8673	8672.7	4.0839	0.05933 .
N_Gemme:Defogl:Dirad	1	1180	1180.0	0.5556	0.46620
Residuals	17	36102	2123.6		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.557
DIPROVE_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Co_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	5695	5695	0.7442	0.40032
DIPROVE_2	1	59978	59978	7.8380	0.01231 *
DIPROVE_3	1	6735	6735	0.8801	0.36132
N_Gemme	1	150	150	0.0196	0.89020
Defogl	1	33296	33296	4.3512	0.05236 .
Dirad	1	399	399	0.0522	0.82207
Vigoria	1	12655	12655	1.6538	0.21569
N_Gemme:Defogl	1	17644	17644	2.3058	0.14727
N_Gemme:Dirad	1	10293	10293	1.3451	0.26217
Defogl:Dirad	1	31995	31995	4.1812	0.05668 .
N_Gemme:Vigoria	1	5219	5219	0.6820	0.42033
Defogl:Vigoria	1	497	497	0.0649	0.80195
Dirad:Vigoria	1	6409	6409	0.8375	0.37291
N_Gemme:Defogl:Dirad	1	17	17	0.0022	0.96299
Residuals	17	130088	7652		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: -0.562
 DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	18835	18834.6	2.7291	0.11688
DIPROVE_2	1	4922	4921.8	0.7132	0.41013
DIPROVE_3	1	16744	16743.6	2.4261	0.13775
N_Gemme	1	3575	3574.9	0.5180	0.48148
Defogl	1	3870	3870.2	0.5608	0.46418
Dirad	1	6593	6593.0	0.9553	0.34208
Vigoria	1	3232	3232.1	0.4683	0.50298
N_Gemme:Defogl	1	12606	12605.5	1.8265	0.19425
N_Gemme:Dirad	1	303	302.8	0.0439	0.83658
Defogl:Dirad	1	9114	9114.0	1.3206	0.26640
N_Gemme:Vigoria	1	9046	9046.4	1.3108	0.26811
Defogl:Vigoria	1	4210	4209.9	0.6100	0.44553
Dirad:Vigoria	1	20915	20915.4	3.0306	0.09977 .
N_Gemme:Defogl:Dirad	1	21111	21110.7	3.0589	0.09833 .
Residuals	17	117324	6901.4		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Co_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	216279	216279	5.8697	0.02686 *
DIPROVE_2	1	23125	23125	0.6276	0.43915
DIPROVE_3	1	66246	66246	1.7979	0.19761
N_Gemme	1	6285	6285	0.1706	0.68477
Defogl	1	113237	113237	3.0732	0.09761 .
Dirad	1	88332	88332	2.3973	0.13996
Vigoria	1	10452	10452	0.2836	0.60121
N_Gemme:Defogl	1	69721	69721	1.8922	0.18681
N_Gemme:Dirad	1	31245	31245	0.8480	0.37001
Defogl:Dirad	1	143443	143443	3.8930	0.06497 .
N_Gemme:Vigoria	1	1826	1826	0.0496	0.82648
Defogl:Vigoria	1	39033	39033	1.0593	0.31779
Dirad:Vigoria	1	53202	53202	1.4439	0.24598
N_Gemme:Defogl:Dirad	1	27423	27423	0.7443	0.40031
Residuals	17	626392	36847		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.507
 DIPROVE_2: n.s
 DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Co_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1519959	1519959	4.4246	0.05062 .
DIPROVE_2	1	93066	93066	0.2709	0.60943
DIPROVE_3	1	1000904	1000904	2.9136	0.10603
N_Gemme	1	12175	12175	0.0354	0.85291
Defogl	1	177346	177346	0.5162	0.48221
Dirad	1	1379437	1379437	4.0155	0.06129 .
Vigoria	1	935824	935824	2.7242	0.11719
N_Gemme:Defogl	1	15531	15531	0.0452	0.83414
N_Gemme:Dirad	1	435954	435954	1.2690	0.27560
Defogl:Dirad	1	146397	146397	0.4262	0.52261
N_Gemme:Vigoria	1	1155914	1155914	3.3648	0.08417 .
Defogl:Vigoria	1	130230	130230	0.3791	0.54624
Dirad:Vigoria	1	692	692	0.0020	0.96473
N_Gemme:Defogl:Dirad	1	501830	501830	1.4608	0.24334
Residuals	17	5839981	343528		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Co_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.00054489	0.00054489	28.9868	4.948e-05 ***
DIPROVE_2	1	0.00004790	0.00004790	2.5484	0.12883
DIPROVE_3	1	0.00000242	0.00000242	0.1289	0.72399
N_Gemme	1	0.00001048	0.00001048	0.5574	0.46552
Defogl	1	0.00005691	0.00005691	3.0276	0.09993 .
Dirad	1	0.00005958	0.00005958	3.1695	0.09290 .
Vigoria	1	0.00011234	0.00011234	5.9763	0.02569 *
N_Gemme:Defogl	1	0.00000175	0.00000175	0.0930	0.76416
N_Gemme:Dirad	1	0.00000430	0.00000430	0.2289	0.63847
Defogl:Dirad	1	0.00001333	0.00001333	0.7092	0.41139
N_Gemme:Vigoria	1	0.00003282	0.00003282	1.7457	0.20392
Defogl:Vigoria	1	0.00005304	0.00005304	2.8216	0.11128
Dirad:Vigoria	1	0.00003676	0.00003676	1.9558	0.17994
N_Gemme:Defogl:Dirad	1	0.00004987	0.00004987	2.6530	0.12174
Residuals	17	0.00031956	0.00001880		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.794
 DIPROVE_2: n.s
 DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Co_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	27.9376	27.9376	30.6547	3.623e-05	***
DIPROVE_2	1	2.6193	2.6193	2.8741	0.10825	
DIPROVE_3	1	0.0221	0.0221	0.0242	0.87816	
N_Gemme	1	0.3917	0.3917	0.4298	0.52086	
Defogl	1	2.7682	2.7682	3.0374	0.09942	.
Dirad	1	3.2806	3.2806	3.5996	0.07491	.
Vigoria	1	5.2064	5.2064	5.7128	0.02869	*
N_Gemme:Defogl	1	0.0992	0.0992	0.1088	0.74552	
N_Gemme:Dirad	1	0.2369	0.2369	0.2600	0.61670	
Defogl:Dirad	1	0.7067	0.7067	0.7754	0.39083	
N_Gemme:Vigoria	1	1.3530	1.3530	1.4846	0.23969	
Defogl:Vigoria	1	2.4491	2.4491	2.6873	0.11953	
Dirad:Vigoria	1	1.5504	1.5504	1.7011	0.20952	
N_Gemme:Defogl:Dirad	1	2.4679	2.4679	2.7080	0.11821	
Residuals	17	15.4932	0.9114			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.802
 DIPROVE_2: n.s
 DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Co_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.001886	0.0018859	0.1653	0.6894
DIPROVE_2	1	0.004656	0.0046556	0.4080	0.5315
DIPROVE_3	1	0.005889	0.0058891	0.5161	0.4823
N_Gemme	1	0.005391	0.0053909	0.4724	0.5011
Defogl	1	0.004435	0.0044348	0.3887	0.5413
Dirad	1	0.004066	0.0040656	0.3563	0.5584
Vigoria	1	0.001997	0.0019966	0.1750	0.6810
N_Gemme:Defogl	1	0.002793	0.0027933	0.2448	0.6271
N_Gemme:Dirad	1	0.000071	0.0000715	0.0063	0.9378
Defogl:Dirad	1	0.003165	0.0031653	0.2774	0.6052
N_Gemme:Vigoria	1	0.000071	0.0000707	0.0062	0.9382
Defogl:Vigoria	1	0.019181	0.0191811	1.6810	0.2121
Dirad:Vigoria	1	0.007049	0.0070489	0.6178	0.4427
N_Gemme:Defogl:Dirad	1	0.002129	0.0021289	0.1866	0.6712
Residuals	17	0.193979	0.0114105		

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

AcTitolB

Analysis of Variance Table

Response: Co_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.00055	0.00055	0.0036	0.95283

DIPROVE_2	1	0.47398	0.47398	3.0901	0.09676	.
DIPROVE_3	1	0.72789	0.72789	4.7455	0.04374	*
N_Gemme	1	0.02686	0.02686	0.1751	0.68085	
Defogl	1	0.01706	0.01706	0.1112	0.74285	
Dirad	1	0.08873	0.08873	0.5785	0.45733	
Vigoria	1	0.00135	0.00135	0.0088	0.92648	
N_Gemme:Defogl	1	0.04578	0.04578	0.2984	0.59196	
N_Gemme:Dirad	1	0.06984	0.06984	0.4553	0.50889	
Defogl:Dirad	1	0.00219	0.00219	0.0143	0.90619	
N_Gemme:Vigoria	1	0.00315	0.00315	0.0205	0.88778	
Defogl:Vigoria	1	0.12554	0.12554	0.8185	0.37828	
Dirad:Vigoria	1	0.05592	0.05592	0.3646	0.55396	
N_Gemme:Defogl:Dirad	1	0.10811	0.10811	0.7048	0.41282	
Residuals	17	2.60756	0.15339			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.467

AcTartB

Analysis of Variance Table

Response: Co_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.11227	0.11227	1.8162	0.19546
DIPROVE_2	1	0.07364	0.07364	1.1913	0.29030
DIPROVE_3	1	0.51819	0.51819	8.3824	0.01006 *
N_Gemme	1	0.11068	0.11068	1.7904	0.19851
Defogl	1	0.02509	0.02509	0.4059	0.53254
Dirad	1	0.11897	0.11897	1.9245	0.18328
Vigoria	1	0.04682	0.04682	0.7574	0.39626
N_Gemme:Defogl	1	0.00334	0.00334	0.0541	0.81885
N_Gemme:Dirad	1	0.10570	0.10570	1.7099	0.20841
Defogl:Dirad	1	0.15370	0.15370	2.4864	0.13326
N_Gemme:Vigoria	1	0.00327	0.00327	0.0529	0.82082
Defogl:Vigoria	1	0.06224	0.06224	1.0069	0.32972
Dirad:Vigoria	1	0.00492	0.00492	0.0796	0.78128
N_Gemme:Defogl:Dirad	1	0.09366	0.09366	1.5151	0.23512
Residuals	17	1.05091	0.06182		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.575

AcMalicoB

Analysis of Variance Table

Response: Co_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.81510	0.81510	9.3889	0.007024 **
DIPROVE_2	1	0.07889	0.07889	0.9088	0.353801
DIPROVE_3	1	0.00597	0.00597	0.0687	0.796321
N_Gemme	1	0.18973	0.18973	2.1855	0.157609

Defogl	1	0.05711	0.05711	0.6579	0.428521
Dirad	1	0.01494	0.01494	0.1721	0.683405
Vigoria	1	0.29925	0.29925	3.4470	0.080784 .
N_Gemme:Defogl	1	0.00033	0.00033	0.0038	0.951385
N_Gemme:Dirad	1	0.12610	0.12610	1.4525	0.244642
Defogl:Dirad	1	0.00016	0.00016	0.0018	0.966565
N_Gemme:Vigoria	1	0.00674	0.00674	0.0776	0.783955
Defogl:Vigoria	1	0.37984	0.37984	4.3753	0.051783 .
Dirad:Vigoria	1	0.01398	0.01398	0.1610	0.693257
N_Gemme:Defogl:Dirad	1	0.00970	0.00970	0.1117	0.742301
Residuals	17	1.47585	0.08681		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.596
DIPROVE_2: n.s
DIPROVE_3: n.s

KB

Analysis of Variance Table

Response: Co_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.18229	0.182285	4.5528	0.04773 *
DIPROVE_2	1	0.05699	0.056992	1.4234	0.24922
DIPROVE_3	1	0.03023	0.030228	0.7550	0.39701
N_Gemme	1	0.03949	0.039495	0.9864	0.33455
Defogl	1	0.01987	0.019867	0.4962	0.49071
Dirad	1	0.00565	0.005650	0.1411	0.71181
Vigoria	1	0.01161	0.011614	0.2901	0.59715
N_Gemme:Defogl	1	0.00459	0.004591	0.1147	0.73904
N_Gemme:Dirad	1	0.00908	0.009080	0.2268	0.63999
Defogl:Dirad	1	0.01368	0.013678	0.3416	0.56657
N_Gemme:Vigoria	1	0.00006	0.000064	0.0016	0.96846
Defogl:Vigoria	1	0.15949	0.159487	3.9833	0.06223 .
Dirad:Vigoria	1	0.00200	0.001998	0.0499	0.82588
N_Gemme:Defogl:Dirad	1	0.00006	0.000065	0.0016	0.96840
Residuals	17	0.68065	0.040038		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.46
DIPROVE_2: n.s
DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Co_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1632.3	1632.3	2.7695	0.114400
DIPROVE_2	1	1774.2	1774.2	3.0103	0.100826
DIPROVE_3	1	26351.0	26351.0	44.7095	3.826e-06 ***
N_Gemme	1	7310.8	7310.8	12.4042	0.002616 **
Defogl	1	2555.8	2555.8	4.3364	0.052723 .
Dirad	1	2605.3	2605.3	4.4204	0.050715 .
Vigoria	1	3663.5	3663.5	6.2159	0.023271 *

N_Gemme:Defogl	1	396.3	396.3	0.6725	0.423541
N_Gemme:Dirad	1	200.6	200.6	0.3403	0.567327
Defogl:Dirad	1	327.6	327.6	0.5558	0.466123
N_Gemme:Vigoria	1	0.9	0.9	0.0015	0.969120
Defogl:Vigoria	1	4157.3	4157.3	7.0537	0.016636 *
Dirad:Vigoria	1	846.7	846.7	1.4366	0.247131
N_Gemme:Defogl:Dirad	1	55.8	55.8	0.0947	0.762028
Residuals	17	10019.5	589.4		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: -0.851

Antoctot1B

Analysis of Variance Table

Response: Co_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	26355.8	26355.8	20.6214	0.0002892	***
DIPROVE_2	1	21.3	21.3	0.0167	0.8987553	
DIPROVE_3	1	10336.3	10336.3	8.0874	0.0112194	*
N_Gemme	1	778.2	778.2	0.6089	0.4459310	
Defogl	1	7514.2	7514.2	5.8793	0.0267519	*
Dirad	1	12607.0	12607.0	9.8641	0.0059607	**
Vigoria	1	7736.8	7736.8	6.0535	0.0248813	*
N_Gemme:Defogl	1	74.5	74.5	0.0583	0.8121016	
N_Gemme:Dirad	1	701.0	701.0	0.5485	0.4690311	
Defogl:Dirad	1	1977.9	1977.9	1.5475	0.2303810	
N_Gemme:Vigoria	1	1112.0	1112.0	0.8700	0.3640043	
Defogl:Vigoria	1	2783.8	2783.8	2.1781	0.1582650	
Dirad:Vigoria	1	5240.9	5240.9	4.1006	0.0588670	.
N_Gemme:Defogl:Dirad	1	486.2	486.2	0.3804	0.5455736	
Residuals	17	21727.3	1278.1			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.74
 DIPROVE_2: n.s
 DIPROVE_3: -0.568

Poliftot1B

Analysis of Variance Table

Response: Co_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	150742	150742	6.2194	0.023237	*
DIPROVE_2	1	33356	33356	1.3762	0.256918	
DIPROVE_3	1	284724	284724	11.7474	0.003211	**
N_Gemme	1	8631	8631	0.3561	0.558533	
Defogl	1	15742	15742	0.6495	0.431416	
Dirad	1	114262	114262	4.7143	0.044359	*
Vigoria	1	164404	164404	6.7831	0.018509	*
N_Gemme:Defogl	1	3578	3578	0.1476	0.705582	
N_Gemme:Dirad	1	927	927	0.0382	0.847277	
Defogl:Dirad	1	12369	12369	0.5103	0.484695	

N_Gemme:Vigoria	1	36278	36278	1.4968	0.237855
Defogl:Vigoria	1	45455	45455	1.8754	0.188674
Dirad:Vigoria	1	26646	26646	1.0994	0.309083
N_Gemme:Defogl:Dirad	1	7165	7165	0.2956	0.593697
Residuals	17	412034	24237		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.518
DIPROVE_2: n.s
DIPROVE_3: -0.639

Antoctot2B

Analysis of Variance Table

Response: Co_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	12696	12696.5	3.5372	0.07725 .
DIPROVE_2	1	1045	1045.2	0.2912	0.59645
DIPROVE_3	1	7202	7202.2	2.0065	0.17469
N_Gemme	1	549	549.3	0.1530	0.70051
Defogl	1	15577	15576.7	4.3396	0.05264 .
Dirad	1	6344	6344.0	1.7674	0.20127
Vigoria	1	276	275.9	0.0769	0.78494
N_Gemme:Defogl	1	24	23.9	0.0067	0.93590
N_Gemme:Dirad	1	1090	1089.8	0.3036	0.58879
Defogl:Dirad	1	1702	1701.5	0.4740	0.50042
N_Gemme:Vigoria	1	634	634.4	0.1767	0.67946
Defogl:Vigoria	1	1535	1535.2	0.4277	0.52186
Dirad:Vigoria	1	918	917.9	0.2557	0.61957
N_Gemme:Defogl:Dirad	1	4367	4366.6	1.2165	0.28542
Residuals	17	61020	3589.4		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	9519	9519	0.1024	0.75290
DIPROVE_2	1	119205	119205	1.2820	0.27325
DIPROVE_3	1	213530	213530	2.2963	0.14805
N_Gemme	1	0	0	0.0000	0.99954
Defogl	1	32890	32890	0.3537	0.55986
Dirad	1	26847	26847	0.2887	0.59801
Vigoria	1	185481	185481	1.9947	0.17589
N_Gemme:Defogl	1	280	280	0.0030	0.95686
N_Gemme:Dirad	1	8359	8359	0.0899	0.76795
Defogl:Dirad	1	5061	5061	0.0544	0.81832
N_Gemme:Vigoria	1	301185	301185	3.2390	0.08968 .
Defogl:Vigoria	1	141089	141089	1.5173	0.23480
Dirad:Vigoria	1	4693	4693	0.0505	0.82493

N_Gemme:Defogl:Dirad 1 15829 15829 0.1702 0.68507
 Residuals 17 1580775 92987

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

CORTIGLIANO 2008 IASMA

PvinacciaA

Analysis of Variance Table

Response: Co_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	198218	198218	2.3020	0.1475799
IASMA_2	1	928	928	0.0108	0.9185476
IASMA_3	1	40275	40275	0.4677	0.5032476
N_Gemme	1	263664	263664	3.0621	0.0981646 .
Defogl	1	826190	826190	9.5951	0.0065382 **
Dirad	1	2132777	2132777	24.7693	0.0001150 ***
Vigoria	1	3397	3397	0.0395	0.8449152
N_Gemme:Defogl	1	64299	64299	0.7468	0.3995361
N_Gemme:Dirad	1	386830	386830	4.4925	0.0490639 *
Defogl:Dirad	1	12083	12083	0.1403	0.7125928
N_Gemme:Vigoria	1	212148	212148	2.4638	0.1349205
Defogl:Vigoria	1	42993	42993	0.4993	0.4893800
Dirad:Vigoria	1	1298071	1298071	15.0753	0.0011957 **
N_Gemme:Defogl:Dirad	1	55479	55479	0.6443	0.4332316
Residuals	17	1463794	86106		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00000172	1.7250e-06	0.0835	0.77611
IASMA_2	1	0.00000770	7.7010e-06	0.3728	0.54956
IASMA_3	1	0.00000015	1.4800e-07	0.0072	0.93345
N_Gemme	1	0.00000751	7.5090e-06	0.3635	0.55452
Defogl	1	0.00007581	7.5814e-05	3.6701	0.07237 .
Dirad	1	0.00001377	1.3769e-05	0.6666	0.42555 .
Vigoria	1	0.00001153	1.1530e-05	0.5582	0.46521
N_Gemme:Defogl	1	0.00000008	7.6000e-08	0.0037	0.95248
N_Gemme:Dirad	1	0.00005600	5.6001e-05	2.7110	0.11802
Defogl:Dirad	1	0.00000421	4.2140e-06	0.2040	0.65722
N_Gemme:Vigoria	1	0.00000000	4.0000e-09	0.0002	0.98874
Defogl:Vigoria	1	0.00000458	4.5790e-06	0.2217	0.64375
Dirad:Vigoria	1	0.00001090	1.0898e-05	0.5275	0.47752
N_Gemme:Defogl:Dirad	1	0.00003148	3.1476e-05	1.5237	0.23385
Residuals	17	0.00035117	2.0657e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

BrixA

Analysis of Variance Table

Response: Co_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.0229	0.0229	0.0228	0.88177
IASMA_2	1	0.4134	0.4134	0.4123	0.52936
IASMA_3	1	0.1970	0.1970	0.1965	0.66316
N_Gemme	1	0.3439	0.3439	0.3430	0.56580
Defogl	1	3.2632	3.2632	3.2547	0.08897 .
Dirad	1	0.6767	0.6767	0.6749	0.42271
Vigoria	1	0.7696	0.7696	0.7676	0.39317
N_Gemme:Defogl	1	0.0104	0.0104	0.0103	0.92024
N_Gemme:Dirad	1	2.8499	2.8499	2.8425	0.11007
Defogl:Dirad	1	0.2040	0.2040	0.2035	0.65761
N_Gemme:Vigoria	1	0.0001	0.0001	0.0001	0.99153
Defogl:Vigoria	1	0.1613	0.1613	0.1608	0.69338
Dirad:Vigoria	1	0.5733	0.5733	0.5718	0.45991
N_Gemme:Defogl:Dirad	1	1.6598	1.6598	1.6554	0.21546
Residuals	17	17.0444	1.0026		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PhA

Analysis of Variance Table

Response: Co_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.007041	0.007041	2.1623	0.1596970
IASMA_2	1	0.002914	0.002914	0.8951	0.3573614
IASMA_3	1	0.063628	0.063628	19.5408	0.0003744 ***
N_Gemme	1	0.001232	0.001232	0.3783	0.5466355
Defogl	1	0.001402	0.001402	0.4306	0.5204694
Dirad	1	0.008752	0.008752	2.6877	0.1194970
Vigoria	1	0.014442	0.014442	4.4352	0.0503728 .
N_Gemme:Defogl	1	0.008365	0.008365	2.5690	0.1273965
N_Gemme:Dirad	1	0.000176	0.000176	0.0542	0.8187136
Defogl:Dirad	1	0.000015	0.000015	0.0046	0.9467265
N_Gemme:Vigoria	1	0.000431	0.000431	0.1323	0.7205176
Defogl:Vigoria	1	0.012564	0.012564	3.8584	0.0660613 .
Dirad:Vigoria	1	0.000673	0.000673	0.2066	0.6552314
N_Gemme:Defogl:Dirad	1	0.002250	0.002250	0.6911	0.4173231
Residuals	17	0.055355	0.003256		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: 0.731

AcTitola

Analysis of Variance Table

Response: Co_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.07235	0.07235	0.6677	0.42516
IASMA_2	1	0.01879	0.01879	0.1734	0.68233
IASMA_3	1	0.79441	0.79441	7.3312	0.01494 *
N_Gemme	1	0.04340	0.04340	0.4005	0.53526
Defogl	1	0.04225	0.04225	0.3899	0.54065
Dirad	1	0.00160	0.00160	0.0148	0.90465
Vigoria	1	0.05503	0.05503	0.5079	0.48573
N_Gemme:Defogl	1	0.01568	0.01568	0.1447	0.70837
N_Gemme:Dirad	1	0.23695	0.23695	2.1867	0.15750
Defogl:Dirad	1	0.01116	0.01116	0.1030	0.75217
N_Gemme:Vigoria	1	0.11169	0.11169	1.0307	0.32422
Defogl:Vigoria	1	0.00315	0.00315	0.0291	0.86659
Dirad:Vigoria	1	0.03177	0.03177	0.2932	0.59520
N_Gemme:Defogl:Dirad	1	0.09143	0.09143	0.8438	0.37117
Residuals	17	1.84213	0.10836		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: -0.549

AcTartA

Analysis of Variance Table

Response: Co_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.07316	0.07316	0.6190	0.4422493
IASMA_2	1	0.02473	0.02473	0.2093	0.6531407
IASMA_3	1	2.40880	2.40880	20.3809	0.0003061 ***
N_Gemme	1	0.14639	0.14639	1.2387	0.2812217
Defogl	1	0.05454	0.05454	0.4615	0.5060867
Dirad	1	0.00013	0.00013	0.0011	0.9739524
Vigoria	1	0.19580	0.19580	1.6567	0.2153031
N_Gemme:Defogl	1	0.02975	0.02975	0.2517	0.6222889
N_Gemme:Dirad	1	0.14151	0.14151	1.1973	0.2891282
Defogl:Dirad	1	0.11968	0.11968	1.0126	0.3283957
N_Gemme:Vigoria	1	0.17929	0.17929	1.5169	0.2348485
Defogl:Vigoria	1	0.27298	0.27298	2.3097	0.1469523
Dirad:Vigoria	1	0.01677	0.01677	0.1419	0.7110632
N_Gemme:Defogl:Dirad	1	0.38835	0.38835	3.2859	0.0875772 .
Residuals	17	2.00921	0.11819		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: -0.738

AcMalicoA

Analysis of Variance Table

Response: Co_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.13488	0.13488	2.8987	0.106862
IASMA_2	1	0.02877	0.02877	0.6183	0.442486
IASMA_3	1	0.61609	0.61609	13.2405	0.002031 **
N_Gemme	1	0.00167	0.00167	0.0359	0.851998
Defogl	1	0.11992	0.11992	2.5773	0.126822
Dirad	1	0.03475	0.03475	0.7469	0.399498
Vigoria	1	0.55459	0.55459	11.9189	0.003042 **
N_Gemme:Defogl	1	0.03023	0.03023	0.6496	0.431376
N_Gemme:Dirad	1	0.04986	0.04986	1.0716	0.315087
Defogl:Dirad	1	0.05201	0.05201	1.1178	0.305194
N_Gemme:Vigoria	1	0.01399	0.01399	0.3006	0.590648
Defogl:Vigoria	1	0.11702	0.11702	2.5148	0.131207
Dirad:Vigoria	1	0.06052	0.06052	1.3006	0.269908
N_Gemme:Defogl:Dirad	1	0.01025	0.01025	0.2204	0.644722
Residuals	17	0.79102	0.04653		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.662

KA

Analysis of Variance Table

Response: Co_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.03822	0.03822	1.8341	0.193378
IASMA_2	1	0.04782	0.04782	2.2946	0.148199
IASMA_3	1	0.63910	0.63910	30.6670	3.615e-05 ***
N_Gemme	1	0.03361	0.03361	1.6129	0.221199
Defogl	1	0.00128	0.00128	0.0613	0.807387
Dirad	1	0.06087	0.06087	2.9209	0.105632
Vigoria	1	0.22997	0.22997	11.0352	0.004035 **
N_Gemme:Defogl	1	0.00379	0.00379	0.1820	0.675052
N_Gemme:Dirad	1	0.00007	0.00007	0.0032	0.955482
Defogl:Dirad	1	0.03035	0.03035	1.4564	0.244030
N_Gemme:Vigoria	1	0.01350	0.01350	0.6477	0.432046
Defogl:Vigoria	1	0.05498	0.05498	2.6382	0.122719
Dirad:Vigoria	1	0.00002	0.00002	0.0010	0.975005
N_Gemme:Defogl:Dirad	1	0.00820	0.00820	0.3935	0.538789
Residuals	17	0.35428	0.02084		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: -0.802

ApaA

Analysis of Variance Table

Response: Co_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	3734.0	3734.0	6.2368	0.023072	*
IASMA_2	1	1908.6	1908.6	3.1879	0.092036	.
IASMA_3	1	21493.1	21493.1	35.8996	1.458e-05	***
N_Gemme	1	57.8	57.8	0.0966	0.759721	
Defogl	1	5765.2	5765.2	9.6295	0.006461	**
Dirad	1	1795.6	1795.6	2.9991	0.101416	
Vigoria	1	7953.7	7953.7	13.2849	0.002004	**
N_Gemme:Defogl	1	4.3	4.3	0.0073	0.933139	
N_Gemme:Dirad	1	10.2	10.2	0.0170	0.897674	
Defogl:Dirad	1	31.2	31.2	0.0521	0.822244	
N_Gemme:Vigoria	1	567.1	567.1	0.9472	0.344088	
Defogl:Vigoria	1	3237.9	3237.9	5.4082	0.032673	*
Dirad:Vigoria	1	224.8	224.8	0.3755	0.548128	
N_Gemme:Defogl:Dirad	1	827.0	827.0	1.3813	0.256086	
Residuals	17	10177.9	598.7			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.518

IASMA_2: n.s

IASMA_3: -0.824

Antoctot1A

Analysis of Variance Table

Response: Co_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	5588	5587.7	1.9882	0.17657	
IASMA_2	1	77	76.6	0.0272	0.87085	
IASMA_3	1	14	14.1	0.0050	0.94442	
N_Gemme	1	19	18.6	0.0066	0.93616	
Defogl	1	1567	1566.6	0.5574	0.46550	
Dirad	1	11006	11006.4	3.9162	0.06426	.
Vigoria	1	286	286.2	0.1018	0.75354	
N_Gemme:Defogl	1	1503	1503.4	0.5349	0.47451	
N_Gemme:Dirad	1	170	170.2	0.0605	0.80858	
Defogl:Dirad	1	239	238.7	0.0849	0.77423	
N_Gemme:Vigoria	1	2347	2347.1	0.8351	0.37357	
Defogl:Vigoria	1	3426	3426.5	1.2192	0.28491	
Dirad:Vigoria	1	15969	15968.6	5.6818	0.02907	*
N_Gemme:Defogl:Dirad	1	459	459.4	0.1635	0.69102	
Residuals	17	47778	2810.5			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Co_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	20734	20734	2.3903	0.14050
IASMA_2	1	669	669	0.0771	0.78458
IASMA_3	1	30	30	0.0035	0.95372
N_Gemme	1	1804	1804	0.2080	0.65411
Defogl	1	59089	59089	6.8119	0.01830 *
Dirad	1	2458	2458	0.2833	0.60142
Vigoria	1	984	984	0.1134	0.74038
N_Gemme:Defogl	1	16344	16344	1.8842	0.18770
N_Gemme:Dirad	1	9810	9810	1.1309	0.30245
Defogl:Dirad	1	13342	13342	1.5381	0.23175
N_Gemme:Vigoria	1	34009	34009	3.9207	0.06412 .
Defogl:Vigoria	1	10700	10700	1.2336	0.28218
Dirad:Vigoria	1	160	160	0.0184	0.89370
N_Gemme:Defogl:Dirad	1	3473	3473	0.4004	0.53532
Residuals	17	147464	8674		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	2056	2056	0.2781	0.60479
IASMA_2	1	766	766	0.1037	0.75141
IASMA_3	1	354	354	0.0479	0.82936
N_Gemme	1	13953	13953	1.8872	0.18737
Defogl	1	24056	24056	3.2536	0.08902 .
Dirad	1	60	60	0.0081	0.92950
Vigoria	1	110	110	0.0149	0.90440
N_Gemme:Defogl	1	18514	18514	2.5040	0.13198
N_Gemme:Dirad	1	183	183	0.0247	0.87700
Defogl:Dirad	1	1631	1631	0.2207	0.64451
N_Gemme:Vigoria	1	1181	1181	0.1598	0.69433
Defogl:Vigoria	1	17484	17484	2.3648	0.14250
Dirad:Vigoria	1	36287	36287	4.9079	0.04067 *
N_Gemme:Defogl:Dirad	1	10075	10075	1.3626	0.25919
Residuals	17	125690	7394		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Co_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	8890	8890	0.1835	0.67379

IASMA_2	1	1862	1862	0.0384	0.84692
IASMA_3	1	21	21	0.0004	0.98383
N_Gemme	1	44152	44152	0.9112	0.35318
Defogl	1	1611	1611	0.0332	0.85748
Dirad	1	282597	282597	5.8318	0.02729 *
Vigoria	1	300	300	0.0062	0.93825
N_Gemme:Defogl	1	101078	101078	2.0859	0.16684
N_Gemme:Dirad	1	83944	83944	1.7323	0.20559
Defogl:Dirad	1	48992	48992	1.0110	0.32876
N_Gemme:Vigoria	1	20212	20212	0.4171	0.52701
Defogl:Vigoria	1	82261	82261	1.6976	0.20998
Dirad:Vigoria	1	16540	16540	0.3413	0.56673
N_Gemme:Defogl:Dirad	1	3	3	0.0001	0.99357
Residuals	17	823779	48458		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Co_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	202236	202236	0.4640	0.50495
IASMA_2	1	131811	131811	0.3024	0.58954
IASMA_3	1	192138	192138	0.4408	0.51565
N_Gemme	1	519885	519885	1.1927	0.29003
Defogl	1	1793498	1793498	4.1145	0.05848 .
Dirad	1	741668	741668	1.7015	0.20948
Vigoria	1	58813	58813	0.1349	0.71791
N_Gemme:Defogl	1	43695	43695	0.1002	0.75540
N_Gemme:Dirad	1	209923	209923	0.4816	0.49708
Defogl:Dirad	1	7392	7392	0.0170	0.89792
N_Gemme:Vigoria	1	1139052	1139052	2.6131	0.12439
Defogl:Vigoria	1	467562	467562	1.0726	0.31486
Dirad:Vigoria	1	55025	55025	0.1262	0.72674
N_Gemme:Defogl:Dirad	1	372310	372310	0.8541	0.36832
Residuals	17	7410231	435896		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Co_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00000458	0.00000458	0.1398	0.713134
IASMA_2	1	0.00012015	0.00012015	3.6701	0.072372 .
IASMA_3	1	0.00003080	0.00003080	0.9408	0.345676
N_Gemme	1	0.00000133	0.00000133	0.0405	0.842799

Defogl	1	0.00043275	0.00043275	13.2188	0.002044	**
Dirad	1	0.00001022	0.00001022	0.3122	0.583605	
Vigoria	1	0.00003239	0.00003239	0.9893	0.333857	
N_Gemme:Defogl	1	0.00001973	0.00001973	0.6028	0.448177	
N_Gemme:Dirad	1	0.00000506	0.00000506	0.1546	0.699021	
Defogl:Dirad	1	0.00003673	0.00003673	1.1219	0.304337	
N_Gemme:Vigoria	1	0.00000763	0.00000763	0.2331	0.635385	
Defogl:Vigoria	1	0.00000980	0.00000980	0.2992	0.591469	
Dirad:Vigoria	1	0.00004841	0.00004841	1.4787	0.240599	
N_Gemme:Defogl:Dirad	1	0.00002985	0.00002985	0.9118	0.353020	
Residuals	17	0.00055654	0.00003274			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

BrixB

Analysis of Variance Table

Response: Co_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.1582	0.1582	0.0980	0.758075	
IASMA_2	1	5.8853	5.8853	3.6449	0.073268	.
IASMA_3	1	2.1995	2.1995	1.3622	0.259270	
N_Gemme	1	0.0642	0.0642	0.0398	0.844290	
Defogl	1	21.0253	21.0253	13.0214	0.002168	**
Dirad	1	0.6689	0.6689	0.4143	0.528395	
Vigoria	1	1.7731	1.7731	1.0981	0.309350	
N_Gemme:Defogl	1	1.0356	1.0356	0.6413	0.434276	
N_Gemme:Dirad	1	0.2136	0.2136	0.1323	0.720566	
Defogl:Dirad	1	1.7926	1.7926	1.1102	0.306789	
N_Gemme:Vigoria	1	0.2070	0.2070	0.1282	0.724690	
Defogl:Vigoria	1	0.4480	0.4480	0.2774	0.605194	
Dirad:Vigoria	1	2.1159	2.1159	1.3104	0.268180	
N_Gemme:Defogl:Dirad	1	1.5456	1.5456	0.9572	0.341606	
Residuals	17	27.4495	1.6147			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Co_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.000193	0.0001925	0.0211	0.88628	
IASMA_2	1	0.003584	0.0035842	0.3924	0.53936	
IASMA_3	1	0.010302	0.0103023	1.1279	0.30308	
N_Gemme	1	0.010505	0.0105049	1.1501	0.29852	
Defogl	1	0.004936	0.0049361	0.5404	0.47228	
Dirad	1	0.000096	0.0000959	0.0105	0.91958	
Vigoria	1	0.006140	0.0061397	0.6722	0.42364	

N_Gemme:Defogl	1	0.003361	0.0033609	0.3680	0.55214
N_Gemme:Dirad	1	0.000009	0.0000086	0.0009	0.97585
Defogl:Dirad	1	0.019765	0.0197648	2.1639	0.15956
N_Gemme:Vigoria	1	0.006641	0.0066408	0.7270	0.40570
Defogl:Vigoria	1	0.030709	0.0307086	3.3620	0.08428 .
Dirad:Vigoria	1	0.005201	0.0052005	0.5694	0.46084
N_Gemme:Defogl:Dirad	1	0.000039	0.0000394	0.0043	0.94840
Residuals	17	0.155278	0.0091340		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Co_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00066	0.00066	0.0048	0.94574
IASMA_2	1	0.00022	0.00022	0.0016	0.96870
IASMA_3	1	0.48397	0.48397	3.5002	0.07868 .
N_Gemme	1	0.04104	0.04104	0.2968	0.59297
Defogl	1	0.13259	0.13259	0.9590	0.34118
Dirad	1	0.00015	0.00015	0.0011	0.97402
Vigoria	1	0.01408	0.01408	0.1018	0.75352
N_Gemme:Defogl	1	0.03843	0.03843	0.2780	0.60485
N_Gemme:Dirad	1	0.12308	0.12308	0.8902	0.35865
Defogl:Dirad	1	0.12649	0.12649	0.9148	0.35224
N_Gemme:Vigoria	1	0.31499	0.31499	2.2781	0.14957
Defogl:Vigoria	1	0.34342	0.34342	2.4837	0.13346
Dirad:Vigoria	1	0.00050	0.00050	0.0036	0.95288
N_Gemme:Defogl:Dirad	1	0.38428	0.38428	2.7792	0.11381
Residuals	17	2.35058	0.13827		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Co_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.11169	0.111692	2.1478	0.16103
IASMA_2	1	0.11770	0.117700	2.2633	0.15083
IASMA_3	1	0.08316	0.083163	1.5992	0.22308
N_Gemme	1	0.24101	0.241011	4.6345	0.04599 *
Defogl	1	0.00341	0.003414	0.0656	0.80086
Dirad	1	0.05560	0.055604	1.0692	0.31561
Vigoria	1	0.20115	0.201146	3.8679	0.06576 .
N_Gemme:Defogl	1	0.00931	0.009309	0.1790	0.67753
N_Gemme:Dirad	1	0.10455	0.104548	2.0104	0.17430
Defogl:Dirad	1	0.19605	0.196054	3.7700	0.06894 .

N_Gemme:Vigoria	1	0.15188	0.151876	2.9205	0.10565
Defogl:Vigoria	1	0.01848	0.018484	0.3554	0.55891
Dirad:Vigoria	1	0.01507	0.015073	0.2898	0.59730
N_Gemme:Defogl:Dirad	1	0.29028	0.290281	5.5819	0.03033 *
Residuals	17	0.88407	0.052004		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Co_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.26211	0.26211	2.6290	0.12332
IASMA_2	1	0.42315	0.42315	4.2444	0.05503 .
IASMA_3	1	0.08187	0.08187	0.8212	0.37750
N_Gemme	1	0.04033	0.04033	0.4046	0.53322
Defogl	1	0.56305	0.56305	5.6477	0.02949 *
Dirad	1	0.01181	0.01181	0.1184	0.73498
Vigoria	1	0.00051	0.00051	0.0051	0.94378
N_Gemme:Defogl	1	0.00274	0.00274	0.0275	0.87023
N_Gemme:Dirad	1	0.04154	0.04154	0.4167	0.52720
Defogl:Dirad	1	0.00797	0.00797	0.0799	0.78084
N_Gemme:Vigoria	1	0.04016	0.04016	0.4028	0.53410
Defogl:Vigoria	1	0.16149	0.16149	1.6198	0.22025
Dirad:Vigoria	1	0.06369	0.06369	0.6389	0.43515
N_Gemme:Defogl:Dirad	1	0.07842	0.07842	0.7866	0.38751
Residuals	17	1.69484	0.09970		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

KB

Analysis of Variance Table

Response: Co_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.02145	0.021445	0.5686	0.46114
IASMA_2	1	0.10545	0.105448	2.7958	0.11282
IASMA_3	1	0.14583	0.145830	3.8664	0.06581 .
N_Gemme	1	0.03631	0.036309	0.9627	0.34028
Defogl	1	0.01263	0.012626	0.3347	0.57046
Dirad	1	0.00236	0.002361	0.0626	0.80544
Vigoria	1	0.00053	0.000531	0.0141	0.90697
N_Gemme:Defogl	1	0.00131	0.001308	0.0347	0.85450
N_Gemme:Dirad	1	0.00017	0.000170	0.0045	0.94729
Defogl:Dirad	1	0.05986	0.059856	1.5870	0.22478
N_Gemme:Vigoria	1	0.03649	0.036486	0.9673	0.33914
Defogl:Vigoria	1	0.14151	0.141508	3.7518	0.06955 .
Dirad:Vigoria	1	0.00035	0.000345	0.0091	0.92492

N_Gemme:Defogl:Dirad 1 0.01033 0.010332 0.2739 0.60746
 Residuals 17 0.64119 0.037717

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Co_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1969.5	1969.5	3.5218	0.0778408	.
IASMA_2	1	74.2	74.2	0.1327	0.7201791	
IASMA_3	1	18664.8	18664.8	33.3763	2.232e-05	***
N_Gemme	1	108.1	108.1	0.1933	0.6656917	
Defogl	1	12246.0	12246.0	21.8982	0.0002154	***
Dirad	1	2208.8	2208.8	3.9497	0.0632357	.
Vigoria	1	8856.9	8856.9	15.8379	0.0009689	***
N_Gemme:Defogl	1	353.7	353.7	0.6324	0.4374333	
N_Gemme:Dirad	1	749.7	749.7	1.3406	0.2629300	
Defogl:Dirad	1	484.4	484.4	0.8662	0.3650391	
N_Gemme:Vigoria	1	24.2	24.2	0.0433	0.8375829	
Defogl:Vigoria	1	5850.1	5850.1	10.4611	0.0048734	**
Dirad:Vigoria	1	429.2	429.2	0.7675	0.3932121	
N_Gemme:Defogl:Dirad	1	371.4	371.4	0.6641	0.4263843	
Residuals	17	9506.8	559.2			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: -0.814

Antoctot1B

Analysis of Variance Table

Response: Co_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	44.6	44.6	0.0315	0.8612459	
IASMA_2	1	1703.6	1703.6	1.2018	0.2882504	
IASMA_3	1	13237.9	13237.9	9.3384	0.0071489	**
N_Gemme	1	2839.4	2839.4	2.0030	0.1750447	
Defogl	1	11382.3	11382.3	8.0294	0.0114638	*
Dirad	1	6775.3	6775.3	4.7795	0.0430771	*
Vigoria	1	30620.0	30620.0	21.6002	0.0002305	***
N_Gemme:Defogl	1	98.9	98.9	0.0698	0.7948681	
N_Gemme:Dirad	1	151.2	151.2	0.1067	0.7479324	
Defogl:Dirad	1	471.4	471.4	0.3325	0.5717444	
N_Gemme:Vigoria	1	2162.2	2162.2	1.5253	0.2336208	
Defogl:Vigoria	1	0.0	0.0	0.0000	0.9974502	
Dirad:Vigoria	1	5648.9	5648.9	3.9849	0.0621839	.
N_Gemme:Defogl:Dirad	1	218.6	218.6	0.1542	0.6994060	
Residuals	17	24098.9	1417.6			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.595

Poliftot1B

Analysis of Variance Table

Response: Co_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	26930	26930	1.1005	0.308848
IASMA_2	1	1063	1063	0.0434	0.837412
IASMA_3	1	264473	264473	10.8073	0.004346 **
N_Gemme	1	85920	85920	3.5110	0.078258 .
Defogl	1	20458	20458	0.8360	0.373336
Dirad	1	87679	87679	3.5829	0.075532 .
Vigoria	1	288586	288586	11.7926	0.003166 **
N_Gemme:Defogl	1	4094	4094	0.1673	0.687617
N_Gemme:Dirad	1	6	6	0.0002	0.987916
Defogl:Dirad	1	14859	14859	0.6072	0.446566
N_Gemme:Vigoria	1	39669	39669	1.6210	0.220087
Defogl:Vigoria	1	28294	28294	1.1562	0.297283
Dirad:Vigoria	1	31703	31703	1.2955	0.270821
N_Gemme:Defogl:Dirad	1	6561	6561	0.2681	0.611276
Residuals	17	416019	24472		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: -0.623

Antoctot2B

Analysis of Variance Table

Response: Co_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	66	66	0.0261	0.873633
IASMA_2	1	5071	5071	1.9911	0.176267
IASMA_3	1	7268	7268	2.8535	0.109432
N_Gemme	1	1552	1552	0.6093	0.445790
Defogl	1	38164	38164	14.9837	0.001227 **
Dirad	1	642	642	0.2520	0.622076
Vigoria	1	13714	13714	5.3844	0.033011 *
N_Gemme:Defogl	1	59	59	0.0232	0.880821
N_Gemme:Dirad	1	31	31	0.0121	0.913772
Defogl:Dirad	1	317	317	0.1246	0.728487
N_Gemme:Vigoria	1	503	503	0.1974	0.662426
Defogl:Vigoria	1	776	776	0.3046	0.588197
Dirad:Vigoria	1	2446	2446	0.9602	0.340880
N_Gemme:Defogl:Dirad	1	1070	1070	0.4200	0.525614
Residuals	17	43300	2547		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	68327	68327	0.7312	0.4044
IASMA_2	1	12863	12863	0.1376	0.7152
IASMA_3	1	85873	85873	0.9189	0.3512
N_Gemme	1	41468	41468	0.4438	0.5143
Defogl	1	14542	14542	0.1556	0.6981
Dirad	1	27819	27819	0.2977	0.5924
Vigoria	1	274594	274594	2.9384	0.1047
N_Gemme:Defogl	1	46	46	0.0005	0.9825
N_Gemme:Dirad	1	11966	11966	0.1280	0.7249
Defogl:Dirad	1	8270	8270	0.0885	0.7697
N_Gemme:Vigoria	1	310632	310632	3.3241	0.0859 .
Defogl:Vigoria	1	167475	167475	1.7922	0.1983
Dirad:Vigoria	1	12937	12937	0.1384	0.7144
N_Gemme:Defogl:Dirad	1	19302	19302	0.2066	0.6552
Residuals	17	1588627	93449		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

CORTIGLIANO 2009M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Co_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1094	1094	0.0275	0.869756
IBIMET_2	1	60549	60549	1.5225	0.230254
IBIMET_3	1	20941	20941	0.5266	0.475695
N_Gemme	1	24942	24942	0.6272	0.436843
Defogl	1	6918	6918	0.1740	0.680652
Vigoria	1	10297	10297	0.2589	0.615921
N_Gemme:Defogl	1	3860	3860	0.0971	0.758307
N_Gemme:Vigoria	1	452400	452400	11.3759	0.002744 **
Defogl:Vigoria	1	8801	8801	0.2213	0.642684
Residuals	22	874902	39768		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	1.2551e-04	1.2551e-04	18.0497	0.0003288	***
IBIMET_2	1	2.3401e-05	2.3401e-05	3.3652	0.0801475	.
IBIMET_3	1	7.2092e-05	7.2092e-05	10.3673	0.0039434	**
N_Gemme	1	2.1700e-07	2.1700e-07	0.0313	0.8612485	
Defogl	1	3.7635e-05	3.7635e-05	5.4122	0.0296043	*
Vigoria	1	5.0000e-09	5.0000e-09	0.0007	0.9792615	
N_Gemme:Defogl	1	7.3540e-06	7.3540e-06	1.0575	0.3149633	
N_Gemme:Vigoria	1	5.3600e-07	5.3600e-07	0.0771	0.7838114	
Defogl:Vigoria	1	3.4190e-06	3.4190e-06	0.4916	0.4905614	
Residuals	22	1.5298e-04	6.9540e-06			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.671
 IBIMET_2: n.s
 IBIMET_3: -0.566

BrixA

Analysis of Variance Table

Response: Co_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	5.8621	5.8621	17.0147	0.0004447	***
IBIMET_2	1	1.0297	1.0297	2.9888	0.0978540	.
IBIMET_3	1	3.3450	3.3450	9.7088	0.0050344	**
N_Gemme	1	0.0011	0.0011	0.0031	0.9562534	
Defogl	1	2.0156	2.0156	5.8501	0.0242978	*
Vigoria	1	0.0004	0.0004	0.0012	0.9728517	
N_Gemme:Defogl	1	0.2864	0.2864	0.8314	0.3717539	
N_Gemme:Vigoria	1	0.0157	0.0157	0.0456	0.8328922	
Defogl:Vigoria	1	0.1496	0.1496	0.4341	0.5168058	
Residuals	22	7.5797	0.3445			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.66
 IBIMET_2: n.s
 IBIMET_3: -0.553

PhA

Analysis of Variance Table

Response: Co_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.015873	0.015873	7.3085	0.012978	*
IBIMET_2	1	0.020125	0.020125	9.2661	0.005954	**
IBIMET_3	1	0.038583	0.038583	17.7649	0.000357	***
N_Gemme	1	0.004891	0.004891	2.2520	0.147655	
Defogl	1	0.003490	0.003490	1.6068	0.218198	
Vigoria	1	0.005205	0.005205	2.3966	0.135868	
N_Gemme:Defogl	1	0.002821	0.002821	1.2987	0.266709	
N_Gemme:Vigoria	1	0.001677	0.001677	0.7722	0.389038	
Defogl:Vigoria	1	0.000496	0.000496	0.2286	0.637299	

Residuals 22 0.047781 0.002172

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.499
IBIMET_2: 0.544
IBIMET_3: -0.668

AcTitola

Analysis of Variance Table

Response: Co_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.71631	0.71631	14.0227	0.001122	**
IBIMET_2	1	0.07752	0.07752	1.5176	0.230987	
IBIMET_3	1	0.50463	0.50463	9.8788	0.004724	**
N_Gemme	1	0.27346	0.27346	5.3533	0.030413	*
Defogl	1	0.19108	0.19108	3.7406	0.066070	.
Vigoria	1	0.09373	0.09373	1.8349	0.189296	
N_Gemme:Defogl	1	0.00053	0.00053	0.0104	0.919629	
N_Gemme:Vigoria	1	0.08131	0.08131	1.5916	0.220304	
Defogl:Vigoria	1	0.00017	0.00017	0.0032	0.955149	
Residuals	22	1.12381	0.05108			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.624
IBIMET_2: n.s
IBIMET_3: 0.557

AcTartA

Analysis of Variance Table

Response: Co_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.61468	0.61468	9.3602	0.005744	**
IBIMET_2	1	0.24217	0.24217	3.6878	0.067869	.
IBIMET_3	1	0.22297	0.22297	3.3954	0.078898	.
N_Gemme	1	0.11465	0.11465	1.7459	0.199974	
Defogl	1	0.07400	0.07400	1.1269	0.299961	
Vigoria	1	0.35275	0.35275	5.3717	0.030158	*
N_Gemme:Defogl	1	0.07474	0.07474	1.1381	0.297619	
N_Gemme:Vigoria	1	0.18506	0.18506	2.8180	0.107362	
Defogl:Vigoria	1	0.01444	0.01444	0.2200	0.643682	
Residuals	22	1.44472	0.06567			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.546
IBIMET_2: n.s
IBIMET_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Co_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.07492	0.07492	3.5228	0.0738562	.
IBIMET_2	1	0.32061	0.32061	15.0743	0.0008028	***
IBIMET_3	1	0.00429	0.00429	0.2019	0.6576259	
N_Gemme	1	0.00458	0.00458	0.2152	0.6472954	
Defogl	1	0.07605	0.07605	3.5757	0.0718742	.
Vigoria	1	0.01291	0.01291	0.6070	0.4442306	
N_Gemme:Defogl	1	0.00348	0.00348	0.1638	0.6895779	
N_Gemme:Vigoria	1	0.00106	0.00106	0.0498	0.8253972	
Defogl:Vigoria	1	0.00003	0.00003	0.0013	0.9711348	
Residuals	22	0.46790	0.02127			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.638
IBIMET_3: n.s

KA

Analysis of Variance Table

Response: Co_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.02732	0.027317	1.3392	0.2596	
IBIMET_2	1	0.06946	0.069456	3.4051	0.0785	.
IBIMET_3	1	0.00169	0.001685	0.0826	0.7765	
N_Gemme	1	0.02793	0.027933	1.3694	0.2544	
Defogl	1	0.03611	0.036111	1.7703	0.1970	
Vigoria	1	0.01380	0.013800	0.6765	0.4196	
N_Gemme:Defogl	1	0.00271	0.002706	0.1327	0.7192	
N_Gemme:Vigoria	1	0.00092	0.000916	0.0449	0.8341	
Defogl:Vigoria	1	0.00556	0.005562	0.2727	0.6068	
Residuals	22	0.44876	0.020398			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: n.s

ApaA

Analysis of Variance Table

Response: Co_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.8	0.8	0.0021	0.963962	
IBIMET_2	1	4359.7	4359.7	11.0612	0.003068	**
IBIMET_3	1	1441.6	1441.6	3.6577	0.068918	.
N_Gemme	1	93.6	93.6	0.2374	0.630913	
Defogl	1	186.6	186.6	0.4734	0.498598	
Vigoria	1	17.3	17.3	0.0440	0.835804	
N_Gemme:Defogl	1	1.9	1.9	0.0049	0.944573	
N_Gemme:Vigoria	1	454.2	454.2	1.1525	0.294662	
Defogl:Vigoria	1	649.8	649.8	1.6486	0.212507	
Residuals	22	8671.1	394.1			

Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.578
IBIMET_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Co_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	23338	23337.7	11.6839	0.002462 **
IBIMET_2	1	14793	14792.9	7.4060	0.012465 *
IBIMET_3	1	11036	11036.4	5.5253	0.028119 *
N_Gemme	1	739	738.8	0.3699	0.549289
Defogl	1	164	163.8	0.0820	0.777267
Vigoria	1	651	650.6	0.3257	0.573988
N_Gemme:Defogl	1	5910	5910.4	2.9590	0.099439 .
N_Gemme:Vigoria	1	3061	3060.7	1.5323	0.228811
Defogl:Vigoria	1	968	968.1	0.4847	0.493596
Residuals	22	43943	1997.4		

Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.589
IBIMET_2: 0.502
IBIMET_3: -0.448

Poliftot1A

Analysis of Variance Table

Response: Co_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	19743	19742.8	5.7378	0.02555 *
IBIMET_2	1	9	9.4	0.0027	0.95889
IBIMET_3	1	1583	1583.1	0.4601	0.50465
N_Gemme	1	544	543.9	0.1581	0.69476
Defogl	1	813	812.9	0.2363	0.63173
Vigoria	1	7765	7764.6	2.2566	0.14726
N_Gemme:Defogl	1	9526	9526.3	2.7686	0.11031
N_Gemme:Vigoria	1	13669	13668.9	3.9725	0.05879 .
Defogl:Vigoria	1	7002	7002.1	2.0350	0.16775
Residuals	22	75698	3440.8		

Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.455
IBIMET_2: n.s
IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IBIMET_1	1	132180	132180	20.8191	0.0001526	***
IBIMET_2	1	28960	28960	4.5613	0.0440812	*
IBIMET_3	1	48677	48677	7.6669	0.0111975	*
N_Gemme	1	12437	12437	1.9589	0.1755705	
Defogl	1	12487	12487	1.9668	0.1747428	
Vigoria	1	5985	5985	0.9427	0.3421303	
N_Gemme:Defogl	1	19	19	0.0030	0.9568618	
N_Gemme:Vigoria	1	5051	5051	0.7955	0.3821003	
Defogl:Vigoria	1	3503	3503	0.5517	0.4654734	
Residuals	22	139678	6349			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.697
 IBIMET_2: 0.414
 IBIMET_3: -0.508

Poliftot2A

Analysis of Variance Table

Response: Co_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	315349	315349	13.0527	0.001543	**
IBIMET_2	1	775	775	0.0321	0.859529	
IBIMET_3	1	41578	41578	1.7210	0.203092	
N_Gemme	1	21666	21666	0.8968	0.353933	
Defogl	1	231835	231835	9.5959	0.005253	**
Vigoria	1	50500	50500	2.0903	0.162337	
N_Gemme:Defogl	1	40309	40309	1.6684	0.209874	
N_Gemme:Vigoria	1	66405	66405	2.7486	0.111533	
Defogl:Vigoria	1	11230	11230	0.4648	0.502494	
Residuals	22	531513	24160			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.61
 IBIMET_2: n.s
 IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Co_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	678866	678866	18.9819	0.0002524	***
IBIMET_2	1	6983	6983	0.1953	0.6628951	
IBIMET_3	1	864523	864523	24.1730	6.449e-05	***
N_Gemme	1	32813	32813	0.9175	0.3485472	
Defogl	1	448	448	0.0125	0.9118800	
Vigoria	1	384774	384774	10.7587	0.0034201	**
N_Gemme:Defogl	1	85402	85402	2.3879	0.1365392	
N_Gemme:Vigoria	1	79335	79335	2.2183	0.1505812	
Defogl:Vigoria	1	10130	10130	0.2833	0.5999077	
Residuals	22	786807	35764			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.681
 IBIMET_2: n.s
 IBIMET_3: 0.724

DensMostoB

Analysis of Variance Table

Response: Co_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1.0802e-04	1.0802e-04	7.8683	0.01032 *
IBIMET_2	1	2.6780e-06	2.6780e-06	0.1951	0.66301
IBIMET_3	1	6.6658e-05	6.6658e-05	4.8557	0.03832 *
N_Gemme	1	1.0000e-09	1.0000e-09	0.0001	0.99246
Defogl	1	8.3670e-06	8.3670e-06	0.6095	0.44329
Vigoria	1	3.2756e-05	3.2756e-05	2.3861	0.13668
N_Gemme:Defogl	1	3.2000e-07	3.2000e-07	0.0233	0.88004
N_Gemme:Vigoria	1	5.9440e-06	5.9440e-06	0.4330	0.51736
Defogl:Vigoria	1	7.0550e-05	7.0550e-05	5.1392	0.03357 *
Residuals	22	3.0201e-04	1.3728e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.513
 IBIMET_2: n.s
 IBIMET_3: -0.425

BrixB

Analysis of Variance Table

Response: Co_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	5.4104	5.4104	8.2173	0.008966 **
IBIMET_2	1	0.1604	0.1604	0.2436	0.626516
IBIMET_3	1	3.3591	3.3591	5.1018	0.034154 *
N_Gemme	1	0.0047	0.0047	0.0071	0.933717
Defogl	1	0.4518	0.4518	0.6862	0.416357
Vigoria	1	1.5102	1.5102	2.2937	0.144133
N_Gemme:Defogl	1	0.0151	0.0151	0.0229	0.881036
N_Gemme:Vigoria	1	0.3069	0.3069	0.4662	0.501888
Defogl:Vigoria	1	3.3328	3.3328	5.0618	0.034794 *
Residuals	22	14.4852	0.6584		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.521
 IBIMET_2: n.s
 IBIMET_3: -0.434

PhB

Analysis of Variance Table

Response: Co_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.078880	0.078880	21.9638	0.0001128 ***
IBIMET_2	1	0.068707	0.068707	19.1311	0.0002420 ***

IBIMET_3	1	0.032983	0.032983	9.1840	0.0061438	**
N_Gemme	1	0.000013	0.000013	0.0037	0.9517901	
Defogl	1	0.000004	0.000004	0.0012	0.9724956	
Vigoria	1	0.075017	0.075017	20.8883	0.0001498	***
N_Gemme:Defogl	1	0.001016	0.001016	0.2828	0.6001974	
N_Gemme:Vigoria	1	0.000266	0.000266	0.0740	0.7881758	
Defogl:Vigoria	1	0.007189	0.007189	2.0017	0.1711195	
Residuals	22	0.079010	0.003591			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.707
 IBIMET_2: -0.682
 IBIMET_3: 0.543

AcTitolB

Analysis of Variance Table

Response: Co_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.27598	0.27598	4.9342	0.036933	*
IBIMET_2	1	0.28405	0.28405	5.0786	0.034524	*
IBIMET_3	1	0.48452	0.48452	8.6627	0.007519	**
N_Gemme	1	0.03666	0.03666	0.6555	0.426814	
Defogl	1	0.04166	0.04166	0.7449	0.397419	
Vigoria	1	0.54083	0.54083	9.6695	0.005109	**
N_Gemme:Defogl	1	0.00156	0.00156	0.0278	0.869061	
N_Gemme:Vigoria	1	0.01168	0.01168	0.2089	0.652142	
Defogl:Vigoria	1	0.02123	0.02123	0.3795	0.544201	
Residuals	22	1.23050	0.05593			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.428
 IBIMET_2: -0.433
 IBIMET_3: 0.532

AcTartB

Analysis of Variance Table

Response: Co_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.33533	0.33533	11.0321	0.003100	**
IBIMET_2	1	0.26905	0.26905	8.8518	0.006984	**
IBIMET_3	1	0.34096	0.34096	11.2174	0.002902	**
N_Gemme	1	0.05082	0.05082	1.6721	0.209392	
Defogl	1	0.06268	0.06268	2.0623	0.165053	
Vigoria	1	0.41019	0.41019	13.4952	0.001333	**
N_Gemme:Defogl	1	0.00699	0.00699	0.2299	0.636317	
N_Gemme:Vigoria	1	0.00242	0.00242	0.0796	0.780457	
Defogl:Vigoria	1	0.01543	0.01543	0.5076	0.483676	
Residuals	22	0.66870	0.03040			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.578
 IBIMET_2: 0.536

IBIMET_3: -0.581

AcMalicoB

Analysis of Variance Table

Response: Co_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.02109	0.021092	0.8027	0.379975	
IBIMET_2	1	0.29941	0.299412	11.3950	0.002725	**
IBIMET_3	1	0.02994	0.029940	1.1395	0.297340	
N_Gemme	1	0.02811	0.028113	1.0699	0.312196	
Defogl	1	0.00764	0.007638	0.2907	0.595201	
Viguria	1	0.04922	0.049221	1.8732	0.184918	
N_Gemme:Defogl	1	0.03952	0.039524	1.5042	0.232988	
N_Gemme:Viguria	1	0.01524	0.015240	0.5800	0.454398	
Defogl:Viguria	1	0.03538	0.035379	1.3465	0.258334	
Residuals	22	0.57806	0.026276			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.584
IBIMET_3: n.s

KB

Analysis of Variance Table

Response: Co_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.03490	0.034895	2.0900	0.162367	
IBIMET_2	1	0.14777	0.147774	8.8505	0.006988	**
IBIMET_3	1	0.03361	0.033610	2.0130	0.169973	
N_Gemme	1	0.01593	0.015934	0.9544	0.339235	
Defogl	1	0.00790	0.007901	0.4732	0.498712	
Viguria	1	0.08476	0.084759	5.0764	0.034559	*
N_Gemme:Defogl	1	0.03986	0.039862	2.3874	0.136581	
N_Gemme:Viguria	1	0.00116	0.001156	0.0692	0.794935	
Defogl:Viguria	1	0.00383	0.003833	0.2296	0.636579	
Residuals	22	0.36733	0.016697			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.536
IBIMET_3: n.s

ApaB

Analysis of Variance Table

Response: Co_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	1939.8	1939.8	9.6572	0.005133	**
IBIMET_2	1	10146.2	10146.2	50.5131	3.97e-07	***
IBIMET_3	1	625.9	625.9	3.1159	0.091412	.
N_Gemme	1	1259.5	1259.5	6.2705	0.020185	*

Defogl	1	91.7	91.7	0.4565	0.506322
Vigoria	1	2017.2	2017.2	10.0428	0.004444 **
N_Gemme:Defogl	1	152.8	152.8	0.7607	0.392528
N_Gemme:Vigoria	1	24.9	24.9	0.1238	0.728260
Defogl:Vigoria	1	323.6	323.6	1.6112	0.217582
Residuals	22	4419.0	200.9		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.552
IBIMET_2: -0.835
IBIMET_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Co_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	4839.2	4839.2	4.2184	0.05206 .
IBIMET_2	1	5507.2	5507.2	4.8007	0.03933 *
IBIMET_3	1	1535.2	1535.2	1.3382	0.25976
N_Gemme	1	1639.1	1639.1	1.4288	0.24469
Defogl	1	789.7	789.7	0.6884	0.41563
Vigoria	1	3819.8	3819.8	3.3297	0.08165 .
N_Gemme:Defogl	1	2490.3	2490.3	2.1708	0.15482
N_Gemme:Vigoria	1	173.6	173.6	0.1513	0.70100
Defogl:Vigoria	1	226.0	226.0	0.1970	0.66149
Residuals	22	25237.9	1147.2		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.423
IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Co_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	25236	25236.4	16.6996	0.0004884 ***
IBIMET_2	1	1752	1751.9	1.1593	0.2932795
IBIMET_3	1	29959	29958.6	19.8245	0.0001998 ***
N_Gemme	1	509	508.7	0.3366	0.5676714
Defogl	1	3066	3065.9	2.0288	0.1683762
Vigoria	1	28589	28588.8	18.9180	0.0002569 ***
N_Gemme:Defogl	1	789	788.9	0.5220	0.4775825
N_Gemme:Vigoria	1	347	346.8	0.2295	0.6366268
Defogl:Vigoria	1	18	18.5	0.0122	0.9129941
Residuals	22	33246	1511.2		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.657
IBIMET_2: n.s
IBIMET_3: -0.688

Antoctot2B

Analysis of Variance Table

Response: Co_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	8545	8545.0	1.6727	0.20931
IBIMET_2	1	22216	22216.0	4.3489	0.04885 *
IBIMET_3	1	24	23.9	0.0047	0.94606
N_Gemme	1	5083	5082.9	0.9950	0.32937
Defogl	1	5774	5774.0	1.1303	0.29924
Vigoria	1	9438	9438.0	1.8475	0.18784
N_Gemme:Defogl	1	7747	7746.9	1.5165	0.23115
N_Gemme:Vigoria	1	115	115.1	0.0225	0.88207
Defogl:Vigoria	1	612	612.0	0.1198	0.73254
Residuals	22	112385	5108.4		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.406
IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	39333	39333	6.0382	0.022353 *
IBIMET_2	1	10568	10568	1.6224	0.216054
IBIMET_3	1	12103	12103	1.8579	0.186654
N_Gemme	1	4051	4051	0.6218	0.438778
Defogl	1	28470	28470	4.3706	0.048334 *
Vigoria	1	73401	73401	11.2680	0.002850 **
N_Gemme:Defogl	1	2404	2404	0.3691	0.549740
N_Gemme:Vigoria	1	2733	2733	0.4196	0.523858
Defogl:Vigoria	1	4	4	0.0006	0.980892
Residuals	22	143310	6514		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.464
IBIMET_2: n.s
IBIMET_3: n.s

CORTIGLIANO 2009M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Co_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	19522	19522	0.6371	0.4332842
DIPROVE_2	1	865	865	0.0282	0.8681294
DIPROVE_3	1	4509	4509	0.1471	0.7049568
N_Gemme	1	20139	20139	0.6573	0.4262156

Defogl	1	24247	24247	0.7913	0.3833275
Vigoria	1	84031	84031	2.7424	0.1119124
N_Gemme:Defogl	1	3886	3886	0.1268	0.7251570
N_Gemme:Vigoria	1	525795	525795	17.1598	0.0004261 ***
Defogl:Vigoria	1	107607	107607	3.5118	0.0742751 .
Residuals	22	674106	30641		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	8.1300e-07	8.1300e-07	0.1196	0.73274
DIPROVE_2	1	2.0829e-05	2.0829e-05	3.0633	0.09402 .
DIPROVE_3	1	3.0700e-07	3.0700e-07	0.0452	0.83362
N_Gemme	1	1.7500e-07	1.7500e-07	0.0257	0.87413
Defogl	1	2.0349e-04	2.0349e-04	29.9276	1.699e-05 ***
Vigoria	1	4.2420e-06	4.2420e-06	0.6238	0.43807
N_Gemme:Defogl	1	1.6736e-05	1.6736e-05	2.4613	0.13095
N_Gemme:Vigoria	1	0.0000e+00	0.0000e+00	0.0001	0.99391
Defogl:Vigoria	1	2.6969e-05	2.6969e-05	3.9663	0.05898 .
Residuals	22	1.4959e-04	6.8000e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

BrixA

Analysis of Variance Table

Response: Co_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.0348	0.0348	0.1053	0.74864
DIPROVE_2	1	0.9674	0.9674	2.9279	0.10113
DIPROVE_3	1	0.0105	0.0105	0.0319	0.85992
N_Gemme	1	0.0005	0.0005	0.0014	0.97087
Defogl	1	9.8782	9.8782	29.8958	1.711e-05 ***
Vigoria	1	0.1406	0.1406	0.4255	0.52097
N_Gemme:Defogl	1	0.7165	0.7165	2.1683	0.15504
N_Gemme:Vigoria	1	0.0007	0.0007	0.0021	0.96349
Defogl:Vigoria	1	1.2668	1.2668	3.8340	0.06302 .
Residuals	22	7.2693	0.3304		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Co_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.021359	0.0213591	8.9188	0.006805	**
DIPROVE_2	1	0.020756	0.0207555	8.6667	0.007507	**
DIPROVE_3	1	0.013631	0.0136313	5.6919	0.026081	*
N_Gemme	1	0.003209	0.0032093	1.3401	0.259432	
Defogl	1	0.016670	0.0166704	6.9609	0.015011	*
Vigoria	1	0.002812	0.0028122	1.1743	0.290260	
N_Gemme:Defogl	1	0.004608	0.0046083	1.9242	0.179283	
N_Gemme:Vigoria	1	0.005082	0.0050824	2.1222	0.159300	
Defogl:Vigoria	1	0.000128	0.0001278	0.0534	0.819427	
Residuals	22	0.052687	0.0023949			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.537
DIPROVE_2: -0.532
DIPROVE_3: -0.453

AcTitola

Analysis of Variance Table

Response: Co_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.11603	0.11603	2.0099	0.170280	
DIPROVE_2	1	0.07805	0.07805	1.3521	0.257370	
DIPROVE_3	1	0.29616	0.29616	5.1304	0.033704	*
N_Gemme	1	0.21709	0.21709	3.7606	0.065406	.
Defogl	1	0.76277	0.76277	13.2132	0.001463	**
Vigoria	1	0.16138	0.16138	2.7955	0.108696	
N_Gemme:Defogl	1	0.00018	0.00018	0.0031	0.955933	
N_Gemme:Vigoria	1	0.15994	0.15994	2.7706	0.110188	
Defogl:Vigoria	1	0.00094	0.00094	0.0163	0.899693	
Residuals	22	1.27001	0.05773			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.435

AcTarta

Analysis of Variance Table

Response: Co_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.06639	0.06639	1.1318	0.298932	
DIPROVE_2	1	0.07373	0.07373	1.2570	0.274306	
DIPROVE_3	1	0.54113	0.54113	9.2253	0.006047	**
N_Gemme	1	0.07796	0.07796	1.3291	0.261340	
Defogl	1	0.44500	0.44500	7.5863	0.011573	*
Vigoria	1	0.41328	0.41328	7.0456	0.014485	*

N_Gemme:Defogl	1	0.16528	0.16528	2.8178	0.107376
N_Gemme:Vigoria	1	0.24701	0.24701	4.2111	0.052249 .
Defogl:Vigoria	1	0.01992	0.01992	0.3396	0.565986
Residuals	22	1.29047	0.05866		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.544

AcMalicoA

Analysis of Variance Table

Response: Co_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.04119	0.041189	1.7613	0.198074
DIPROVE_2	1	0.03437	0.034372	1.4698	0.238238
DIPROVE_3	1	0.06512	0.065116	2.7844	0.109356
N_Gemme	1	0.00299	0.002993	0.1280	0.723937
Defogl	1	0.24359	0.243591	10.4161	0.003873 **
Vigoria	1	0.01425	0.014246	0.6092	0.443413
N_Gemme:Defogl	1	0.01784	0.017842	0.7629	0.391849
N_Gemme:Vigoria	1	0.00011	0.000114	0.0049	0.944888
Defogl:Vigoria	1	0.03188	0.031881	1.3633	0.255473
Residuals	22	0.51449	0.023386		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Co_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.02920	0.0292032	1.2332	0.2788
DIPROVE_2	1	0.00658	0.0065784	0.2778	0.6034
DIPROVE_3	1	0.00169	0.0016877	0.0713	0.7920
N_Gemme	1	0.02607	0.0260738	1.1010	0.3054
Defogl	1	0.01346	0.0134550	0.5682	0.4590
Vigoria	1	0.00286	0.0028594	0.1207	0.7315
N_Gemme:Defogl	1	0.00004	0.0000449	0.0019	0.9657
N_Gemme:Vigoria	1	0.00952	0.0095179	0.4019	0.5326
Defogl:Vigoria	1	0.02384	0.0238413	1.0068	0.3266
Residuals	22	0.52098	0.0236810		

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

ApaA

Analysis of Variance Table

Response: Co_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	905.7	905.7	2.4985	0.128227	
DIPROVE_2	1	2195.5	2195.5	6.0567	0.022171	*
DIPROVE_3	1	3661.2	3661.2	10.1003	0.004351	**
N_Gemme	1	250.5	250.5	0.6911	0.414714	
Defogl	1	143.5	143.5	0.3958	0.535761	
Vigoria	1	324.5	324.5	0.8952	0.354338	
N_Gemme:Defogl	1	226.0	226.0	0.6236	0.438138	
N_Gemme:Vigoria	1	179.4	179.4	0.4950	0.489080	
Defogl:Vigoria	1	15.8	15.8	0.0435	0.836635	
Residuals	22	7974.6	362.5			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: 0.465

DIPROVE_3: -0.561

Antoctot1A

Analysis of Variance Table

Response: Co_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	5867	5866.7	2.8792	0.103838	
DIPROVE_2	1	3092	3091.6	1.5172	0.231043	
DIPROVE_3	1	16827	16826.9	8.2580	0.008822	**
N_Gemme	1	293	293.0	0.1438	0.708162	
Defogl	1	4988	4987.7	2.4478	0.131964	
Vigoria	1	19690	19690.4	9.6633	0.005121	**
N_Gemme:Defogl	1	3662	3662.2	1.7973	0.193725	
N_Gemme:Vigoria	1	2751	2750.9	1.3500	0.257723	
Defogl:Vigoria	1	2605	2605.3	1.2786	0.270338	
Residuals	22	44828	2037.6			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: 0.522

Poliftot1A

Analysis of Variance Table

Response: Co_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1710	1709.8	0.5081	0.48345	
DIPROVE_2	1	4310	4309.9	1.2808	0.26994	
DIPROVE_3	1	1340	1340.4	0.3983	0.53445	
N_Gemme	1	519	519.5	0.1544	0.69817	
Defogl	1	22594	22593.6	6.7143	0.01667	*
Vigoria	1	1329	1328.7	0.3949	0.53622	
N_Gemme:Defogl	1	14575	14574.8	4.3313	0.04927	*
N_Gemme:Vigoria	1	14412	14412.1	4.2829	0.05044	.
Defogl:Vigoria	1	1533	1533.3	0.4556	0.50670	
Residuals	22	74030	3365.0			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	17961	17961	2.6684	0.116591
DIPROVE_2	1	4761	4761	0.7073	0.409390
DIPROVE_3	1	75172	75172	11.1681	0.002953 **
N_Gemme	1	7706	7706	1.1449	0.296220
Defogl	1	87747	87747	13.0364	0.001551 **
Vigoria	1	36397	36397	5.4075	0.029668 *
N_Gemme:Defogl	1	216	216	0.0322	0.859334
N_Gemme:Vigoria	1	6813	6813	1.0121	0.325333
Defogl:Vigoria	1	4125	4125	0.6128	0.442076
Residuals	22	148080	6731		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.58

Poliftot2A

Analysis of Variance Table

Response: Co_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	10125	10125	0.5535	0.46476
DIPROVE_2	1	5191	5191	0.2838	0.59956
DIPROVE_3	1	61643	61643	3.3700	0.07995 .
N_Gemme	1	16073	16073	0.8787	0.35874
Defogl	1	638982	638982	34.9324	6.009e-06 ***
Vigoria	1	10089	10089	0.5515	0.46555
N_Gemme:Defogl	1	10166	10166	0.5558	0.46386
N_Gemme:Vigoria	1	67695	67695	3.7008	0.06742 .
Defogl:Vigoria	1	88771	88771	4.8530	0.03837 *
Residuals	22	402424	18292		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Co_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	41885	41885	0.9282	0.3458033
DIPROVE_2	1	618404	618404	13.7042	0.0012443 **
DIPROVE_3	1	164734	164734	3.6506	0.0691684 .
N_Gemme	1	14730	14730	0.3264	0.5735568
Defogl	1	737758	737758	16.3491	0.0005426 ***
Vigoria	1	101045	101045	2.2392	0.1487578
N_Gemme:Defogl	1	28852	28852	0.6394	0.4324837
N_Gemme:Vigoria	1	126495	126495	2.8032	0.1082356
Defogl:Vigoria	1	103422	103422	2.2919	0.1442869
Residuals	22	992754	45125		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.62
DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Co_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	2.2261e-05	2.2261e-05	1.6191	0.216498
DIPROVE_2	1	1.9079e-05	1.9079e-05	1.3877	0.251375
DIPROVE_3	1	2.2900e-07	2.2900e-07	0.0166	0.898540
N_Gemme	1	3.0000e-09	3.0000e-09	0.0002	0.988835
Defogl	1	8.9019e-05	8.9019e-05	6.4748	0.018473 *
Vigoria	1	1.3034e-04	1.3034e-04	9.4800	0.005488 **
N_Gemme:Defogl	1	3.7210e-06	3.7210e-06	0.2706	0.608094
N_Gemme:Vigoria	1	1.6300e-07	1.6300e-07	0.0119	0.914229
Defogl:Vigoria	1	3.0024e-05	3.0024e-05	2.1838	0.153644
Residuals	22	3.0247e-04	1.3748e-05		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Co_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1.1328	1.1328	1.7184	0.203415
DIPROVE_2	1	1.0028	1.0028	1.5211	0.230467
DIPROVE_3	1	0.0111	0.0111	0.0169	0.897896
N_Gemme	1	0.0050	0.0050	0.0076	0.931544
Defogl	1	4.5099	4.5099	6.8412	0.015791 *
Vigoria	1	6.2812	6.2812	9.5281	0.005389 **
N_Gemme:Defogl	1	0.1752	0.1752	0.2658	0.611287
N_Gemme:Vigoria	1	0.0048	0.0048	0.0073	0.932637
Defogl:Vigoria	1	1.4107	1.4107	2.1399	0.157651
Residuals	22	14.5031	0.6592		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Co_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.067418	0.067418	19.7833	0.0002020	***
DIPROVE_2	1	0.052646	0.052646	15.4486	0.0007145	***
DIPROVE_3	1	0.025013	0.025013	7.3398	0.0128112	*
N_Gemme	1	0.000012	0.000012	0.0034	0.9538532	
Defogl	1	0.040463	0.040463	11.8735	0.0023048	**
Vigoria	1	0.079055	0.079055	23.1980	8.225e-05	***
N_Gemme:Defogl	1	0.000039	0.000039	0.0113	0.9161945	
N_Gemme:Vigoria	1	0.003199	0.003199	0.9387	0.3431324	
Defogl:Vigoria	1	0.000269	0.000269	0.0789	0.7814106	
Residuals	22	0.074972	0.003408			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.688
DIPROVE_2: -0.642
DIPROVE_3: 0.5

AcTitolB

Analysis of Variance Table

Response: Co_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.26535	0.26535	4.0587	0.05633	.
DIPROVE_2	1	0.26051	0.26051	3.9846	0.05844	.
DIPROVE_3	1	0.19404	0.19404	2.9680	0.09896	.
N_Gemme	1	0.02120	0.02120	0.3243	0.57481	
Defogl	1	0.24433	0.24433	3.7372	0.06619	.
Vigoria	1	0.47603	0.47603	7.2811	0.01313	*
N_Gemme:Defogl	1	0.01210	0.01210	0.1851	0.67124	
N_Gemme:Vigoria	1	0.01265	0.01265	0.1935	0.66428	
Defogl:Vigoria	1	0.00412	0.00412	0.0631	0.80403	
Residuals	22	1.43833	0.06538			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTartB

Analysis of Variance Table

Response: Co_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.17400	0.17400	4.9836	0.036088	*

DIPROVE_2	1	0.24939	0.24939	7.1431	0.013905	*
DIPROVE_3	1	0.19074	0.19074	5.4632	0.028924	*
N_Gemme	1	0.06934	0.06934	1.9859	0.172746	
Defogl	1	0.33498	0.33498	9.5947	0.005255	**
Vigoria	1	0.30595	0.30595	8.7630	0.007230	**
N_Gemme:Defogl	1	0.03408	0.03408	0.9760	0.333925	
N_Gemme:Vigoria	1	0.03388	0.03388	0.9703	0.335304	
Defogl:Vigoria	1	0.00213	0.00213	0.0610	0.807204	
Residuals	22	0.76810	0.03491			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.43
DIPROVE_2: 0.495
DIPROVE_3: -0.446

AcMalicoB

Analysis of Variance Table

Response: Co_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.09524	0.095242	3.2184	0.08657	.
DIPROVE_2	1	0.00498	0.004976	0.1681	0.68574	
DIPROVE_3	1	0.12963	0.129627	4.3804	0.04810	*
N_Gemme	1	0.03685	0.036847	1.2452	0.27652	
Defogl	1	0.08399	0.083992	2.8383	0.10618	
Vigoria	1	0.00143	0.001426	0.0482	0.82824	
N_Gemme:Defogl	1	0.02811	0.028111	0.9499	0.34033	
N_Gemme:Vigoria	1	0.00950	0.009503	0.3211	0.57667	
Defogl:Vigoria	1	0.06287	0.062865	2.1244	0.15910	
Residuals	22	0.65104	0.029593			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.407

KB

Analysis of Variance Table

Response: Co_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.170049	0.170049	12.8130	0.0016719	**
DIPROVE_2	1	0.000435	0.000435	0.0327	0.8580600	
DIPROVE_3	1	0.028496	0.028496	2.1471	0.1569846	
N_Gemme	1	0.012823	0.012823	0.9662	0.3363256	
Defogl	1	0.006561	0.006561	0.4944	0.4893580	
Vigoria	1	0.191211	0.191211	14.4076	0.0009913	***
N_Gemme:Defogl	1	0.003722	0.003722	0.2805	0.6016873	
N_Gemme:Vigoria	1	0.010841	0.010841	0.8168	0.3759000	
Defogl:Vigoria	1	0.020939	0.020939	1.5777	0.2222624	
Residuals	22	0.291974	0.013272			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.607

DIPROVE_2: n.s
DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Co_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	5037.4	5037.4	21.2831	0.0001349	***
DIPROVE_2	1	3624.6	3624.6	15.3138	0.0007450	***
DIPROVE_3	1	4360.4	4360.4	18.4225	0.0002956	***
N_Gemme	1	1528.9	1528.9	6.4597	0.0185937	*
Defogl	1	468.4	468.4	1.9791	0.1734553	
Vigoria	1	266.5	266.5	1.1261	0.3001274	
N_Gemme:Defogl	1	439.1	439.1	1.8553	0.1869552	
N_Gemme:Vigoria	1	5.9	5.9	0.0250	0.8758966	
Defogl:Vigoria	1	62.2	62.2	0.2629	0.6132567	
Residuals	22	5207.1	236.7			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.701
DIPROVE_2: 0.641
DIPROVE_3: -0.675

Antoctot1B

Analysis of Variance Table

Response: Co_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	5469.0	5469.0	4.9915	0.03595	*
DIPROVE_2	1	139.0	139.0	0.1269	0.72508	
DIPROVE_3	1	7077.9	7077.9	6.4599	0.01859	*
N_Gemme	1	1094.7	1094.7	0.9992	0.32838	
Defogl	1	3363.7	3363.7	3.0700	0.09368	.
Vigoria	1	1347.8	1347.8	1.2301	0.27936	
N_Gemme:Defogl	1	3000.7	3000.7	2.7387	0.11214	
N_Gemme:Vigoria	1	570.0	570.0	0.5202	0.47833	
Defogl:Vigoria	1	90.5	90.5	0.0826	0.77648	
Residuals	22	24104.6	1095.7			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.43
DIPROVE_2: n.s
DIPROVE_3: -0.476

Poliftot1B

Analysis of Variance Table

Response: Co_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	10406	10406	7.7683	0.01074	*
DIPROVE_2	1	30452	30452	22.7321	9.257e-05	***
DIPROVE_3	1	3925	3925	2.9302	0.10100	

N_Gemme	1	924	924	0.6895	0.41524
Defogl	1	41470	41470	30.9571	1.361e-05 ***
Vigoria	1	2770	2770	2.0675	0.16454
N_Gemme:Defogl	1	36	36	0.0266	0.87191
N_Gemme:Vigoria	1	2238	2238	1.6708	0.20956
Defogl:Vigoria	1	1820	1820	1.3584	0.25629
Residuals	22	29471	1340		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.511
DIPROVE_2: -0.713
DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Co_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	15176	15175.6	2.9759	0.09853 .
DIPROVE_2	1	1192	1192.5	0.2338	0.63347
DIPROVE_3	1	18704	18703.7	3.6678	0.06856 .
N_Gemme	1	3895	3895.1	0.7638	0.39157
Defogl	1	6144	6144.3	1.2049	0.28421
Vigoria	1	2756	2755.5	0.5404	0.47005
N_Gemme:Defogl	1	11812	11811.9	2.3163	0.14227
N_Gemme:Vigoria	1	15	14.7	0.0029	0.95762
Defogl:Vigoria	1	58	57.7	0.0113	0.91627
Residuals	22	112188	5099.4		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	23261	23261	3.3169	0.082201 .
DIPROVE_2	1	33063	33063	4.7147	0.040967 *
DIPROVE_3	1	2651	2651	0.3781	0.544948
N_Gemme	1	4038	4038	0.5758	0.456000
Defogl	1	94787	94787	13.5162	0.001323 **
Vigoria	1	2545	2545	0.3629	0.553077
N_Gemme:Defogl	1	1358	1358	0.1936	0.664232
N_Gemme:Vigoria	1	194	194	0.0276	0.869542
Defogl:Vigoria	1	199	199	0.0283	0.867890
Residuals	22	154282	7013		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.42
DIPROVE_3: n.s

CORTIGLIANO 2009M IASMA

PvinacciaA

Analysis of Variance Table

Response: Co_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	7283	7283	0.2068	0.653713
IASMA_2	1	85244	85244	2.4209	0.133994
IASMA_3	1	98601	98601	2.8003	0.108409
N_Gemme	1	3914	3914	0.1112	0.741974
Defogl	1	36161	36161	1.0270	0.321888
Vigoria	1	76793	76793	2.1809	0.153903
N_Gemme:Defogl	1	34273	34273	0.9734	0.334573
N_Gemme:Vigoria	1	346313	346313	9.8353	0.004801 **
Defogl:Vigoria	1	1478	1478	0.0420	0.839539
Residuals	22	774646	35211		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Co_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	8.3646e-05	8.3646e-05	9.6413	0.0051638 **
IASMA_2	1	7.2100e-07	7.2100e-07	0.0831	0.7757964
IASMA_3	1	8.2270e-06	8.2270e-06	0.9482	0.3407520
N_Gemme	1	4.8000e-08	4.8000e-08	0.0056	0.9412712
Defogl	1	1.2906e-04	1.2906e-04	14.8756	0.0008545 ***
Vigoria	1	6.1000e-08	6.1000e-08	0.0071	0.9338256
N_Gemme:Defogl	1	8.4260e-06	8.4260e-06	0.9712	0.3351024
N_Gemme:Vigoria	1	2.0000e-09	2.0000e-09	0.0002	0.9891625
Defogl:Vigoria	1	2.1000e-06	2.1000e-06	0.2421	0.6275686
Residuals	22	1.9087e-04	8.6760e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.552
IASMA_2: n.s
IASMA_3: n.s

BrixA

Analysis of Variance Table

Response: Co_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	4.0394	4.0394	9.7148	0.0050231 **
IASMA_2	1	0.0680	0.0680	0.1636	0.6898044
IASMA_3	1	0.3838	0.3838	0.9229	0.3471425
N_Gemme	1	0.0000	0.0000	0.0001	0.9938434

Defogl	1	6.2339	6.2339	14.9927	0.0008236	***
Viguria	1	0.0004	0.0004	0.0009	0.9759043	
N_Gemme:Defogl	1	0.3360	0.3360	0.8081	0.3784087	
N_Gemme:Viguria	1	0.0009	0.0009	0.0023	0.9624307	
Defogl:Viguria	1	0.0754	0.0754	0.1812	0.6744476	
Residuals	22	9.1475	0.4158			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.553

IASMA_2: n.s

IASMA_3: n.s

PhA

Analysis of Variance Table

Response: Co_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.066287	0.066287	27.3471	3.028e-05	***
IASMA_2	1	0.000855	0.000855	0.3527	0.55865	
IASMA_3	1	0.000011	0.000011	0.0046	0.94673	
N_Gemme	1	0.004466	0.004466	1.8424	0.18843	
Defogl	1	0.000001	0.000001	0.0006	0.98071	
Viguria	1	0.009936	0.009936	4.0990	0.05521	.
N_Gemme:Defogl	1	0.000403	0.000403	0.1662	0.68746	
N_Gemme:Viguria	1	0.005530	0.005530	2.2816	0.14515	
Defogl:Viguria	1	0.000128	0.000128	0.0530	0.82007	
Residuals	22	0.053326	0.002424			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.744

IASMA_2: n.s

IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Co_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.90077	0.90077	16.8769	0.0004633	***
IASMA_2	1	0.00732	0.00732	0.1372	0.7146005	
IASMA_3	1	0.00024	0.00024	0.0046	0.9467285	
N_Gemme	1	0.27495	0.27495	5.1514	0.0333765	*
Defogl	1	0.22646	0.22646	4.2429	0.0514397	.
Viguria	1	0.28808	0.28808	5.3974	0.0298047	*
N_Gemme:Defogl	1	0.00508	0.00508	0.0952	0.7605486	
N_Gemme:Viguria	1	0.16582	0.16582	3.1069	0.0918499	.
Defogl:Viguria	1	0.01963	0.01963	0.3678	0.5504079	
Residuals	22	1.17421	0.05337			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.659

IASMA_2: n.s

IASMA_3: n.s

AcTartA

Analysis of Variance Table

Response: Co_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.93117	0.93117	15.9317	0.0006158	***
IASMA_2	1	0.11899	0.11899	2.0358	0.1676715	
IASMA_3	1	0.01668	0.01668	0.2854	0.5985448	
N_Gemme	1	0.18894	0.18894	3.2326	0.0859231	.
Defogl	1	0.06704	0.06704	1.1470	0.2957828	
Viguria	1	0.34874	0.34874	5.9668	0.0230698	*
N_Gemme:Defogl	1	0.04472	0.04472	0.7651	0.3911780	
N_Gemme:Viguria	1	0.27357	0.27357	4.6806	0.0416365	*
Defogl:Viguria	1	0.06449	0.06449	1.1034	0.3049208	
Residuals	22	1.28584	0.05845			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.648

IASMA_2: n.s

IASMA_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Co_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.01389	0.01389	0.6628	0.4243056	
IASMA_2	1	0.03350	0.03350	1.5981	0.2194071	
IASMA_3	1	0.01360	0.01360	0.6486	0.4292399	
N_Gemme	1	0.01752	0.01752	0.8359	0.3704954	
Defogl	1	0.40229	0.40229	19.1895	0.0002381	***
Viguria	1	0.01327	0.01327	0.6330	0.4347631	
N_Gemme:Defogl	1	0.00001	0.00001	0.0003	0.9871528	
N_Gemme:Viguria	1	0.00470	0.00470	0.2244	0.6404066	
Defogl:Viguria	1	0.00584	0.00584	0.2784	0.6030178	
Residuals	22	0.46121	0.02096			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

KA

Analysis of Variance Table

Response: Co_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.00193	0.001930	0.0976	0.75771	
IASMA_2	1	0.02752	0.027525	1.3912	0.25079	
IASMA_3	1	0.10364	0.103636	5.2383	0.03206	*
N_Gemme	1	0.02615	0.026155	1.3220	0.26258	
Defogl	1	0.02077	0.020771	1.0498	0.31668	
Viguria	1	0.00180	0.001801	0.0910	0.76572	

N_Gemme:Defogl	1	0.00569	0.005688	0.2875	0.59722
N_Gemme:Vigoria	1	0.01146	0.011460	0.5793	0.45468
Defogl:Vigoria	1	0.00002	0.000021	0.0011	0.97432
Residuals	22	0.43526	0.019784		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.439

ApaA

Analysis of Variance Table

Response: Co_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	3456.1	3456.1	9.1710	0.006174	**
IASMA_2	1	110.1	110.1	0.2921	0.594312	
IASMA_3	1	52.0	52.0	0.1380	0.713823	
N_Gemme	1	48.4	48.4	0.1284	0.723492	
Defogl	1	2328.0	2328.0	6.1775	0.021023	*
Vigoria	1	96.3	96.3	0.2556	0.618179	
N_Gemme:Defogl	1	0.6	0.6	0.0016	0.968374	
N_Gemme:Vigoria	1	527.3	527.3	1.3993	0.249475	
Defogl:Vigoria	1	967.2	967.2	2.5664	0.123415	
Residuals	22	8290.7	376.8			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.542
IASMA_2: n.s
IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Co_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	23144	23143.8	15.6366	0.0006742	***
IASMA_2	1	12729	12728.8	8.6000	0.0077062	**
IASMA_3	1	35	35.0	0.0237	0.8791782	
N_Gemme	1	159	159.0	0.1075	0.7461562	
Defogl	1	19	19.1	0.0129	0.9106696	
Vigoria	1	31047	31047.1	20.9764	0.0001463	***
N_Gemme:Defogl	1	878	878.0	0.5932	0.4493814	
N_Gemme:Vigoria	1	3272	3272.0	2.2106	0.1512551	
Defogl:Vigoria	1	758	758.0	0.5121	0.4817566	
Residuals	22	32562	1480.1			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.645
IASMA_2: -0.53
IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Co_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	20174	20173.6	6.3393	0.01959 *
IASMA_2	1	4747	4747.2	1.4918	0.23487
IASMA_3	1	9546	9545.7	2.9996	0.09728 .
N_Gemme	1	393	393.3	0.1236	0.72851
Defogl	1	6860	6860.1	2.1557	0.15619
Vigoria	1	8313	8313.3	2.6124	0.12028
N_Gemme:Defogl	1	4734	4733.8	1.4875	0.23551
N_Gemme:Vigoria	1	9424	9424.0	2.9614	0.09931 .
Defogl:Vigoria	1	2151	2150.6	0.6758	0.41985
Residuals	22	70010	3182.3		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.473
IASMA_2: n.s
IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Co_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	107721	107721	24.3307	6.203e-05 ***
IASMA_2	1	36829	36829	8.3185	0.0086122 **
IASMA_3	1	26331	26331	5.9473	0.0232697 *
N_Gemme	1	5980	5980	1.3507	0.2576175
Defogl	1	32764	32764	7.4003	0.0124941 *
Vigoria	1	69565	69565	15.7124	0.0006586 ***
N_Gemme:Defogl	1	10971	10971	2.4779	0.1297267
N_Gemme:Vigoria	1	961	961	0.2171	0.6458701
Defogl:Vigoria	1	454	454	0.1024	0.7519353
Residuals	22	97402	4427		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.725
IASMA_2: -0.524
IASMA_3: -0.461

Poliftot2A

Analysis of Variance Table

Response: Co_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	234090	234090	12.6267	0.0017804 **
IASMA_2	1	6337	6337	0.3418	0.5647296
IASMA_3	1	86052	86052	4.6416	0.0424188 *
N_Gemme	1	15916	15916	0.8585	0.3642174
Defogl	1	410126	410126	22.1219	0.0001083 ***
Vigoria	1	7606	7606	0.4102	0.5284622
N_Gemme:Defogl	1	96687	96687	5.2152	0.0324054 *
N_Gemme:Vigoria	1	38852	38852	2.0956	0.1618229

```
Defogl:Vigoria    1    7627    7627    0.4114 0.5278922
Residuals        22 407866    18539
```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
IASMA_1: 0.604
IASMA_2: n.s
IASMA_3: -0.417
```

PvinacciaB

Analysis of Variance Table

Response: Co_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	681795	681795	13.4715	0.001343	**
IASMA_2	1	56303	56303	1.1125	0.302989	
IASMA_3	1	246906	246906	4.8786	0.037911	*
N_Gemme	1	7537	7537	0.1489	0.703268	
Defogl	1	153140	153140	3.0259	0.095921	.
Vigoria	1	405558	405558	8.0134	0.009730	**
N_Gemme:Defogl	1	152428	152428	3.0118	0.096648	.
N_Gemme:Vigoria	1	94568	94568	1.8686	0.185448	
Defogl:Vigoria	1	18417	18417	0.3639	0.552518	
Residuals	22	1113426	50610			

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
IASMA_1: -0.616
IASMA_2: n.s
IASMA_3: 0.426
```

DensMostoB

Analysis of Variance Table

Response: Co_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	1.2393e-04	1.2393e-04	10.4264	0.003859	**
IASMA_2	1	6.8000e-08	6.8000e-08	0.0057	0.940555	
IASMA_3	1	3.4200e-07	3.4200e-07	0.0288	0.866895	
N_Gemme	1	0.0000e+00	0.0000e+00	0.0000	0.997253	
Defogl	1	2.8311e-05	2.8311e-05	2.3818	0.137015	
Vigoria	1	5.4484e-05	5.4484e-05	4.5838	0.043608	*
N_Gemme:Defogl	1	4.9000e-08	4.9000e-08	0.0041	0.949630	
N_Gemme:Vigoria	1	8.9700e-07	8.9700e-07	0.0754	0.786149	
Defogl:Vigoria	1	1.2772e-04	1.2772e-04	10.7456	0.003436	**
Residuals	22	2.6150e-04	1.1886e-05			

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
IASMA_1: -0.567
IASMA_2: n.s
IASMA_3: n.s
```

BrixB

Analysis of Variance Table

Response: Co_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	6.0224	6.0224	10.5176	0.003733	**
IASMA_2	1	0.0008	0.0008	0.0014	0.970948	
IASMA_3	1	0.0166	0.0166	0.0290	0.866446	
N_Gemme	1	0.0026	0.0026	0.0046	0.946508	
Defogl	1	1.4088	1.4088	2.4603	0.131027	
Vigoria	1	2.8725	2.8725	5.0166	0.035535	*
N_Gemme:Defogl	1	0.0023	0.0023	0.0041	0.949690	
N_Gemme:Vigoria	1	0.0461	0.0461	0.0806	0.779172	
Defogl:Vigoria	1	6.0674	6.0674	10.5962	0.003627	**
Residuals	22	12.5971	0.5726			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.569

IASMA_2: n.s

IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Co_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.146451	0.146451	48.6286	5.331e-07	***
IASMA_2	1	0.001678	0.001678	0.5570	0.46336	
IASMA_3	1	0.000244	0.000244	0.0809	0.77880	
N_Gemme	1	0.000644	0.000644	0.2137	0.64844	
Defogl	1	0.000093	0.000093	0.0310	0.86184	
Vigoria	1	0.103964	0.103964	34.5210	6.521e-06	***
N_Gemme:Defogl	1	0.000575	0.000575	0.1908	0.66647	
N_Gemme:Vigoria	1	0.002686	0.002686	0.8920	0.35519	
Defogl:Vigoria	1	0.020495	0.020495	6.8054	0.01603	*
Residuals	22	0.066256	0.003012			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.83

IASMA_2: n.s

IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Co_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.79773	0.79773	13.7765	0.001215	**
IASMA_2	1	0.13370	0.13370	2.3089	0.142879	
IASMA_3	1	0.00467	0.00467	0.0806	0.779163	
N_Gemme	1	0.07438	0.07438	1.2844	0.269277	
Defogl	1	0.00000	0.00000	0.0000	0.996498	
Vigoria	1	0.51000	0.51000	8.8075	0.007106	**
N_Gemme:Defogl	1	0.01161	0.01161	0.2005	0.658691	
N_Gemme:Vigoria	1	0.02300	0.02300	0.3972	0.535056	
Defogl:Vigoria	1	0.09968	0.09968	1.7214	0.203038	
Residuals	22	1.27392	0.05791			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.621
IASMA_2: n.s
IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Co_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.75139	0.75139	28.0257	2.593e-05	***
IASMA_2	1	0.17639	0.17639	6.5791	0.0176605	*
IASMA_3	1	0.00000	0.00000	0.0000	0.9958560	
N_Gemme	1	0.01669	0.01669	0.6224	0.4385742	
Defogl	1	0.00773	0.00773	0.2884	0.5966234	
Vigoria	1	0.47479	0.47479	17.7089	0.0003629	***
N_Gemme:Defogl	1	0.00121	0.00121	0.0451	0.8337964	
N_Gemme:Vigoria	1	0.06067	0.06067	2.2629	0.1467275	
Defogl:Vigoria	1	0.08387	0.08387	3.1281	0.0908203	.
Residuals	22	0.58984	0.02681			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.748
IASMA_2: -0.48
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Co_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.01968	0.019685	0.5957	0.44843	
IASMA_2	1	0.02757	0.027574	0.8344	0.37089	
IASMA_3	1	0.01314	0.013143	0.3977	0.53476	
N_Gemme	1	0.01012	0.010124	0.3064	0.58551	
Defogl	1	0.16124	0.161242	4.8795	0.03790	*
Vigoria	1	0.03685	0.036855	1.1153	0.30239	
N_Gemme:Defogl	1	0.04816	0.048162	1.4575	0.24015	
N_Gemme:Vigoria	1	0.02585	0.025849	0.7822	0.38603	
Defogl:Vigoria	1	0.03400	0.033999	1.0289	0.32145	
Residuals	22	0.72699	0.033045			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

KB

Analysis of Variance Table

Response: Co_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	0.084549	0.084549	6.9756	0.0149181	*
IASMA_2	1	0.000049	0.000049	0.0040	0.9499814	
IASMA_3	1	0.024953	0.024953	2.0588	0.1653971	
N_Gemme	1	0.023668	0.023668	1.9527	0.1762308	
Defogl	1	0.008062	0.008062	0.6651	0.4234903	
Vigoria	1	0.249492	0.249492	20.5842	0.0001625	***
N_Gemme:Defogl	1	0.026257	0.026257	2.1664	0.1552220	
N_Gemme:Vigoria	1	0.009553	0.009553	0.7882	0.3842614	
Defogl:Vigoria	1	0.043813	0.043813	3.6148	0.0704469	.
Residuals	22	0.266653	0.012121			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.491
IASMA_2: n.s
IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Co_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	7329.9	7329.9	26.9130	3.347e-05	***
IASMA_2	1	234.5	234.5	0.8612	0.3634846	
IASMA_3	1	28.8	28.8	0.1059	0.7479187	
N_Gemme	1	784.6	784.6	2.8809	0.1037394	
Defogl	1	1295.1	1295.1	4.7551	0.0401892	*
Vigoria	1	4715.4	4715.4	17.3134	0.0004073	***
N_Gemme:Defogl	1	67.1	67.1	0.2465	0.6244699	
N_Gemme:Vigoria	1	33.1	33.1	0.1216	0.7306260	
Defogl:Vigoria	1	520.3	520.3	1.9102	0.1808120	
Residuals	22	5991.8	272.4			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.742
IASMA_2: n.s
IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Co_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	5478.8	5478.8	4.7129	0.04100	*
IASMA_2	1	722.0	722.0	0.6211	0.43906	
IASMA_3	1	1048.2	1048.2	0.9016	0.35266	
N_Gemme	1	2390.2	2390.2	2.0560	0.16567	
Defogl	1	162.4	162.4	0.1397	0.71217	
Vigoria	1	7928.1	7928.1	6.8197	0.01594	*
N_Gemme:Defogl	1	2419.6	2419.6	2.0814	0.16319	
N_Gemme:Vigoria	1	394.7	394.7	0.3395	0.56602	
Defogl:Vigoria	1	138.4	138.4	0.1190	0.73335	
Residuals	22	25575.6	1162.5			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.42
IASMA_2: n.s
IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Co_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	47204	47204	25.6948	4.458e-05 ***
IASMA_2	1	933	933	0.5079	0.483540
IASMA_3	1	494	494	0.2686	0.609418
N_Gemme	1	967	967	0.5266	0.475678
Defogl	1	6022	6022	3.2779	0.083901 .
Vigoria	1	20733	20733	11.2857	0.002832 **
N_Gemme:Defogl	1	4153	4153	2.2608	0.146903
N_Gemme:Vigoria	1	2087	2087	1.1360	0.298052
Defogl:Vigoria	1	501	501	0.2728	0.606668
Residuals	22	40416	1837		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.734
IASMA_2: n.s
IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Co_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	10141	10141.1	1.8789	0.18429
IASMA_2	1	8357	8356.7	1.5483	0.22649
IASMA_3	1	10	10.4	0.0019	0.96542
N_Gemme	1	10411	10411.5	1.9290	0.17877
Defogl	1	274	274.4	0.0508	0.82368
Vigoria	1	17444	17443.6	3.2318	0.08596 .
N_Gemme:Defogl	1	5542	5542.3	1.0268	0.32192
N_Gemme:Vigoria	1	56	55.9	0.0104	0.91988
Defogl:Vigoria	1	958	958.3	0.1775	0.67758
Residuals	22	118745	5397.5		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Co_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	97177	97177	15.4889	0.0007056 ***

IASMA_2	1	3640	3640	0.5803	0.4543023
IASMA_3	1	18256	18256	2.9099	0.1021200
N_Gemme	1	1417	1417	0.2258	0.6393373
Defogl	1	16868	16868	2.6886	0.1152922
Vigoria	1	28753	28753	4.5828	0.0436283 *
N_Gemme:Defogl	1	6937	6937	1.1057	0.3044304
N_Gemme:Vigoria	1	149	149	0.0237	0.8789478
Defogl:Vigoria	1	5152	5152	0.8212	0.3746359
Residuals	22	138027	6274		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.643
 IASMA_2: n.s
 IASMA_3: n.s

DONNA OLIMPIA 2008M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Do_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	27952	27952	0.7142	0.417827
IBIMET_3	1	0	0	0.0000	0.997415
N_Gemme	1	27895	27895	0.7127	0.418287
Defogl	1	480983	480983	12.2886	0.005674 **
Dirad	1	168157	168157	4.2963	0.064984 .
Vigoria	1	6411	6411	0.1638	0.694199
N_Gemme:Defogl	1	18193	18193	0.4648	0.510865
N_Gemme:Dirad	1	6436	6436	0.1644	0.693643
Defogl:Dirad	1	217402	217402	5.5544	0.040173 *
N_Gemme:Vigoria	1	3324	3324	0.0849	0.776692
Defogl:Vigoria	1	56533	56533	1.4444	0.257123
Dirad:Vigoria	1	168679	168679	4.3096	0.064635 .
N_Gemme:Defogl:Dirad	1	2344	2344	0.0599	0.811641
Residuals	10	391404	39140		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: n.s

DensMostoA

Analysis of Variance Table

Response: Do_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	1.0300e-07	1.0300e-07	0.0473	0.832258
IBIMET_3	1	1.2900e-07	1.2900e-07	0.0593	0.812520
N_Gemme	1	3.6400e-06	3.6400e-06	1.6717	0.225102
Defogl	1	3.1859e-05	3.1859e-05	14.6296	0.003347 **
Dirad	1	2.6202e-05	2.6202e-05	12.0319	0.006034 **
Vigoria	1	1.6926e-05	1.6926e-05	7.7723	0.019189 *
N_Gemme:Defogl	1	6.7650e-06	6.7650e-06	3.1064	0.108464
N_Gemme:Dirad	1	1.2000e-07	1.2000e-07	0.0552	0.819038
Defogl:Dirad	1	4.8840e-06	4.8840e-06	2.2426	0.165136

N_Gemme:Vigoria	1	2.8900e-07	2.8900e-07	0.1328	0.723104
Defogl:Vigoria	1	8.5500e-07	8.5500e-07	0.3925	0.545002
Dirad:Vigoria	1	7.1740e-06	7.1740e-06	3.2942	0.099587 .
N_Gemme:Defogl:Dirad	1	1.8000e-08	1.8000e-08	0.0081	0.929926
Residuals	10	2.1777e-05	2.1780e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Do_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.01871	0.01871	0.1618	0.695957
IBIMET_3	1	0.00551	0.00551	0.0477	0.831575
N_Gemme	1	0.13172	0.13172	1.1392	0.310921
Defogl	1	1.73259	1.73259	14.9843	0.003104 **
Dirad	1	1.24064	1.24064	10.7296	0.008351 **
Vigoria	1	0.74279	0.74279	6.4240	0.029638 *
N_Gemme:Defogl	1	0.41042	0.41042	3.5495	0.088934 .
N_Gemme:Dirad	1	0.01215	0.01215	0.1051	0.752454
Defogl:Dirad	1	0.33307	0.33307	2.8805	0.120506
N_Gemme:Vigoria	1	0.00862	0.00862	0.0746	0.790370
Defogl:Vigoria	1	0.04528	0.04528	0.3916	0.545484
Dirad:Vigoria	1	0.30514	0.30514	2.6390	0.135330
N_Gemme:Defogl:Dirad	1	0.00028	0.00028	0.0024	0.961780
Residuals	10	1.15627	0.11563		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Do_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.009708	0.0097081	1.4912	0.25005
IBIMET_3	1	0.002458	0.0024575	0.3775	0.55268
N_Gemme	1	0.001409	0.0014092	0.2164	0.65173
Defogl	1	0.013909	0.0139093	2.1364	0.17454
Dirad	1	0.024951	0.0249513	3.8325	0.07874 .
Vigoria	1	0.001867	0.0018675	0.2868	0.60396
N_Gemme:Defogl	1	0.010911	0.0109113	1.6760	0.22455
N_Gemme:Dirad	1	0.016088	0.0160875	2.4710	0.14704
Defogl:Dirad	1	0.000919	0.0009186	0.1411	0.71504
N_Gemme:Vigoria	1	0.000499	0.0004985	0.0766	0.78763
Defogl:Vigoria	1	0.000521	0.0005206	0.0800	0.78312
Dirad:Vigoria	1	0.001484	0.0014845	0.2280	0.64326
N_Gemme:Defogl:Dirad	1	0.002070	0.0020704	0.3180	0.58523
Residuals	10	0.065105	0.0065105		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

AcTitola

Analysis of Variance Table

Response: Do_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.21171	0.21171	4.2334	0.0666638	.
IBIMET_3	1	0.11610	0.11610	2.3214	0.1585817	
N_Gemme	1	0.00305	0.00305	0.0611	0.8098155	
Defogl	1	1.29843	1.29843	25.9629	0.0004672	***
Dirad	1	0.23402	0.23402	4.6793	0.0558095	.
Vigoria	1	0.09877	0.09877	1.9749	0.1902254	
N_Gemme:Defogl	1	0.16008	0.16008	3.2009	0.1038784	
N_Gemme:Dirad	1	0.01187	0.01187	0.2374	0.6366016	
Defogl:Dirad	1	0.01769	0.01769	0.3538	0.5651833	
N_Gemme:Vigoria	1	0.00043	0.00043	0.0085	0.9283141	
Defogl:Vigoria	1	0.02744	0.02744	0.5487	0.4758832	
Dirad:Vigoria	1	0.11361	0.11361	2.2717	0.1626811	
N_Gemme:Defogl:Dirad	1	0.01503	0.01503	0.3006	0.5955295	
Residuals	10	0.50011	0.05001			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

AcTarta

Analysis of Variance Table

Response: Do_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.50754	0.50754	15.4606	0.002811	**
IBIMET_3	1	0.02156	0.02156	0.6569	0.436531	
N_Gemme	1	0.08802	0.08802	2.6812	0.132583	
Defogl	1	0.87752	0.87752	26.7308	0.000419	***
Dirad	1	0.01603	0.01603	0.4882	0.500642	
Vigoria	1	0.04712	0.04712	1.4353	0.258526	
N_Gemme:Defogl	1	0.02142	0.02142	0.6526	0.437984	
N_Gemme:Dirad	1	0.00156	0.00156	0.0475	0.831775	
Defogl:Dirad	1	0.00028	0.00028	0.0086	0.927794	
N_Gemme:Vigoria	1	0.00251	0.00251	0.0763	0.787945	
Defogl:Vigoria	1	0.02232	0.02232	0.6800	0.428817	
Dirad:Vigoria	1	0.03400	0.03400	1.0358	0.332801	
N_Gemme:Defogl:Dirad	1	0.01348	0.01348	0.4108	0.535997	
Residuals	10	0.32828	0.03283			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: -0.779
IBIMET_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Do_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.19008	0.19008	10.9883	0.0078149	**
IBIMET_3	1	0.40492	0.40492	23.4078	0.0006834	***
N_Gemme	1	0.04912	0.04912	2.8394	0.1228830	
Defogl	1	0.13388	0.13388	7.7393	0.0193855	*
Dirad	1	0.04576	0.04576	2.6453	0.1349164	
Viguria	1	0.08733	0.08733	5.0483	0.0484391	*
N_Gemme:Defogl	1	0.00504	0.00504	0.2912	0.6012619	
N_Gemme:Dirad	1	0.12347	0.12347	7.1376	0.0234267	*
Defogl:Dirad	1	0.02916	0.02916	1.6855	0.2233374	
N_Gemme:Viguria	1	0.05880	0.05880	3.3989	0.0950361	.
Defogl:Viguria	1	0.01171	0.01171	0.6772	0.4297496	
Dirad:Viguria	1	0.05251	0.05251	3.0353	0.1120857	
N_Gemme:Defogl:Dirad	1	0.01507	0.01507	0.8713	0.3725884	
Residuals	10	0.17299	0.01730			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.724

IBIMET_3: 0.837

KA

Analysis of Variance Table

Response: Do_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.08294	0.082936	2.1552	0.1728
IBIMET_3	1	0.06274	0.062742	1.6305	0.2305
N_Gemme	1	0.00700	0.007000	0.1819	0.6788
Defogl	1	0.01438	0.014380	0.3737	0.5546
Dirad	1	0.04970	0.049705	1.2917	0.2822
Viguria	1	0.00157	0.001574	0.0409	0.8438
N_Gemme:Defogl	1	0.02068	0.020678	0.5373	0.4804
N_Gemme:Dirad	1	0.10407	0.104072	2.7045	0.1311
Defogl:Dirad	1	0.00687	0.006870	0.1785	0.6816
N_Gemme:Viguria	1	0.00105	0.001050	0.0273	0.8721
Defogl:Viguria	1	0.00597	0.005965	0.1550	0.7021
Dirad:Viguria	1	0.00049	0.000487	0.0126	0.9127
N_Gemme:Defogl:Dirad	1	0.00397	0.003965	0.1030	0.7548
Residuals	10	0.38481	0.038481		

IBIMET_2: n.s

IBIMET_3: n.s

ApaA

Analysis of Variance Table

Response: Do_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	55.51	55.51	0.2362	0.63742
IBIMET_3	1	73.83	73.83	0.3141	0.58748
N_Gemme	1	36.33	36.33	0.1546	0.70244
Defogl	1	927.35	927.35	3.9460	0.07506
Dirad	1	167.74	167.74	0.7138	0.41795

Vigoria	1	27.22	27.22	0.1158	0.74064
N_Gemme:Defogl	1	40.50	40.50	0.1724	0.68679
N_Gemme:Dirad	1	218.83	218.83	0.9312	0.35732
Defogl:Dirad	1	8.86	8.86	0.0377	0.84990
N_Gemme:Vigoria	1	34.65	34.65	0.1474	0.70905
Defogl:Vigoria	1	34.72	34.72	0.1478	0.70873
Dirad:Vigoria	1	6.09	6.09	0.0259	0.87535
N_Gemme:Defogl:Dirad	1	224.93	224.93	0.9571	0.35098
Residuals	10	2350.06	235.01		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Do_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	16297	16297	1.6063	0.23374
IBIMET_3	1	1818	1818	0.1792	0.68099
N_Gemme	1	263	263	0.0260	0.87521
Defogl	1	27577	27577	2.7181	0.13023
Dirad	1	86976	86976	8.5727	0.01509 *
Vigoria	1	14014	14014	1.3813	0.26711
N_Gemme:Defogl	1	4715	4715	0.4647	0.51091
N_Gemme:Dirad	1	800	800	0.0789	0.78453
Defogl:Dirad	1	786	786	0.0775	0.78637
N_Gemme:Vigoria	1	22555	22555	2.2232	0.16681
Defogl:Vigoria	1	2068	2068	0.2038	0.66127
Dirad:Vigoria	1	17313	17313	1.7065	0.22069
N_Gemme:Defogl:Dirad	1	2058	2058	0.2028	0.66208
Residuals	10	101457	10146		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	4568	4568	0.0784	0.78519
IBIMET_3	1	27615	27615	0.4739	0.50684
N_Gemme	1	15415	15415	0.2646	0.61818
Defogl	1	1238	1238	0.0213	0.88698
Dirad	1	259055	259055	4.4459	0.06119 .
Vigoria	1	47610	47610	0.8171	0.38729
N_Gemme:Defogl	1	54793	54793	0.9404	0.35505
N_Gemme:Dirad	1	8905	8905	0.1528	0.70404
Defogl:Dirad	1	16166	16166	0.2774	0.60987
N_Gemme:Vigoria	1	136374	136374	2.3405	0.15705
Defogl:Vigoria	1	10169	10169	0.1745	0.68495
Dirad:Vigoria	1	1788	1788	0.0307	0.86444

N_Gemme:Defogl:Dirad 1 619 619 0.0106 0.91994
 Residuals 10 582681 58268

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	4391	4391	0.5097	0.49158
IBIMET_3	1	389	389	0.0451	0.83603
N_Gemme	1	879	879	0.1020	0.75599
Defogl	1	42316	42316	4.9121	0.05101 .
Dirad	1	54251	54251	6.2975	0.03094 *
Vigoria	1	4190	4190	0.4864	0.50143
N_Gemme:Defogl	1	6727	6727	0.7809	0.39762
N_Gemme:Dirad	1	94	94	0.0109	0.91903
Defogl:Dirad	1	625	625	0.0726	0.79310
N_Gemme:Vigoria	1	12671	12671	1.4709	0.25308
Defogl:Vigoria	1	15	15	0.0018	0.96741
Dirad:Vigoria	1	8553	8553	0.9928	0.34255
N_Gemme:Defogl:Dirad	1	1759	1759	0.2042	0.66100
Residuals	10	86148	8615		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	7498	7498	0.1033	0.7546
IBIMET_3	1	60234	60234	0.8296	0.3838
N_Gemme	1	27663	27663	0.3810	0.5509
Defogl	1	9627	9627	0.1326	0.7233
Dirad	1	144789	144789	1.9941	0.1883
Vigoria	1	25176	25176	0.3467	0.5690
N_Gemme:Defogl	1	63652	63652	0.8767	0.3712
N_Gemme:Dirad	1	6656	6656	0.0917	0.7683
Defogl:Dirad	1	26055	26055	0.3588	0.5625
N_Gemme:Vigoria	1	121660	121660	1.6756	0.2246
Defogl:Vigoria	1	55248	55248	0.7609	0.4035
Dirad:Vigoria	1	6285	6285	0.0866	0.7746
N_Gemme:Defogl:Dirad	1	129	129	0.0018	0.9672
Residuals	10	726085	72608		

IBIMET_2: n.s

IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	11390	11390	1.7401	0.2165368
IBIMET_3	1	87869	87869	13.4238	0.0043612 **
N_Gemme	1	59820	59820	9.1388	0.0128286 *
Defogl	1	101262	101262	15.4698	0.0028058 **
Dirad	1	161314	161314	24.6441	0.0005664 ***
Vigoria	1	14455	14455	2.2083	0.1681048
N_Gemme:Defogl	1	3095	3095	0.4729	0.5072983
N_Gemme:Dirad	1	4744	4744	0.7247	0.4145327
Defogl:Dirad	1	19742	19742	3.0159	0.1130965
N_Gemme:Vigoria	1	22104	22104	3.3768	0.0959750 .
Defogl:Vigoria	1	12886	12886	1.9687	0.1908689
Dirad:Vigoria	1	8396	8396	1.2827	0.2838293
N_Gemme:Defogl:Dirad	1	509	509	0.0778	0.7859673
Residuals	10	65458	6546		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: -0.757

DensMostoB

Analysis of Variance Table

Response: Do_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	8.0480e-07	8.0480e-07	0.2545	0.62484
IBIMET_3	1	8.9923e-06	8.9923e-06	2.8437	0.12263
N_Gemme	1	1.0000e-10	1.0000e-10	0.0000	0.99481
Defogl	1	1.4383e-05	1.4383e-05	4.5486	0.05875 .
Dirad	1	4.8935e-06	4.8935e-06	1.5475	0.24187
Vigoria	1	2.2351e-06	2.2351e-06	0.7068	0.42015
N_Gemme:Defogl	1	3.0442e-06	3.0442e-06	0.9627	0.34965
N_Gemme:Dirad	1	3.2731e-06	3.2731e-06	1.0351	0.33295
Defogl:Dirad	1	7.2400e-08	7.2400e-08	0.0229	0.88278
N_Gemme:Vigoria	1	1.9443e-06	1.9443e-06	0.6149	0.45114
Defogl:Vigoria	1	7.0762e-06	7.0762e-06	2.2378	0.16555
Dirad:Vigoria	1	9.5220e-07	9.5220e-07	0.3011	0.59523
N_Gemme:Defogl:Dirad	1	1.0403e-06	1.0403e-06	0.3290	0.57893
Residuals	10	3.1622e-05	3.1622e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Do_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.00378	0.00378	0.0316	0.86237

IBIMET_3	1	0.50657	0.50657	4.2417	0.06644	.
N_Gemme	1	0.04826	0.04826	0.4041	0.53923	
Defogl	1	0.69318	0.69318	5.8042	0.03673	*
Dirad	1	0.30494	0.30494	2.5534	0.14114	
Vigoria	1	0.12041	0.12041	1.0082	0.33901	
N_Gemme:Defogl	1	0.14328	0.14328	1.1997	0.29905	
N_Gemme:Dirad	1	0.06705	0.06705	0.5614	0.47094	
Defogl:Dirad	1	0.00434	0.00434	0.0364	0.85259	
N_Gemme:Vigoria	1	0.11980	0.11980	1.0031	0.34018	
Defogl:Vigoria	1	0.17495	0.17495	1.4650	0.25398	
Dirad:Vigoria	1	0.02007	0.02007	0.1680	0.69051	
N_Gemme:Defogl:Dirad	1	0.04415	0.04415	0.3696	0.55675	
Residuals	10	1.19426	0.11943			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Do_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.001977	0.0019766	0.5720	0.46690	
IBIMET_3	1	0.001466	0.0014659	0.4242	0.52952	
N_Gemme	1	0.000310	0.0003102	0.0898	0.77062	
Defogl	1	0.006966	0.0069660	2.0159	0.18607	
Dirad	1	0.000887	0.0008866	0.2566	0.62347	
Vigoria	1	0.018191	0.0181908	5.2643	0.04468	*
N_Gemme:Defogl	1	0.016813	0.0168134	4.8657	0.05193	.
N_Gemme:Dirad	1	0.000178	0.0001779	0.0515	0.82509	
Defogl:Dirad	1	0.000482	0.0004817	0.1394	0.71667	
N_Gemme:Vigoria	1	0.011325	0.0113246	3.2772	0.10035	
Defogl:Vigoria	1	0.000718	0.0007178	0.2077	0.65828	
Dirad:Vigoria	1	0.015468	0.0154676	4.4762	0.06046	.
N_Gemme:Defogl:Dirad	1	0.003590	0.0035898	1.0389	0.33211	
Residuals	10	0.034555	0.0034555			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_2	1	0.08723	0.087227	2.3978	0.15255	
IBIMET_3	1	0.26244	0.262443	7.2143	0.02286	*
N_Gemme	1	0.00099	0.000987	0.0271	0.87242	
Defogl	1	0.12710	0.127095	3.4937	0.09113	.
Dirad	1	0.06936	0.069362	1.9067	0.19741	
Vigoria	1	0.14332	0.143317	3.9397	0.07526	.
N_Gemme:Defogl	1	0.00015	0.000149	0.0041	0.95016	
N_Gemme:Dirad	1	0.07649	0.076490	2.1026	0.17768	

Defogl:Dirad	1	0.00273	0.002734	0.0752	0.78954
N_Gemme:Vigoria	1	0.03603	0.036029	0.9904	0.34311
Defogl:Vigoria	1	0.00585	0.005847	0.1607	0.69691
Dirad:Vigoria	1	0.10309	0.103087	2.8338	0.12321
N_Gemme:Defogl:Dirad	1	0.00978	0.009784	0.2690	0.61532
Residuals	10	0.36378	0.036378		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: 0.647

AcTartB

Analysis of Variance Table

Response: Do_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.050780	0.050780	1.9394	0.19392
IBIMET_3	1	0.022426	0.022426	0.8565	0.37652
N_Gemme	1	0.019193	0.019193	0.7330	0.41196
Defogl	1	0.087575	0.087575	3.3446	0.09736 .
Dirad	1	0.019329	0.019329	0.7382	0.41036
Vigoria	1	0.000069	0.000069	0.0026	0.96020
N_Gemme:Defogl	1	0.000436	0.000436	0.0167	0.89985
N_Gemme:Dirad	1	0.066662	0.066662	2.5459	0.14166
Defogl:Dirad	1	0.001328	0.001328	0.0507	0.82634
N_Gemme:Vigoria	1	0.003370	0.003370	0.1287	0.72722
Defogl:Vigoria	1	0.007489	0.007489	0.2860	0.60448
Dirad:Vigoria	1	0.119986	0.119986	4.5824	0.05797 .
N_Gemme:Defogl:Dirad	1	0.003218	0.003218	0.1229	0.73317
Residuals	10	0.261838	0.026184		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.62800	0.62800	21.3216	0.0009542 ***
IBIMET_3	1	0.85210	0.85210	28.9303	0.0003107 ***
N_Gemme	1	0.14580	0.14580	4.9501	0.0502777 .
Defogl	1	0.02908	0.02908	0.9874	0.3438255
Dirad	1	0.00838	0.00838	0.2845	0.6054410
Vigoria	1	0.01792	0.01792	0.6083	0.4534991
N_Gemme:Defogl	1	0.05430	0.05430	1.8436	0.2043816
N_Gemme:Dirad	1	0.02935	0.02935	0.9966	0.3416842
Defogl:Dirad	1	0.00060	0.00060	0.0205	0.8890352
N_Gemme:Vigoria	1	0.01214	0.01214	0.4123	0.5352461
Defogl:Vigoria	1	0.00440	0.00440	0.1495	0.7071588
Dirad:Vigoria	1	0.25530	0.25530	8.6678	0.0146819 *
N_Gemme:Defogl:Dirad	1	0.05841	0.05841	1.9831	0.1893847
Residuals	10	0.29454	0.02945		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: 0.825

IBIMET_3: 0.862

KB

Analysis of Variance Table

Response: Do_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	0.127678	0.127678	4.6988	0.05539 .
IBIMET_3	1	0.086716	0.086716	3.1913	0.10433
N_Gemme	1	0.004548	0.004548	0.1674	0.69107
Defogl	1	0.002800	0.002800	0.1030	0.75481
Dirad	1	0.006232	0.006232	0.2293	0.64231
Vigoria	1	0.085145	0.085145	3.1335	0.10712
N_Gemme:Defogl	1	0.082236	0.082236	3.0265	0.11254
N_Gemme:Dirad	1	0.007290	0.007290	0.2683	0.61575
Defogl:Dirad	1	0.001213	0.001213	0.0446	0.83692
N_Gemme:Vigoria	1	0.058777	0.058777	2.1631	0.17211
Defogl:Vigoria	1	0.002711	0.002711	0.0998	0.75861
Dirad:Vigoria	1	0.135215	0.135215	4.9762	0.04978 *
N_Gemme:Defogl:Dirad	1	0.037808	0.037808	1.3914	0.26547
Residuals	10	0.271722	0.027172		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

ApaB

Analysis of Variance Table

Response: Do_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	308.88	308.88	1.6036	0.234102
IBIMET_3	1	326.85	326.85	1.6968	0.221900
N_Gemme	1	48.33	48.33	0.2509	0.627284
Defogl	1	3064.08	3064.08	15.9070	0.002566 **
Dirad	1	64.63	64.63	0.3355	0.575232
Vigoria	1	187.91	187.91	0.9755	0.346599
N_Gemme:Defogl	1	593.69	593.69	3.0821	0.109683
N_Gemme:Dirad	1	176.36	176.36	0.9156	0.361197
Defogl:Dirad	1	32.66	32.66	0.1695	0.689217
N_Gemme:Vigoria	1	397.70	397.70	2.0647	0.181290
Defogl:Vigoria	1	21.40	21.40	0.1111	0.745768
Dirad:Vigoria	1	992.75	992.75	5.1538	0.046555 *
N_Gemme:Defogl:Dirad	1	183.82	183.82	0.9543	0.351666
Residuals	10	1926.25	192.62		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	185	185.5	0.0249	0.8777
IBIMET_3	1	3244	3243.9	0.4361	0.5239
N_Gemme	1	2217	2216.5	0.2980	0.5971
Defogl	1	124	124.4	0.0167	0.8996
Dirad	1	1610	1609.8	0.2164	0.6517
Vigoria	1	1601	1600.6	0.2152	0.6527
N_Gemme:Defogl	1	93	92.6	0.0125	0.9134
N_Gemme:Dirad	1	406	405.9	0.0546	0.8200
Defogl:Dirad	1	127	127.3	0.0171	0.8985
N_Gemme:Vigoria	1	2521	2520.5	0.3389	0.5734
Defogl:Vigoria	1	217	217.0	0.0292	0.8678
Dirad:Vigoria	1	3143	3142.8	0.4225	0.5303
N_Gemme:Defogl:Dirad	1	7311	7310.9	0.9829	0.3449
Residuals	10	74380	7438.0		

IBIMET_2: n.s
 IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Do_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	14558	14558	0.2923	0.60058
IBIMET_3	1	24031	24031	0.4825	0.50310
N_Gemme	1	25765	25765	0.5174	0.48844
Defogl	1	56798	56798	1.1405	0.31065
Dirad	1	53694	53694	1.0782	0.32356
Vigoria	1	245	245	0.0049	0.94548
N_Gemme:Defogl	1	13285	13285	0.2668	0.61675
N_Gemme:Dirad	1	16530	16530	0.3319	0.57727
Defogl:Dirad	1	19910	19910	0.3998	0.54138
N_Gemme:Vigoria	1	64618	64618	1.2975	0.28122
Defogl:Vigoria	1	30224	30224	0.6069	0.45400
Dirad:Vigoria	1	71955	71955	1.4448	0.25705
N_Gemme:Defogl:Dirad	1	203239	203239	4.0810	0.07096
Residuals	10	498015	49802		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
 IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Do_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	11591	11591.3	1.7366	0.2170
IBIMET_3	1	9887	9886.8	1.4812	0.2515
N_Gemme	1	444	443.6	0.0665	0.8018
Defogl	1	18696	18696.5	2.8010	0.1251

Dirad	1	2150	2149.8	0.3221	0.5829
Vigoria	1	9322	9322.1	1.3966	0.2646
N_Gemme:Defogl	1	2540	2540.0	0.3805	0.5511
N_Gemme:Dirad	1	2932	2932.0	0.4393	0.5225
Defogl:Dirad	1	0	0.3	0.0000	0.9948
N_Gemme:Vigoria	1	677	676.8	0.1014	0.7567
Defogl:Vigoria	1	14	13.9	0.0021	0.9644
Dirad:Vigoria	1	13013	13012.9	1.9495	0.1929
N_Gemme:Defogl:Dirad	1	2631	2630.7	0.3941	0.5442
Residuals	10	66749	6674.9		

IBIMET_2: n.s
IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Do_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_2	1	4451	4451	0.0759	0.7885
IBIMET_3	1	52853	52853	0.9013	0.3648
N_Gemme	1	2881	2881	0.0491	0.8290
Defogl	1	3836	3836	0.0654	0.8033
Dirad	1	81944	81944	1.3974	0.2645
Vigoria	1	34770	34770	0.5929	0.4591
N_Gemme:Defogl	1	58774	58774	1.0023	0.3404
N_Gemme:Dirad	1	36	36	0.0006	0.9806
Defogl:Dirad	1	24001	24001	0.4093	0.5367
N_Gemme:Vigoria	1	141718	141718	2.4167	0.1511
Defogl:Vigoria	1	35073	35073	0.5981	0.4572
Dirad:Vigoria	1	158306	158306	2.6996	0.1314
N_Gemme:Defogl:Dirad	1	217175	217175	3.7035	0.0832 .
Residuals	10	586406	58641		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_2: n.s
IBIMET_3: n.s

DONNA OLIMPIA 2008M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Do_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	71305	71305	1.7535	0.218081
DIPROVE_2	1	17811	17811	0.4380	0.524671
DIPROVE_3	1	11068	11068	0.2722	0.614463
N_Gemme	1	32320	32320	0.7948	0.395866
Defogl	1	543188	543188	13.3577	0.005278 **
Dirad	1	72768	72768	1.7895	0.213809
Vigoria	1	34690	34690	0.8531	0.379781
N_Gemme:Defogl	1	17155	17155	0.4219	0.532230
N_Gemme:Dirad	1	17054	17054	0.4194	0.533414
Defogl:Dirad	1	199032	199032	4.8945	0.054238 .
N_Gemme:Vigoria	1	12432	12432	0.3057	0.593785
Defogl:Vigoria	1	173128	173128	4.2575	0.069103 .

Dirad:Vigoria	1	5681	5681	0.1397	0.717220
N_Gemme:Defogl:Dirad	1	2096	2096	0.0515	0.825464
Residuals	9	365983	40665		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Do_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1.5327e-05	1.5327e-05	10.2282	0.010863 *
DIPROVE_2	1	1.0534e-06	1.0534e-06	0.7029	0.423508
DIPROVE_3	1	4.6490e-07	4.6490e-07	0.3103	0.591100
N_Gemme	1	9.8180e-07	9.8180e-07	0.6552	0.439148
Defogl	1	2.4442e-05	2.4442e-05	16.3105	0.002935 **
Dirad	1	1.8315e-05	1.8315e-05	12.2220	0.006766 **
Vigoria	1	2.5292e-05	2.5292e-05	16.8781	0.002644 **
N_Gemme:Defogl	1	7.8194e-06	7.8194e-06	5.2181	0.048221 *
N_Gemme:Dirad	1	7.2030e-07	7.2030e-07	0.4807	0.505621
Defogl:Dirad	1	5.6058e-06	5.6058e-06	3.7408	0.085107 .
N_Gemme:Vigoria	1	4.1500e-08	4.1500e-08	0.0277	0.871452
Defogl:Vigoria	1	6.9823e-06	6.9823e-06	4.6595	0.059204 .
Dirad:Vigoria	1	6.5000e-09	6.5000e-09	0.0043	0.948960
N_Gemme:Defogl:Dirad	1	2.0050e-07	2.0050e-07	0.1338	0.722963
Residuals	9	1.3487e-05	1.4985e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.729
DIPROVE_2: n.s
DIPROVE_3: n.s

Brixa

Analysis of Variance Table

Response: Do_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.69810	0.69810	8.5065	0.017127 *
DIPROVE_2	1	0.04663	0.04663	0.5682	0.470251
DIPROVE_3	1	0.00604	0.00604	0.0736	0.792313
N_Gemme	1	0.02680	0.02680	0.3266	0.581643
Defogl	1	1.27556	1.27556	15.5430	0.003393 **
Dirad	1	0.86735	0.86735	10.5689	0.009982 **
Vigoria	1	1.26433	1.26433	15.4061	0.003484 **
N_Gemme:Defogl	1	0.49296	0.49296	6.0068	0.036704 *
N_Gemme:Dirad	1	0.02574	0.02574	0.3137	0.589098
Defogl:Dirad	1	0.31092	0.31092	3.7887	0.083441 .
N_Gemme:Vigoria	1	0.00681	0.00681	0.0830	0.779786
Defogl:Vigoria	1	0.37975	0.37975	4.6274	0.059926 .
Dirad:Vigoria	1	0.00122	0.00122	0.0149	0.905453
N_Gemme:Defogl:Dirad	1	0.00238	0.00238	0.0290	0.868456
Residuals	9	0.73860	0.08207		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.697
DIPROVE_2: n.s
DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Do_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.025567	0.0255671	6.9474	0.02710 *
DIPROVE_2	1	0.005793	0.0057929	1.5741	0.24121
DIPROVE_3	1	0.021914	0.0219141	5.9548	0.03735 *
N_Gemme	1	0.006989	0.0069894	1.8993	0.20145
Defogl	1	0.001368	0.0013681	0.3718	0.55712
Dirad	1	0.014931	0.0149311	4.0573	0.07481 .
Vigoria	1	0.016547	0.0165467	4.4963	0.06299 .
N_Gemme:Defogl	1	0.010240	0.0102402	2.7826	0.12964
N_Gemme:Dirad	1	0.006514	0.0065142	1.7701	0.21609
Defogl:Dirad	1	0.006716	0.0067162	1.8250	0.20970
N_Gemme:Vigoria	1	0.000310	0.0003101	0.0843	0.77817
Defogl:Vigoria	1	0.000186	0.0001859	0.0505	0.82720
Dirad:Vigoria	1	0.000053	0.0000535	0.0145	0.90670
N_Gemme:Defogl:Dirad	1	0.001649	0.0016489	0.4481	0.52006
Residuals	9	0.033121	0.0036801		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.66
DIPROVE_2: n.s
DIPROVE_3: -0.631

AcTitola

Analysis of Variance Table

Response: Do_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.21948	0.21948	4.5261	0.062278 .
DIPROVE_2	1	0.08410	0.08410	1.7344	0.220393
DIPROVE_3	1	0.26523	0.26523	5.4695	0.044114 *
N_Gemme	1	0.00001	0.00001	0.0001	0.991856
Defogl	1	0.78624	0.78624	16.2138	0.002988 **
Dirad	1	0.01691	0.01691	0.3488	0.569321
Vigoria	1	0.80095	0.80095	16.5173	0.002824 **
N_Gemme:Defogl	1	0.12055	0.12055	2.4861	0.149312
N_Gemme:Dirad	1	0.00002	0.00002	0.0005	0.983334
Defogl:Dirad	1	0.00519	0.00519	0.1071	0.750952
N_Gemme:Vigoria	1	0.03327	0.03327	0.6862	0.428884
Defogl:Vigoria	1	0.00007	0.00007	0.0014	0.971216
Dirad:Vigoria	1	0.00140	0.00140	0.0290	0.868618
N_Gemme:Defogl:Dirad	1	0.03846	0.03846	0.7932	0.396316
Residuals	9	0.43643	0.04849		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: 0.615

AcTartA

Analysis of Variance Table

Response: Do_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.01711	0.01711	0.2062	0.66053
DIPROVE_2	1	0.07823	0.07823	0.9429	0.35691
DIPROVE_3	1	0.00249	0.00249	0.0300	0.86632
N_Gemme	1	0.00156	0.00156	0.0187	0.89411
Defogl	1	0.55312	0.55312	6.6663	0.02960 *
Dirad	1	0.00108	0.00108	0.0131	0.91149
Vigoria	1	0.41941	0.41941	5.0549	0.05115 .
N_Gemme:Defogl	1	0.07178	0.07178	0.8652	0.37657
N_Gemme:Dirad	1	0.00010	0.00010	0.0012	0.97294
Defogl:Dirad	1	0.01240	0.01240	0.1495	0.70801
N_Gemme:Vigoria	1	0.00038	0.00038	0.0046	0.94756
Defogl:Vigoria	1	0.03314	0.03314	0.3994	0.54313
Dirad:Vigoria	1	0.02432	0.02432	0.2931	0.60141
N_Gemme:Defogl:Dirad	1	0.01978	0.01978	0.2384	0.63702
Residuals	9	0.74675	0.08297		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Do_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.06592	0.065925	1.2056	0.3007
DIPROVE_2	1	0.00344	0.003437	0.0628	0.8077
DIPROVE_3	1	0.12192	0.121924	2.2298	0.1696
N_Gemme	1	0.01933	0.019328	0.3535	0.5668
Defogl	1	0.16211	0.162110	2.9647	0.1192
Dirad	1	0.00200	0.001998	0.0365	0.8526
Vigoria	1	0.11033	0.110332	2.0178	0.1892
N_Gemme:Defogl	1	0.00893	0.008935	0.1634	0.6955
N_Gemme:Dirad	1	0.17242	0.172415	3.1531	0.1095
Defogl:Dirad	1	0.00831	0.008314	0.1520	0.7057
N_Gemme:Vigoria	1	0.12905	0.129053	2.3601	0.1588
Defogl:Vigoria	1	0.03778	0.037775	0.6908	0.4274
Dirad:Vigoria	1	0.01591	0.015905	0.2909	0.6027
N_Gemme:Defogl:Dirad	1	0.03026	0.030256	0.5533	0.4759
Residuals	9	0.49213	0.054681		

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Do_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.118040	0.118040	3.5797	0.09104 .
DIPROVE_2	1	0.042657	0.042657	1.2936	0.28475
DIPROVE_3	1	0.040859	0.040859	1.2391	0.29449
N_Gemme	1	0.052875	0.052875	1.6035	0.23720
Defogl	1	0.008285	0.008285	0.2513	0.62822
Dirad	1	0.049918	0.049918	1.5138	0.24974
Vigoria	1	0.033532	0.033532	1.0169	0.33960
N_Gemme:Defogl	1	0.032967	0.032967	0.9998	0.34349
N_Gemme:Dirad	1	0.038560	0.038560	1.1694	0.30765
Defogl:Dirad	1	0.030299	0.030299	0.9188	0.36282
N_Gemme:Vigoria	1	0.000562	0.000562	0.0170	0.89902
Defogl:Vigoria	1	0.000697	0.000697	0.0211	0.88762
Dirad:Vigoria	1	0.000126	0.000126	0.0038	0.95207
N_Gemme:Defogl:Dirad	1	0.000087	0.000087	0.0026	0.96022
Residuals	9	0.296773	0.032975		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

ApaA

Analysis of Variance Table

Response: Do_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1546.61	1546.61	7.0002	0.02666 *
DIPROVE_2	1	12.36	12.36	0.0560	0.81831
DIPROVE_3	1	0.34	0.34	0.0016	0.96937
N_Gemme	1	88.11	88.11	0.3988	0.54342
Defogl	1	69.84	69.84	0.3161	0.58768
Dirad	1	94.28	94.28	0.4267	0.52994
Vigoria	1	26.67	26.67	0.1207	0.73627
N_Gemme:Defogl	1	87.25	87.25	0.3949	0.54534
N_Gemme:Dirad	1	27.64	27.64	0.1251	0.73173
Defogl:Dirad	1	0.18	0.18	0.0008	0.97772
N_Gemme:Vigoria	1	26.41	26.41	0.1195	0.73749
Defogl:Vigoria	1	15.48	15.48	0.0701	0.79720
Dirad:Vigoria	1	4.39	4.39	0.0199	0.89095
N_Gemme:Defogl:Dirad	1	218.62	218.62	0.9895	0.34585
Residuals	9	1988.44	220.94		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.661
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Do_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	2283	2283	0.2327	0.64105
DIPROVE_2	1	16720	16720	1.7039	0.22416
DIPROVE_3	1	22782	22782	2.3216	0.16192
N_Gemme	1	4552	4552	0.4639	0.51296
Defogl	1	30352	30352	3.0931	0.11249
Dirad	1	46313	46313	4.7196	0.05788 .
Vigoria	1	57230	57230	5.8321	0.03893 *
N_Gemme:Defogl	1	3654	3654	0.3724	0.55679
N_Gemme:Dirad	1	78	78	0.0080	0.93077
Defogl:Dirad	1	2893	2893	0.2948	0.60032
N_Gemme:Vigoria	1	8053	8053	0.8206	0.38860
Defogl:Vigoria	1	14255	14255	1.4527	0.25883
Dirad:Vigoria	1	17	17	0.0018	0.96728
N_Gemme:Defogl:Dirad	1	1199	1199	0.1222	0.73469
Residuals	9	88316	9813		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	59379	59379	1.0953	0.32259
DIPROVE_2	1	3643	3643	0.0672	0.80129
DIPROVE_3	1	93895	93895	1.7320	0.22069
N_Gemme	1	46129	46129	0.8509	0.38036
Defogl	1	17667	17667	0.3259	0.58206
Dirad	1	229245	229245	4.2287	0.06989 .
Vigoria	1	64841	64841	1.1961	0.30251
N_Gemme:Defogl	1	46244	46244	0.8530	0.37980
N_Gemme:Dirad	1	25	25	0.0005	0.98334
Defogl:Dirad	1	416	416	0.0077	0.93213
N_Gemme:Vigoria	1	111256	111256	2.0523	0.18579
Defogl:Vigoria	1	4352	4352	0.0803	0.78334
Dirad:Vigoria	1	628	628	0.0116	0.91666
N_Gemme:Defogl:Dirad	1	1370	1370	0.0253	0.87721
Residuals	9	487906	54212		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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DIPROVE_1	1	10367	10367	1.4336	0.26176
DIPROVE_2	1	12769	12769	1.7658	0.21661
DIPROVE_3	1	21395	21395	2.9586	0.11954
N_Gemme	1	140	140	0.0194	0.89224
Defogl	1	33633	33633	4.6510	0.05939 .
Dirad	1	29799	29799	4.1207	0.07294 .
Vigoria	1	28043	28043	3.8779	0.08044 .
N_Gemme:Defogl	1	5996	5996	0.8292	0.38625
N_Gemme:Dirad	1	1629	1629	0.2252	0.64638
Defogl:Dirad	1	2080	2080	0.2877	0.60471
N_Gemme:Vigoria	1	2968	2968	0.4104	0.53772
Defogl:Vigoria	1	6418	6418	0.8875	0.37074
Dirad:Vigoria	1	1264	1264	0.1748	0.68572
N_Gemme:Defogl:Dirad	1	1426	1426	0.1971	0.66753
Residuals	9	65083	7231		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	125402	125402	1.9218	0.1990
DIPROVE_2	1	0	0	0.0000	0.9986
DIPROVE_3	1	68234	68234	1.0457	0.3332
N_Gemme	1	25280	25280	0.3874	0.5491
Defogl	1	65328	65328	1.0012	0.3432
Dirad	1	174171	174171	2.6692	0.1367
Vigoria	1	9054	9054	0.1388	0.7181
N_Gemme:Defogl	1	58595	58595	0.8980	0.3681
N_Gemme:Dirad	1	2694	2694	0.0413	0.8435
Defogl:Dirad	1	3161	3161	0.0485	0.8307
N_Gemme:Vigoria	1	114340	114340	1.7523	0.2182
Defogl:Vigoria	1	40977	40977	0.6280	0.4485
Dirad:Vigoria	1	3367	3367	0.0516	0.8254
N_Gemme:Defogl:Dirad	1	2892	2892	0.0443	0.8379
Residuals	9	587261	65251		

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	94051	94051	14.1120	0.0045086 **
DIPROVE_2	1	2627	2627	0.3942	0.5456923
DIPROVE_3	1	15160	15160	2.2747	0.1657715
N_Gemme	1	19437	19437	2.9164	0.1218627
Defogl	1	105045	105045	15.7615	0.0032543 **

Dirad	1	211449	211449	31.7270	0.0003206	***
Vigoria	1	3449	3449	0.5175	0.4901495	
N_Gemme:Defogl	1	10754	10754	1.6136	0.2358351	
N_Gemme:Dirad	1	17079	17079	2.5627	0.1438771	
Defogl:Dirad	1	10841	10841	1.6267	0.2341008	
N_Gemme:Vigoria	1	19971	19971	2.9965	0.1174912	
Defogl:Vigoria	1	2860	2860	0.4292	0.5287766	
Dirad:Vigoria	1	118	118	0.0176	0.8972412	
N_Gemme:Defogl:Dirad	1	221	221	0.0332	0.8594794	
Residuals	9	59982	6665			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.781
DIPROVE_2: n.s
DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Do_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1.5480e-07	1.5480e-07	0.0559	0.81841
DIPROVE_2	1	1.7590e-07	1.7590e-07	0.0635	0.80668
DIPROVE_3	1	1.5013e-05	1.5013e-05	5.4210	0.04487 *
N_Gemme	1	6.1200e-08	6.1200e-08	0.0221	0.88509
Defogl	1	1.5276e-05	1.5276e-05	5.5160	0.04340 *
Dirad	1	5.6000e-09	5.6000e-09	0.0020	0.96503
Vigoria	1	1.3266e-05	1.3266e-05	4.7900	0.05638 .
N_Gemme:Defogl	1	4.4590e-07	4.4590e-07	0.1610	0.69759
N_Gemme:Dirad	1	1.3307e-06	1.3307e-06	0.4805	0.50570
Defogl:Dirad	1	5.9260e-07	5.9260e-07	0.2140	0.65465
N_Gemme:Vigoria	1	2.7070e-07	2.7070e-07	0.0978	0.76166
Defogl:Vigoria	1	8.0902e-06	8.0902e-06	2.9212	0.12159
Dirad:Vigoria	1	1.5760e-07	1.5760e-07	0.0569	0.81680
N_Gemme:Defogl:Dirad	1	5.6720e-07	5.6720e-07	0.2048	0.66158
Residuals	9	2.4925e-05	2.7695e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.613

BrixB

Analysis of Variance Table

Response: Do_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.07682	0.07682	0.7812	0.39977
DIPROVE_2	1	0.00811	0.00811	0.0825	0.78043
DIPROVE_3	1	0.61090	0.61090	6.2124	0.03428 *
N_Gemme	1	0.01370	0.01370	0.1394	0.71756
Defogl	1	0.67902	0.67902	6.9052	0.02746 *
Dirad	1	0.00266	0.00266	0.0271	0.87288
Vigoria	1	0.61407	0.61407	6.2447	0.03392 *
N_Gemme:Defogl	1	0.01825	0.01825	0.1855	0.67678

N_Gemme:Dirad	1	0.03685	0.03685	0.3748	0.55556
Defogl:Dirad	1	0.01966	0.01966	0.2000	0.66532
N_Gemme:Vigoria	1	0.01404	0.01404	0.1428	0.71426
Defogl:Vigoria	1	0.35196	0.35196	3.5791	0.09106 .
Dirad:Vigoria	1	0.07248	0.07248	0.7371	0.41287
N_Gemme:Defogl:Dirad	1	0.04148	0.04148	0.4218	0.53225
Residuals	9	0.88502	0.09834		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: -0.639

PhB

Analysis of Variance Table

Response: Do_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.0124498	0.0124498	6.2457	0.03391	*
DIPROVE_2	1	0.0059736	0.0059736	2.9968	0.11748	
DIPROVE_3	1	0.0076725	0.0076725	3.8491	0.08139	.
N_Gemme	1	0.0018693	0.0018693	0.9378	0.35815	
Defogl	1	0.0016817	0.0016817	0.8436	0.38231	
Dirad	1	0.0002821	0.0002821	0.1415	0.71550	
Vigoria	1	0.0179232	0.0179232	8.9916	0.01499	*
N_Gemme:Defogl	1	0.0186526	0.0186526	9.3575	0.01359	*
N_Gemme:Dirad	1	0.0008420	0.0008420	0.4224	0.53198	
Defogl:Dirad	1	0.0048921	0.0048921	2.4542	0.15165	
N_Gemme:Vigoria	1	0.0088918	0.0088918	4.4608	0.06386	.
Defogl:Vigoria	1	0.0012940	0.0012940	0.6492	0.44118	
Dirad:Vigoria	1	0.0089839	0.0089839	4.5070	0.06273	.
N_Gemme:Defogl:Dirad	1	0.0035756	0.0035756	1.7938	0.21331	
Residuals	9	0.0179400	0.0019933			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.64
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.13399	0.133991	2.1434	0.17722	
DIPROVE_2	1	0.00449	0.004489	0.0718	0.79476	
DIPROVE_3	1	0.03393	0.033928	0.5427	0.48006	
N_Gemme	1	0.10475	0.104751	1.6757	0.22773	
Defogl	1	0.02108	0.021080	0.3372	0.57569	
Dirad	1	0.01027	0.010270	0.1643	0.69471	
Vigoria	1	0.23288	0.232875	3.7252	0.08566	.
N_Gemme:Defogl	1	0.00700	0.007000	0.1120	0.74558	
N_Gemme:Dirad	1	0.01750	0.017500	0.2799	0.60953	
Defogl:Dirad	1	0.00919	0.009192	0.1470	0.71027	
N_Gemme:Vigoria	1	0.00523	0.005229	0.0837	0.77896	

Defogl:Vigoria	1	0.11971	0.119713	1.9150	0.19976
Dirad:Vigoria	1	0.01751	0.017509	0.2801	0.60944
N_Gemme:Defogl:Dirad	1	0.00819	0.008191	0.1310	0.72573
Residuals	9	0.56262	0.062513		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

AcTartB

Analysis of Variance Table

Response: Do_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.012344	0.012344	0.4106	0.53764
DIPROVE_2	1	0.161089	0.161089	5.3583	0.04587 *
DIPROVE_3	1	0.007184	0.007184	0.2390	0.63665
N_Gemme	1	0.008817	0.008817	0.2933	0.60127
Defogl	1	0.064746	0.064746	2.1536	0.17629
Dirad	1	0.050456	0.050456	1.6783	0.22739
Vigoria	1	0.000096	0.000096	0.0032	0.95624
N_Gemme:Defogl	1	0.005573	0.005573	0.1854	0.67693
N_Gemme:Dirad	1	0.012178	0.012178	0.4051	0.54032
Defogl:Dirad	1	0.009316	0.009316	0.3099	0.59133
N_Gemme:Vigoria	1	0.001282	0.001282	0.0426	0.84100
Defogl:Vigoria	1	0.003375	0.003375	0.1123	0.74527
Dirad:Vigoria	1	0.056464	0.056464	1.8782	0.20375
N_Gemme:Defogl:Dirad	1	0.000209	0.000209	0.0069	0.93542
Residuals	9	0.270571	0.030063		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.611
DIPROVE_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.10368	0.10368	1.1010	0.32140
DIPROVE_2	1	0.33227	0.33227	3.5286	0.09303 .
DIPROVE_3	1	0.05134	0.05134	0.5452	0.47909
N_Gemme	1	0.13871	0.13871	1.4730	0.25576
Defogl	1	0.00003	0.00003	0.0003	0.98689
Dirad	1	0.03200	0.03200	0.3399	0.57423
Vigoria	1	0.02697	0.02697	0.2864	0.60553
N_Gemme:Defogl	1	0.24673	0.24673	2.6202	0.13996
N_Gemme:Dirad	1	0.06392	0.06392	0.6788	0.43130
Defogl:Dirad	1	0.01497	0.01497	0.1590	0.69935
N_Gemme:Vigoria	1	0.04413	0.04413	0.4687	0.51085
Defogl:Vigoria	1	0.03088	0.03088	0.3279	0.58089
Dirad:Vigoria	1	0.28881	0.28881	3.0670	0.11381
N_Gemme:Defogl:Dirad	1	0.16842	0.16842	1.7885	0.21392

Residuals 9 0.84748 0.09416

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

KB

Analysis of Variance Table

Response: Do_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.037001	0.037001	2.9555	0.11970
DIPROVE_2	1	0.090210	0.090210	7.2057	0.02503 *
DIPROVE_3	1	0.055932	0.055932	4.4677	0.06369 .
N_Gemme	1	0.122343	0.122343	9.7724	0.01220 *
Defogl	1	0.000019	0.000019	0.0015	0.96997
Dirad	1	0.006093	0.006093	0.4867	0.50303
Vigoria	1	0.090244	0.090244	7.2084	0.02501 *
N_Gemme:Defogl	1	0.111968	0.111968	8.9437	0.01519 *
N_Gemme:Dirad	1	0.037075	0.037075	2.9614	0.11938
Defogl:Dirad	1	0.031256	0.031256	2.4966	0.14855
N_Gemme:Vigoria	1	0.052305	0.052305	4.1780	0.07130 .
Defogl:Vigoria	1	0.000049	0.000049	0.0039	0.95130
Dirad:Vigoria	1	0.101443	0.101443	8.1030	0.01920 *
N_Gemme:Defogl:Dirad	1	0.061481	0.061481	4.9109	0.05391 .
Residuals	9	0.112673	0.012519		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: 0.667

DIPROVE_3: n.s

ApaB

Analysis of Variance Table

Response: Do_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	882.33	882.33	2.9826	0.1182
DIPROVE_2	1	185.59	185.59	0.6273	0.4487
DIPROVE_3	1	6.55	6.55	0.0221	0.8850
N_Gemme	1	37.81	37.81	0.1278	0.7290
Defogl	1	1504.07	1504.07	5.0843	0.0506 .
Dirad	1	75.31	75.31	0.2546	0.6260
Vigoria	1	672.68	672.68	2.2739	0.1658
N_Gemme:Defogl	1	1030.80	1030.80	3.4844	0.0948 .
N_Gemme:Dirad	1	23.96	23.96	0.0810	0.7824
Defogl:Dirad	1	6.36	6.36	0.0215	0.8866
N_Gemme:Vigoria	1	304.46	304.46	1.0292	0.3369
Defogl:Vigoria	1	103.14	103.14	0.3486	0.5694
Dirad:Vigoria	1	726.40	726.40	2.4555	0.1516
N_Gemme:Defogl:Dirad	1	103.41	103.41	0.3495	0.5689
Residuals	9	2662.46	295.83		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	4898	4898.5	0.9157	0.36359
DIPROVE_2	1	221	221.1	0.0413	0.84340
DIPROVE_3	1	100	99.7	0.0186	0.89441
N_Gemme	1	5	4.5	0.0008	0.97744
Defogl	1	1011	1011.3	0.1891	0.67393
Dirad	1	493	492.8	0.0921	0.76838
Vigoria	1	3767	3766.7	0.7042	0.42312
N_Gemme:Defogl	1	7	6.7	0.0013	0.97250
N_Gemme:Dirad	1	4958	4958.5	0.9270	0.36081
Defogl:Dirad	1	431	430.9	0.0806	0.78296
N_Gemme:Vigoria	1	698	697.7	0.1304	0.72632
Defogl:Vigoria	1	4987	4987.2	0.9323	0.35948
Dirad:Vigoria	1	21970	21970.2	4.1072	0.07333 .
N_Gemme:Defogl:Dirad	1	5490	5489.9	1.0263	0.33749
Residuals	9	48142	5349.2		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Do_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	37954	37954	0.7453	0.41038
DIPROVE_2	1	6448	6448	0.1266	0.73017
DIPROVE_3	1	2705	2705	0.0531	0.82287
N_Gemme	1	93	93	0.0018	0.96681
Defogl	1	36423	36423	0.7152	0.41963
Dirad	1	21822	21822	0.4285	0.52909
Vigoria	1	11119	11119	0.2183	0.65141
N_Gemme:Defogl	1	26525	26525	0.5209	0.48880
N_Gemme:Dirad	1	29262	29262	0.5746	0.46781
Defogl:Dirad	1	28312	28312	0.5560	0.47492
N_Gemme:Vigoria	1	88314	88314	1.7342	0.22042
Defogl:Vigoria	1	2044	2044	0.0401	0.84568
Dirad:Vigoria	1	206838	206838	4.0617	0.07468 .
N_Gemme:Defogl:Dirad	1	136687	136687	2.6841	0.13578
Residuals	9	458322	50925		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s

DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Do_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	23685	23685.4	3.2265	0.1060
DIPROVE_2	1	226	226.0	0.0308	0.8646
DIPROVE_3	1	43	42.6	0.0058	0.9409
N_Gemme	1	2325	2324.5	0.3166	0.5874
Defogl	1	8773	8772.5	1.1950	0.3027
Dirad	1	103	102.7	0.0140	0.9084
Vigoria	1	153	153.5	0.0209	0.8882
N_Gemme:Defogl	1	4351	4350.6	0.5926	0.4611
N_Gemme:Dirad	1	5203	5203.5	0.7088	0.4216
Defogl:Dirad	1	1176	1176.1	0.1602	0.6983
N_Gemme:Vigoria	1	10	10.4	0.0014	0.9707
Defogl:Vigoria	1	1313	1313.4	0.1789	0.6822
Dirad:Vigoria	1	23092	23092.4	3.1457	0.1099
N_Gemme:Defogl:Dirad	1	4123	4123.1	0.5617	0.4727
Residuals	9	66069	7341.0		

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Do_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	16694	16694	0.2704	0.6156
DIPROVE_2	1	6180	6180	0.1001	0.7589
DIPROVE_3	1	590	590	0.0096	0.9243
N_Gemme	1	2188	2188	0.0354	0.8549
Defogl	1	18708	18708	0.3030	0.5954
Dirad	1	27594	27594	0.4470	0.5205
Vigoria	1	436	436	0.0071	0.9348
N_Gemme:Defogl	1	91520	91520	1.4825	0.2543
N_Gemme:Dirad	1	2891	2891	0.0468	0.8335
Defogl:Dirad	1	51744	51744	0.8382	0.3838
N_Gemme:Vigoria	1	182983	182983	2.9641	0.1192
Defogl:Vigoria	1	19187	19187	0.3108	0.5908
Dirad:Vigoria	1	261246	261246	4.2318	0.0698
N_Gemme:Defogl:Dirad	1	164661	164661	2.6673	0.1369
Residuals	9	555604	61734		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

DONNA OLIMPIA 2008M IASMA

PvinacciaA

Analysis of Variance Table

Response: Do_2008M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	206	206	0.0035	0.9538
IASMA_2	1	172865	172865	2.9708	0.1189
IASMA_3	1	59516	59516	1.0228	0.3383
N_Gemme	1	79734	79734	1.3703	0.2718
Defogl	1	184524	184524	3.1711	0.1086
Dirad	1	137424	137424	2.3617	0.1587
Viguria	1	33870	33870	0.5821	0.4650
N_Gemme:Defogl	1	3810	3810	0.0655	0.8038
N_Gemme:Dirad	1	2647	2647	0.0455	0.8359
Defogl:Dirad	1	107787	107787	1.8524	0.2066
N_Gemme:Viguria	1	6	6	0.0001	0.9920
Defogl:Viguria	1	143029	143029	2.4580	0.1514
Dirad:Viguria	1	116295	116295	1.9986	0.1911
N_Gemme:Defogl:Dirad	1	10307	10307	0.1771	0.6837
Residuals	9	523694	58188		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

DensMostoA

Analysis of Variance Table

Response: Do_2008M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	1.0537e-06	1.0537e-06	0.8399	0.383314
IASMA_2	1	8.9501e-06	8.9501e-06	7.1342	0.025581 *
IASMA_3	1	7.2930e-06	7.2930e-06	5.8133	0.039179 *
N_Gemme	1	7.3035e-06	7.3035e-06	5.8216	0.039069 *
Defogl	1	1.5874e-05	1.5874e-05	12.6537	0.006146 **
Dirad	1	2.6432e-05	2.6432e-05	21.0690	0.001309 **
Viguria	1	2.0203e-05	2.0203e-05	16.1043	0.003050 **
N_Gemme:Defogl	1	6.0270e-06	6.0270e-06	4.8042	0.056082 .
N_Gemme:Dirad	1	2.6739e-06	2.6739e-06	2.1314	0.178318
Defogl:Dirad	1	4.4190e-06	4.4190e-06	3.5224	0.093279 .
N_Gemme:Viguria	1	2.1751e-06	2.1751e-06	1.7338	0.220470
Defogl:Viguria	1	2.6240e-07	2.6240e-07	0.2092	0.658257
Dirad:Viguria	1	6.7417e-06	6.7417e-06	5.3739	0.045623 *
N_Gemme:Defogl:Dirad	1	3.9700e-08	3.9700e-08	0.0317	0.862711
Residuals	9	1.1291e-05	1.2545e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: -0.665
IASMA_3: 0.626

BrixA

Analysis of Variance Table

Response: Do_2008M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
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IASMA_1	1	0.03534	0.03534	0.5315	0.484526	
IASMA_2	1	0.50024	0.50024	7.5239	0.022737	*
IASMA_3	1	0.42145	0.42145	6.3388	0.032893	*
N_Gemme	1	0.32657	0.32657	4.9117	0.053893	.
Defogl	1	0.77751	0.77751	11.6941	0.007632	**
Dirad	1	1.21514	1.21514	18.2763	0.002065	**
Vigoria	1	1.00274	1.00274	15.0816	0.003711	**
N_Gemme:Defogl	1	0.39010	0.39010	5.8673	0.038468	*
N_Gemme:Dirad	1	0.15024	0.15024	2.2597	0.167030	
Defogl:Dirad	1	0.30829	0.30829	4.6369	0.059711	.
N_Gemme:Vigoria	1	0.09741	0.09741	1.4650	0.256960	
Defogl:Vigoria	1	0.02515	0.02515	0.3782	0.553787	
Dirad:Vigoria	1	0.29441	0.29441	4.4280	0.064667	.
N_Gemme:Defogl:Dirad	1	0.00022	0.00022	0.0033	0.955246	
Residuals	9	0.59839	0.06649			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: -0.675
 IASMA_3: 0.643

PhA

Analysis of Variance Table

Response: Do_2008M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.007577	0.0075767	1.2301	0.2962
IASMA_2	1	0.000605	0.0006051	0.0982	0.7611
IASMA_3	1	0.004627	0.0046272	0.7512	0.4086
N_Gemme	1	0.007756	0.0077558	1.2591	0.2909
Defogl	1	0.001370	0.0013698	0.2224	0.6485
Dirad	1	0.016482	0.0164824	2.6759	0.1363
Vigoria	1	0.012898	0.0128976	2.0939	0.1818
N_Gemme:Defogl	1	0.016841	0.0168411	2.7341	0.1326
N_Gemme:Dirad	1	0.015975	0.0159748	2.5935	0.1418
Defogl:Dirad	1	0.001742	0.0017419	0.2828	0.6077
N_Gemme:Vigoria	1	0.002761	0.0027606	0.4482	0.5200
Defogl:Vigoria	1	0.002468	0.0024681	0.4007	0.5425
Dirad:Vigoria	1	0.003365	0.0033652	0.5463	0.4787
N_Gemme:Defogl:Dirad	1	0.001996	0.0019961	0.3241	0.5831
Residuals	9	0.055437	0.0061596		

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Do_2008M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.02857	0.02857	0.4053	0.54023
IASMA_2	1	0.22074	0.22074	3.1313	0.11059
IASMA_3	1	0.44040	0.44040	6.2471	0.03389 *
N_Gemme	1	0.01458	0.01458	0.2069	0.66001
Defogl	1	0.28212	0.28212	4.0019	0.07650 .

Dirad	1	0.05510	0.05510	0.7817	0.39964
Vigoria	1	0.72918	0.72918	10.3435	0.01055 *
N_Gemme:Defogl	1	0.21848	0.21848	3.0992	0.11218
N_Gemme:Dirad	1	0.00664	0.00664	0.0942	0.76591
Defogl:Dirad	1	0.03078	0.03078	0.4366	0.52532
N_Gemme:Vigoria	1	0.00147	0.00147	0.0209	0.88820
Defogl:Vigoria	1	0.00277	0.00277	0.0393	0.84732
Dirad:Vigoria	1	0.07993	0.07993	1.1339	0.31468
N_Gemme:Defogl:Dirad	1	0.06309	0.06309	0.8949	0.36884
Residuals	9	0.63447	0.07050		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: -0.64

AcTartA

Analysis of Variance Table

Response: Do_2008M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.24400	0.24400	3.5954	0.09044 .
IASMA_2	1	0.09851	0.09851	1.4516	0.25900
IASMA_3	1	0.19503	0.19503	2.8738	0.12427
N_Gemme	1	0.00319	0.00319	0.0470	0.83313
Defogl	1	0.09585	0.09585	1.4124	0.26508
Dirad	1	0.00352	0.00352	0.0518	0.82505
Vigoria	1	0.51924	0.51924	7.6512	0.02189 *
N_Gemme:Defogl	1	0.07562	0.07562	1.1143	0.31866
N_Gemme:Dirad	1	0.01135	0.01135	0.1673	0.69210
Defogl:Dirad	1	0.00162	0.00162	0.0239	0.88063
N_Gemme:Vigoria	1	0.01005	0.01005	0.1481	0.70933
Defogl:Vigoria	1	0.00170	0.00170	0.0250	0.87780
Dirad:Vigoria	1	0.09589	0.09589	1.4130	0.26497
N_Gemme:Defogl:Dirad	1	0.01532	0.01532	0.2257	0.64604
Residuals	9	0.61077	0.06786		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Do_2008M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.22685	0.226852	5.6428	0.04154 *
IASMA_2	1	0.19846	0.198463	4.9366	0.05340 .
IASMA_3	1	0.12020	0.120196	2.9898	0.11785
N_Gemme	1	0.00000	0.000002	0.0000	0.99508
Defogl	1	0.12915	0.129146	3.2124	0.10668
Dirad	1	0.03445	0.034447	0.8568	0.37878
Vigoria	1	0.03558	0.035578	0.8850	0.37141
N_Gemme:Defogl	1	0.00590	0.005899	0.1467	0.71058

N_Gemme:Dirad	1	0.04389	0.043892	1.0918	0.32333
Defogl:Dirad	1	0.09016	0.090165	2.2428	0.16847
N_Gemme:Vigoria	1	0.09604	0.096042	2.3890	0.15660
Defogl:Vigoria	1	0.00097	0.000965	0.0240	0.88028
Dirad:Vigoria	1	0.00516	0.005160	0.1284	0.72841
N_Gemme:Defogl:Dirad	1	0.03121	0.031207	0.7762	0.40121
Residuals	9	0.36182	0.040202		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.621
IASMA_2: n.s
IASMA_3: n.s

KA

Analysis of Variance Table

Response: Do_2008M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.04916	0.049157	1.2754	0.2880
IASMA_2	1	0.00294	0.002945	0.0764	0.7885
IASMA_3	1	0.00789	0.007886	0.2046	0.6617
N_Gemme	1	0.05579	0.055793	1.4476	0.2596
Defogl	1	0.00131	0.001310	0.0340	0.8578
Dirad	1	0.03539	0.035390	0.9182	0.3630
Vigoria	1	0.02427	0.024267	0.6296	0.4479
N_Gemme:Defogl	1	0.05661	0.056608	1.4688	0.2564
N_Gemme:Dirad	1	0.10505	0.105049	2.7256	0.1331
Defogl:Dirad	1	0.02035	0.020348	0.5280	0.4859
N_Gemme:Vigoria	1	0.01887	0.018866	0.4895	0.5018
Defogl:Vigoria	1	0.00532	0.005317	0.1379	0.7189
Dirad:Vigoria	1	0.01621	0.016207	0.4205	0.5329
N_Gemme:Defogl:Dirad	1	0.00022	0.000223	0.0058	0.9411
Residuals	9	0.34687	0.038541		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

ApaA

Analysis of Variance Table

Response: Do_2008M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	258.77	258.77	1.0262	0.3375
IASMA_2	1	86.35	86.35	0.3425	0.5728
IASMA_3	1	0.45	0.45	0.0018	0.9671
N_Gemme	1	44.43	44.43	0.1762	0.6845
Defogl	1	888.07	888.07	3.5219	0.0933
Dirad	1	206.76	206.76	0.8200	0.3888
Vigoria	1	28.26	28.26	0.1121	0.7455
N_Gemme:Defogl	1	26.85	26.85	0.1065	0.7517
N_Gemme:Dirad	1	92.29	92.29	0.3660	0.5601
Defogl:Dirad	1	1.15	1.15	0.0046	0.9475
N_Gemme:Vigoria	1	29.02	29.02	0.1151	0.7422
Defogl:Vigoria	1	17.24	17.24	0.0684	0.7996
Dirad:Vigoria	1	11.82	11.82	0.0469	0.8334

N_Gemme:Defogl:Dirad	1	245.80	245.80	0.9748	0.3493
Residuals	9	2269.37	252.15		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Do_2008M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	26	26	0.0026	0.96009
IASMA_2	1	22441	22441	2.2543	0.16749
IASMA_3	1	16272	16272	1.6345	0.23306
N_Gemme	1	1862	1862	0.1870	0.67559
Defogl	1	491	491	0.0493	0.82920
Dirad	1	69785	69785	7.0100	0.02658 *
Vigoria	1	46334	46334	4.6543	0.05932 .
N_Gemme:Defogl	1	2796	2796	0.2809	0.60896
N_Gemme:Dirad	1	3251	3251	0.3265	0.58170
Defogl:Dirad	1	79	79	0.0080	0.93082
N_Gemme:Vigoria	1	27205	27205	2.7328	0.13270
Defogl:Vigoria	1	7756	7756	0.7791	0.40038
Dirad:Vigoria	1	8283	8283	0.8320	0.38547
N_Gemme:Defogl:Dirad	1	2523	2523	0.2534	0.62678
Residuals	9	89595	9955		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2008M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	16190	16190	0.2809	0.6089
IASMA_2	1	51807	51807	0.8988	0.3679
IASMA_3	1	10570	10570	0.1834	0.6785
N_Gemme	1	35829	35829	0.6216	0.4507
Defogl	1	7813	7813	0.1356	0.7213
Dirad	1	282200	282200	4.8962	0.0542 .
Vigoria	1	33133	33133	0.5749	0.4677
N_Gemme:Defogl	1	16826	16826	0.2919	0.6021
N_Gemme:Dirad	1	9755	9755	0.1692	0.6904
Defogl:Dirad	1	29307	29307	0.5085	0.4939
N_Gemme:Vigoria	1	139037	139037	2.4123	0.1548
Defogl:Vigoria	1	3032	3032	0.0526	0.8237
Dirad:Vigoria	1	6560	6560	0.1138	0.7436
N_Gemme:Defogl:Dirad	1	6203	6203	0.1076	0.7504
Residuals	9	518734	57637		

Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2008M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	12	12	0.0015	0.96956
IASMA_2	1	27796	27796	3.5612	0.09176 .
IASMA_3	1	20064	20064	2.5706	0.14333
N_Gemme	1	561	561	0.0719	0.79465
Defogl	1	6557	6557	0.8401	0.38326
Dirad	1	47262	47262	6.0553	0.03611 *
Vigoria	1	18586	18586	2.3813	0.15719
N_Gemme:Defogl	1	4940	4940	0.6329	0.44678
N_Gemme:Dirad	1	1799	1799	0.2305	0.64264
Defogl:Dirad	1	43	43	0.0055	0.94233
N_Gemme:Vigoria	1	18098	18098	2.3187	0.16216
Defogl:Vigoria	1	1468	1468	0.1881	0.67468
Dirad:Vigoria	1	2588	2588	0.3315	0.57886
N_Gemme:Defogl:Dirad	1	2990	2990	0.3831	0.55128
Residuals	9	70245	7805		

Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2008M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	22429	22429	0.3197	0.5856
IASMA_2	1	61489	61489	0.8765	0.3736
IASMA_3	1	37728	37728	0.5378	0.4820
N_Gemme	1	38411	38411	0.5475	0.4782
Defogl	1	2554	2554	0.0364	0.8529
Dirad	1	201781	201781	2.8763	0.1241
Vigoria	1	38	38	0.0005	0.9820
N_Gemme:Defogl	1	13956	13956	0.1989	0.6661
N_Gemme:Dirad	1	6108	6108	0.0871	0.7746
Defogl:Dirad	1	39101	39101	0.5574	0.4744
N_Gemme:Vigoria	1	121035	121035	1.7253	0.2215
Defogl:Vigoria	1	42092	42092	0.6000	0.4584
Dirad:Vigoria	1	55284	55284	0.7880	0.3978
N_Gemme:Defogl:Dirad	1	7372	7372	0.1051	0.7532
Residuals	9	631380	70153		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2008M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	10672	10672	1.7648	0.2167247
IASMA_2	1	33432	33432	5.5285	0.0432143 *
IASMA_3	1	4370	4370	0.7227	0.4172972
N_Gemme	1	96472	96472	15.9533	0.0031380 **
Defogl	1	96723	96723	15.9947	0.0031135 **
Dirad	1	210000	210000	34.7270	0.0002311 ***
Vigoria	1	1968	1968	0.3255	0.5822977
N_Gemme:Defogl	1	172	172	0.0284	0.8698546
N_Gemme:Dirad	1	1078	1078	0.1783	0.6827370
Defogl:Dirad	1	18298	18298	3.0258	0.1159453
N_Gemme:Vigoria	1	15681	15681	2.5932	0.1417837
Defogl:Vigoria	1	26274	26274	4.3448	0.0667851 .
Dirad:Vigoria	1	3468	3468	0.5734	0.4682608
N_Gemme:Defogl:Dirad	1	13	13	0.0021	0.9643600
Residuals	9	54424	6047		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: 0.617
IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Do_2008M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	8.8013e-06	8.8013e-06	4.9932	0.05231 .
IASMA_2	1	5.1130e-07	5.1130e-07	0.2901	0.60323
IASMA_3	1	1.6300e-05	1.6300e-05	9.2475	0.01400 *
N_Gemme	1	1.4100e-08	1.4100e-08	0.0080	0.93065
Defogl	1	3.8689e-06	3.8689e-06	2.1949	0.17261
Dirad	1	2.5993e-06	2.5993e-06	1.4747	0.25551
Vigoria	1	9.2833e-06	9.2833e-06	5.2666	0.04739 *
N_Gemme:Defogl	1	7.0264e-06	7.0264e-06	3.9862	0.07698 .
N_Gemme:Dirad	1	2.1440e-07	2.1440e-07	0.1217	0.73526
Defogl:Dirad	1	6.9000e-08	6.9000e-08	0.0392	0.84751
N_Gemme:Vigoria	1	5.3497e-06	5.3497e-06	3.0350	0.11546
Defogl:Vigoria	1	7.8480e-06	7.8480e-06	4.4523	0.06406 .
Dirad:Vigoria	1	1.7760e-06	1.7760e-06	1.0075	0.34171
N_Gemme:Defogl:Dirad	1	8.0690e-07	8.0690e-07	0.4578	0.51567
Residuals	9	1.5864e-05	1.7627e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: 0.712

BrixB

Analysis of Variance Table

Response: Do_2008M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.34764	0.34764	3.8560	0.08116 .
IASMA_2	1	0.06448	0.06448	0.7152	0.41964
IASMA_3	1	0.55307	0.55307	6.1346	0.03517 *
N_Gemme	1	0.02582	0.02582	0.2863	0.60554
Defogl	1	0.26662	0.26662	2.9573	0.11961
Dirad	1	0.16376	0.16376	1.8164	0.21069
Vigoria	1	0.42421	0.42421	4.7052	0.05819 .
N_Gemme:Defogl	1	0.21388	0.21388	2.3723	0.15789
N_Gemme:Dirad	1	0.00023	0.00023	0.0025	0.96107
Defogl:Dirad	1	0.01799	0.01799	0.1996	0.66561
N_Gemme:Vigoria	1	0.19869	0.19869	2.2038	0.17183
Defogl:Vigoria	1	0.29138	0.29138	3.2319	0.10576
Dirad:Vigoria	1	0.00671	0.00671	0.0744	0.79115
N_Gemme:Defogl:Dirad	1	0.05917	0.05917	0.6564	0.43874
Residuals	9	0.81140	0.09016		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: 0.637

PhB

Analysis of Variance Table

Response: Do_2008M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.0001284	0.0001284	0.0383	0.84922
IASMA_2	1	0.0107116	0.0107116	3.1950	0.10750
IASMA_3	1	0.0002323	0.0002323	0.0693	0.79829
N_Gemme	1	0.0023678	0.0023678	0.7062	0.42246
Defogl	1	0.0017956	0.0017956	0.5356	0.48289
Dirad	1	0.0014294	0.0014294	0.4264	0.53011
Vigoria	1	0.0158830	0.0158830	4.7375	0.05749 .
N_Gemme:Defogl	1	0.0118436	0.0118436	3.5326	0.09288 .
N_Gemme:Dirad	1	0.0004465	0.0004465	0.1332	0.72359
Defogl:Dirad	1	0.0007976	0.0007976	0.2379	0.63739
N_Gemme:Vigoria	1	0.0161460	0.0161460	4.8159	0.05584 .
Defogl:Vigoria	1	0.0004208	0.0004208	0.1255	0.73128
Dirad:Vigoria	1	0.0160888	0.0160888	4.7989	0.05619 .
N_Gemme:Defogl:Dirad	1	0.0044588	0.0044588	1.3299	0.27852
Residuals	9	0.0301737	0.0033526		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2008M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00635	0.006353	0.0998	0.7593
IASMA_2	1	0.09606	0.096060	1.5093	0.2504
IASMA_3	1	0.00112	0.001122	0.0176	0.8973
N_Gemme	1	0.03175	0.031747	0.4988	0.4979
Defogl	1	0.15972	0.159717	2.5095	0.1476
Dirad	1	0.02515	0.025146	0.3951	0.5453
Vigoria	1	0.20063	0.200630	3.1523	0.1096
N_Gemme:Defogl	1	0.05556	0.055561	0.8730	0.3745
N_Gemme:Dirad	1	0.05116	0.051164	0.8039	0.3933
Defogl:Dirad	1	0.01244	0.012437	0.1954	0.6689
N_Gemme:Vigoria	1	0.00025	0.000252	0.0040	0.9512
Defogl:Vigoria	1	0.05650	0.056504	0.8878	0.3707
Dirad:Vigoria	1	0.01036	0.010364	0.1628	0.6960
N_Gemme:Defogl:Dirad	1	0.00847	0.008469	0.1331	0.7237
Residuals	9	0.57281	0.063645		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Do_2008M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.045662	0.045662	1.8311	0.20901
IASMA_2	1	0.001700	0.001700	0.0682	0.79989
IASMA_3	1	0.102977	0.102977	4.1294	0.07269 .
N_Gemme	1	0.004608	0.004608	0.1848	0.67740
Defogl	1	0.102943	0.102943	4.1281	0.07273 .
Dirad	1	0.019030	0.019030	0.7631	0.40506
Vigoria	1	0.005055	0.005055	0.2027	0.66320
N_Gemme:Defogl	1	0.015161	0.015161	0.6080	0.45557
N_Gemme:Dirad	1	0.025630	0.025630	1.0278	0.33715
Defogl:Dirad	1	0.010325	0.010325	0.4140	0.53597
N_Gemme:Vigoria	1	0.001586	0.001586	0.0636	0.80655
Defogl:Vigoria	1	0.007290	0.007290	0.2924	0.60184
Dirad:Vigoria	1	0.095684	0.095684	3.8370	0.08180 .
N_Gemme:Defogl:Dirad	1	0.001613	0.001613	0.0647	0.80498
Residuals	9	0.224434	0.024937		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2008M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.20174	0.20174	2.9711	0.11886
IASMA_2	1	0.00207	0.00207	0.0306	0.86510
IASMA_3	1	0.00842	0.00842	0.1240	0.73282

N_Gemme	1	0.01359	0.01359	0.2001	0.66518
Defogl	1	0.25374	0.25374	3.7369	0.08525 .
Dirad	1	0.00047	0.00047	0.0069	0.93564
Vigoria	1	0.00187	0.00187	0.0275	0.87194
N_Gemme:Defogl	1	0.18602	0.18602	2.7396	0.13227
N_Gemme:Dirad	1	0.04563	0.04563	0.6720	0.43351
Defogl:Dirad	1	0.02606	0.02606	0.3838	0.55095
N_Gemme:Vigoria	1	0.12131	0.12131	1.7866	0.21414
Defogl:Vigoria	1	0.08002	0.08002	1.1784	0.30589
Dirad:Vigoria	1	0.70289	0.70289	10.3517	0.01053 *
N_Gemme:Defogl:Dirad	1	0.13538	0.13538	1.9938	0.19158
Residuals	9	0.61111	0.06790		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

KB

Analysis of Variance Table

Response: Do_2008M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.010043	0.010043	0.4536	0.51755
IASMA_2	1	0.042231	0.042231	1.9074	0.20057
IASMA_3	1	0.002288	0.002288	0.1034	0.75518
N_Gemme	1	0.069744	0.069744	3.1502	0.10966
Defogl	1	0.019362	0.019362	0.8745	0.37411
Dirad	1	0.013922	0.013922	0.6288	0.44819
Vigoria	1	0.080668	0.080668	3.6436	0.08863 .
N_Gemme:Defogl	1	0.082206	0.082206	3.7130	0.08610 .
N_Gemme:Dirad	1	0.004869	0.004869	0.2199	0.65027
Defogl:Dirad	1	0.008060	0.008060	0.3641	0.56116
N_Gemme:Vigoria	1	0.120480	0.120480	5.4418	0.04455 *
Defogl:Vigoria	1	0.000000	0.000000	0.0000	0.99801
Dirad:Vigoria	1	0.184026	0.184026	8.3119	0.01809 *
N_Gemme:Defogl:Dirad	1	0.072932	0.072932	3.2941	0.10292
Residuals	9	0.199259	0.022140		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Do_2008M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	1123.76	1123.76	5.8477	0.03872 *
IASMA_2	1	1040.97	1040.97	5.4169	0.04494 *
IASMA_3	1	56.51	56.51	0.2941	0.60078
N_Gemme	1	56.65	56.65	0.2948	0.60035
Defogl	1	1072.97	1072.97	5.5834	0.04240 *
Dirad	1	9.39	9.39	0.0488	0.83001

Vigoria	1	854.96	854.96	4.4489	0.06415	.
N_Gemme:Defogl	1	456.98	456.98	2.3780	0.15745	
N_Gemme:Dirad	1	13.46	13.46	0.0700	0.79726	
Defogl:Dirad	1	35.64	35.64	0.1855	0.67686	
N_Gemme:Vigoria	1	380.89	380.89	1.9820	0.19277	
Defogl:Vigoria	1	174.60	174.60	0.9085	0.36540	
Dirad:Vigoria	1	1119.22	1119.22	5.8241	0.03904	*
N_Gemme:Defogl:Dirad	1	199.81	199.81	1.0397	0.33451	
Residuals	9	1729.54	192.17			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.628

IASMA_2: 0.613

IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2008M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	16456	16455.9	2.9364	0.1208
IASMA_2	1	135	135.3	0.0241	0.8799
IASMA_3	1	3752	3752.0	0.6695	0.4344
N_Gemme	1	1154	1154.2	0.2059	0.6607
Defogl	1	174	174.2	0.0311	0.8640
Dirad	1	2474	2474.4	0.4415	0.5230
Vigoria	1	706	705.5	0.1259	0.7309
N_Gemme:Defogl	1	613	613.2	0.1094	0.7484
N_Gemme:Dirad	1	1516	1515.9	0.2705	0.6156
Defogl:Dirad	1	207	207.0	0.0369	0.8519
N_Gemme:Vigoria	1	6914	6914.1	1.2337	0.2955
Defogl:Vigoria	1	1821	1820.6	0.3249	0.5826
Dirad:Vigoria	1	3881	3880.7	0.6925	0.4269
N_Gemme:Defogl:Dirad	1	6938	6938.4	1.2381	0.2947
Residuals	9	50437	5604.1		

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Do_2008M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	48537	48537	1.2764	0.28777	
IASMA_2	1	23820	23820	0.6264	0.44903	
IASMA_3	1	298823	298823	7.8584	0.02060	*
N_Gemme	1	19098	19098	0.5022	0.49647	
Defogl	1	29408	29408	0.7734	0.40205	
Dirad	1	39558	39558	1.0403	0.33439	
Vigoria	1	9	9	0.0002	0.98789	
N_Gemme:Defogl	1	5092	5092	0.1339	0.72288	
N_Gemme:Dirad	1	1516	1516	0.0399	0.84619	
Defogl:Dirad	1	3894	3894	0.1024	0.75627	
N_Gemme:Vigoria	1	6564	6564	0.1726	0.68753	

Defogl:Vigoria	1	48449	48449	1.2741	0.28819
Dirad:Vigoria	1	60166	60166	1.5822	0.24009
N_Gemme:Defogl:Dirad	1	165702	165702	4.3576	0.06645 .
Residuals	9	342233	38026		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: -0.683

Antoctot2B

Analysis of Variance Table

Response: Do_2008M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	5151	5150.7	0.7199	0.41818
IASMA_2	1	12112	12112.1	1.6928	0.22556
IASMA_3	1	2647	2646.9	0.3699	0.55808
N_Gemme	1	412	411.8	0.0576	0.81578
Defogl	1	29547	29547.0	4.1294	0.07269 .
Dirad	1	275	275.2	0.0385	0.84888
Vigoria	1	477	476.8	0.0666	0.80211
N_Gemme:Defogl	1	70	69.7	0.0097	0.92355
N_Gemme:Dirad	1	6158	6158.2	0.8607	0.37776
Defogl:Dirad	1	164	163.6	0.0229	0.88314
N_Gemme:Vigoria	1	1464	1463.9	0.2046	0.66174
Defogl:Vigoria	1	966	966.0	0.1350	0.72179
Dirad:Vigoria	1	12655	12654.6	1.7686	0.21628
N_Gemme:Defogl:Dirad	1	4152	4152.1	0.5803	0.46569
Residuals	9	64397	7155.2		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Do_2008M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	11653	11653	0.2294	0.64337
IASMA_2	1	99684	99684	1.9625	0.19476
IASMA_3	1	382211	382211	7.5249	0.02273 *
N_Gemme	1	3751	3751	0.0738	0.79195
Defogl	1	8633	8633	0.1700	0.68978
Dirad	1	31210	31210	0.6144	0.45325
Vigoria	1	2624	2624	0.0517	0.82528
N_Gemme:Defogl	1	144	144	0.0028	0.95869
N_Gemme:Dirad	1	12577	12577	0.2476	0.63069
Defogl:Dirad	1	5870	5870	0.1156	0.74168
N_Gemme:Vigoria	1	47155	47155	0.9284	0.36046
Defogl:Vigoria	1	21465	21465	0.4226	0.53188
Dirad:Vigoria	1	131034	131034	2.5798	0.14270
N_Gemme:Defogl:Dirad	1	187077	187077	3.6831	0.08718 .

Residuals 9 457136 50793

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: -0.675

DONNA OLIMPIA 2009M IBIMET

PvinacciaA

Analysis of Variance Table

Response: Do_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	553	553	0.0166	0.9002841
IBIMET_2	1	112313	112313	3.3724	0.0994759 .
IBIMET_3	1	285220	285220	8.5644	0.0168528 *
N_Gemme	1	719333	719333	21.5996	0.0012063 **
Defogl	1	338452	338452	10.1628	0.0110436 *
Dirad	1	1274630	1274630	38.2736	0.0001615 ***
Vigoria	1	100789	100789	3.0264	0.1159139
N_Gemme:Defogl	1	21957	21957	0.6593	0.4377555
N_Gemme:Dirad	1	46319	46319	1.3908	0.2685017
Defogl:Dirad	1	38729	38729	1.1629	0.3089153
N_Gemme:Vigoria	1	1220	1220	0.0366	0.8524734
Defogl:Vigoria	1	5	5	0.0002	0.9904372
Dirad:Vigoria	1	8113	8113	0.2436	0.6334202
N_Gemme:Defogl:Dirad	1	47193	47193	1.4171	0.2643352
Residuals	9	299728	33303		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s

IBIMET_2: n.s

IBIMET_3: 0.698

DensMostoA

Analysis of Variance Table

Response: Do_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1.6786e-06	1.6786e-06	1.0281	0.337077
IBIMET_2	1	1.1795e-05	1.1795e-05	7.2249	0.024882 *
IBIMET_3	1	2.9653e-06	2.9653e-06	1.8163	0.210695
N_Gemme	1	1.7594e-05	1.7594e-05	10.7763	0.009488 **
Defogl	1	1.5082e-05	1.5082e-05	9.2380	0.014031 *
Dirad	1	1.1533e-05	1.1533e-05	7.0642	0.026138 *
Vigoria	1	1.7218e-06	1.7218e-06	1.0546	0.331249
N_Gemme:Defogl	1	3.1430e-07	3.1430e-07	0.1925	0.671194
N_Gemme:Dirad	1	1.8500e-07	1.8500e-07	0.1133	0.744122
Defogl:Dirad	1	8.7390e-07	8.7390e-07	0.5353	0.483017
N_Gemme:Vigoria	1	4.9000e-09	4.9000e-09	0.0030	0.957385
Defogl:Vigoria	1	3.3780e-07	3.3780e-07	0.2069	0.659966
Dirad:Vigoria	1	8.4310e-07	8.4310e-07	0.5164	0.490600
N_Gemme:Defogl:Dirad	1	1.6398e-06	1.6398e-06	1.0044	0.342432
Residuals	9	1.4693e-05	1.6326e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.667
IBIMET_3: n.s

BrixA

Analysis of Variance Table

Response: Do_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.02782	0.02782	0.3534	0.566832
IBIMET_2	1	0.50951	0.50951	6.4718	0.031505 *
IBIMET_3	1	0.13346	0.13346	1.6953	0.225244
N_Gemme	1	0.87274	0.87274	11.0855	0.008806 **
Defogl	1	0.75465	0.75465	9.5855	0.012804 *
Dirad	1	0.60865	0.60865	7.7310	0.021384 *
Vigoria	1	0.15066	0.15066	1.9137	0.199905
N_Gemme:Defogl	1	0.01932	0.01932	0.2454	0.632174
N_Gemme:Dirad	1	0.00743	0.00743	0.0943	0.765730
Defogl:Dirad	1	0.05506	0.05506	0.6994	0.424645
N_Gemme:Vigoria	1	0.00022	0.00022	0.0028	0.958856
Defogl:Vigoria	1	0.01610	0.01610	0.2044	0.661859
Dirad:Vigoria	1	0.02562	0.02562	0.3254	0.582347
N_Gemme:Defogl:Dirad	1	0.05684	0.05684	0.7220	0.417507
Residuals	9	0.70855	0.07873		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.647
IBIMET_3: n.s

PhA

Analysis of Variance Table

Response: Do_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.072763	0.072763	81.0178	8.53e-06 ***
IBIMET_2	1	0.000059	0.000059	0.0662	0.80268
IBIMET_3	1	0.002681	0.002681	2.9850	0.11811
N_Gemme	1	0.007220	0.007220	8.0389	0.01955 *
Defogl	1	0.002176	0.002176	2.4229	0.15400
Dirad	1	0.008323	0.008323	9.2678	0.01392 *
Vigoria	1	0.000216	0.000216	0.2401	0.63583
N_Gemme:Defogl	1	0.000579	0.000579	0.6450	0.44260
N_Gemme:Dirad	1	0.002296	0.002296	2.5567	0.14429
Defogl:Dirad	1	0.000497	0.000497	0.5529	0.47611
N_Gemme:Vigoria	1	0.004323	0.004323	4.8134	0.05589 .
Defogl:Vigoria	1	0.002121	0.002121	2.3621	0.15869
Dirad:Vigoria	1	0.001181	0.001181	1.3149	0.28108
N_Gemme:Defogl:Dirad	1	0.000615	0.000615	0.6845	0.42944
Residuals	9	0.008083	0.000898		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.949
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTitola

Analysis of Variance Table

Response: Do_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.100453	0.100453	6.5081	0.031139	*
IBIMET_2	1	0.025009	0.025009	1.6203	0.234948	
IBIMET_3	1	0.003863	0.003863	0.2503	0.628901	
N_Gemme	1	0.230452	0.230452	14.9303	0.003824	**
Defogl	1	0.084672	0.084672	5.4856	0.043866	*
Dirad	1	0.200213	0.200213	12.9712	0.005734	**
Vigoria	1	0.000084	0.000084	0.0054	0.942822	
N_Gemme:Defogl	1	0.038920	0.038920	2.5215	0.146766	
N_Gemme:Dirad	1	0.041357	0.041357	2.6794	0.136080	
Defogl:Dirad	1	0.013313	0.013313	0.8625	0.377276	
N_Gemme:Vigoria	1	0.067607	0.067607	4.3801	0.065876	.
Defogl:Vigoria	1	0.022403	0.022403	1.4514	0.259022	
Dirad:Vigoria	1	0.022259	0.022259	1.4421	0.260446	
N_Gemme:Defogl:Dirad	1	0.007040	0.007040	0.4561	0.516415	
Residuals	9	0.138917	0.015435			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.648
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTarta

Analysis of Variance Table

Response: Do_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.278126	0.278126	43.7441	9.77e-05	***
IBIMET_2	1	0.093682	0.093682	14.7345	0.003975	**
IBIMET_3	1	0.006546	0.006546	1.0296	0.336762	
N_Gemme	1	0.141792	0.141792	22.3012	0.001085	**
Defogl	1	0.005361	0.005361	0.8431	0.382456	
Dirad	1	0.097155	0.097155	15.2806	0.003570	**
Vigoria	1	0.001002	0.001002	0.1576	0.700670	
N_Gemme:Defogl	1	0.005407	0.005407	0.8504	0.380491	
N_Gemme:Dirad	1	0.073157	0.073157	11.5063	0.007973	**
Defogl:Dirad	1	0.000083	0.000083	0.0131	0.911517	
N_Gemme:Vigoria	1	0.023364	0.023364	3.6747	0.087482	.
Defogl:Vigoria	1	0.015110	0.015110	2.3765	0.157567	
Dirad:Vigoria	1	0.001638	0.001638	0.2577	0.623944	
N_Gemme:Defogl:Dirad	1	0.038054	0.038054	5.9852	0.036971	*
Residuals	9	0.057222	0.006358			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.911
 IBIMET_2: 0.788
 IBIMET_3: n.s

AcMalicoA

Analysis of Variance Table

Response: Do_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.54806	0.54806	23.1817	0.000954	***
IBIMET_2	1	0.10271	0.10271	4.3442	0.066800	.
IBIMET_3	1	0.10714	0.10714	4.5320	0.062138	.
N_Gemme	1	0.00000	0.00000	0.0000	0.996432	
Defogl	1	0.05443	0.05443	2.3024	0.163487	
Dirad	1	0.00500	0.00500	0.2117	0.656375	
Vigoria	1	0.10786	0.10786	4.5622	0.061425	.
N_Gemme:Defogl	1	0.01418	0.01418	0.5999	0.458471	
N_Gemme:Dirad	1	0.08196	0.08196	3.4668	0.095520	.
Defogl:Dirad	1	0.01673	0.01673	0.7076	0.422020	
N_Gemme:Vigoria	1	0.06033	0.06033	2.5519	0.144626	
Defogl:Vigoria	1	0.01706	0.01706	0.7218	0.417580	
Dirad:Vigoria	1	0.00039	0.00039	0.0163	0.901193	
N_Gemme:Defogl:Dirad	1	0.08672	0.08672	3.6683	0.087718	.
Residuals	9	0.21278	0.02364			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.849
IBIMET_2: n.s
IBIMET_3: n.s

KA

Analysis of Variance Table

Response: Do_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.46924	0.46924	79.3112	9.306e-06	***
IBIMET_2	1	0.02164	0.02164	3.6578	0.08810	.
IBIMET_3	1	0.05243	0.05243	8.8616	0.01553	*
N_Gemme	1	0.02399	0.02399	4.0547	0.07489	.
Defogl	1	0.00213	0.00213	0.3594	0.56364	
Dirad	1	0.03988	0.03988	6.7405	0.02892	*
Vigoria	1	0.03084	0.03084	5.2126	0.04832	*
N_Gemme:Defogl	1	0.00045	0.00045	0.0758	0.78928	
N_Gemme:Dirad	1	0.03753	0.03753	6.3434	0.03284	*
Defogl:Dirad	1	0.00054	0.00054	0.0920	0.76849	
N_Gemme:Vigoria	1	0.00014	0.00014	0.0229	0.88307	
Defogl:Vigoria	1	0.00004	0.00004	0.0062	0.93909	
Dirad:Vigoria	1	0.00355	0.00355	0.6000	0.45845	
N_Gemme:Defogl:Dirad	1	0.00178	0.00178	0.3008	0.59670	
Residuals	9	0.05325	0.00592			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.948
IBIMET_2: n.s
IBIMET_3: 0.704

ApaA

Analysis of Variance Table

Response: Do_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	170.16	170.16	2.0280	0.1881534
IBIMET_2	1	1806.28	1806.28	21.5282	0.0012195 **
IBIMET_3	1	0.24	0.24	0.0029	0.9583130
N_Gemme	1	115.13	115.13	1.3722	0.2715097
Defogl	1	2518.90	2518.90	30.0216	0.0003906 ***
Dirad	1	133.71	133.71	1.5936	0.2385350
Viguria	1	129.62	129.62	1.5449	0.2452951
N_Gemme:Defogl	1	35.71	35.71	0.4256	0.5304848
N_Gemme:Dirad	1	416.98	416.98	4.9697	0.0527579 .
Defogl:Dirad	1	78.36	78.36	0.9340	0.3590858
N_Gemme:Viguria	1	17.06	17.06	0.2033	0.6627047
Defogl:Viguria	1	3.80	3.80	0.0453	0.8362752
Dirad:Viguria	1	269.52	269.52	3.2123	0.1066850
N_Gemme:Defogl:Dirad	1	22.15	22.15	0.2640	0.6197872
Residuals	9	755.13	83.90		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: 0.84
 IBIMET_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Do_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	3	3	0.0012	0.973610
IBIMET_2	1	55747	55747	20.1260	0.001519 **
IBIMET_3	1	5349	5349	1.9311	0.198051
N_Gemme	1	9793	9793	3.5355	0.092761 .
Defogl	1	12101	12101	4.3688	0.066166 .
Dirad	1	14006	14006	5.0566	0.051113 .
Viguria	1	5	5	0.0019	0.966220
N_Gemme:Defogl	1	1031	1031	0.3722	0.556882
N_Gemme:Dirad	1	4949	4949	1.7869	0.214114
Defogl:Dirad	1	780	780	0.2817	0.608410
N_Gemme:Viguria	1	504	504	0.1820	0.679682
Defogl:Viguria	1	5890	5890	2.1264	0.178771
Dirad:Viguria	1	2452	2452	0.8852	0.371353
N_Gemme:Defogl:Dirad	1	3874	3874	1.3985	0.267270
Residuals	9	24929	2770		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: 0.831
 IBIMET_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	7762.3	7762.3	3.0759	0.11336
IBIMET_2	1	16818.5	16818.5	6.6646	0.02962 *
IBIMET_3	1	4.2	4.2	0.0017	0.96846
N_Gemme	1	10893.9	10893.9	4.3169	0.06752 .
Defogl	1	12467.4	12467.4	4.9404	0.05333 .
Dirad	1	24807.0	24807.0	9.8301	0.01202 *
Vigoria	1	210.5	210.5	0.0834	0.77928
N_Gemme:Defogl	1	805.1	805.1	0.3190	0.58599
N_Gemme:Dirad	1	514.9	514.9	0.2040	0.66218
Defogl:Dirad	1	88.3	88.3	0.0350	0.85573
N_Gemme:Vigoria	1	200.1	200.1	0.0793	0.78462
Defogl:Vigoria	1	3817.1	3817.1	1.5126	0.24992
Dirad:Vigoria	1	48.2	48.2	0.0191	0.89312
N_Gemme:Defogl:Dirad	1	867.7	867.7	0.3438	0.57204
Residuals	9	22712.1	2523.6		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: 0.652
IBIMET_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0	0	0.0000	0.99951
IBIMET_2	1	121739	121739	12.6875	0.00610 **
IBIMET_3	1	35556	35556	3.7057	0.08636 .
N_Gemme	1	1409	1409	0.1468	0.71046
Defogl	1	6299	6299	0.6565	0.43871
Dirad	1	3853	3853	0.4015	0.54207
Vigoria	1	1910	1910	0.1991	0.66601
N_Gemme:Defogl	1	5886	5886	0.6134	0.45362
N_Gemme:Dirad	1	21831	21831	2.2752	0.16573
Defogl:Dirad	1	542	542	0.0565	0.81745
N_Gemme:Vigoria	1	5084	5084	0.5298	0.48519
Defogl:Vigoria	1	10634	10634	1.1083	0.31990
Dirad:Vigoria	1	8906	8906	0.9282	0.36051
N_Gemme:Defogl:Dirad	1	1004	1004	0.1046	0.75376
Residuals	9	86356	9595		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: 0.765
IBIMET_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	26503	26502.9	4.8936	0.05426 .

IBIMET_2	1	15139	15138.9	2.7953	0.12887
IBIMET_3	1	4999	4999.2	0.9231	0.36177
N_Gemme	1	70	70.5	0.0130	0.91168
Defogl	1	1731	1730.5	0.3195	0.58571
Dirad	1	5258	5258.3	0.9709	0.35020
Vigoria	1	274	274.3	0.0506	0.82698
N_Gemme:Defogl	1	5199	5199.2	0.9600	0.35279
N_Gemme:Dirad	1	4412	4412.2	0.8147	0.39026
Defogl:Dirad	1	689	689.1	0.1272	0.72954
N_Gemme:Vigoria	1	3554	3554.0	0.6562	0.43879
Defogl:Vigoria	1	4064	4064.4	0.7505	0.40883
Dirad:Vigoria	1	1	0.8	0.0002	0.99047
N_Gemme:Defogl:Dirad	1	3812	3812.4	0.7039	0.42319
Residuals	9	48743	5415.9		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: n.s
 IBIMET_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	189332	189332	2.9798	0.1183867
IBIMET_2	1	2430	2430	0.0382	0.8492874
IBIMET_3	1	787936	787936	12.4010	0.0065001 **
N_Gemme	1	1175103	1175103	18.4944	0.0019891 **
Defogl	1	471415	471415	7.4194	0.0234592 *
Dirad	1	1644555	1644555	25.8829	0.0006561 ***
Vigoria	1	21848	21848	0.3439	0.5720257
N_Gemme:Defogl	1	76774	76774	1.2083	0.3002004
N_Gemme:Dirad	1	316014	316014	4.9736	0.0526832 .
Defogl:Dirad	1	26326	26326	0.4143	0.5358351
N_Gemme:Vigoria	1	54431	54431	0.8567	0.3788233
Defogl:Vigoria	1	17846	17846	0.2809	0.6089590
Dirad:Vigoria	1	971	971	0.0153	0.9043192
N_Gemme:Defogl:Dirad	1	90948	90948	1.4314	0.2621028
Residuals	9	571844	63538		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: n.s
 IBIMET_3: 0.761

DensMostoB

Analysis of Variance Table

Response: Do_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	2.4303e-05	2.4303e-05	2.7781	0.1299
IBIMET_2	1	2.7070e-06	2.7068e-06	0.3094	0.5916
IBIMET_3	1	4.5080e-06	4.5084e-06	0.5154	0.4910
N_Gemme	1	1.0550e-06	1.0548e-06	0.1206	0.7364

Defogl	1	8.1000e-08	8.1100e-08	0.0093	0.9254
Dirad	1	1.6790e-06	1.6793e-06	0.1920	0.6716
Vigoria	1	8.4360e-06	8.4362e-06	0.9644	0.3518
N_Gemme:Defogl	1	3.9560e-06	3.9559e-06	0.4522	0.5182
N_Gemme:Dirad	1	7.3080e-06	7.3084e-06	0.8354	0.3845
Defogl:Dirad	1	1.4150e-06	1.4154e-06	0.1618	0.6969
N_Gemme:Vigoria	1	5.7900e-07	5.7930e-07	0.0662	0.8027
Defogl:Vigoria	1	9.9100e-07	9.9100e-07	0.1133	0.7441
Dirad:Vigoria	1	7.4100e-06	7.4104e-06	0.8471	0.3814
N_Gemme:Defogl:Dirad	1	1.1890e-06	1.1888e-06	0.1359	0.7209
Residuals	9	7.8732e-05	8.7480e-06		

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: n.s

BrixB

Analysis of Variance Table

Response: Do_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	1.2499	1.24986	2.8298	0.1268
IBIMET_2	1	0.1301	0.13013	0.2946	0.6004
IBIMET_3	1	0.2454	0.24544	0.5557	0.4750
N_Gemme	1	0.0736	0.07364	0.1667	0.6926
Defogl	1	0.0003	0.00031	0.0007	0.9795
Dirad	1	0.0652	0.06524	0.1477	0.7096
Vigoria	1	0.3665	0.36653	0.8299	0.3861
N_Gemme:Defogl	1	0.2089	0.20891	0.4730	0.5089
N_Gemme:Dirad	1	0.3486	0.34863	0.7893	0.3974
Defogl:Dirad	1	0.1000	0.10000	0.2264	0.6455
N_Gemme:Vigoria	1	0.0420	0.04202	0.0951	0.7648
Defogl:Vigoria	1	0.0338	0.03377	0.0765	0.7884
Dirad:Vigoria	1	0.3519	0.35186	0.7967	0.3953
N_Gemme:Defogl:Dirad	1	0.0901	0.09008	0.2040	0.6622
Residuals	9	3.9751	0.44167		

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: n.s

PhB

Analysis of Variance Table

Response: Do_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.090220	0.090220	23.3689	0.0009285 ***
IBIMET_2	1	0.005195	0.005195	1.3456	0.2758890
IBIMET_3	1	0.000005	0.000005	0.0014	0.9714644
N_Gemme	1	0.010807	0.010807	2.7992	0.1286417
Defogl	1	0.002026	0.002026	0.5247	0.4872398
Dirad	1	0.005164	0.005164	1.3375	0.2772499
Vigoria	1	0.001109	0.001109	0.2873	0.6049627
N_Gemme:Defogl	1	0.000628	0.000628	0.1627	0.6960836
N_Gemme:Dirad	1	0.004461	0.004461	1.1555	0.3103684
Defogl:Dirad	1	0.002457	0.002457	0.6365	0.4455368
N_Gemme:Vigoria	1	0.000007	0.000007	0.0019	0.9658450

Defogl:Vigoria	1	0.000357	0.000357	0.0926	0.7678562
Dirad:Vigoria	1	0.003952	0.003952	1.0236	0.3380875
N_Gemme:Defogl:Dirad	1	0.004410	0.004410	1.1424	0.3129756
Residuals	9	0.034746	0.003861		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.85
 IBIMET_2: n.s
 IBIMET_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.37572	0.37572	16.6103	0.0027763	**
IBIMET_2	1	0.60121	0.60121	26.5794	0.0005987	***
IBIMET_3	1	0.04717	0.04717	2.0854	0.1826031	
N_Gemme	1	0.07087	0.07087	3.1329	0.1105042	
Defogl	1	0.09915	0.09915	4.3832	0.0657965	.
Dirad	1	0.17859	0.17859	7.8955	0.0203806	*
Vigoria	1	0.00503	0.00503	0.2224	0.6484323	
N_Gemme:Defogl	1	0.03424	0.03424	1.5137	0.2497595	
N_Gemme:Dirad	1	0.00730	0.00730	0.3226	0.5839487	
Defogl:Dirad	1	0.00199	0.00199	0.0878	0.7737680	
N_Gemme:Vigoria	1	0.00140	0.00140	0.0621	0.8088529	
Defogl:Vigoria	1	0.00016	0.00016	0.0069	0.9356701	
Dirad:Vigoria	1	0.00070	0.00070	0.0307	0.8647198	
N_Gemme:Defogl:Dirad	1	0.03624	0.03624	1.6023	0.2373567	
Residuals	9	0.20358	0.02262			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.805
 IBIMET_2: -0.864
 IBIMET_3: n.s

AcTartB

Analysis of Variance Table

Response: Do_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IBIMET_1	1	0.65707	0.65707	16.1538	0.003022	**
IBIMET_2	1	0.58055	0.58055	14.2725	0.004363	**
IBIMET_3	1	0.00004	0.00004	0.0010	0.975063	
N_Gemme	1	0.02756	0.02756	0.6776	0.431692	
Defogl	1	0.01295	0.01295	0.3185	0.586317	
Dirad	1	0.08047	0.08047	1.9782	0.193156	
Vigoria	1	0.01930	0.01930	0.4745	0.508272	
N_Gemme:Defogl	1	0.00842	0.00842	0.2070	0.659886	
N_Gemme:Dirad	1	0.00228	0.00228	0.0561	0.818028	
Defogl:Dirad	1	0.04207	0.04207	1.0343	0.335711	
N_Gemme:Vigoria	1	0.00236	0.00236	0.0581	0.814933	
Defogl:Vigoria	1	0.02479	0.02479	0.6096	0.454999	
Dirad:Vigoria	1	0.04160	0.04160	1.0228	0.338265	
N_Gemme:Defogl:Dirad	1	0.00113	0.00113	0.0279	0.871136	

Residuals 9 0.36608 0.04068

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: -0.801
IBIMET_2: -0.783
IBIMET_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.06084	0.06084	1.1079	0.319973
IBIMET_2	1	0.58103	0.58103	10.5809	0.009952 **
IBIMET_3	1	0.07510	0.07510	1.3677	0.272252
N_Gemme	1	0.01112	0.01112	0.2025	0.663364
Defogl	1	0.01062	0.01062	0.1934	0.670468
Dirad	1	0.01287	0.01287	0.2344	0.639852
Vigoria	1	0.11871	0.11871	2.1617	0.175560
N_Gemme:Defogl	1	0.10224	0.10224	1.8618	0.205555
N_Gemme:Dirad	1	0.04707	0.04707	0.8572	0.378686
Defogl:Dirad	1	0.06592	0.06592	1.2005	0.301676
N_Gemme:Vigoria	1	0.01439	0.01439	0.2621	0.621006
Defogl:Vigoria	1	0.00148	0.00148	0.0269	0.873350
Dirad:Vigoria	1	0.03423	0.03423	0.6234	0.450099
N_Gemme:Defogl:Dirad	1	0.03039	0.03039	0.5534	0.475902
Residuals	9	0.49421	0.05491		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: -0.735
IBIMET_3: n.s

KB

Analysis of Variance Table

Response: Do_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	0.62255	0.62255	60.5459	2.761e-05 ***
IBIMET_2	1	0.01720	0.01720	1.6729	0.22807
IBIMET_3	1	0.01856	0.01856	1.8048	0.21202
N_Gemme	1	0.01339	0.01339	1.3024	0.28324
Defogl	1	0.00375	0.00375	0.3644	0.56096
Dirad	1	0.03199	0.03199	3.1115	0.11157
Vigoria	1	0.00222	0.00222	0.2155	0.65351
N_Gemme:Defogl	1	0.00473	0.00473	0.4602	0.51457
N_Gemme:Dirad	1	0.03067	0.03067	2.9823	0.11825
Defogl:Dirad	1	0.03837	0.03837	3.7321	0.08542 .
N_Gemme:Vigoria	1	0.00904	0.00904	0.8794	0.37284
Defogl:Vigoria	1	0.00007	0.00007	0.0066	0.93724
Dirad:Vigoria	1	0.01572	0.01572	1.5293	0.24752
N_Gemme:Defogl:Dirad	1	0.00166	0.00166	0.1617	0.69697
Residuals	9	0.09254	0.01028		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: 0.933
 IBIMET_2: n.s
 IBIMET_3: n.s

ApaB

Analysis of Variance Table

Response: Do_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	86.97	86.97	0.5179	0.49001
IBIMET_2	1	126.84	126.84	0.7553	0.40738
IBIMET_3	1	122.55	122.55	0.7297	0.41513
N_Gemme	1	445.35	445.35	2.6519	0.13787
Defogl	1	1556.87	1556.87	9.2706	0.01391 *
Dirad	1	200.13	200.13	1.1917	0.30334
Vigoria	1	114.97	114.97	0.6846	0.42940
N_Gemme:Defogl	1	526.37	526.37	3.1344	0.11044
N_Gemme:Dirad	1	763.44	763.44	4.5460	0.06181 .
Defogl:Dirad	1	33.05	33.05	0.1968	0.66777
N_Gemme:Vigoria	1	19.67	19.67	0.1171	0.74004
Defogl:Vigoria	1	35.52	35.52	0.2115	0.65650
Dirad:Vigoria	1	246.79	246.79	1.4696	0.25628
N_Gemme:Defogl:Dirad	1	166.76	166.76	0.9930	0.34505
Residuals	9	1511.43	167.94		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: n.s
 IBIMET_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	420	420	0.2587	0.623229
IBIMET_2	1	37516	37516	23.1235	0.000962 ***
IBIMET_3	1	427	427	0.2633	0.620210
N_Gemme	1	669	669	0.4125	0.536720
Defogl	1	18996	18996	11.7084	0.007607 **
Dirad	1	9177	9177	5.6564	0.041342 *
Vigoria	1	21520	21520	13.2643	0.005384 **
N_Gemme:Defogl	1	2642	2642	1.6281	0.233905
N_Gemme:Dirad	1	8331	8331	5.1349	0.049685 *
Defogl:Dirad	1	1001	1001	0.6168	0.452432
N_Gemme:Vigoria	1	5216	5216	3.2148	0.106567
Defogl:Vigoria	1	1475	1475	0.9089	0.365315
Dirad:Vigoria	1	579	579	0.3569	0.564960
N_Gemme:Defogl:Dirad	1	4880	4880	3.0081	0.116875
Residuals	9	14602	1622		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
 IBIMET_2: 0.848

IBIMET_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Do_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	2167.8	2167.8	2.0419	0.18679
IBIMET_2	1	3998.5	3998.5	3.7662	0.08422 .
IBIMET_3	1	938.0	938.0	0.8835	0.37178
N_Gemme	1	3411.2	3411.2	3.2131	0.10665
Defogl	1	20371.4	20371.4	19.1883	0.00177 **
Dirad	1	6607.1	6607.1	6.2234	0.03416 *
Vigoria	1	1224.0	1224.0	1.1529	0.31089
N_Gemme:Defogl	1	22.2	22.2	0.0209	0.88833
N_Gemme:Dirad	1	7570.1	7570.1	7.1304	0.02561 *
Defogl:Dirad	1	2160.3	2160.3	2.0349	0.18748
N_Gemme:Vigoria	1	2405.7	2405.7	2.2660	0.16650
Defogl:Vigoria	1	1095.9	1095.9	1.0322	0.33617
Dirad:Vigoria	1	755.2	755.2	0.7113	0.42085
N_Gemme:Defogl:Dirad	1	1006.3	1006.3	0.9478	0.35571
Residuals	9	9554.9	1061.7		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s

IBIMET_2: n.s

IBIMET_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Do_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	373	373	0.0474	0.832449
IBIMET_2	1	157407	157407	20.0209	0.001545 **
IBIMET_3	1	27788	27788	3.5344	0.092806 .
N_Gemme	1	19119	19119	2.4317	0.153333
Defogl	1	11240	11240	1.4297	0.262368
Dirad	1	7029	7029	0.8941	0.369067
Vigoria	1	62638	62638	7.9670	0.019963 *
N_Gemme:Defogl	1	29408	29408	3.7405	0.085120 .
N_Gemme:Dirad	1	71954	71954	9.1520	0.014357 *
Defogl:Dirad	1	4856	4856	0.6177	0.452109
N_Gemme:Vigoria	1	6914	6914	0.8795	0.372834
Defogl:Vigoria	1	4749	4749	0.6041	0.456972
Dirad:Vigoria	1	8826	8826	1.1226	0.316961
N_Gemme:Defogl:Dirad	1	1495	1495	0.1902	0.673053
Residuals	9	70759	7862		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s

IBIMET_2: 0.831

IBIMET_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Do_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IBIMET_1	1	32047	32047	3.4157	0.09764 .
IBIMET_2	1	30395	30395	3.2396	0.10541
IBIMET_3	1	55568	55568	5.9227	0.03776 *
N_Gemme	1	2373	2373	0.2529	0.62713
Defogl	1	2294	2294	0.2445	0.63279
Dirad	1	5	5	0.0005	0.98193
Vigoria	1	1008	1008	0.1075	0.75054
N_Gemme:Defogl	1	11796	11796	1.2573	0.29119
N_Gemme:Dirad	1	75785	75785	8.0775	0.01934 *
Defogl:Dirad	1	7699	7699	0.8206	0.38862
N_Gemme:Vigoria	1	7	7	0.0007	0.97890
Defogl:Vigoria	1	2227	2227	0.2374	0.63774
Dirad:Vigoria	1	82	82	0.0087	0.92754
N_Gemme:Defogl:Dirad	1	3963	3963	0.4224	0.53198
Residuals	9	84441	9382		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IBIMET_1: n.s
IBIMET_2: n.s
IBIMET_3: -0.63

DONNA OLIMPIA 2009M DIPROVE

PvinacciaA

Analysis of Variance Table

Response: Do_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	165755	165755	5.4566	0.0443148 *
DIPROVE_2	1	1273	1273	0.0419	0.8423432
DIPROVE_3	1	6271	6271	0.2064	0.6603298
N_Gemme	1	942703	942703	31.0333	0.0003470 ***
Defogl	1	381126	381126	12.5465	0.0062933 **
Dirad	1	1273547	1273547	41.9245	0.0001148 ***
Vigoria	1	44772	44772	1.4739	0.2556320
N_Gemme:Defogl	1	13072	13072	0.4303	0.5282349
N_Gemme:Dirad	1	85511	85511	2.8150	0.1276985
Defogl:Dirad	1	44196	44196	1.4549	0.2584935
N_Gemme:Vigoria	1	3958	3958	0.1303	0.7264601
Defogl:Vigoria	1	5453	5453	0.1795	0.6817356
Dirad:Vigoria	1	20959	20959	0.6900	0.4276613
N_Gemme:Defogl:Dirad	1	32563	32563	1.0720	0.3275224
Residuals	9	273394	30377		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.614
DIPROVE_2: n.s
DIPROVE_3: n.s

DensMostoA

Analysis of Variance Table

Response: Do_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	2.5096e-05	2.5096e-05	16.4429	0.002863	**
DIPROVE_2	1	2.2000e-09	2.2000e-09	0.0014	0.970512	
DIPROVE_3	1	2.7050e-07	2.7050e-07	0.1772	0.683653	
N_Gemme	1	2.3059e-05	2.3059e-05	15.1086	0.003692	**
Defogl	1	3.1764e-06	3.1764e-06	2.0812	0.183007	
Dirad	1	1.0188e-05	1.0188e-05	6.6754	0.029518	*
Viguria	1	1.6200e-08	1.6200e-08	0.0106	0.920083	
N_Gemme:Defogl	1	8.9500e-08	8.9500e-08	0.0586	0.814084	
N_Gemme:Dirad	1	1.1480e-07	1.1480e-07	0.0752	0.790103	
Defogl:Dirad	1	2.7828e-06	2.7828e-06	1.8233	0.209892	
N_Gemme:Viguria	1	1.6300e-08	1.6300e-08	0.0107	0.919931	
Defogl:Viguria	1	2.8810e-07	2.8810e-07	0.1888	0.674170	
Dirad:Viguria	1	1.7447e-06	1.7447e-06	1.1432	0.312819	
N_Gemme:Defogl:Dirad	1	6.8120e-07	6.8120e-07	0.4463	0.520843	
Residuals	9	1.3736e-05	1.5262e-06			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.804
 DIPROVE_2: n.s
 DIPROVE_3: n.s

BrixA

Analysis of Variance Table

Response: Do_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	1.09200	1.09200	14.5393	0.004134	**
DIPROVE_2	1	0.00329	0.00329	0.0439	0.838759	
DIPROVE_3	1	0.04633	0.04633	0.6168	0.452419	
N_Gemme	1	1.14795	1.14795	15.2841	0.003568	**
Defogl	1	0.16531	0.16531	2.2010	0.172071	
Dirad	1	0.52947	0.52947	7.0495	0.026256	*
Viguria	1	0.00001	0.00001	0.0002	0.989950	
N_Gemme:Defogl	1	0.00741	0.00741	0.0987	0.760571	
N_Gemme:Dirad	1	0.00086	0.00086	0.0114	0.917319	
Defogl:Dirad	1	0.16307	0.16307	2.1711	0.174717	
N_Gemme:Viguria	1	0.00214	0.00214	0.0285	0.869562	
Defogl:Viguria	1	0.02672	0.02672	0.3558	0.565563	
Dirad:Viguria	1	0.06897	0.06897	0.9183	0.362954	
N_Gemme:Defogl:Dirad	1	0.01713	0.01713	0.2281	0.644330	
Residuals	9	0.67596	0.07511			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.786
 DIPROVE_2: n.s
 DIPROVE_3: n.s

PhA

Analysis of Variance Table

Response: Do_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.0184756	0.0184756	15.0835	0.003710	**
DIPROVE_2	1	0.0123087	0.0123087	10.0489	0.011366	*
DIPROVE_3	1	0.0198090	0.0198090	16.1722	0.003011	**
N_Gemme	1	0.0122256	0.0122256	9.9810	0.011564	*
Defogl	1	0.0000472	0.0000472	0.0386	0.848664	
Dirad	1	0.0082523	0.0082523	6.7373	0.028945	*
Vigoria	1	0.0172334	0.0172334	14.0695	0.004548	**
N_Gemme:Defogl	1	0.0006124	0.0006124	0.5000	0.497399	
N_Gemme:Dirad	1	0.0006984	0.0006984	0.5702	0.469484	
Defogl:Dirad	1	0.0010958	0.0010958	0.8946	0.368939	
N_Gemme:Vigoria	1	0.0011102	0.0011102	0.9064	0.365950	
Defogl:Vigoria	1	0.0023809	0.0023809	1.9438	0.196719	
Dirad:Vigoria	1	0.0061518	0.0061518	5.0223	0.051754	.
N_Gemme:Defogl:Dirad	1	0.0017081	0.0017081	1.3945	0.267917	
Residuals	9	0.0110239	0.0012249			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.791
DIPROVE_2: -0.726
DIPROVE_3: -0.802

AcTitola

Analysis of Variance Table

Response: Do_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.01044	0.01044	0.7148	0.419750	
DIPROVE_2	1	0.32831	0.32831	22.4752	0.001058	**
DIPROVE_3	1	0.00013	0.00013	0.0087	0.927617	
N_Gemme	1	0.12425	0.12425	8.5060	0.017129	*
Defogl	1	0.05078	0.05078	3.4762	0.095136	.
Dirad	1	0.19167	0.19167	13.1213	0.005552	**
Vigoria	1	0.06154	0.06154	4.2131	0.070319	.
N_Gemme:Defogl	1	0.06438	0.06438	4.4074	0.065184	.
N_Gemme:Dirad	1	0.00758	0.00758	0.5186	0.489730	
Defogl:Dirad	1	0.00171	0.00171	0.1171	0.740030	
N_Gemme:Vigoria	1	0.01369	0.01369	0.9370	0.358337	
Defogl:Vigoria	1	0.00008	0.00008	0.0057	0.941386	
Dirad:Vigoria	1	0.00278	0.00278	0.1905	0.672787	
N_Gemme:Defogl:Dirad	1	0.00776	0.00776	0.5310	0.484701	
Residuals	9	0.13147	0.01461			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.845
DIPROVE_3: n.s

AcTarta

Analysis of Variance Table

Response: Do_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.184213	0.184213	42.8365	0.0001058	***
DIPROVE_2	1	0.238934	0.238934	55.5613	3.876e-05	***

DIPROVE_3	1	0.035932	0.035932	8.3557	0.0178664	*
N_Gemme	1	0.115262	0.115262	26.8028	0.0005816	***
Defogl	1	0.029415	0.029415	6.8400	0.0280241	*
Dirad	1	0.090849	0.090849	21.1259	0.0012974	**
Vigoria	1	0.034291	0.034291	7.9740	0.0199223	*
N_Gemme:Defogl	1	0.000218	0.000218	0.0507	0.8269532	
N_Gemme:Dirad	1	0.005531	0.005531	1.2862	0.2860595	
Defogl:Dirad	1	0.006357	0.006357	1.4782	0.2549783	
N_Gemme:Vigoria	1	0.005452	0.005452	1.2677	0.2893257	
Defogl:Vigoria	1	0.000108	0.000108	0.0252	0.8774356	
Dirad:Vigoria	1	0.009423	0.009423	2.1912	0.1729315	
N_Gemme:Defogl:Dirad	1	0.043011	0.043011	10.0018	0.0115028	*
Residuals	9	0.038703	0.004300			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.909

DIPROVE_2: 0.928

DIPROVE_3: 0.694

AcMalicoA

Analysis of Variance Table

Response: Do_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.59046	0.59046	14.1991	0.004429	**
DIPROVE_2	1	0.00058	0.00058	0.0138	0.908920	
DIPROVE_3	1	0.03640	0.03640	0.8753	0.373920	
N_Gemme	1	0.03661	0.03661	0.8803	0.372606	
Defogl	1	0.00482	0.00482	0.1160	0.741237	
Dirad	1	0.00605	0.00605	0.1456	0.711641	
Vigoria	1	0.10917	0.10917	2.6252	0.139633	
N_Gemme:Defogl	1	0.03102	0.03102	0.7461	0.410153	
N_Gemme:Dirad	1	0.02129	0.02129	0.5119	0.492462	
Defogl:Dirad	1	0.05054	0.05054	1.2154	0.298865	
N_Gemme:Vigoria	1	0.00915	0.00915	0.2201	0.650112	
Defogl:Vigoria	1	0.01980	0.01980	0.4761	0.507602	
Dirad:Vigoria	1	0.00693	0.00693	0.1666	0.692715	
N_Gemme:Defogl:Dirad	1	0.11829	0.11829	2.8445	0.125962	
Residuals	9	0.37426	0.04158			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.782

DIPROVE_2: n.s

DIPROVE_3: n.s

KA

Analysis of Variance Table

Response: Do_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.216857	0.216857	16.7540	0.002704	**
DIPROVE_2	1	0.058548	0.058548	4.5233	0.062344	.
DIPROVE_3	1	0.038798	0.038798	2.9975	0.117440	
N_Gemme	1	0.065393	0.065393	5.0522	0.051195	.
Defogl	1	0.007906	0.007906	0.6108	0.454548	

Dirad	1	0.047438	0.047438	3.6650	0.087838	.
Vigoria	1	0.132543	0.132543	10.2401	0.010831	*
N_Gemme:Defogl	1	0.000473	0.000473	0.0366	0.852603	
N_Gemme:Dirad	1	0.017974	0.017974	1.3886	0.268856	
Defogl:Dirad	1	0.006331	0.006331	0.4892	0.501976	
N_Gemme:Vigoria	1	0.000008	0.000008	0.0006	0.980972	
Defogl:Vigoria	1	0.027117	0.027117	2.0950	0.181702	
Dirad:Vigoria	1	0.000156	0.000156	0.0121	0.914927	
N_Gemme:Defogl:Dirad	1	0.001388	0.001388	0.1073	0.750786	
Residuals	9	0.116492	0.012944			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.807
 DIPROVE_2: n.s
 DIPROVE_3: n.s

ApaA

Analysis of Variance Table

Response: Do_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	20.54	20.54	0.2865	0.605466	
DIPROVE_2	1	1600.35	1600.35	22.3232	0.001082	**
DIPROVE_3	1	725.25	725.25	10.1166	0.011173	*
N_Gemme	1	21.07	21.07	0.2939	0.600925	
Defogl	1	1420.93	1420.93	19.8206	0.001595	**
Dirad	1	258.26	258.26	3.6025	0.090173	.
Vigoria	1	90.24	90.24	1.2588	0.290922	
N_Gemme:Defogl	1	44.46	44.46	0.6202	0.451219	
N_Gemme:Dirad	1	0.00	0.00	0.0000	0.999222	
Defogl:Dirad	1	663.27	663.27	9.2519	0.013979	*
N_Gemme:Vigoria	1	36.43	36.43	0.5082	0.493985	
Defogl:Vigoria	1	220.58	220.58	3.0769	0.113310	
Dirad:Vigoria	1	577.29	577.29	8.0527	0.019475	*
N_Gemme:Defogl:Dirad	1	148.85	148.85	2.0764	0.183467	
Residuals	9	645.21	71.69			

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: 0.844
 DIPROVE_3: -0.727

Antoctot1A

Analysis of Variance Table

Response: Do_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	51362	51362	15.2567	0.003587	**
DIPROVE_2	1	666	666	0.1978	0.667032	
DIPROVE_3	1	6895	6895	2.0481	0.186187	
N_Gemme	1	7120	7120	2.1148	0.179847	
Defogl	1	7288	7288	2.1650	0.175269	
Dirad	1	6980	6980	2.0733	0.183755	
Vigoria	1	17590	17590	5.2249	0.048103	*
N_Gemme:Defogl	1	1	1	0.0002	0.989896	

N_Gemme:Dirad	1	1229	1229	0.3651	0.560599
Defogl:Dirad	1	1695	1695	0.5036	0.495887
N_Gemme:Vigoria	1	1846	1846	0.5482	0.477918
Defogl:Vigoria	1	62	62	0.0185	0.894857
Dirad:Vigoria	1	1	1	0.0003	0.985565
N_Gemme:Defogl:Dirad	1	8381	8381	2.4896	0.149059
Residuals	9	30299	3367		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.793
DIPROVE_2: n.s
DIPROVE_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	32317	32317	12.8274	0.005917	**
DIPROVE_2	1	619	619	0.2455	0.632121	
DIPROVE_3	1	41	41	0.0162	0.901378	
N_Gemme	1	8840	8840	3.5087	0.093826	.
Defogl	1	7710	7710	3.0603	0.114158	
Dirad	1	20400	20400	8.0970	0.019229	*
Vigoria	1	1404	1404	0.5572	0.474455	
N_Gemme:Defogl	1	121	121	0.0481	0.831214	
N_Gemme:Dirad	1	2323	2323	0.9220	0.362043	
Defogl:Dirad	1	587	587	0.2332	0.640712	
N_Gemme:Vigoria	1	66	66	0.0264	0.874622	
Defogl:Vigoria	1	3295	3295	1.3080	0.282270	
Dirad:Vigoria	1	60	60	0.0238	0.880792	
N_Gemme:Defogl:Dirad	1	1559	1559	0.6189	0.451685	
Residuals	9	22675	2519			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.767
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	112735	112735	10.4761	0.01021	*
DIPROVE_2	1	2847	2847	0.2645	0.61941	
DIPROVE_3	1	23365	23365	2.1712	0.17471	
N_Gemme	1	4	4	0.0003	0.98582	
Defogl	1	72	72	0.0067	0.93645	
Dirad	1	241	241	0.0224	0.88433	
Vigoria	1	40485	40485	3.7621	0.08436	.
N_Gemme:Defogl	1	985	985	0.0916	0.76906	
N_Gemme:Dirad	1	4493	4493	0.4175	0.53430	
Defogl:Dirad	1	10861	10861	1.0093	0.34131	
N_Gemme:Vigoria	1	3763	3763	0.3497	0.56883	

Defogl:Vigoria	1	4625	4625	0.4298	0.52850
Dirad:Vigoria	1	394	394	0.0366	0.85252
N_Gemme:Defogl:Dirad	1	9288	9288	0.8631	0.37712
Residuals	9	96850	10761		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.733

DIPROVE_2: n.s

DIPROVE_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	46399	46399	8.8181	0.01571	*
DIPROVE_2	1	3643	3643	0.6923	0.42690	
DIPROVE_3	1	22	22	0.0042	0.94958	
N_Gemme	1	583	583	0.1108	0.74684	
Defogl	1	641	641	0.1217	0.73518	
Dirad	1	4310	4310	0.8192	0.38900	
Vigoria	1	66	66	0.0126	0.91325	
N_Gemme:Defogl	1	3394	3394	0.6450	0.44259	
N_Gemme:Dirad	1	6874	6874	1.3065	0.28253	
Defogl:Dirad	1	7189	7189	1.3662	0.27249	
N_Gemme:Vigoria	1	95	95	0.0181	0.89588	
Defogl:Vigoria	1	2266	2266	0.4306	0.52813	
Dirad:Vigoria	1	531	531	0.1010	0.75791	
N_Gemme:Defogl:Dirad	1	1080	1080	0.2052	0.66126	
Residuals	9	47356	5262			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.703

DIPROVE_2: n.s

DIPROVE_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	50054	50054	0.9019	0.3670726	
DIPROVE_2	1	227365	227365	4.0969	0.0736363	.
DIPROVE_3	1	96	96	0.0017	0.9677729	
N_Gemme	1	1449531	1449531	26.1189	0.0006359	***
Defogl	1	237830	237830	4.2854	0.0683499	.
Dirad	1	2120556	2120556	38.2100	0.0001625	***
Vigoria	1	33751	33751	0.6082	0.4554984	
N_Gemme:Defogl	1	8140	8140	0.1467	0.7106303	
N_Gemme:Dirad	1	664991	664991	11.9824	0.0071432	**
Defogl:Dirad	1	62304	62304	1.1226	0.3169556	
N_Gemme:Vigoria	1	9258	9258	0.1668	0.6925073	
Defogl:Vigoria	1	159	159	0.0029	0.9584790	
Dirad:Vigoria	1	4851	4851	0.0874	0.7741967	
N_Gemme:Defogl:Dirad	1	79409	79409	1.4309	0.2621853	

Residuals 9 499476 55497

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

DensMostoB

Analysis of Variance Table

Response: Do_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	1.6400e-06	1.6402e-06	0.2442	0.6330
DIPROVE_2	1	3.0820e-06	3.0820e-06	0.4589	0.5152
DIPROVE_3	1	6.2400e-07	6.2370e-07	0.0929	0.7675
N_Gemme	1	4.7730e-06	4.7735e-06	0.7108	0.4210
Defogl	1	1.8000e-08	1.8200e-08	0.0027	0.9597
Dirad	1	3.1760e-06	3.1756e-06	0.4729	0.5090
Vigoria	1	2.0292e-05	2.0292e-05	3.0218	0.1162
N_Gemme:Defogl	1	1.0277e-05	1.0277e-05	1.5304	0.2474
N_Gemme:Dirad	1	1.4202e-05	1.4201e-05	2.1148	0.1799
Defogl:Dirad	1	3.3200e-06	3.3202e-06	0.4944	0.4997
N_Gemme:Vigoria	1	2.8480e-06	2.8481e-06	0.4241	0.5312
Defogl:Vigoria	1	1.8253e-05	1.8253e-05	2.7181	0.1336
Dirad:Vigoria	1	1.1800e-07	1.1850e-07	0.0176	0.8973
N_Gemme:Defogl:Dirad	1	1.2880e-06	1.2879e-06	0.1918	0.6718
Residuals	9	6.0439e-05	6.7154e-06		

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

BrixB

Analysis of Variance Table

Response: Do_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.10836	0.10836	0.3125	0.5898
DIPROVE_2	1	0.15094	0.15094	0.4352	0.5260
DIPROVE_3	1	0.01340	0.01340	0.0386	0.8485
N_Gemme	1	0.29283	0.29283	0.8444	0.3821
Defogl	1	0.00032	0.00032	0.0009	0.9764
Dirad	1	0.12859	0.12859	0.3708	0.5576
Vigoria	1	0.95411	0.95411	2.7512	0.1316
N_Gemme:Defogl	1	0.51387	0.51387	1.4818	0.2545
N_Gemme:Dirad	1	0.68465	0.68465	1.9742	0.1936
Defogl:Dirad	1	0.20009	0.20009	0.5770	0.4669
N_Gemme:Vigoria	1	0.16619	0.16619	0.4792	0.5062
Defogl:Vigoria	1	0.84771	0.84771	2.4444	0.1524
Dirad:Vigoria	1	0.00481	0.00481	0.0139	0.9088
N_Gemme:Defogl:Dirad	1	0.09444	0.09444	0.2723	0.6144
Residuals	9	3.12119	0.34680		

DIPROVE_1: n.s

DIPROVE_2: n.s

DIPROVE_3: n.s

PhB

Analysis of Variance Table

Response: Do_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.005656	0.005656	1.5933	0.238584
DIPROVE_2	1	0.024348	0.024348	6.8589	0.027859 *
DIPROVE_3	1	0.006205	0.006205	1.7481	0.218730
N_Gemme	1	0.013487	0.013487	3.7993	0.083075 .
Defogl	1	0.004418	0.004418	1.2445	0.293509
Dirad	1	0.003445	0.003445	0.9704	0.350313
Vigoria	1	0.047267	0.047267	13.3154	0.005326 **
N_Gemme:Defogl	1	0.000773	0.000773	0.2176	0.651938
N_Gemme:Dirad	1	0.000192	0.000192	0.0540	0.821522
Defogl:Dirad	1	0.002062	0.002062	0.5810	0.465433
N_Gemme:Vigoria	1	0.001647	0.001647	0.4640	0.512921
Defogl:Vigoria	1	0.017055	0.017055	4.8046	0.056073 .
Dirad:Vigoria	1	0.000197	0.000197	0.0554	0.819206
N_Gemme:Defogl:Dirad	1	0.006847	0.006847	1.9290	0.198274
Residuals	9	0.031948	0.003550		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.658
DIPROVE_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.07138	0.07138	4.8582	0.0549682 .
DIPROVE_2	1	0.02832	0.02832	1.9276	0.1984159
DIPROVE_3	1	0.15698	0.15698	10.6851	0.0097013 **
N_Gemme	1	0.09381	0.09381	6.3852	0.0324001 *
Defogl	1	0.23860	0.23860	16.2404	0.0029732 **
Dirad	1	0.05929	0.05929	4.0356	0.0754657 .
Vigoria	1	0.41561	0.41561	28.2891	0.0004818 ***
N_Gemme:Defogl	1	0.00080	0.00080	0.0543	0.8210093
N_Gemme:Dirad	1	0.03125	0.03125	2.1268	0.1787389
Defogl:Dirad	1	0.00980	0.00980	0.6671	0.4351528
N_Gemme:Vigoria	1	0.02090	0.02090	1.4227	0.2634527
Defogl:Vigoria	1	0.23200	0.23200	15.7911	0.0032360 **
Dirad:Vigoria	1	0.08941	0.08941	6.0856	0.0357520 *
N_Gemme:Defogl:Dirad	1	0.08298	0.08298	5.6478	0.0414644 *
Residuals	9	0.13222	0.01469		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: 0.737

AcTartB

Analysis of Variance Table

Response: Do_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.00277	0.00277	0.0506	0.82697
DIPROVE_2	1	0.00509	0.00509	0.0930	0.76739
DIPROVE_3	1	0.26009	0.26009	4.7539	0.05714 .
N_Gemme	1	0.13877	0.13877	2.5365	0.14571
Defogl	1	0.05609	0.05609	1.0252	0.33773
Dirad	1	0.01319	0.01319	0.2410	0.63523
Vigoria	1	0.28693	0.28693	5.2444	0.04777 *
N_Gemme:Defogl	1	0.00021	0.00021	0.0039	0.95152
N_Gemme:Dirad	1	0.00176	0.00176	0.0321	0.86176
Defogl:Dirad	1	0.00440	0.00440	0.0805	0.78303
N_Gemme:Vigoria	1	0.00183	0.00183	0.0334	0.85908
Defogl:Vigoria	1	0.34313	0.34313	6.2716	0.03362 *
Dirad:Vigoria	1	0.24932	0.24932	4.5569	0.06155 .
N_Gemme:Defogl:Dirad	1	0.01072	0.01072	0.1959	0.66850
Residuals	9	0.49240	0.05471		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
 DIPROVE_2: n.s
 DIPROVE_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	0.41811	0.41811	7.6719	0.02176 *
DIPROVE_2	1	0.20405	0.20405	3.7442	0.08499 .
DIPROVE_3	1	0.14504	0.14504	2.6613	0.13725
N_Gemme	1	0.00726	0.00726	0.1333	0.72347
Defogl	1	0.03937	0.03937	0.7225	0.41737
Dirad	1	0.00197	0.00197	0.0361	0.85361
Vigoria	1	0.00097	0.00097	0.0178	0.89686
N_Gemme:Defogl	1	0.02334	0.02334	0.4283	0.52918
N_Gemme:Dirad	1	0.06640	0.06640	1.2184	0.29832
Defogl:Dirad	1	0.00006	0.00006	0.0011	0.97407
N_Gemme:Vigoria	1	0.04511	0.04511	0.8277	0.38666
Defogl:Vigoria	1	0.03105	0.03105	0.5698	0.46963
Dirad:Vigoria	1	0.10762	0.10762	1.9748	0.19351
N_Gemme:Defogl:Dirad	1	0.07937	0.07937	1.4564	0.25827
Residuals	9	0.49049	0.05450		

 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: 0.678
 DIPROVE_2: n.s
 DIPROVE_3: n.s

KB

Analysis of Variance Table

Response: Do_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	0.083337	0.083337	4.5534	0.06163	.
DIPROVE_2	1	0.127113	0.127113	6.9452	0.02712	*
DIPROVE_3	1	0.094612	0.094612	5.1694	0.04907	*
N_Gemme	1	0.041947	0.041947	2.2919	0.16435	
Defogl	1	0.030006	0.030006	1.6395	0.23241	
Dirad	1	0.028150	0.028150	1.5381	0.24626	
Vigoria	1	0.156415	0.156415	8.5462	0.01694	*
N_Gemme:Defogl	1	0.003747	0.003747	0.2047	0.66164	
N_Gemme:Dirad	1	0.012589	0.012589	0.6878	0.42835	
Defogl:Dirad	1	0.037811	0.037811	2.0659	0.18447	
N_Gemme:Vigoria	1	0.001342	0.001342	0.0733	0.79266	
Defogl:Vigoria	1	0.115180	0.115180	6.2932	0.03339	*
Dirad:Vigoria	1	0.000859	0.000859	0.0470	0.83328	
N_Gemme:Defogl:Dirad	1	0.004634	0.004634	0.2532	0.62691	
Residuals	9	0.164721	0.018302			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: -0.66
DIPROVE_3: -0.604

ApaB

Analysis of Variance Table

Response: Do_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	22.65	22.65	0.1653	0.69378	
DIPROVE_2	1	1033.64	1033.64	7.5446	0.02260	*
DIPROVE_3	1	459.64	459.64	3.3550	0.10023	
N_Gemme	1	82.87	82.87	0.6049	0.45667	
Defogl	1	1403.89	1403.89	10.2471	0.01081	*
Dirad	1	111.92	111.92	0.8169	0.38964	
Vigoria	1	33.55	33.55	0.2449	0.63255	
N_Gemme:Defogl	1	245.76	245.76	1.7938	0.21330	
N_Gemme:Dirad	1	608.24	608.24	4.4396	0.06438	.
Defogl:Dirad	1	270.75	270.75	1.9762	0.19336	
N_Gemme:Vigoria	1	6.55	6.55	0.0478	0.83177	
Defogl:Vigoria	1	20.64	20.64	0.1507	0.70690	
Dirad:Vigoria	1	152.86	152.86	1.1157	0.31837	
N_Gemme:Defogl:Dirad	1	270.73	270.73	1.9761	0.19337	
Residuals	9	1233.03	137.00			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: 0.675
DIPROVE_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	43686	43686	26.5631	0.000600	***

DIPROVE_2	1	242	242	0.1471	0.710191	
DIPROVE_3	1	22061	22061	13.4141	0.005215	**
N_Gemme	1	507	507	0.3083	0.592268	
Defogl	1	7764	7764	4.7206	0.057859	.
Dirad	1	9693	9693	5.8935	0.038129	*
Viguria	1	12	12	0.0074	0.933399	
N_Gemme:Defogl	1	1934	1934	1.1760	0.306367	
N_Gemme:Dirad	1	846	846	0.5141	0.491538	
Defogl:Dirad	1	1278	1278	0.7770	0.400977	
N_Gemme:Viguria	1	237	237	0.1442	0.712914	
Defogl:Viguria	1	4613	4613	2.8049	0.128297	
Dirad:Viguria	1	8739	8739	5.3139	0.046601	*
N_Gemme:Defogl:Dirad	1	11037	11037	6.7110	0.029187	*
Residuals	9	14802	1645			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.864
DIPROVE_2: n.s
DIPROVE_3: -0.774

Poliftot1B

Analysis of Variance Table

Response: Do_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	23901.4	23901.4	14.8098	0.003916	**
DIPROVE_2	1	2201.3	2201.3	1.3640	0.272853	
DIPROVE_3	1	156.7	156.7	0.0971	0.762421	
N_Gemme	1	1510.4	1510.4	0.9359	0.358620	
Defogl	1	5079.1	5079.1	3.1471	0.109808	
Dirad	1	9653.5	9653.5	5.9815	0.037016	*
Viguria	1	147.3	147.3	0.0912	0.769460	
N_Gemme:Defogl	1	180.7	180.7	0.1120	0.745570	
N_Gemme:Dirad	1	3234.3	3234.3	2.0040	0.190543	
Defogl:Dirad	1	1014.2	1014.2	0.6284	0.448328	
N_Gemme:Viguria	1	116.4	116.4	0.0721	0.794361	
Defogl:Viguria	1	3.0	3.0	0.0019	0.966461	
Dirad:Viguria	1	0.9	0.9	0.0006	0.981218	
N_Gemme:Defogl:Dirad	1	1564.1	1564.1	0.9692	0.350612	
Residuals	9	14525.0	1613.9			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.789
DIPROVE_2: n.s
DIPROVE_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Do_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
DIPROVE_1	1	148522	148522	13.8473	0.004762	**
DIPROVE_2	1	19071	19071	1.7781	0.215151	
DIPROVE_3	1	79793	79793	7.4394	0.023319	*
N_Gemme	1	24281	24281	2.2638	0.166689	

Defogl	1	2561	2561	0.2388	0.636774
Dirad	1	1423	1423	0.1327	0.724061
Vigoria	1	1356	1356	0.1264	0.730338
N_Gemme:Defogl	1	12712	12712	1.1852	0.304597
N_Gemme:Dirad	1	278	278	0.0259	0.875686
Defogl:Dirad	1	10643	10643	0.9923	0.345219
N_Gemme:Vigoria	1	4076	4076	0.3801	0.552848
Defogl:Vigoria	1	11318	11318	1.0552	0.331122
Dirad:Vigoria	1	56226	56226	5.2421	0.047809 *
N_Gemme:Defogl:Dirad	1	15766	15766	1.4699	0.256223
Residuals	9	96532	10726		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: -0.779
DIPROVE_2: n.s
DIPROVE_3: -0.673

Poliftot2B

Analysis of Variance Table

Response: Do_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DIPROVE_1	1	76343	76343	4.0614	0.07469 .
DIPROVE_2	1	24346	24346	1.2952	0.28448
DIPROVE_3	1	369	369	0.0196	0.89166
N_Gemme	1	13095	13095	0.6967	0.42551
Defogl	1	540	540	0.0287	0.86919
Dirad	1	1658	1658	0.0882	0.77322
Vigoria	1	1049	1049	0.0558	0.81851
N_Gemme:Defogl	1	3704	3704	0.1971	0.66759
N_Gemme:Dirad	1	59	59	0.0031	0.95654
Defogl:Dirad	1	5	5	0.0002	0.98789
N_Gemme:Vigoria	1	12862	12862	0.6842	0.42952
Defogl:Vigoria	1	441	441	0.0235	0.88167
Dirad:Vigoria	1	5586	5586	0.2972	0.59891
N_Gemme:Defogl:Dirad	1	457	457	0.0243	0.87956
Residuals	9	169177	18797		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

DIPROVE_1: n.s
DIPROVE_2: n.s
DIPROVE_3: n.s

DONNA OLIMPIA 2009M IASMA

PvinacciaA

Analysis of Variance Table

Response: Do_2009M[, 11]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	682317	682317	27.3921	0.0005392 ***
IASMA_2	1	81056	81056	3.2541	0.1047421
IASMA_3	1	304925	304925	12.2414	0.0067365 **
N_Gemme	1	512144	512144	20.5604	0.0014173 **
Defogl	1	142463	142463	5.7193	0.0404577 *
Dirad	1	1059407	1059407	42.5306	0.0001087 ***

Vigoria	1	75524	75524	3.0319	0.1156239
N_Gemme:Defogl	1	10596	10596	0.4254	0.5305729
N_Gemme:Dirad	1	136675	136675	5.4869	0.0438463 *
Defogl:Dirad	1	18541	18541	0.7443	0.4106792
N_Gemme:Vigoria	1	4787	4787	0.1922	0.6714488
Defogl:Vigoria	1	22054	22054	0.8854	0.3713007
Dirad:Vigoria	1	0	0	0.0000	0.9987892
N_Gemme:Defogl:Dirad	1	19882	19882	0.7982	0.3948961
Residuals	9	224183	24909		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.868
IASMA_2: n.s
IASMA_3: -0.759

DensMostoA

Analysis of Variance Table

Response: Do_2009M[, 12]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	9.9664e-06	9.9664e-06	6.1197	0.035349 *
IASMA_2	1	6.8963e-06	6.8963e-06	4.2346	0.069727 .
IASMA_3	1	2.7980e-07	2.7980e-07	0.1718	0.688207
N_Gemme	1	1.1270e-05	1.1270e-05	6.9204	0.027329 *
Defogl	1	2.6383e-05	2.6383e-05	16.2003	0.002996 **
Dirad	1	6.7731e-06	6.7731e-06	4.1589	0.071841 .
Vigoria	1	1.7920e-07	1.7920e-07	0.1100	0.747720
N_Gemme:Defogl	1	3.9000e-08	3.9000e-08	0.0239	0.880438
N_Gemme:Dirad	1	7.3360e-07	7.3360e-07	0.4505	0.518973
Defogl:Dirad	1	1.4800e-08	1.4800e-08	0.0091	0.926220
N_Gemme:Vigoria	1	1.4247e-06	1.4247e-06	0.8748	0.374045
Defogl:Vigoria	1	1.3607e-06	1.3607e-06	0.8355	0.384512
Dirad:Vigoria	1	3.2220e-07	3.2220e-07	0.1978	0.666986
N_Gemme:Defogl:Dirad	1	9.6100e-07	9.6100e-07	0.5901	0.462054
Residuals	9	1.4657e-05	1.6286e-06		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.636
IASMA_2: n.s
IASMA_3: n.s

BrixA

Analysis of Variance Table

Response: Do_2009M[, 13]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.53747	0.53747	6.4644	0.031580 *
IASMA_2	1	0.29988	0.29988	3.6068	0.090009 .
IASMA_3	1	0.00084	0.00084	0.0101	0.922324
N_Gemme	1	0.56824	0.56824	6.8345	0.028073 *
Defogl	1	1.16424	1.16424	14.0028	0.004611 **
Dirad	1	0.33677	0.33677	4.0505	0.075016 .
Vigoria	1	0.02439	0.02439	0.2933	0.601256
N_Gemme:Defogl	1	0.00610	0.00610	0.0734	0.792543
N_Gemme:Dirad	1	0.02798	0.02798	0.3365	0.576062

Defogl:Dirad	1	0.00038	0.00038	0.0045	0.947888
N_Gemme:Vigoria	1	0.08727	0.08727	1.0497	0.332335
Defogl:Vigoria	1	0.06626	0.06626	0.7970	0.395243
Dirad:Vigoria	1	0.04404	0.04404	0.5297	0.485254
N_Gemme:Defogl:Dirad	1	0.03449	0.03449	0.4149	0.535582
Residuals	9	0.74829	0.08314		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.647

IASMA_2: n.s

IASMA_3: n.s

PhA

Analysis of Variance Table

Response: Do_2009M[, 14]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.005903	0.0059032	1.2562	0.29139
IASMA_2	1	0.001458	0.0014584	0.3103	0.59105
IASMA_3	1	0.002068	0.0020681	0.4401	0.52371
N_Gemme	1	0.006843	0.0068426	1.4561	0.25831
Defogl	1	0.021127	0.0211265	4.4957	0.06301 .
Dirad	1	0.008008	0.0080081	1.7041	0.22414
Vigoria	1	0.003975	0.0039748	0.8458	0.38173
N_Gemme:Defogl	1	0.000442	0.0004419	0.0940	0.76608
N_Gemme:Dirad	1	0.000549	0.0005486	0.1167	0.74044
Defogl:Dirad	1	0.000048	0.0000483	0.0103	0.92148
N_Gemme:Vigoria	1	0.010315	0.0103146	2.1949	0.17261
Defogl:Vigoria	1	0.006539	0.0065388	1.3914	0.26840
Dirad:Vigoria	1	0.003343	0.0033426	0.7113	0.42086
N_Gemme:Defogl:Dirad	1	0.000223	0.0002229	0.0474	0.83244
Residuals	9	0.042294	0.0046993		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

AcTitola

Analysis of Variance Table

Response: Do_2009M[, 15]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.138132	0.138132	6.0600	0.03606 *
IASMA_2	1	0.061137	0.061137	2.6822	0.13590
IASMA_3	1	0.164432	0.164432	7.2139	0.02497 *
N_Gemme	1	0.088206	0.088206	3.8697	0.08071 .
Defogl	1	0.004176	0.004176	0.1832	0.67869
Dirad	1	0.139957	0.139957	6.1401	0.03511 *
Vigoria	1	0.017490	0.017490	0.7673	0.40382
N_Gemme:Defogl	1	0.046794	0.046794	2.0529	0.18572
N_Gemme:Dirad	1	0.016503	0.016503	0.7240	0.41690
Defogl:Dirad	1	0.053875	0.053875	2.3636	0.15858
N_Gemme:Vigoria	1	0.003631	0.003631	0.1593	0.69912
Defogl:Vigoria	1	0.032620	0.032620	1.4311	0.26215

```

Dirad:Vigoria      1 0.010627 0.010627  0.4662 0.51192
N_Gemme:Defogl:Dirad 1 0.013836 0.013836  0.6070 0.45591
Residuals         9 0.205145 0.022794

```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

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IASMA_1: 0.634
IASMA_2: n.s
IASMA_3: -0.667

```

AcTartA

Analysis of Variance Table

Response: Do_2009M[, 16]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.071879	0.071879	3.2932	0.10296
IASMA_2	1	0.124136	0.124136	5.6874	0.04090 *
IASMA_3	1	0.085150	0.085150	3.9012	0.07968 .
N_Gemme	1	0.073520	0.073520	3.3684	0.09965 .
Defogl	1	0.047488	0.047488	2.1757	0.17431
Dirad	1	0.099361	0.099361	4.5523	0.06166 .
Vigoria	1	0.005812	0.005812	0.2663	0.61827
N_Gemme:Defogl	1	0.017809	0.017809	0.8159	0.38991
N_Gemme:Dirad	1	0.030917	0.030917	1.4165	0.26443
Defogl:Dirad	1	0.000084	0.000084	0.0038	0.95198
N_Gemme:Vigoria	1	0.007048	0.007048	0.3229	0.58377
Defogl:Vigoria	1	0.048737	0.048737	2.2329	0.16931
Dirad:Vigoria	1	0.000000	0.000000	0.0000	0.99932
N_Gemme:Defogl:Dirad	1	0.029317	0.029317	1.3432	0.27630
Residuals	9	0.196438	0.021826		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

IASMA_1: n.s
IASMA_2: 0.622
IASMA_3: n.s

```

AcMalicoA

Analysis of Variance Table

Response: Do_2009M[, 17]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00221	0.002211	0.0391	0.84757
IASMA_2	1	0.10923	0.109228	1.9335	0.19780
IASMA_3	1	0.00674	0.006745	0.1194	0.73764
N_Gemme	1	0.00834	0.008340	0.1476	0.70974
Defogl	1	0.28818	0.288181	5.1011	0.05029 .
Dirad	1	0.05191	0.051913	0.9189	0.36280
Vigoria	1	0.14222	0.142218	2.5174	0.14706
N_Gemme:Defogl	1	0.15602	0.156015	2.7616	0.13091
N_Gemme:Dirad	1	0.00009	0.000091	0.0016	0.96892
Defogl:Dirad	1	0.07784	0.077844	1.3779	0.27058
N_Gemme:Vigoria	1	0.01151	0.011506	0.2037	0.66245
Defogl:Vigoria	1	0.00119	0.001188	0.0210	0.88790
Dirad:Vigoria	1	0.01891	0.018914	0.3348	0.57704
N_Gemme:Defogl:Dirad	1	0.03253	0.032526	0.5758	0.46739
Residuals	9	0.50844	0.056494		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

KA

Analysis of Variance Table

Response: Do_2009M[, 18]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.022588	0.022588	0.8228	0.38800
IASMA_2	1	0.045520	0.045520	1.6581	0.22998
IASMA_3	1	0.032292	0.032292	1.1763	0.30631
N_Gemme	1	0.027942	0.027942	1.0178	0.33939
Defogl	1	0.114395	0.114395	4.1671	0.07161
Dirad	1	0.066960	0.066960	2.4391	0.15278
Vigoria	1	0.083552	0.083552	3.0435	0.11502
N_Gemme:Defogl	1	0.028137	0.028137	1.0250	0.33779
N_Gemme:Dirad	1	0.000150	0.000150	0.0055	0.94262
Defogl:Dirad	1	0.006495	0.006495	0.2366	0.63831
N_Gemme:Vigoria	1	0.004536	0.004536	0.1652	0.69389
Defogl:Vigoria	1	0.009087	0.009087	0.3310	0.57917
Dirad:Vigoria	1	0.043285	0.043285	1.5767	0.24085
N_Gemme:Defogl:Dirad	1	0.005416	0.005416	0.1973	0.66741
Residuals	9	0.247071	0.027452		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

ApaA

Analysis of Variance Table

Response: Do_2009M[, 19]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	305.19	305.19	1.1710	0.30734
IASMA_2	1	1151.41	1151.41	4.4178	0.06492
IASMA_3	1	681.98	681.98	2.6166	0.14020
N_Gemme	1	16.09	16.09	0.0617	0.80933
Defogl	1	1251.88	1251.88	4.8033	0.05610
Dirad	1	240.54	240.54	0.9229	0.36181
Vigoria	1	31.15	31.15	0.1195	0.73748
N_Gemme:Defogl	1	57.71	57.71	0.2214	0.64914
N_Gemme:Dirad	1	0.09	0.09	0.0004	0.98530
Defogl:Dirad	1	17.57	17.57	0.0674	0.80099
N_Gemme:Vigoria	1	315.20	315.20	1.2094	0.30000
Defogl:Vigoria	1	49.27	49.27	0.1890	0.67396
Dirad:Vigoria	1	7.61	7.61	0.0292	0.86812
N_Gemme:Defogl:Dirad	1	1.36	1.36	0.0052	0.94392
Residuals	9	2345.68	260.63		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Antoctot1A

Analysis of Variance Table

Response: Do_2009M[, 20]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	21439	21438.5	3.5187	0.09343 .
IASMA_2	1	22463	22462.6	3.6868	0.08704 .
IASMA_3	1	869	869.4	0.1427	0.71438
N_Gemme	1	4356	4356.2	0.7150	0.41971
Defogl	1	20637	20637.1	3.3872	0.09884 .
Dirad	1	1285	1285.3	0.2110	0.65691
Vigoria	1	5673	5673.1	0.9311	0.35978
N_Gemme:Defogl	1	359	359.0	0.0589	0.81364
N_Gemme:Dirad	1	358	357.9	0.0587	0.81393
Defogl:Dirad	1	265	265.3	0.0435	0.83936
N_Gemme:Vigoria	1	304	304.3	0.0499	0.82814
Defogl:Vigoria	1	2199	2199.1	0.3609	0.56281
Dirad:Vigoria	1	5335	5335.3	0.8757	0.37381
N_Gemme:Defogl:Dirad	1	1038	1037.8	0.1703	0.68948
Residuals	9	54834	6092.7		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot1A

Analysis of Variance Table

Response: Do_2009M[, 21]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	27049.7	27049.7	7.9534	0.02004 *
IASMA_2	1	10786.1	10786.1	3.1714	0.10863
IASMA_3	1	3537.1	3537.1	1.0400	0.33445
N_Gemme	1	3736.0	3736.0	1.0985	0.32193
Defogl	1	16427.8	16427.8	4.8302	0.05554 .
Dirad	1	8161.1	8161.1	2.3996	0.15578
Vigoria	1	249.6	249.6	0.0734	0.79257
N_Gemme:Defogl	1	437.9	437.9	0.1288	0.72800
N_Gemme:Dirad	1	80.2	80.2	0.0236	0.88135
Defogl:Dirad	1	243.4	243.4	0.0716	0.79511
N_Gemme:Vigoria	1	417.7	417.7	0.1228	0.73406
Defogl:Vigoria	1	141.3	141.3	0.0416	0.84300
Dirad:Vigoria	1	64.5	64.5	0.0190	0.89348
N_Gemme:Defogl:Dirad	1	75.5	75.5	0.0222	0.88487
Residuals	9	30609.3	3401.0		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: -0.685
IASMA_2: n.s
IASMA_3: n.s

Antoctot2A

Analysis of Variance Table

Response: Do_2009M[, 22]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	10527	10527	0.5267	0.4864
IASMA_2	1	36089	36089	1.8057	0.2119
IASMA_3	1	7777	7777	0.3891	0.5482
N_Gemme	1	647	647	0.0324	0.8612
Defogl	1	35655	35655	1.7840	0.2145
Dirad	1	1946	1946	0.0974	0.7621
Vigoria	1	10721	10721	0.5364	0.4826
N_Gemme:Defogl	1	396	396	0.0198	0.8911
N_Gemme:Dirad	1	2275	2275	0.1138	0.7435
Defogl:Dirad	1	47	47	0.0023	0.9625
N_Gemme:Vigoria	1	52	52	0.0026	0.9604
Defogl:Vigoria	1	8302	8302	0.4154	0.5353
Dirad:Vigoria	1	16571	16571	0.8291	0.3863
N_Gemme:Defogl:Dirad	1	128	128	0.0064	0.9381
Residuals	9	179876	19986		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

Poliftot2A

Analysis of Variance Table

Response: Do_2009M[, 23]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	4906	4906.4	0.5979	0.4592
IASMA_2	1	7838	7837.8	0.9551	0.3540
IASMA_3	1	5974	5974.3	0.7280	0.4157
N_Gemme	1	31	31.5	0.0038	0.9520
Defogl	1	17657	17656.6	2.1515	0.1765
Dirad	1	232	231.8	0.0283	0.8702
Vigoria	1	203	203.1	0.0247	0.8785
N_Gemme:Defogl	1	235	235.0	0.0286	0.8694
N_Gemme:Dirad	1	1020	1020.4	0.1243	0.7325
Defogl:Dirad	1	3496	3496.1	0.4260	0.5303
N_Gemme:Vigoria	1	56	56.3	0.0069	0.9358
Defogl:Vigoria	1	1072	1072.2	0.1307	0.7261
Dirad:Vigoria	1	1264	1264.1	0.1540	0.7038
N_Gemme:Defogl:Dirad	1	6604	6604.1	0.8047	0.3930
Residuals	9	73860	8206.7		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PvinacciaB

Analysis of Variance Table

Response: Do_2009M[, 24]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
IASMA_1	1	984464	984464	12.6105	0.006205	**
IASMA_2	1	81129	81129	1.0392	0.334622	
IASMA_3	1	198109	198109	2.5377	0.145622	
N_Gemme	1	937469	937469	12.0085	0.007101	**
Defogl	1	26816	26816	0.3435	0.572220	
Dirad	1	1761101	1761101	22.5588	0.001045	**
Vigoria	1	4033	4033	0.0517	0.825279	
N_Gemme:Defogl	1	8668	8668	0.1110	0.746597	
N_Gemme:Dirad	1	563108	563108	7.2131	0.024971	*
Defogl:Dirad	1	6295	6295	0.0806	0.782862	
N_Gemme:Vigoria	1	48568	48568	0.6221	0.450531	
Defogl:Vigoria	1	34455	34455	0.4414	0.523125	
Dirad:Vigoria	1	24926	24926	0.3193	0.585843	
N_Gemme:Defogl:Dirad	1	66028	66028	0.8458	0.381737	
Residuals	9	702603	78067			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: 0.764
IASMA_2: n.s
IASMA_3: n.s

DensMostoB

Analysis of Variance Table

Response: Do_2009M[, 25]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	9.5960e-06	9.5957e-06	1.0910	0.3235
IASMA_2	1	6.7200e-07	6.7190e-07	0.0764	0.7885
IASMA_3	1	3.0780e-06	3.0780e-06	0.3500	0.5687
N_Gemme	1	3.3970e-06	3.3973e-06	0.3863	0.5497
Defogl	1	2.8200e-07	2.8210e-07	0.0321	0.8618
Dirad	1	1.2920e-06	1.2915e-06	0.1468	0.7105
Vigoria	1	1.4385e-05	1.4385e-05	1.6356	0.2329
N_Gemme:Defogl	1	5.7620e-06	5.7623e-06	0.6552	0.4391
N_Gemme:Dirad	1	1.1248e-05	1.1249e-05	1.2789	0.2873
Defogl:Dirad	1	1.8900e-07	1.8890e-07	0.0215	0.8867
N_Gemme:Vigoria	1	8.0090e-06	8.0094e-06	0.9107	0.3649
Defogl:Vigoria	1	2.7600e-07	2.7570e-07	0.0313	0.8634
Dirad:Vigoria	1	5.8200e-06	5.8200e-06	0.6617	0.4369
N_Gemme:Defogl:Dirad	1	1.1890e-06	1.1889e-06	0.1352	0.7216
Residuals	9	7.9156e-05	8.7951e-06		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

BrixB

Analysis of Variance Table

Response: Do_2009M[, 26]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.5204	0.52036	1.1562	0.3102
IASMA_2	1	0.0250	0.02495	0.0554	0.8191
IASMA_3	1	0.1486	0.14858	0.3301	0.5797
N_Gemme	1	0.1979	0.19787	0.4396	0.5239

Defogl	1	0.0465	0.04649	0.1033	0.7553
Dirad	1	0.0396	0.03958	0.0879	0.7735
Vigoria	1	0.6268	0.62685	1.3928	0.2682
N_Gemme:Defogl	1	0.2936	0.29360	0.6524	0.4401
N_Gemme:Dirad	1	0.5405	0.54053	1.2010	0.3016
Defogl:Dirad	1	0.0242	0.02423	0.0538	0.8217
N_Gemme:Vigoria	1	0.3640	0.36404	0.8089	0.3919
Defogl:Vigoria	1	0.0090	0.00901	0.0200	0.8906
Dirad:Vigoria	1	0.3163	0.31634	0.7029	0.4235
N_Gemme:Defogl:Dirad	1	0.0785	0.07851	0.1744	0.6860
Residuals	9	4.0506	0.45006		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

PhB

Analysis of Variance Table

Response: Do_2009M[, 27]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.019772	0.0197716	2.4491	0.15204
IASMA_2	1	0.001198	0.0011984	0.1484	0.70898
IASMA_3	1	0.003499	0.0034990	0.4334	0.52680
N_Gemme	1	0.005533	0.0055329	0.6853	0.42916
Defogl	1	0.004494	0.0044937	0.5566	0.47466
Dirad	1	0.001270	0.0012699	0.1573	0.70090
Vigoria	1	0.016189	0.0161892	2.0053	0.19041
N_Gemme:Defogl	1	0.000140	0.0001405	0.0174	0.89795
N_Gemme:Dirad	1	0.002338	0.0023375	0.2895	0.60356
Defogl:Dirad	1	0.001774	0.0017742	0.2198	0.65037
N_Gemme:Vigoria	1	0.001144	0.0011437	0.1417	0.71535
Defogl:Vigoria	1	0.003493	0.0034931	0.4327	0.52714
Dirad:Vigoria	1	0.030749	0.0307493	3.8088	0.08275
N_Gemme:Defogl:Dirad	1	0.001294	0.0012944	0.1603	0.69819
Residuals	9	0.072658	0.0080732		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTitolB

Analysis of Variance Table

Response: Do_2009M[, 28]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.12439	0.124388	2.2263	0.16988
IASMA_2	1	0.19239	0.192390	3.4434	0.09648
IASMA_3	1	0.00002	0.000019	0.0003	0.98570
N_Gemme	1	0.05888	0.058875	1.0538	0.33144
Defogl	1	0.07092	0.070921	1.2694	0.28903
Dirad	1	0.06989	0.069892	1.2509	0.29234
Vigoria	1	0.17031	0.170305	3.0481	0.11478
N_Gemme:Defogl	1	0.03087	0.030873	0.5526	0.47623
N_Gemme:Dirad	1	0.16161	0.161614	2.8926	0.12320

Defogl:Dirad	1	0.03475	0.034749	0.6219	0.45060
N_Gemme:Vigoria	1	0.01875	0.018745	0.3355	0.57664
Defogl:Vigoria	1	0.02845	0.028446	0.5091	0.49360
Dirad:Vigoria	1	0.14943	0.149427	2.6745	0.13640
N_Gemme:Defogl:Dirad	1	0.04984	0.049840	0.8920	0.36959
Residuals	9	0.50285	0.055872		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcTartB

Analysis of Variance Table

Response: Do_2009M[, 29]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00047	0.000472	0.0051	0.9446
IASMA_2	1	0.12762	0.127621	1.3828	0.2698
IASMA_3	1	0.01674	0.016739	0.1814	0.6802
N_Gemme	1	0.11796	0.117960	1.2781	0.2875
Defogl	1	0.01355	0.013551	0.1468	0.7105
Dirad	1	0.08494	0.084945	0.9204	0.3624
Vigoria	1	0.17953	0.179533	1.9452	0.1966
N_Gemme:Defogl	1	0.05480	0.054799	0.5937	0.4607
N_Gemme:Dirad	1	0.05990	0.059903	0.6491	0.4412
Defogl:Dirad	1	0.23397	0.233971	2.5351	0.1458
N_Gemme:Vigoria	1	0.07238	0.072381	0.7842	0.3989
Defogl:Vigoria	1	0.00207	0.002070	0.0224	0.8842
Dirad:Vigoria	1	0.05126	0.051258	0.5554	0.4751
N_Gemme:Defogl:Dirad	1	0.02085	0.020855	0.2260	0.6459
Residuals	9	0.83064	0.092293		

IASMA_1: n.s
IASMA_2: n.s
IASMA_3: n.s

AcMalicoB

Analysis of Variance Table

Response: Do_2009M[, 30]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.00770	0.00770	0.0852	0.77705
IASMA_2	1	0.34630	0.34630	3.8292	0.08206
IASMA_3	1	0.06882	0.06882	0.7609	0.40570
N_Gemme	1	0.00601	0.00601	0.0665	0.80233
Defogl	1	0.14889	0.14889	1.6464	0.23151
Dirad	1	0.00403	0.00403	0.0446	0.83751
Vigoria	1	0.01343	0.01343	0.1485	0.70889
N_Gemme:Defogl	1	0.01768	0.01768	0.1955	0.66881
N_Gemme:Dirad	1	0.06388	0.06388	0.7063	0.42243
Defogl:Dirad	1	0.13727	0.13727	1.5179	0.24915
N_Gemme:Vigoria	1	0.00067	0.00067	0.0074	0.93311
Defogl:Vigoria	1	0.00510	0.00510	0.0564	0.81759
Dirad:Vigoria	1	0.00085	0.00085	0.0094	0.92505
N_Gemme:Defogl:Dirad	1	0.02567	0.02567	0.2839	0.60709

Residuals 9 0.81392 0.09044

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

KB

Analysis of Variance Table

Response: Do_2009M[, 31]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	0.01000	0.009999	0.2176	0.65195
IASMA_2	1	0.00009	0.000086	0.0019	0.96646
IASMA_3	1	0.01522	0.015218	0.3312	0.57906
N_Gemme	1	0.02616	0.026156	0.5692	0.46985
Defogl	1	0.03808	0.038077	0.8287	0.38639
Dirad	1	0.05495	0.054951	1.1959	0.30255
Vigoria	1	0.05188	0.051884	1.1291	0.31564
N_Gemme:Defogl	1	0.01633	0.016331	0.3554	0.56576
N_Gemme:Dirad	1	0.01054	0.010537	0.2293	0.64345
Defogl:Dirad	1	0.00909	0.009093	0.1979	0.66693
N_Gemme:Vigoria	1	0.04152	0.041515	0.9035	0.36667
Defogl:Vigoria	1	0.02251	0.022508	0.4898	0.50169
Dirad:Vigoria	1	0.19127	0.191270	4.1625	0.07174 .
N_Gemme:Defogl:Dirad	1	0.00128	0.001282	0.0279	0.87105
Residuals	9	0.41355	0.045950		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

ApaB

Analysis of Variance Table

Response: Do_2009M[, 32]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	764.71	764.71	4.4047	0.06525 .
IASMA_2	1	529.94	529.94	3.0525	0.11456
IASMA_3	1	30.41	30.41	0.1752	0.68538
N_Gemme	1	105.94	105.94	0.6102	0.45476
Defogl	1	1535.37	1535.37	8.8437	0.01560 *
Dirad	1	78.58	78.58	0.4526	0.51800
Vigoria	1	47.62	47.62	0.2743	0.61313
N_Gemme:Defogl	1	100.68	100.68	0.5799	0.46583
N_Gemme:Dirad	1	651.18	651.18	3.7508	0.08476 .
Defogl:Dirad	1	29.24	29.24	0.1684	0.69110
N_Gemme:Vigoria	1	470.90	470.90	2.7124	0.13398
Defogl:Vigoria	1	0.61	0.61	0.0035	0.95399
Dirad:Vigoria	1	0.57	0.57	0.0033	0.95563
N_Gemme:Defogl:Dirad	1	48.50	48.50	0.2794	0.60989
Residuals	9	1562.50	173.61		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s
 IASMA_3: n.s

Antoctot1B

Analysis of Variance Table

Response: Do_2009M[, 33]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	4076	4075.7	1.1556	0.31035
IASMA_2	1	21213	21212.7	6.0144	0.03661 *
IASMA_3	1	2561	2561.1	0.7262	0.41623
N_Gemme	1	8	8.5	0.0024	0.96200
Defogl	1	24849	24849.2	7.0455	0.02629 *
Dirad	1	7524	7524.2	2.1333	0.17814
Vigoria	1	5864	5864.3	1.6627	0.22939
N_Gemme:Defogl	1	1787	1786.5	0.5065	0.49468
N_Gemme:Dirad	1	28	27.5	0.0078	0.93153
Defogl:Dirad	1	3034	3033.7	0.8601	0.37790
N_Gemme:Vigoria	1	12604	12604.1	3.5736	0.09128 .
Defogl:Vigoria	1	2866	2865.9	0.8126	0.39084
Dirad:Vigoria	1	5579	5579.0	1.5818	0.24015
N_Gemme:Defogl:Dirad	1	3715	3715.2	1.0534	0.33152
Residuals	9	31743	3527.0		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: 0.633
 IASMA_3: n.s

Poliftot1B

Analysis of Variance Table

Response: Do_2009M[, 34]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	5067.7	5067.7	3.8487	0.08141 .
IASMA_2	1	2365.4	2365.4	1.7965	0.21299
IASMA_3	1	106.1	106.1	0.0806	0.78291
N_Gemme	1	3179.3	3179.3	2.4146	0.15463
Defogl	1	23184.5	23184.5	17.6079	0.00232 **
Dirad	1	7090.9	7090.9	5.3853	0.04544 *
Vigoria	1	1190.5	1190.5	0.9041	0.36651
N_Gemme:Defogl	1	149.3	149.3	0.1134	0.74406
N_Gemme:Dirad	1	3433.9	3433.9	2.6079	0.14079
Defogl:Dirad	1	779.2	779.2	0.5918	0.46145
N_Gemme:Vigoria	1	29.0	29.0	0.0220	0.88532
Defogl:Vigoria	1	2182.1	2182.1	1.6572	0.23010
Dirad:Vigoria	1	629.5	629.5	0.4781	0.50673
N_Gemme:Defogl:Dirad	1	2050.8	2050.8	1.5576	0.24352
Residuals	9	11850.4	1316.7		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s
 IASMA_2: n.s

IASMA_3: n.s

Antoctot2B

Analysis of Variance Table

Response: Do_2009M[, 35]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	857	857	0.0312	0.86375
IASMA_2	1	98982	98982	3.5996	0.09028
IASMA_3	1	19576	19576	0.7119	0.42068
N_Gemme	1	15330	15330	0.5575	0.47433
Defogl	1	43543	43543	1.5835	0.23992
Dirad	1	25	25	0.0009	0.97657
Vigoria	1	6676	6676	0.2428	0.63399
N_Gemme:Defogl	1	10257	10257	0.3730	0.55648
N_Gemme:Dirad	1	568	568	0.0207	0.88889
Defogl:Dirad	1	15383	15383	0.5594	0.47358
N_Gemme:Vigoria	1	20214	20214	0.7351	0.41348
Defogl:Vigoria	1	3915	3915	0.1424	0.71467
Dirad:Vigoria	1	1688	1688	0.0614	0.80991
N_Gemme:Defogl:Dirad	1	59	59	0.0022	0.96396
Residuals	9	247484	27498		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Poliftot2B

Analysis of Variance Table

Response: Do_2009M[, 36]

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
IASMA_1	1	992	991.5	0.0432	0.8400
IASMA_2	1	21934	21933.7	0.9552	0.3539
IASMA_3	1	10291	10290.8	0.4482	0.5200
N_Gemme	1	6098	6097.9	0.2656	0.6187
Defogl	1	30351	30351.3	1.3218	0.2799
Dirad	1	4473	4473.2	0.1948	0.6694
Vigoria	1	723	723.3	0.0315	0.8631
N_Gemme:Defogl	1	56	55.6	0.0024	0.9618
N_Gemme:Dirad	1	1513	1512.6	0.0659	0.8032
Defogl:Dirad	1	6090	6090.4	0.2652	0.6190
N_Gemme:Vigoria	1	10558	10557.7	0.4598	0.5148
Defogl:Vigoria	1	1095	1094.5	0.0477	0.8320
Dirad:Vigoria	1	7133	7132.5	0.3106	0.5909
N_Gemme:Defogl:Dirad	1	1733	1733.2	0.0755	0.7897
Residuals	9	206653	22961.4		

IASMA_1: n.s

IASMA_2: n.s

IASMA_3: n.s

Allegato 3.1 : CODICI E LORO STRUTTURA ALL'INTERNO DEL GEODB TUSCANIA

Codici piante e campioni

La struttura del database si basa fundamentalmente sul disegno sperimentale e sull'utilizzo di due codici che riconoscono le piante:

1. **ID_POSIZIONE**. Questo codice è ottenibile attraverso la composizione dei quattro campi **ID_VIGNETO, FILA, FORO, POSIZIONE**. Questo codice identifica ogni pianta in campo indipendentemente che essa faccia parte o no di quelle scelte effettivamente ogni anno come campioni all'interno di ogni parcella. Il codice è composto da una serie di informazioni che permettono a chi sta cercando la posizione in campo delle singole viti, di poterle rintracciare semplicemente scomponendo le varie parti del codice. Il codice è così composto:

- Sigla del vigneto
- Numero della fila
- Numero del foro (campata)
- Posizione della pianta nel foro (1,2,3,4,5,...)
- Ogni parte del codice è divisa dalla seguente da underscore (_)

Di conseguenza la pianta con il codice: Br_2_3_5 è la pianta che si trova nel vigneto di Brolio, nella fila 2, foro 3, posizione 5.

L'elenco delle piante dei vigneti, identificate univocamente dalla composizione dei campi **ID_VIGNETO, FILA, FORO, POSIZIONE**, è contenuto nella tabella **PASSAPORTO**, che forma l'asse portante di tutto il GeoDB Tuscania, in quanto la maggior parte delle altre tabelle sono collegate a questa attraverso i campi che individuano la loro posizione nel vigneto.

2. **ID_CAMPIONE**. Questo codice identifica le piante che di anno in anno vengono scelte come campioni per le varie analisi. Va precisato che queste possono variare di anno in anno a causa, per esempio, delle loro condizioni sanitarie. Si è quindi creato questo codice che riconosce in maniera univoca il campione e che, correlato all'insieme dei campi che identificano la posizione, permette di sapere la precisa posizione del campione.

Il codice è così composto:

- Sigla del vigneto

- Sigla del blocco
- Sigla dalla parcella
- Sigla della pianta
- Sigla dell'anno
- Ogni parte del codice è divisa dalla seguente da underscore (_)

Di conseguenza la pianta con il codice. Br_1_8_3_07 è la pianta campione per il vigneto di Brolio del blocco 1, parcella 8, pianta 3 per l'anno 2007.

Questo tipo di codice, che va a sostituire il precedentemente creato ed adesso eliminato ID_PIANTA_TESI, è stato così ideato al fine di identificare in maniera univoca il campione e al tempo stesso non dare indicazioni sui trattamenti effettuati sulle singole piante, come invece avveniva precedentemente. Questa scelta nasce dal fatto che gli operatori che vanno in campo ad esaminare le piante campione, non devono sapere il tipo di trattamenti effettuati, perché ciò condizionerebbe la loro valutazione. Ogni pianta campione quindi ha il suo codice ID_CAMPIONE e il codice ID_POSIZIONE (ovvero la composizione di ID_VIGNETO, FILA, FORO e POSIZIONE) per poter così essere identificata in campo.

Codici per i blocchi e le parcelle

Per quanto riguarda i blocchi e le parcelle, anche per questi si è creato un codice che li rende univoci. Infatti, precedentemente, all'interno di ogni blocco la parcella erano genericamente identificate con un numero da 1 a 8, per cui, considerati i 4 vigneti, avevamo 32 parcelle identificate genericamente con 1, 2, 3... Adesso si è deciso di usare un codice, sia per i blocchi, sia per le parcelle, che segue il modello concettuale dell'ID_CAMPIONE.

Il codice che identifica i blocchi, **ID_BLOCCO**, sarà così composto:

- Sigla del vigneto
- Sigla del blocco
- Ogni parte del codice è divisa dalla seguente da underscore (_)

L'elenco dei blocchi è contenuto nella tabella BLOCCHI.

Il codice che identifica le parcelle, **ID_PARCELLA**, sarà così composto:

- Sigla del vigneto
- Sigla del blocco
- Sigla dalla parcella

- Ogni parte del codice è divisa dalla seguente da underscore (_)

L'elenco delle parcelle è contenuto nella tabella PARCELLE.

Da notare è il fatto che, malgrado siano stati effettuati cambiamenti formali nei codici, in sostanza la numerazione dei blocchi, delle parcelle e dell'ID_POSIZIONE è rimasta la stessa, per cui il Blocco 1 del vigneto di Brolio che prima si chiamava genericamente blocco 1 come quelli di Donna Olimpia e Mortelle, adesso si chiamerà univocamente BR_1. Questo fatto permetterà, a chi ne avrà bisogno, di potersi aggiornare i propri dati con le nuove codifiche senza rischio di errori.

Codici per stazioni meteo

L'elenco delle stazioni meteo presenti nei quattro vigneti sperimentali sarà contenuto nella tabella STAZIONI_METEO. Il codice delle stazioni viene identificato dal campo **ID_STAZ_METEO**. Per le stazioni slave il codice che le identifica sarà quello della parcella in cui si trovano seguito da suffisso S preceduto da underscore. Per cui la slave per la parcella Br_1_8 sarà Br_1_8_S. La loro posizione sarà individuata, per chi le cercasse in campo senza saperne la collocazione, dall'ID_POSIZIONE della pianta più vicina. Per le stazioni Master si è deciso di usare una codificazione analoga a quella usata per le slave, ma visto che, trovandosi all'esterno del vigneto, non appartengono a nessuna parcella, avranno un codice standard tipo Br_0_0_M. Tale scelta è stata fatta perché, da un punto di vista computazionale, è necessario che i codici di master e slave abbiano le stesse dimensioni e tipo di campo. Ovviamente per le stazioni master non ci sarà un ID_POSIZIONE associato.

Allegato 3.2 – SPECIFICHE TECNICHE DEL GEODATABASE TUSCANIA (LINEA A.1)

LE TABELLE DEL GEODATABASE TUSCANIA

Il GeoDB Tuscania è un classico geodatabase relazionale organizzato per tabelle in maniera da poter raccogliere tutti i dati dei vigneti e delle sperimentazioni su di essi svolte evitando ridondanze.

Ogni tabella del GeoDb è caratterizzata dall'aver una chiave primaria, cioè un campo o un insieme di campi che permette di identificare in maniera univoca un record nella tabella. In altre parole, nessun record nella tabella può avere il campo o un insieme di campi identificato dalla chiave primaria identico a quello di un qualsiasi altro record della stessa (vincolo di unicità). La chiave primaria per sua stessa natura non può contenere campi nulli o duplicati e può essere ad un campo o multicampo. La chiave primaria ad un campo è quella che è definita da un solo campo della tabella, ad esempio il campo ID_VIGNETO della tabella VIGNETO. Altro esempio sono i campi contatore come ID_FLUORESCENZA nella tabella FLUORESCENZA. Il campo contatore è un campo particolare che si aggiorna automaticamente in maniera sequenziale ogni qualvolta alla tabella viene aggiunto un record. In tal modo si può essere certi della univocità dei valori nel campo. La chiave primaria multicampo è quella che viene identificata da duo o più campi. Ad esempio nella tabella passaporto abbiamo la chiave primaria composta da (ID_VIGNETO, FILA, FORO, POSIZIONE).

Al fine di evitare ridondanze e razionalizzare le risorse utilizzate dal GeoDB si è fatto uso delle cosiddette categorie. Le categorie sono tabelle che rappresentano sottotipi di una tabella padre. Ad esempio nella tabella PASSAPORTO abbiamo la colonna ID_CLONE che contiene l'informazione relativa alla selezione clonale. Invece di indicare il nome completo del clone per ogni pianta, si è adottato un codice che identifica tale tipo di clone. Nella tabella CLONE, collegata alla tabella PASSAPORTO attraverso il campo ID_CLONE, abbiamo la estrinsecazione di tale codice attraverso il campo descrittivo correlato. Nel GeoDB Tuscania le categorie sono state usate frequentemente e sono successivamente riportate sotto alla tabella padre corrispondente.

Al momento i dati contenuti nel GeoDB riguardano principalmente:

- **La struttura e le caratteristiche del vigneto e delle piante che ne fanno parte.** Le tabelle fondamentali sono PASSAPORTO (contiene le informazioni delle singole piante), VIGNETO (contiene le informazioni caratterizzanti i vigneti), CAMPIONI (contiene l'elenco dei campioni scelti ogni anno per le varie parcelle), BLOCCHI (contiene l'elenco dei blocchi di ogni vigneto) PARCELLE (contiene l'elenco di tutti le parcelle dei vigneti) TESI (contiene l'elenco di tutti le tesi e dei trattamenti che li caratterizzano). La tabella CAMPIONI, attraverso i campi che compongono l'ID_POSIZIONE (**ID_VIGNETO, FILA, FORO, POSIZIONE**) è collegata alla tabella PASSAPORTO che contiene essa stessa, o attraverso il collegamento alle altre tabelle del database, le informazione e analisi rintracciabili per tale pianta o altre informazioni correlate su blocco, parcella, vigneto, meteo...
- **I risultati delle analisi svolte in campo e in laboratorio da IASMA e DIPROVE.** Questi dati sono raccolti in 10 tabelle con relative categorie. Queste tabelle raccolgono le analisi svolte negli anni di sperimentazione e contengono le indicazioni di posizione relative alle piante su cui sono state svolte.
- **I dati raccolti dalle stazioni meteo dell'IBIMET.** La tabella STAZIONI_METEO contiene l'elenco delle stazioni meteo di tutti i vigneti e loro posizione identificata dall'ID_POSIZIONE della pianta più vicina. I dati raccolti dalle stazioni master si trovano nella stazione METEO_MASTER, mentre quelli raccolti dalle stazioni slave in METEO_SLAVE.
- **I risultati delle analisi svolte sulle uve vinificate.** La tabella VINIFICAZIONI contiene le analisi svolte su uve vinificate presso la cantina sperimentale del Consorzio Toscana. Sono presenti inoltre due tabelle categorie che identificano la fate e il tipo di analisi.
- **Le informazioni spaziali riguardanti le entità fisiche presenti nei vari vigneti.**

Queste informazioni, che nei comuni Sistemi Informativi Territoriali sono gestite come semplici shapefile, nel GeoDB sono immagazzinate come tabelle in cui ogni record registra un oggetto (punto, linea o poligono) e, attraverso uno specifico campo, la sua posizione geografica.

Per una maggiore comprensione della struttura del GeoDB Toscana e delle relazioni esistenti fra le varie tabelle rimandiamo alla consultazione del "Diagramma

GeoDatabase Toscana”. Inoltre si riportano di seguito in dettaglio i contenuti e le specifiche tecniche di tutte le tabelle incluse nel DB.

L'attuale struttura del GeoDB in futuro potrà comunque essere modificata e ampliata anche per includere i dati provenienti da eventuali nuovi partner del Consorzio.

ELENCO DELLE COLONNE, CONTENUTO E LORO FORMATO NELLE TABELLE DEL GEODB TUSCANIA

Nota: in rosso sono identificate le primary key

Descrizione della tabella contenente le informazioni sulle piante: PASSAPORTO e CAMPIONI

TABELLA: PASSAPORTO		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer
TAG	Codice del trasponder inserito nel palo che identifica l'inizio del foro	text_16
ID_STATO	Stato della pianta (presente, assente, morta, ecc.)	text_2
ID_BLOCCO	Codice identificativo del blocco	integer
ID_PARCELLA	Codice identificativo della parcella	integer
X_COORD	Coordinate X (UTM32N WGS84)	Double
Y_COORD	Coordinate Y (UTM32N WGS84)	Double
QUOTA	Quota sul livello del mare	Double
ID_CLONE	Selezione clonale	Integer
ID_PORTINNESTO	Portinnesto	Integer
ID_VITIGNO	Nome varietale	text_2
NOTE	Commenti, annotazioni	text_250

TABELLA: CAMPIONI		
NOME COLONNA	CONTENUTO	FORMATO
ID_VIGNETO	Codice identificativo del vigneto	text_2
ID_BLOCCO	Codice identificativo del blocco	Integer
ID_PARCELLA	Codice identificativo della parcella	Integer
SIGLA_PIANTA	Numero identificativo della pianta nella parcella di appartenenza	Integer
ANNO	Anno a cui è riferito il campione	Integer
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

Categorie per la tabella PASSAPORTO e loro dati attualmente previsti

TABELLA: STATO			
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI

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ID_STATO	Stato della pianta	text_2	P	A	M
DESCRIZIONE	Campo descrittivo	text_25	presente	assente	morta

TABELLA: CLONE					
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI		
ID_CLONE	Selezione clonale	Integer	1	2	3
DESCRIZIONE	Campo descrittivo	text_50	Sangiovese clone R24	Sangiovese clone R23	Cabernet Sauvignon 191

TABELLA: PORTINNESTO					
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI		
ID_PORTINNESTO	Portinnesto	Integer	1	2	
DESCRIZIONE	Campo descrittivo	text_100	420A	101_14	

TABELLA: VITIGNO					
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI		
ID_VITIGNO	Nome varietale	text_2	SA	CB	
DESCRIZIONE	Campo descrittivo	text_25	San Giovese	Cabernet Suavignon	

Descrizione della tabella contenente le informazioni fondamentali sui: VIGNETO

TABELLA: VIGNETO		
NOME COLONNA	CONTENUTO	FORMATO
VIGNETO	Nome dell'impianto	text_20
ID_VIGNETO	Codice identificativo del vigneto	text_2
AZIENDA	Ragione sociale dell'azienda	text_30
NOME_COMUNE_CAT	Nome del comune catastale	text_50
CODICE_COMUNE_CAT	Codice del comune catastale	text_10
CODICE_PART_CAT	Codice della particella catastale	text_50
LONGITUDINE	Coordinate X del centroide del vigneto (GB)	Double
LATITUDINE	Coordinate Y del centroide del vigneto (GB)	Double
QUOTA_MEDIA	quota media sul livello del mare	Double
PENDENZA_MEDIA	pendenza media del campo	Double
ESPOSIZIONE_MEDIA	esposizione media del campo	Double
ID_TIPO_TERRENO	tipo di terreno per modello IRR13 su bilancio evapotrasporativo	text_2
LIMO_PERC	Percentuale di limo nel terreno	Double
SABBIA_PERC	Percentuale di sabbia nel terreno	Double
ARGILLA_PERC	Percentuale di argilla nel terreno	Double
ID_FORMA_ALLEVAMENTO	Forma di allevamento	text_2
ID_SISTEMA_ALLEVAMENTO	Sistema di allevamento	text_2
CEPPI_ETTARO	Numero medio di ceppi per ettaro	Integer
SESTO_PIANTE	Distanza tra le piante	Float

SESTO_FILE	Distanza fra le file	Float
PIANTE_X_FORO	Numero di piante per foro (=distanza tra due pali)	Integer
PRODUZ_ETTARO	Produzione media per ettaro	Integer
ID_STATO_IRRIGAZIONE	Stato dell'irrigazione	text_2
ID_TIPO_IRRIGAZIONE	Tipo (sistema) di irrigazione	text_2
DISTANZA_EROGATORI_GOCCIA	Distanza tra erogatori sulla fila in metri	Float
PORTATA_EROGATORI	Portata di un erogatore in litri/ora	Float
NOTE	Commenti, annotazioni	text_250

Categorie per la tabella VIGNETO e loro dati attualmente previsti:

TABELLA: TIPO_TERRENO						
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI			
ID_TIPO_TERRENO	tipo di terreno per modello IRR13	text_2	A	M	S	T
DESCRIZIONE	campo descrittivo	text_50	argilloso	medio impasto	sabboso	torboso

TABELLA: FORMA_ALLEVAMENTO				
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI	
ID_FORMA_ALLEVAMENTO	forma di allevamento	text_2	S	T
DESCRIZIONE	campo descrittivo	text_25	spalliera	tendone

TABELLA: SISTEMA_ALLEVAMENTO					
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI		
ID_SISTEMA_ALLEVAMENTO	sistema di allevamento	text_2	G	C	P
DESCRIZIONE	campo descrittivo	text_100	guyot	cordone speronato	pergola semplice

TABELLA: STATO_IRRIGAZIONE				
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI	
ID_STATO_IRRIGAZIONE	stato dell'irrigazione	text_2	A	P
DESCRIZIONE	campo descrittivo	text_25	assente	presente

TABELLA: TIPO_IRRIGAZIONE						
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI			
ID_TIPO_IRRIGAZIONE	Tipo di irrigazione	text_2	P	SC	G	SU
DESCRIZIONE	campo descrittivo	text_25	pioggia	scorrimento	goccia	subirrigazione

Descrizione della tabella contenente le informazioni fondamentali su: BLOCCHI

TABELLA: BLOCCHI

NOME COLONNA	CONTENUTO	FORMATO
ID_BLOCCO	codice identificativo del blocco	integer
ID_VIGNETO	codice identificativo del vigneto	text_2
ID_VIGORE	codice identificativo del vigore	text_2

Categorie per la tabella BLOCCHI e loro dati attualmente previsti

TABELLA: VIGORE					
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI		
ID_VIGORE	Codice vigore	text_2	L	M	H
DESCRIZIONE	campo descrittivo	text_25	basso	medio	alto

Descrizione della tabella contenente le informazioni fondamentali su: PARCELLE

TABELLA: PARCELLE		
NOME COLONNA	CONTENUTO	FORMATO
ID_PARCELLA	codice identificativo della parcella	integer
ID_BLOCCO	codice identificativo del blocco	integer
ID_TESI	codice identificativo della tesi	Integer
ID_VIGNETO	codice identificativo del vigneto	text_2

Descrizione della tabella contenente le informazioni fondamentali su: TESI

TABELLA: TESI		
NOME COLONNA	CONTENUTO	FORMATO
ID_TESI	Codice identificativo della tesi	Integer
ID_SELEZIONE_GEMME	Tipo di selezione del germoglio	text_2
ID_DEFOGLIAZIONE	Effettuazione defogliazione	text_2
ID_DIRADAMENTO	Effettuazione diradamento	text_2

Categorie per la tabella TESI e loro dati attualmente previsti

TABELLA: SELEZIONE_GEMME				
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI	
ID_SELEZIONE_GEMME	Tipo di selezione del germoglio	text_2	C0	C1
DESCRIZIONE	campo descrittivo	text_25	1 gemma con selezione dei germogli	3 gemme con selezione germogli

TABELLA: DEFOGLIAZIONE				
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI	
ID_DEFOGLIAZIONE	Effettuazione defogliazione	text_2	A0	A1

DESCRIZIONE	campo descrittivo	text_25	senza defogliazione	con defogliazione
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TABELLA: DIRADAMENTO				
NOME COLONNA	CONTENUTO	FORMATO	DATI ATTUALMENTE PREVISTI	
ID_DIRADAMENTO	Effettuazione diradamento	text_2	D0	D1
DESCRIZIONE	campo descrittivo	text_100	nessun diradamento	Diradamento al 50 %

Descrizione delle tabelle contenenti le ANALISI IASMA

TABELLA: FLUORESCENZA		
NOME COLONNA	CONTENUTO	FORMATO
ID_FLUORESCENZA	Campo contatore	Integer
ESPOSIZIONE	Esposizione	Text_4
FOGLIA_NR	Numero della foglia	
DATA_CAMP	Data di campionamento	Date
DATA_CAMP_STR	Data di campionamento	text_8
ORA_CAMP	Ora di campionamento	Time
ORA_CAMP_STR	Ora di campionamento	text_4
DATA_MISURA	Data misurazione	Date
DATA_MISURA_STR	Data misurazione	text_8
Fo	Fluorescenza basale dopo adattamento al buio	Float
Fm	Fluorescenza massima dopo adattamento al buio	Float
FvFm	Efficienza quantica	Float
Time	Ora misurazione	Time
Nr	Numero progressivo delle misure	Integer
ML	parametro dello strumento PAM2000	Float
Leaf_T	Temperatura fogliare	Float
PAR	Flusso luminoso foto sinteticamente attivo	Float
Ft	Fluorescenza basale senza adattamento al buio	Float
ETR	Velocità del trasporto elettronico	Float
Yield	Resa quantica in condizioni di luce	Float
qP	quenching fotochimico	Float
qN	quenching non fotochimico	Float
Fm1	parametro dello strumento PAM2000	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer

TABELLA: NDVI		
NOME COLONNA	CONTENUTO	FORMATO
ID_NDVI	Campo contatore	Integer
ID_MISURE	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
ORA_CAMP	Ora di campionamento	Time
ORA_CAMP_STR	Ora di campionamento	text_4
GPS_data	Localizzazione	text_50
RED_inc	Rosso incidente	Float
RED_refl	Rosso riflesso	Float
GREEN_inc	Verde incidente	Float
GREEN_refl	Verde riflesso	Float
FAR_RED_inc	Rosso lontano incidente	Float
FAR_RED_refl	Rosso lontano riflesso	Float
NIR_inc	Infrarosso vicino incidente	Float
NIR_refl	Infrarosso vicino riflesso	Float
NDVI_RED	Valore di NDVIred (Normalized Difference Vegetation Index)	Float
NDVI_GREEN	Valore di NDVIgreen (Normalized Difference Vegetation Index)	Float
RED_FARRED	Rapporto Red/Far Red	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: SPAD		
NOME COLONNA	CONTENUTO	FORMATO
ID_SPAD	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
ORA	Ora di campionamento	Time
ORA_STR	Ora di campionamento	text_4
SPAD_value	Valore di SPAD	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: LAI		
NOME COLONNA	CONTENUTO	FORMATO
ID_LAI	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer

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DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
LAI	Valore di Leaf Area Index	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: POTENZIALE_IDRICO		
NOME COLONNA	CONTENUTO	FORMATO
ID_POTENZIALE_IDRICO	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
LWP	Potenziale idrico fogliare	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: PRELIEVI		
NOME COLONNA	CONTENUTO	FORMATO
ID_PRELIEVI	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
ID_MATERIALE	ID materiale raccolto	Integer
ID_DESTINAZIONE	ID destinazione del materiale raccolto	Integer
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: DATI_PRODUTTIVI		
NOME COLONNA	CONTENUTO	FORMATO
ID_DATI_PRODUTTIVI	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
Nr_Grapp_ante	Numero di grappoli pre-diradamento	Integer
Nr_Grapp_post	Numero di grappoli post-diradamento	Integer

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Perc_diradamento	Livello di diradamento espresso in %	Float
Nr_Germogli	Numero di germogli	Integer
Peso_netto_per_pianta	Peso dei grappoli alla raccolta	Float
Fertilita	Fertilità della pianta	Float
Peso_medio_grappolo	Peso medio del grappolo	Float
Peso_uva	Peso del campione di uve	Float
Peso_uva_diraspata	Peso della massa diraspata	Float
Peso_mosto	Peso del mosto ottenuto	Float
Vinaccia	Peso della vinaccia	Float
Resa_lavorazione	Peso mosto/peso uve	Float
Indice_avvizzimento	Indice che esprime avvizzimento delle uve	Float
Peso_100_acini	Peso di 100 acini	Float
Botritis_perc	Percentuale di Botritis	Float
Peronospora_perc	Percentuale di Peronospora	Float
Oidio_perc	Percentuale di oidio	Float
Invaiaura_perc	Percentuale di invaiatura	Float
Densità mosto	Densità mosto analizzato al FOSS	Float
Contenuto zuccheri	Derivato dal valore in gradi brix	Float
Brix	Zuccheri solubili (°Brix)	Float
pH	pH	Float
Ac_titolabile	Acidità totale (g/l)	Float
Ac_Tartarico	Acido Tartarico (g/l)	Float
Ac_Malico	Acido malico (g/l)	Float
Potassio	Concentrazione di potassio	Float
APA	Azoto prontamente assimilabile	Float
Antociani_tot_1	Espresso come mg/l sull'estratto da vinacce	Float
Antociani_tot_2	Espresso come mg/Kg peso fresco	Float
Polifenoli_tot_1	Espresso come mg/l sull'estratto da vinacce	Float
Polifenoli_tot_2	Espresso come mg/Kg peso fresco	Float
Legno di potatura	Peso legno di potatura	Float
Indice_Ravaz	indice di Ravaz = peso uva/peso legno	Float
Zuccheri_solubili_perc_ss	Perc. Sul peso secco mg/l	Float
Zuccheri_solubili_perc_sf	Perc. Sul peso fresco mg/l	Float
Zuccheri_strutturali_perc_ss	Perc. Sul peso secco mg/l	Float
Zuccheri_strutturali_perc_sf	Perc. Sul peso fresco mg/l	Float
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro m	Integer

TABELLA: FENOLOGIA		
NOME COLONNA	CONTENUTO	FORMATO
ID_FENOLOGIA	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
ID_Fase	Stadio di sviluppo del fenomeno fenologico	Integer

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Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer
ID_SUBCODE	Campo col subcode (usato come FK per la FASE)	Integer

TABELLA: POTATURA		
NOME COLONNA	CONTENUTO	FORMATO
ID_POTATURA	Campo contatore	Integer
ID_VIGNETO	Codice identificativo del vigneto	Text 2
FILA	Numero della fila (ex. 1,2,3,ecc...)	Integer
FORO	Numero del foro o campata (ese foro 1,2,3, ecc...)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 ad un numero massimo di piante previste nel foro)	Integer
GERMOGLI	Numero germogli	Integer
NODI_C	Nodi_C	Integer
NODI_FEMMINELLE	Nodi femminelle	Integer
NODI	Nodi	Integer
PESO_LEGNO	Peso del legno	Integer
SF	SF	Double
SF_FEMM	SF Femminelle	Double
SF_TOT	SF totali	Double
PMG	PMG	Double

TABELLA: RILIEVI_STRUMENTALI		
NOME COLONNA	CONTENUTO	FORMATO
ID_RILIEVO_STRUMENTALE	Campo contatore	Integer
ID_RILIEVO	Data del rilievo (formato YYYYMMDD)	Text 8
NDVI	NDVI	Double
CT_1	CT 1	Double
CT_2	CT 2	Double
CT_3	CT 3	Double
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: POINT_QUADRAT		
NOME COLONNA	CONTENUTO	FORMATO
ID_POINT_QUADRAT	Campo contatore	Integer
ID_RILIEVO	Data rilievo (formato YYYYMMDD)	Text 8
SF_FG_170_0	SF FG 170 0	Integer
SF_FG_170_40	SF FG 170 40	Integer

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SF_FG_170_80	SF FG 170 80	Integer
SF_FG_130_0	SF FG 130 0	Integer
SF_FG_130_40	SF FG 130 40	Integer
SF_FG_130_80	SF FG 130 80	Integer
SF_FG_90_0	SF FG 90 0	Integer
SF_FG_90_40	SF FG 90 40	Integer
SF_FG_90_80	SF FG 90 80	Integer
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: PROPRIETA_OTTICHE		
NOME COLONNA	CONTENUTO	FORMATO
ID_PROPRIETA_OTTICHE	Campo contatore	Integer
ID_Misure	Codice identificativo del tipo di misura	Integer
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
Spettro_nomefile		text_25
Note	Commenti, annotazioni	text_250
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

Categorie per le tabelle delle ANALISI IASMA

TABELLA: MISURE		
NOME COLONNA	CONTENUTO	FORMATO
ID_MISURE	Codice identificativo del tipo di misura	Integer
DESCRIZIONE	Campo descrittivo	text_100

TABELLA: MATERIALE		
NOME COLONNA	CONTENUTO	FORMATO
ID_MATERIALE	ID materiale raccolto	Integer
DESCRIZIONE	Campo descrittivo	text_250

TABELLA: DESTINAZIONE		
NOME COLONNA	CONTENUTO	FORMATO
ID_DESTINAZIONE	ID destinazione del materiale raccolto	Integer
DESCRIZIONE	Campo descrittivo	text_250

TABELLA: FENOMENO		
NOME COLONNA	CONTENUTO	FORMATO
ID_FENOMENO	Fenomeno fenologico oggetto di indagine	Integer
DESCRIZIONE	Campo descrittivo	text_100

TABELLA: FASE		
NOME COLONNA	CONTENUTO	FORMATO
ID_FASE	Stato di sviluppo del fenomeno fenologico	Integer
ID_FENOMENO	Fenomeno fenologico oggetto di indagine	Integer
ID_SUBCODE	Stato di sviluppo del fenomeno fenologico	Integer
DESCRIZIONE	Campo descrittivo	text_250

Descrizione della tabella contenente le informazioni fondamentali su: STAZIONI METEO

TABELLA: STAZIONI_METEO		
NOME COLONNA	CONTENUTO	FORMATO
ID_STAZ_METEO	Codice identificativo della stazione meteo	text_7
TIPO_STAZIONE	Master o slave	text_1
ID_VIGNETO	Codice identificativo del vigneto	text_2
FILA	Numero della fila (es. 1, 2, ecc.)	Integer
FORO	Numero del foro o campata (es. foro 1, 2, 3, ecc.)	Integer
POSIZIONE	Numero della posizione della pianta nel foro (da 1 al numero massimo di piante previste nel foro)	Integer

TABELLA: METEO_SLAVE		
NOME COLONNA	CONTENUTO	FORMATO
ID_STAZ_METEO	Codice identificativo della stazione meteo	text_8
DATA	Data di campionamento	Date
DATA_STR	Data di campionamento	text_8
ORA	Ora di campionamento	Time
TCJC	Temperatura giunto riferimento (°C)	Double
TAIR	Temperatura aria (°C)	Double
TGRAP	Temperatura interna grappolo (°C)	Double
TLEAF	Temperatura fogliare a infrarosso (°C)	Double
TSOIL1	Temperatura terreno a 30cm (°C)	Double
TSOIL2	Temperatura terreno a 60cm (°C)	Double
T_HEATSOIL1	Temperatura riscaldatore a 30cm (°C) x calcolo pot. Idrico	Double
T_HEATSOIL2	Temperatura riscaldatore a 60cm (°C) x calcolo pot. Idrico	Double
VV	Velocità del vento (m/s)	Double
BAGN_LEAF	Bagnatura fogliare (on/off)	Double
BAT	Livello della batteria	Double
RAD GRAP	Radiazione Grappolo	Double
POT_IDR_30	Potenziale Idrico a 30 cm (MPa)	Double
POT_IDR_60	Potenziale Idrico a 60 cm (MPa)	Double

TABELLA: METEO_MASTER		
NOME COLONNA	CONTENUTO	FORMATO
ID_STAZ_METEO	Codice identificativo della stazione meteo	text_8
DATA	Data di campionamento	Date

DATA_STR	Data di campionamento	text_8
ORA	Ora di campionamento	Time
ORA_STR	Ora di campionamento	text_4
TEMPMEDIA	Temperatura dell'aria (°C) media	Double
TMAX	Temperatura dell'aria (°C) massima	Double
TMIN	Temperatura dell'aria (°C) minima	Double
TDEVSTD	Temperatura dell'aria (°C) deviazione standard	Double
TULTIMO	Temperatura dell'aria (°C) ultimo valore registrato	Double
UMIDMEDIA	Umidità dell'aria (%) media	Double
URMAX	Umidità dell'aria (%) massima	Double
URMIN	Umidità dell'aria (%) minima	Double
URDEVSTD	Umidità dell'aria (%) deviazione standard	Double
URULTIMO	Umidità dell'aria (%) ultimo valore registrato	Double
PRESMEDIA	Pressione atmosferica (hPa) media	Double
PMAX	Pressione atmosferica (hPa) massima	Double
PMIN	Pressione atmosferica (hPa) minima	Double
PDEVSTD	Pressione atmosferica (hPa) deviazione standard	Double
PULTIMO	Pressione atmosferica (hPa) ultimo valore registrato	Double
RADGMEDIA	Radiazione solare globale (W/m2) media	Double
RMAX	Radiazione solare globale (W/m2) massima	Double
RMIN	Radiazione solare globale (W/m2) minima	Double
RDEVSTD	Radiazione solare globale (W/m2) deviazione standard	Double
RULTIMO	Radiazione solare globale (W/m2) ultimo valore registrato	Double
PIOGSOMMA	Precipitazioni (mm) somma nell'intervallo (5minuti)	Double
VVENT1VMED	Velocità del vento (m/s) media	Double
VDEVSTD	Velocità del vento (m/s) deviazione standard	Double
VRAF	Velocità del vento (m/s) massima o di raffica	Double
VMIN	Velocità del vento (m/s) minima	Double
VULTIMO	Velocità del vento (m/s) ultimo valore registrato	Double
DVENT1DPREV	Direzione del vento (°) prevalente	Double
DRAF	Direzione del vento (°) di raffica	Double
DULTIMO	Direzione del vento (°) ultimo valore registrato	Double
NUMCAMP	Numero di campionamenti fatti nell'intervallo considerato (5minuti)	Double
BATTERIA	Livello della batteria	Double

Descrizione della tabella contenente le informazioni fondamentali su:
VINIFICAZIONI

TABELLA: VINIFICAZIONI		
NOME COLONNA	CONTENUTO	FORMATO
ID_VIGNETO	Codice identificativo del vigneto	text_2
ID_BLOCCO	Codice identificativo del blocco	text_3
ID_PARCELLA	Codice identificativo della parcella	integer
ANNO_VENDEMMIA	Anno di vendemmia delle uve analizzate	integer
DATA	Data in cui è stata effettuata l'analisi	Date

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ID_FASE_VINIFICAZIONI	Fase in cui è stata effettuata l'analisi	text_2
ID_ANALISI_VINIFICAZIONI	Codice identificativo del tipo di analisi	Integer
CODICE_TUSCANIA	????????????????	text_8
ZUCCHERI RIDUTTORI	????????????????(g/l)	Float
ALCOOL_PROBABILE	????????????????? (%vol)	Float
ALCOOL_COMPLESSIVO	????????????????? (%vol)	Float
ALCOOL DA SVOLGERE	????????????????? (%vol)	Float
ALCOOL	????????????????? (%vol)	Float
ACIDITA_TOTALE	Acidità totale(g/l)	Float
ACIDITA_VOLATILE	Acidità volatile(g/l)	Float
ACIDO_MALICO	Acido malico(g/l)	Float
ACIDO_LATTICO	Acido L-lattico(g/l)	Float
PH	pH	Float
APA	Azoto prontamente assimilabile (mg/l)	Float
ANTOCIANI_TOTALI	Antociani totali(mg/l)	Float
POLIFENOLI_TOTALI	Polifenoli totali(mg/l)	Float
INTENSITA_COLORANTE	Intensità colorante	Float
TONALITA_COLORANTE	Tonalità colorante	Float

Categorie per la tabella: VINIFICAZIONI

TABELLA: ID_FASE_VINIFICAZIONI		
NOME COLONNA	CONTENUTO	FORMATO
ID_FASE_VINIFICAZIONI	Fase in cui è stata effettuata l'analisi	text_2
DESCRIZIONE	Campo descrittivo	text_100

TABELLA: ID_ANALISI_VINIFICAZIONI		
NOME COLONNA	CONTENUTO	FORMATO
ID_ANALISI_VINIFICAZIONI	Codice identificativo del tipo di analisi	Integer
DESCRIZIONE	Campo descrittivo	text_100

Allegato 3.3 – SPECIFICHE TECNICHE
DEL GEODATABASE TUSCANIA (LINEA
ENOLOGICA)



ALLEGATO 3.3

SPECIFICHE TECNICHE GEODB

Linea enologica (B1.1, B1.2, B4)

Consiglio Nazionale delle Ricerche
Istituto di Biometeorologia (CNR IBIMET), Via G. Caproni 8, 50145, Firenze – Italia

 Consiglio Nazionale delle Ricerche

6 dicembre 2010

STRUTTURA DEI DATABASE ENOLOGICI

I dati riguardanti le linee enologiche del Progetto di Ricerca del Consorzio Tuscania sono stati raccolti in tre database (DB) relazionali. Ogni database ha una struttura ad hoc per raccogliere i dati di una singola linea di ricerca, di cui riprende anche il nome (B1.1, B1.2, B4).

Ogni tabella del GeoDb è caratterizzata dall'aver una chiave primaria, cioè un campo o un insieme di campi che permette di identificare in maniera univoca un record nella tabella. In altre parole, nessun record nella tabella può avere il campo o un insieme di campi identificato dalla chiave primaria identico a quello di un qualsiasi altro record della stessa (vincolo di unicità). La chiave primaria per sua stessa natura non può contenere campi nulli o duplicati e può essere ad un campo o multicampo. La chiave primaria ad un campo è quella che è definita da un solo campo della tabella, ad esempio il campo *asm* della tabella *b11_asm_legenda*. Altro esempio sono i campi contatore come *id_b11_analisi_chimiche* nella tabella *b11_analisi_chimiche*. Il campo contatore è un campo particolare che si aggiorna automaticamente in maniera sequenziale ogni qualvolta alla tabella viene aggiunto un record. In tal modo si può essere certi della univocità dei valori nel campo. La chiave primaria multicampo è quella che viene identificata da più campi. Ad esempio nella tabella *b11_campioni* abbiamo la chiave primaria composta da (*vasca, data, fase, anno_vendemmia*).

Al fine di evitare ridondanze e risparmiare spazio sul GeoDB si è fatto uso delle cosiddette “categorie”. Le “categorie” sono tabelle che rappresentano sottotipi di una tabella padre. Ad esempio nella tabella *b12_vasca_tesi* la colonna *raccolta_cernita* contiene l'informazione relativa alla tipo di raccolta/cernita usata per le uve di ogni vasca. Invece di scrivere la completa descrizione del tipo di raccolta effettuata, si è adottato un codice che la identifica. Nella tabella *b12_tesi_vinif_legenda*, collegata alla tabella *b12_vasca_tesi* attraverso il campo *raccolta_cernita*, è presente la estrinsecazione di tale codice attraverso il campo descrittivo correlato.

Le strutture dei tre DB sono molto simili tra loro, tali che il linguaggio delle QUERY necessario per estrarre i dati prevede solo delle piccole variazioni per essere usato sull'uno o sull'altro DB. Le tabelle hanno come prefisso una sigla che identifica il DB di cui fanno parte; ad esempio la tabella *b11_campioni* contiene i dati dei campioni del DB B11. Le strutture dei tre DB sono state studiate affinché sia possibile rintracciare ogni informazione ritenuta necessaria relativa ad un parametro attraverso le relazioni che legano i vari elementi. Al momento i dati contenuti nei tre DB possono essere raggruppabili in quattro aree di pertinenza:

- **Le analisi svolte dai vari Partner di Progetto su uve, mosti e vini.** Le tabelle **xxx_campioni**, presenti in tutti e tre i DB, presentano un elenco di tutti i campioni analizzati dai vari Partner di Progetto. I campioni vengono identificati univocamente tramite l'utilizzo di una chiave primaria multicampo (*vasca, data, fase, anno_vendemmia*). Collegate ad esse tramite i campi che identificano i campioni, le tabelle *xxx_analisi_chimiche*, *xxx_analisi_aromi*, *xxx_microbiol_popolazioni*, *xxx_microbiol_chim*, *xxx_colore_polifenoli* contengono i risultati delle analisi svolte dai singoli Partner di Progetto.
- **Dati relativi alle vasche.** Nelle tabelle *xxx_vasca* sono censite le vasche con le necessarie informazioni associate per ogni linea di ricerca; esse sono identificate univocamente attraverso la chiave primaria multicampo costituita da *vasca* e *anno_vendemmia*. Il campo *anno_vendemmia* è stato implementato in quanto per motivi tecnici di cantina è necessario ogni anno cambiare il tipo di tesi associato alla singola vasca. Nelle tabelle *xxx_gestione_ferm_vasche* sono presenti parametri tecnici quali temperatura, densità dei mosti... relativi alle singole vasche e necessari per analisi statistiche.
- **Informazioni sulle tesi.** Le tre linee di ricerca sono caratterizzate da un rigido disegno sperimentale che lega univocamente, per ciascun anno, ogni vasca ad una determinata tesi sperimentale. Per i tre DB si può quindi ricavare la tesi sperimentale di ogni vasca e indirettamente, tramite il disegno relazionale di ciascun DB, per ogni campione analizzato.
- **Informazioni relative a replica-giorno.** In base al disegno sperimentale delle tre linee di ricerca, ogni tesi viene replicata tre volte effettuando il riempimento delle vasche con uve raccolte in tre giorni diversi, le cosiddette *replica_giorno*. Visto che le caratteristiche di tali uve influiscono sulle successive analisi, analisi relative ai grappoli raccolti sono state incorporate nei tre DB; esse sono contenute nelle tabelle *b11_asm_caratterizzazione_uva*, *b12_analisi_uva*, *b4_analisi_uva*.

Per una maggiore comprensione della struttura dei tre DB e delle relazioni esistenti fra le varie tabelle rimandiamo alla consultazione dei singoli diagrammi. Inoltre si riportano di seguito in dettaglio i contenuti e le specifiche tecniche delle tabelle incluse nei singoli DB.

STRUTTURA TABELLE LINEA B1.1

Diagramma del DB contenente i dati per la linea di ricerca B1.1

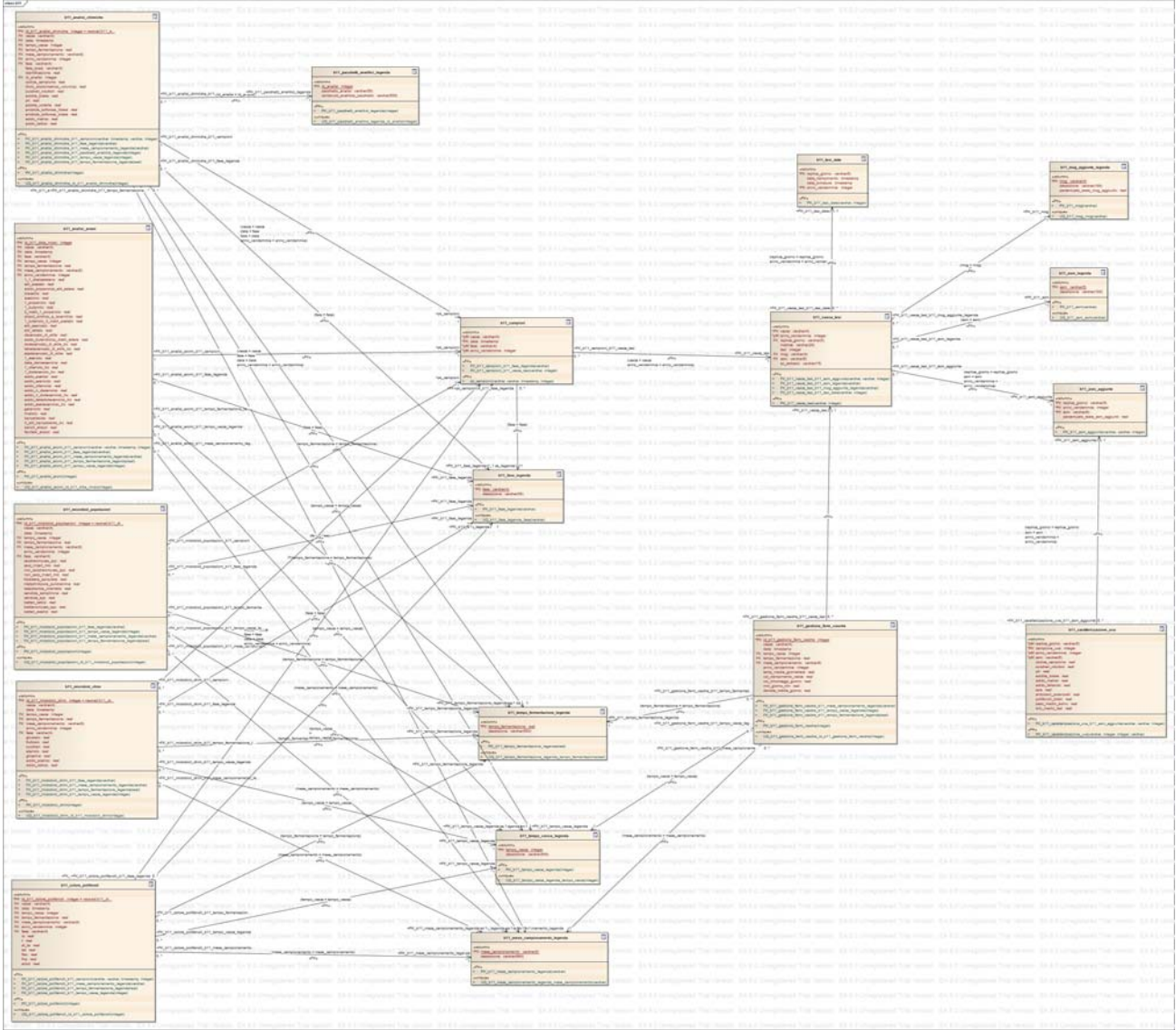


TABELLA: b11_analisi_chimiche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_analisi_chimiche	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
tempo_fermentazione	Fase di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number

anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
fase_isvea	Identificativo del campione generato dal laboratorio di analisi	Text_4
identificazione	Identificativo del campione generato al momento dell'invio del campione al laboratorio di analisi	Number
id_analisi	Gruppo tipologico di analisi chimiche	Number
codice_campione	Identificativo del campione generato dal laboratorio di analisi al ricevimento del campione	Number
titolo_alcolometrico_volumico	Titolo Alcolometrico Volumico (% vol.)	Number
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
acidita_totale	Acidità Totale (g/L acido tartarico)	Number
pH	pH	Number
acidita_volatile	Acidità Volatile (g/L acido acetico)	Number
anidride_solforosa_libera	Anidride solforosa libera (mg/L)	Number
anidride_solforosa_totale	Anidride solforosa totale (mg/L)	Number
acido_malico	Acido malico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b11_analisi_aromi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_analisi_aromi	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
1_1_dietossietano	1_1_dietossietano (ppm)	Number
etil_acetato	Etil_acetato (ppm)	Number
acido_propanoico_etil_estere	Acido_propanoico_etil_estere (ppm)	Number
diacetile	Diacetile (ppm)	Number
acetoino	Acetoino (ppm)	Number
1_propanolo	1_propanolo (ppm)	Number
1_butanolo	1_butanolo (ppm)	Number
2_metil_1_propanolo	2_metil_1_propanolo (ppm)	Number
alcool_amilico_e_isoamilico	Alcool_amilico_e_isoamilico (ppm)	Number
1_butanolo_3_metil_acetato	1_butanolo_3_metil_acetato (ppm)	Number
etil_esanoato	Etil_esanoato (ppm)	Number
etil_lattato	Etil_lattato (ppm)	Number
decanoato_di_etile	Decanoato_di_etile (ppm)	Number
acido_butandioico_dietil_estere	Acido_butandioico_dietil_estere (ppm)	Number
esadecanoato_di_etile	Esadecanoato_di_etile (ppm)	Number
1_esanolo	1_esanolo (ppm)	Number
beta_damascenone	Beta_damascenone (ppm)	Number

acido_acetico	Acido_acetico (ppm)	Number
acido_esanoico	Acido_esanoico (ppm)	Number
acido_ottanoico	Acido_ottanoico (ppm)	Number
acido_n_decanoico	Acido_n_decanoico (ppm)	Number
geraniolo	Geraniolo (ppm)	Number
linalolo	Linalolo (ppm)	Number
benzaldeide	Benzaldeide (ppm)	Number
benzil_alcool	Benzil_alcool (ppm)	Number
feniletil_alcool	Feniletil_alcool (ppm)	Number

TABELLA: b11_microbiol_popolazioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_microbiol_popolazioni	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi.	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
saccharomyces_spp	Popolazione di lieviti spp. Saccharomyces (UFC/ml)	Number
sacc_incert_mis	Incertezza di misura di spp. Saccharomyces (+/-)	Number
non_saccharomyces_spp	Popolazione di lieviti spp. NON Saccharomyces (UFC/ml)	Number
non_sacc_incert_mis	Incertezza di misura di spp. NON Saccharomyces (+/-)	Number
klockera_apiculata	Popolazione di Klockera apiculata (UFC/ml)	Number
metschnikowia_pulcherrima	Popolazione di Metschnikowia pulcherrima (UFC/mL)	Number
issatchenkia_orientalis	Popolazione di Issatchenkia orientalis (UFC/mL)	Number
candida_zemplanina	Popolazione di Candida zemplanina (UFC/mL)	Number
candida_spp	Popolazione di lieviti spp. Candida (UFC/ml)	Number
batteri_lattici	Popolazione di Batteri lattici (UFC/mL)	Number
brettanomyces_spp.	Popolazione di spp. Brettanomyces (UFC/mL)	Number
batteri_acetici	Popolazione di Batteri acetici (UFC/mL)	Number

TABELLA: b11_microbiol_chim		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_microbiol_chim	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number

fase	Fase in cui viene effettuato il prelievo del campione	Text_4
glucosio	Glucosio residuo (g/L)	Number
fruttosio	Fruttosio residuo (g/L)	Number
zuccheri	Zuccheri residui (g/L)	Number
etanolo	Etanolo (%v/v)	Number
glicerina	Glicerolo (g/L)	Number
acido_acetico	Acido lattico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b11_colore_polifenoli		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_colore_polifenoli	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
ic	Intensità colorante (a.u.)	Number
t	Tonalità	Number
al_ta	Antociani liberi più complessi tannino-antociano (a.u.)	Number
tat	Complessi tannino-antociano-tannino (a.u.)	Number
flav	Flavonoidi; (mg/l)	Number
fna	Flavonoidi non antocianici (mg/l)	Number
antot	Antociani totali (mg/l)	Number

TABELLA: b11_campioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
anno_vendemmia	Anno di vendemmia del campione analizzato	Number

TABELLA: b11_pacchetti_analitici_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_analisi	Gruppo tipologico di analisi chimiche	Number
pacchetto_analisi	Pacchetto analitico richiesto per il campione in analisi	Text_50
contenuto_analitico_legenda	Campo descrittivo	Text_500

TABELLA: b11_fase_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
fase	Fase in cui viene effettuata l'analisi	Text_4
descrizione	Campo descrittivo	Text_50

TABELLA: b11_tempo_fermentazione_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_fermentazione	Fase di fermentazione	Number
descrizione	Campo descrittivo	Text_50

TABELLA: b11_tempo_vasca_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
descrizione	Campo descrittivo	Text_50

TABELLA: b11_mese_campionamento_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
mese_campionamento	Mese del campionamento e delle relative analisi	Number
descrizione	Campo descrittivo	Text_50

TABELLA: b11_tesi_date		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_5
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
data_riempimento	Data in cui la vasca è stata riempita	Data
data_svinatura	Data in cui è stata effettuata la svinatura della vasca	Data

TABELLA: b11_vasca_tesi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
vasca	Vasca da cui viene estratto il campione	Text_4
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
replica_giorno	Replica giorno	Text_5
matrice	Matrice	Text_20
tesi	Codice identificativo della tesi	Number
mog	Material Other than Grapes (%)	Text_4
asm	Acini a Scarsa maturazione (%)	Text_6
ex_serbatoi	Campo utile solo per i pressati, in quanto indica da quali serbatoi provenivano	Text_5

TABELLA: b11_gestione_ferm_vasche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b11_parsec	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data a cui si riferiscono i parametri della vasca	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
temp_media_giornaliera	Temperatura media giornaliera (°C)	Number
vol_riempimento_vasca	Livello di riempimento della vasca all'ammostamento (Kg)	Number
vol_rimontaggi_giorno	Volume giornaliero di mosto rimontato dalla pompa (l)	Number
min_rimontaggi_giorno	Somma giornaliera dei minuti di rimontaggio del mosto	Number
densita_media_giorno	Densità media calcolata sui valori misurati al mattino e alla sera (kg/m ³)	Number

TABELLA: b11_mog_aggiunte_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
mog	Material Other than Grapes (%)	Text_4
descrizione	Campo descrittivo	Text_100
percentuale_reale_mog_aggiunto	Percentuale effettiva di uva mog aggiunto nelle tesi	Number

TABELLA: b11_asm_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
asm	Acini a Scarsa maturazione (%)	Text_6
descrizione	Campo descrittivo	Text_100

TABELLA: b11_asm_aggiunte		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_5
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
asm	Acini a Scarsa maturazione (%)	Text_6
percentuale_reale_asm_aggiunti	Percentuale effettiva di uva asm aggiunta nelle tesi	Text_100

TABELLA: b11_caratterizzazione_uva		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_5
campione_uva	Identificativo del campione	Number
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
asm	Acini a Scarsa maturazione (%)	Text_6
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
pH	pH	Number

acidita_totale	Acidità Totale (g/L acido tartarico)	Number
acido_malico	Acido malico (g/L)	Number
acido_tartarico	Acido tartarico (g/L)	Number
apa	Azoto prontamente assimilabile (mg/L)	Number
antociani_potenziali	Antociani potenziali – pH 1,0 (mg/Kg)	Number
polifenoli_totali	Polifenoli totali – pH 1,00 (mg/Kg ac.gallico)	Number
peso_medio_acino	Peso medio acino (g)	Number
brix_medio_tesi	Grado brix medio delle relative tesi asm (°Brix)	Number

STRUTTURA TABELLE LINEA B1.2

Diagramma del DB contenente i dati per la linea di ricerca B1.2

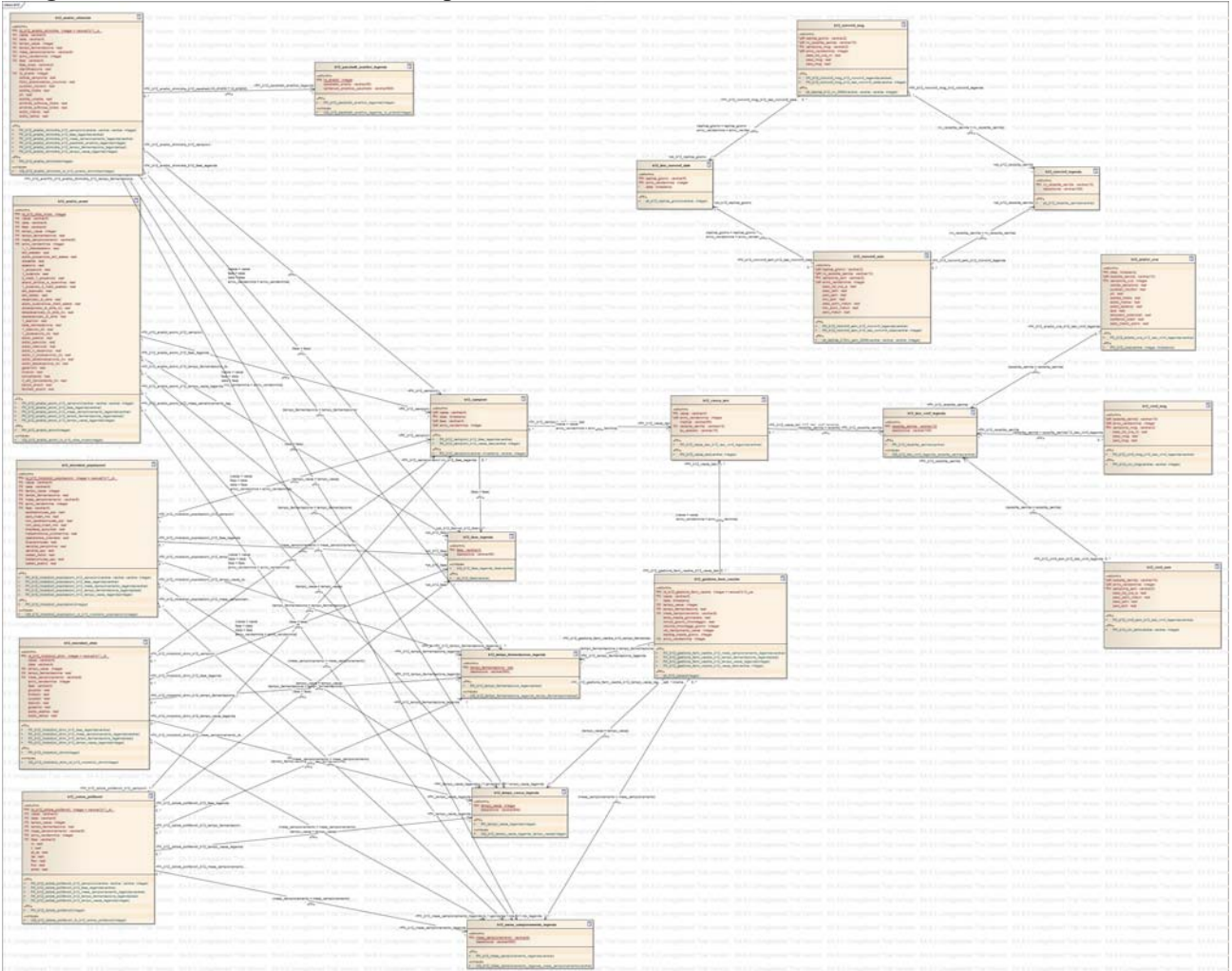


TABELLA: b12_analisi_chimiche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_analisi_chimiche	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
tempo_fermentazione	Fase di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4

fase_isvea	Identificativo del campione generato dal laboratorio di analisi	Text_4
identificazione	Identificativo del campione generato al momento dell'invio del campione al laboratorio di analisi	Number
id_analisi	Gruppo tipologico di analisi chimiche	Number
codice_campione	Identificativo del campione generato dal laboratorio di analisi al ricevimento del campione	Number
titolo_alcolometrico_volumico	Titolo Alcolometrico Volumico (% vol.)	Number
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
acidita_totale	Acidità Totale (g/L acido tartarico)	Number
pH	pH	Number
acidita_volatile	Acidità Volatile (g/L acido acetico)	Number
anidride_solforosa_libera	Anidride solforosa libera (mg/L)	Number
anidride_solforosa_totale	Anidride solforosa totale (mg/L)	Number
acido_malico	Acido malico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b12_analisi_aromi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_analisi_aromi	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
1_1_dietossietano	1_1_dietossietano (ppm)	Number
etil_acetato	Etil_acetato (ppm)	Number
acido_propanoico_etil_estere	Acido_propanoico_etil_estere (ppm)	Number
diacetile	Diacetile (ppm)	Number
acetoino	Acetoino (ppm)	Number
1_propanolo	1_propanolo (ppm)	Number
1_butanolo	1_butanolo (ppm)	Number
2_metil_1_propanolo	2_metil_1_propanolo (ppm)	Number
alcool_amilico_e_isoamilico	Alcool_amilico_e_isoamilico (ppm)	Number
1_butanolo_3_metil_acetato	1_butanolo_3_metil_acetato (ppm)	Number
etil_esanoato	Etil_esanoato (ppm)	Number
etil_lattato	Etil_lattato (ppm)	Number
decanoato_di_etile	Decanoato_di_etile (ppm)	Number
acido_butandioico_dietil_estere	Acido_butandioico_dietil_estere (ppm)	Number
esadecanoato_di_etile	Esadecanoato_di_etile (ppm)	Number
1_esanolo	1_esanolo (ppm)	Number
beta_damascenone	Beta_damascenone (ppm)	Number
acido_acetico	Acido_acetico (ppm)	Number
acido_esanoico	Acido_esanoico (ppm)	Number

acido_ottanoico	Acido_ottanoico (ppm)	Number
acido_n_decanoico	Acido_n_decanoico (ppm)	Number
geraniolo	Geraniolo (ppm)	Number
linalolo	Linalolo (ppm)	Number
benzaldeide	Benzaldeide (ppm)	Number
benzil_alcool	Benzil_alcool (ppm)	Number
feniletil_alcool	Feniletil_alcool (ppm)	Number

TABELLA: b12_microbiol_popolazioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_microbiol_popolazioni	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi.	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
saccharomyces_spp	Popolazione di lieviti spp. Saccharomyces (UFC/ml)	Number
sacc_incert_mis	Incertezza di misura di spp. Saccharomyces (+/-)	Number
non_saccharomyces_spp	Popolazione di lieviti spp. NON Saccharomyces (UFC/ml)	Number
non_sacc_incert_mis	Incertezza di misura di spp. NON Saccharomyces (+/-)	Number
klockera_apiculata	Popolazione di Klockera apiculata (UFC/ml)	Number
metschnikowia_pulcherrima	Popolazione di Metschnikowia pulcherrima (UFC/mL)	Number
issatchenkia_orientalis	Popolazione di Issatchenkia orientalis (UFC/mL)	Number
candida_zemplanina	Popolazione di Candida zemplanina (UFC/mL)	Number
candida_spp	Popolazione di lieviti spp. Candida (UFC/ml)	Number
batteri_lattici	Popolazione di Batteri lattici (UFC/mL)	Number
brettanomyces_spp.	Popolazione di spp. Brettanomyces (UFC/mL)	Number
batteri_acetici	Popolazione di Batteri acetici (UFC/mL)	Number

TABELLA: b12_microbiol_chim		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_microbiol_chim	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
glucosio	Glucosio residuo (g/L)	Number
fruttosio	Fruttosio residuo (g/L)	Number

zuccheri	Zuccheri residui (g/L)	Number
etanolo	Etanolo (%v/v)	Number
glicerina	Glicerolo (g/L)	Number
acido_acetico	Acido lattico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b12_colore_polifenoli		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_colore_polifenoli	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
ic	Intensità colorante (a.u.)	Number
t	Tonalità	Number
al_ta	Antociani liberi più complessi tannino-antociano (a.u.)	Number
tat	Complessi tannino-antociano-tannino (a.u.)	Number
flav	Flavonoidi; (mg/l)	Number
fna	Flavonoidi non antocianici (mg/l)	Number
antot	Antociani totali (mg/l)	Number

TABELLA: b12_campioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
anno_vendemmia	Anno di vendemmia del campione analizzato	Number

TABELLA: b12_pacchetti_analitici_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_analisi	Gruppo tipologico di analisi chimiche	Number
pacchetto_analisi	Pacchetto analitico richiesto per il campione in analisi	Text_50
contenuto_analitico_legenda	Campo descrittivo	Text_500

TABELLA: b12_fase_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
fase	Fase in cui viene effettuata l'analisi	Text_4
descrizione	Campo descrittivo	Text_50

TABELLA: b12_tempo_fermentazione_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_fermentazione	Fase di fermentazione	Number
Descrizione	Campo descrittivo	Text_50

TABELLA: b12_tempo_vasca_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
descrizione	Campo descrittivo	Text_50

TABELLA: b12_mese_campionamento_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
mese_campionamento	Mese del campionamento e delle relative analisi	Number
Descrizione	Campo descrittivo	Text_50

TABELLA: b12_vasca_tesi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
vasca	Vasca da cui viene estratto il campione	Text_4
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
matrice	Matrice	Text_20
raccolta_cernita	Tipo di raccolta/cernita	Text_10
ex_serbatoi	Campo utile solo per i pressati, in quanto indica da quali serbatoi provenivano	Text_5

TABELLA: b12_gestione_ferm_vasche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b12_parsec	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data a cui si riferiscono i parametri della vasca	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
temp_media_giornaliera	Temperatura media giornaliera (°C)	Number
minuti_giorno_rimontaggio	somma giornaliera dei minuti di rimontaggio del mosto	Number
volume_rimontaggi_giorno	volume giornaliero di mosto rimontato dalla pompa (l)	Number
vol_riempimento_vasca	livello di riempimento della vasca all'ammontamento (Kg)	Number
densita_media_giorno	densità media calcolata sui valori misurati al mattino e alla sera (kg/m ³)	Number

TABELLA: b12_tesi_vinif_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
raccolta_cernita	Tipo di raccolta/cernita	Text_10
Descrizione	Campo descrittivo	Text_250

TABELLA: b12_analisi_uva		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
data	Data in cui viene effettuato il prelievo	Date
raccolta_cernita	Tipo di raccolta/cernita	Text_10
campione_uva	Identificativo del campione	Number
codice_campione	Identificativo del campione generato dal laboratorio di analisi al ricevimento del campione	Number
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
pH	pH	Number
acidita_totale	Acidità Totale (g/L acido tartarico)	Number
acido_malico	Acido malico (g/L)	Number
acido_tartarico	Acido tartarico (g/L)	Number
apa	Azoto prontamente assimilabile (mg/L)	Number
antociani_potenziali	Antociani potenziali – pH 1,0 (mg/Kg)	Number
polifenoli_totali	Polifenoli totali – pH 1,00 (mg/Kg ac.gallico)	Number
peso_medio_acino	Peso medio acino (g)	Number

TABELLA: b12_vinif_mog		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
raccolta_cernita	Tipo di raccolta/cernita	Text_10
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
campione_mog	Codice identificativo del campione	Text_2
peso_tot_uva_m	Peso totale del campione (Kg)	Number
peso_mog	Peso del Material Other than Grapes (MOG) (Kg)	Number
perc_mog	Percentuale di Material Other than Grapes (MOG) (%)	Number

TABELLA: b12_vinif_asm		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
raccolta_cernita	Tipo di raccolta/cernita	Text_10
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
campione_mog	Codice identificativo del campione	Text_2
peso_tot_uva_a	Peso totale del campione (Kg)	Number
peso_acini_maturi	Peso degli acini maturi (Kg)	Number
peso_asm	Peso degli Acini a Scarsa maturazione (ASM) (Kg)	Number
perc_asm	Percentuale di Acini a Scarsa maturazione (ASM) (%)	Number

TABELLA: b12_nonvinif_mog		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_2
nv_raccolta_cernita	Tipo di raccolta/cernita	Text_10
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
campione_mog	Codice identificativo del campione	Text_2
peso_tot_uva_m	Peso totale del campione (Kg)	Number
peso_mog	Peso del Material Other than Grapes (MOG) (Kg)	Number
perc_mog	Percentuale di Material Other than Grapes (MOG) (%)	Number

TABELLA: b12_tesi_nonvinif_date		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_2
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
Data	Data in cui viene effettuato il prelievo per la tesi non vinificata	Date

TABELLA: b12_nonvinif_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
nv_raccolta_cernita	Tipo di raccolta/cernita	Text_10
Descrizione	Campo descrittivo	Text_250

TABELLA: b12_nonvinif_asm		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_2
nv_raccolta_cernita	Tipo di raccolta/cernita	Text_10
campione_asm	Codice identificativo del campione	Text_2
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
peso_tot_uva_a	Peso totale del campione (Kg)	Number
peso_asm	Peso degli Acini a Scarsa maturazione (ASM) (Kg)	Number
perc_ASM	Percentuale di Acini a Scarsa maturazione (ASM) (%)	Number
brix_ASM	Gradi brix degli Acini a Scarsa maturazione (ASM) (°)	Number
peso_acini_maturi	Peso degli acini maturi (Kg)	Number
brix_acini_maturi	Gradi brix degli acini maturi (°)	Number
perc_maturi	percentuale di acini maturi (brix > 21.5) sulle uve campionate (%)	Number

STRUTTURA TABELLE LINEA B4

Diagramma del DB contenente i dati per la linea di ricerca B4

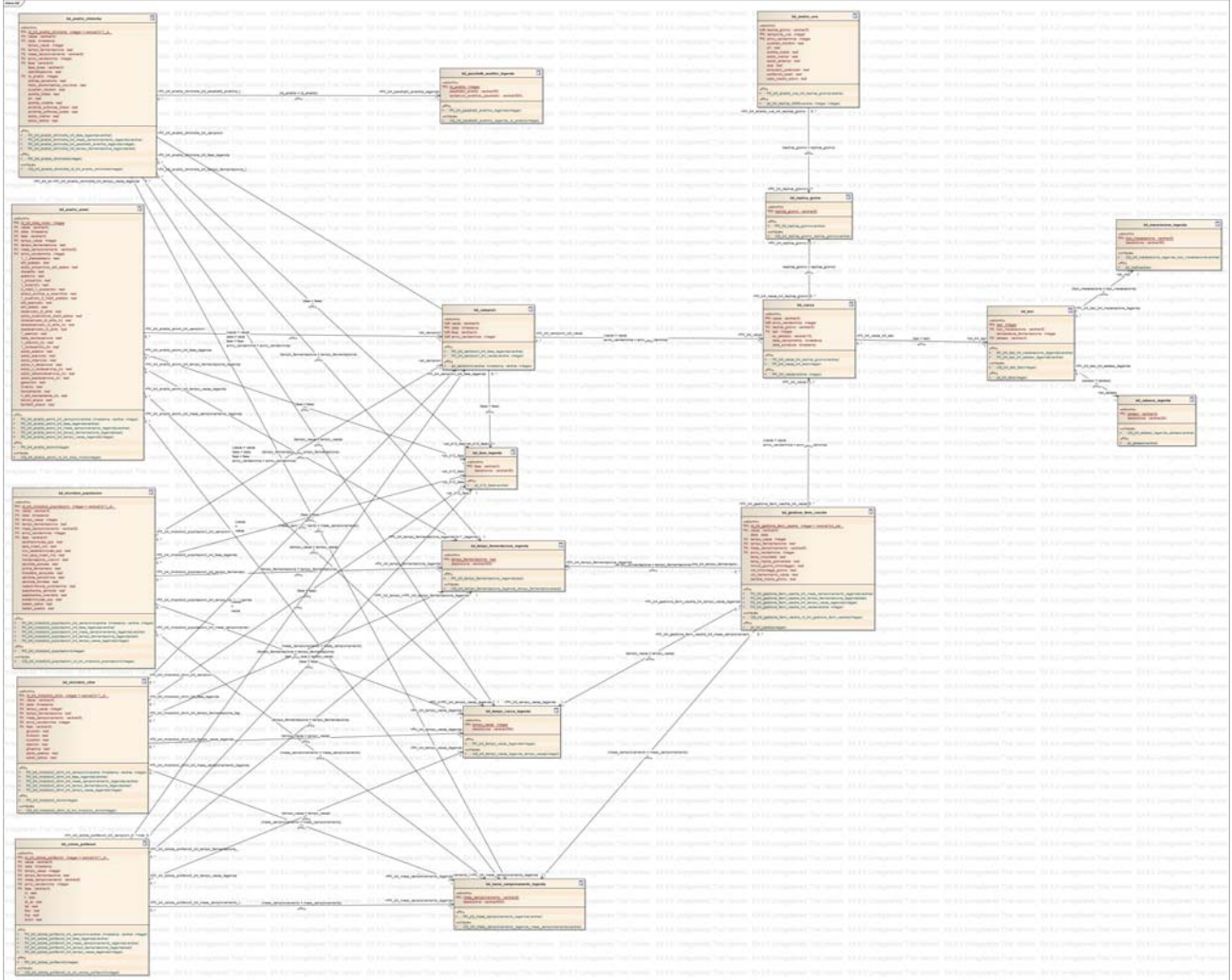


TABELLA: b4_analisi_chimiche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_analisi_chimiche	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
tempo_fermentazione	Fase di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
fase_isvea	Identificativo del campione generato dal laboratorio di analisi	Text_4

identificazione	Identificativo del campione generato al momento dell'invio del campione al laboratorio di analisi	Number
id_analisi	Gruppo tipologico di analisi chimiche	Number
codice_campione	Identificativo del campione generato dal laboratorio di analisi al ricevimento del campione	Number
titolo_alcolometrico_volumico	Titolo Alcolometrico Volumico (% vol.)	Number
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
acidita_totale	Acidità Totale (g/L acido tartarico)	Number
pH	pH	Number
acidita_volatile	Acidità Volatile (g/L acido acetico)	Number
anidride_solforosa_libera	Anidride solforosa libera (mg/L)	Number
anidride_solforosa_totale	Anidride solforosa totale (mg/L)	Number
acido_malico	Acido malico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b4_analisi_aromi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_analisi_aromi	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
1_1_dietossietano	1_1_dietossietano (ppm)	Number
etil_acetato	Etil_acetato (ppm)	Number
acido_propanoico_etil_estere	Acido_propanoico_etil_estere (ppm)	Number
diacetile	Diacetile (ppm)	Number
acetoino	Acetoino (ppm)	Number
1_propanolo	1_propanolo (ppm)	Number
1_butanolo	1_butanolo (ppm)	Number
2_metil_1_propanolo	2_metil_1_propanolo (ppm)	Number
alcool_amilico_e_isoamilico	Alcool_amilico_e_isoamilico (ppm)	Number
1_butanolo_3_metil_acetato	1_butanolo_3_metil_acetato (ppm)	Number
etil_esanoato	Etil_esanoato (ppm)	Number
etil_lattato	Etil_lattato (ppm)	Number
decanoato_di_etile	Decanoato_di_etile (ppm)	Number
acido_butandioico_dietil_estere	Acido_butandioico_dietil_estere (ppm)	Number
esadecanoato_di_etile	Esadecanoato_di_etile (ppm)	Number
1_esanolo	1_esanolo (ppm)	Number
beta_damascenone	Beta_damascenone (ppm)	Number
acido_acetico	Acido_acetico (ppm)	Number
acido_esanoico	Acido_esanoico (ppm)	Number
acido_ottanoico	Acido_ottanoico (ppm)	Number

acido_n_decanoico	Acido_n_decanoico (ppm)	Number
geraniolo	Geraniolo (ppm)	Number
linalolo	Linalolo (ppm)	Number
benzaldeide	Benzaldeide (ppm)	Number
benzil_alcool	Benzil_alcool (ppm)	Number
fenilettil_alcool	Fenilettil_alcool (ppm)	Number

TABELLA: b4_microbiol_popolazioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_microbiol_popolazioni	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi.	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
saccharomyces_spp	Popolazione di lieviti spp. Saccharomyces (UFC/ml)	Number
sacc_incert_mis	Incertezza di misura di spp. Saccharomyces (+/-)	Number
non_saccharomyces_spp	Popolazione di lieviti spp. NON Saccharomyces (UFC/ml)	Number
non_sacc_incert_mis	Incertezza di misura di spp. NON Saccharomyces (+/-)	Number
hanseniasporum_uvarum	Popolazione di Hanseniasporum Uvarum (UFC/ml)	Number
candida_sorbosa	Popolazione di Candida Sorbosa (UFC/ml)	Number
pichia_fermentans	Popolazione di Pichia Fermentans (UFC/ml)	Number
klockera_apiculata	Popolazione di Klockera apiculata (UFC/ml)	Number
metschnikowia_pulcherrima	Popolazione di Metschnikowia pulcherrima (UFC/mL)	Number
issatchenkia_terricola	Popolazione di Issatchenkia terricola (UFC/mL)	Number
issatchenkia_orientalis	Popolazione di Issatchenkia orientalis (UFC/mL)	Number
candida_zemplanina	Popolazione di Candida Zemplanina (UFC/mL)	Number
candida_formata	Popolazione di Candida Formata (UFC/mL)	Number
batteri_lattici	Popolazione di Batteri lattici (UFC/mL)	Number
brettanomyces_spp.	Popolazione di spp. Brettanomyces (UFC/mL)	Number
batteri_acetici	Popolazione di Batteri acetici (UFC/mL)	Number

TABELLA: b4_microbiol_chim		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_microbiol_chim	Campo contatore	Number
Vasca	Vasca da cui viene estratto il campione	Text_4
Data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number

Fase	Fase in cui viene effettuato il prelievo del campione	Text_4
glucosio	Glucosio residuo (g/L)	Number
fruttosio	Fruttosio residuo (g/L)	Number
zuccheri	Zuccheri residui (g/L)	Number
etanolo	Etanolo (%v/v)	Number
glicerina	Glicerolo (g/L)	Number
acido_acetico	Acido lattico (g/L)	Number
acido_lattico	Acido lattico (g/L)	Number

TABELLA: b4_colore_polifenoli		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_colore_polifenoli	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data in cui viene effettuato il prelievo	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia del campione analizzato	Number
fase	Fase in cui viene effettuato il prelievo del campione	Text_4
ic	Intensità colorante (a.u.)	Number
t	Tonalità	Number
al_ta	Antociani liberi più complessi tannino-antociano (a.u.)	Number
tat	Complessi tannino-antociano-tannino (a.u.)	Number
flav	Flavonoidi; (mg/l)	Number
fna	Flavonoidi non antocianici (mg/l)	Number
antot	Antociani totali (mg/l)	Number

TABELLA: b4_campioni		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
Vasca	Vasca da cui viene estratto il campione	Text_4
Data	Data in cui viene effettuato il prelievo	Date
Fase	Fase in cui viene effettuato il prelievo del campione	Text_4
anno_vendemmia	Anno di vendemmia del campione analizzato	Number

TABELLA: b4_pacchetti_analitici_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_analisi	Gruppo tipologico di analisi chimiche	Number
pacchetto_analisi	Pacchetto analitico richiesto per il campione in analisi	Text_50
contenuto_analitico_legenda	Campo descrittivo	Text_500

TABELLA: b4_fase_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
Fase	Fase in cui viene effettuata l'analisi	Text_4
Descrizione	Campo descrittivo	Text_50

TABELLA: b4_tempo_fermentazione_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_fermentazione	Fase di fermentazione	Number
Descrizione	Campo descrittivo	Text_50

TABELLA: b4_tempo_vasca_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno (0) di messa in vasca	Number
Descrizione	Campo descrittivo	Text_50

TABELLA: b4_mese_campionamento_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
mese_campionamento	Mese del campionamento e delle relative analisi	Number
Descrizione	Campo descrittivo	Text_50

TABELLA: b4_vasca		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
vasca	Vasca da cui viene estratto il campione	Text_4
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
replica_giorno	Replica giorno	Text_5
tesi	Codice identificativo della tesi, combinazione dei trattamenti sperimentali per la linea di ricerca	Number
ex_serbatoi	Campo utile solo per i pressati, in quanto indica da quali serbatoi provenivano	Text_5
data_riempimento	Data in cui la vasca è stata riempita	Data
data_svinatura	Data in cui è stata effettuata la svinatura della vasca	Data

TABELLA: b4_replica_giorno		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_5

TABELLA: b4_analisi_uva		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
replica_giorno	Replica giorno	Text_5
campione_uva	Identificativo del campione	Text_4

anno_vendemmia	Anno di vendemmia/sperimentazione	Number
zuccheri_riduttori	Zuccheri Riduttori (g/L)	Number
pH	pH	Number
acidità_totale	Acidità Totale (g/L acido tartarico)	Number
acido_malico	Acido malico (g/L)	Number
acido_tartarico	Acido tartarico (g/L)	Number
apa	Azoto prontamente assimilabile (mg/L)	Number
antociani_potenziali	Antociani potenziali – pH 1,0 (mg/Kg)	Number
polifenoli_totali	Polifenoli totali – pH 1,00 (mg/Kg ac.gallico)	Number
peso_medio_acino	Peso medio acino (g)	Number

TABELLA: B4_gestione_ferm_vasche		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
id_b4_parsec	Campo contatore	Number
vasca	Vasca da cui viene estratto il campione	Text_4
data	Data a cui si riferiscono i parametri della vasca	Date
tempo_vasca	Tempo progressivo, espresso in giorni, di permanenza del pigiato in vasca a partire dal giorno di messa in vasca	Number
tempo_fermentazione	Tempo di fermentazione	Number
mese_campionamento	Mese del campionamento e delle relative analisi	Number
anno_vendemmia	Anno di vendemmia/sperimentazione	Number
temp_inpostata	Temperatura originariamente impostata	
temp_media_giornaliera	Temperatura media giornaliera (°C)	Number
vol_riempimento_vasca	somma giornaliera dei minuti di rimontaggio del mosto	Number
vol_rimontaggi_giorno	volume giornaliero di mosto rimontato dalla pompa (l)	Number
minuti_giorno_rimontaggio	livello di riempimento della vasca all'ammontamento (Kg)	Number
densita_media_giorno	densità media calcolata sui valori misurati al mattino e alla sera (kg/m ³)	

TABELLA: b4_tesi		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tesi	Codice identificativo della tesi, combinazione dei trattamenti sperimentali per la linea di ricerca	Number
tipo_macerazione	Tecnica di macerazione prefermentativa	Text_5
temperatura	Temperatura di sperimentazione	Number
salasso	Salasso	Text_4

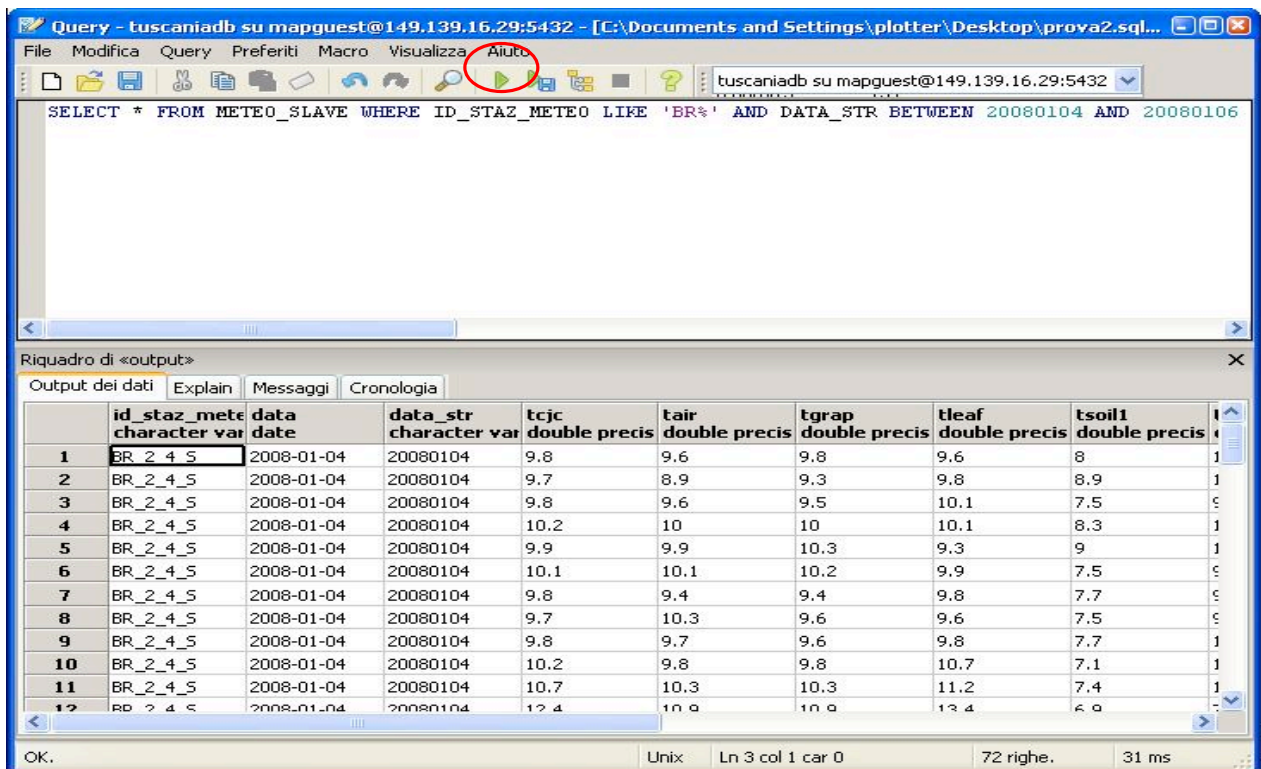
TABELLA: b4_macerazione_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
tipo_macerazione	Tecnica di macerazione prefermentativa	Text_5
descrizione	Campo descrittivo	Text_250

TABELLA: b4_salasso_legenda		
NOME COLONNA	CONTENUTO COLONNA	FORMATO
salasso	Salasso	Text_4
descrizione	Campo descrittivo	Text_250

Allegato 3.4 – RELAZIONI NEL DATABASE E INTERROGAZIONE DEI DATI (QUERY)

Il GeoDb Toscana è un database relazionale e come tale organizza i dati in tabelle, basandosi sulle relazioni fra essi. Le relazioni sono un aspetto fondamentale perché è in base ad esse che è possibile interrogare il GeoDB e poter estrarre le informazioni di cui si ha bisogno attraverso le cosiddette query. Il termine **query**, in informatica viene utilizzato per indicare l'interrogazione di un database in modo da ottenere dei dati contenuti in uno o più database. In PostgreSQL le query vengono gestite tramite il linguaggio SQL.

Di seguito (fig. 3a) è riportato un esempio di query SQL effettuata su tabelle delle stazioni meteo del DB Toscana attraverso l'interfaccia pgAdmin III di PostgreSQL.



The screenshot shows the pgAdmin III interface with a query window open. The query text is: `SELECT * FROM METEO_SLAVE WHERE ID_STAZ_METEO LIKE 'BR%' AND DATA_STR BETWEEN 20080104 AND 20080106`. Below the query window, the results are displayed in a table with the following columns: `id_staz_meteo`, `data`, `data_str`, `tcjc`, `tair`, `tgrap`, `tleaf`, and `tsoil1`. The results table contains 12 rows of data for station BR_2_4_5 on 2008-01-04.

	id_staz_meteo character var	data date	data_str character var	tcjc double precis	tair double precis	tgrap double precis	tleaf double precis	tsoil1 double precis
1	BR_2_4_5	2008-01-04	20080104	9.8	9.6	9.8	9.6	8
2	BR_2_4_5	2008-01-04	20080104	9.7	8.9	9.3	9.8	8.9
3	BR_2_4_5	2008-01-04	20080104	9.8	9.6	9.5	10.1	7.5
4	BR_2_4_5	2008-01-04	20080104	10.2	10	10	10.1	8.3
5	BR_2_4_5	2008-01-04	20080104	9.9	9.9	10.3	9.3	9
6	BR_2_4_5	2008-01-04	20080104	10.1	10.1	10.2	9.9	7.5
7	BR_2_4_5	2008-01-04	20080104	9.8	9.4	9.4	9.8	7.7
8	BR_2_4_5	2008-01-04	20080104	9.7	10.3	9.6	9.6	7.5
9	BR_2_4_5	2008-01-04	20080104	9.8	9.7	9.6	9.8	7.7
10	BR_2_4_5	2008-01-04	20080104	10.2	9.8	9.8	10.7	7.1
11	BR_2_4_5	2008-01-04	20080104	10.7	10.3	10.3	11.2	7.4
12	BR_2_4_5	2008-01-04	20080104	12.4	10.9	10.9	13.4	6.9

Fig. 3a – Finestra per l'editing della query e relativa tabella dei risultati.

Nella finestra in alto viene scritta l'interrogazione al GeoDB in linguaggio SQL. La query scritta chiede la estrazione di tutti i dati (cioè tutte le colonne) della tabella METEO_SLAVE che rispondono alla condizione di avere il campo della colonna ID_STAZ_METEO che comincia per Br e che abbiano un valore della colonna

DATA_STR compreso fra 20080104 e 20080106. In altre parole si è richiesto di estrarre i dati riguardanti le stazioni slave di Brolio compresi nel periodo fra il 4 e il 6 gennaio 2008. Nella finestra più in basso vengono riportati i risultati delle query. Il cerchio rosso segnala il pulsante che permette di salvare i risultati delle query direttamente in un file Excel.

Adesso vediamo un esempio più complesso di query che prende in considerazione relazioni esistenti fra tabelle differenti.

Nelle figura 3b è stata estratta una parte del diagramma complessivo del GeoDB Toscana dove si può vedere ad esempio le relazioni che intercorrono fra le tabelle CAMPIONI, PASSAPORTO, DATI PRODUTTIVI e VIGNETO.

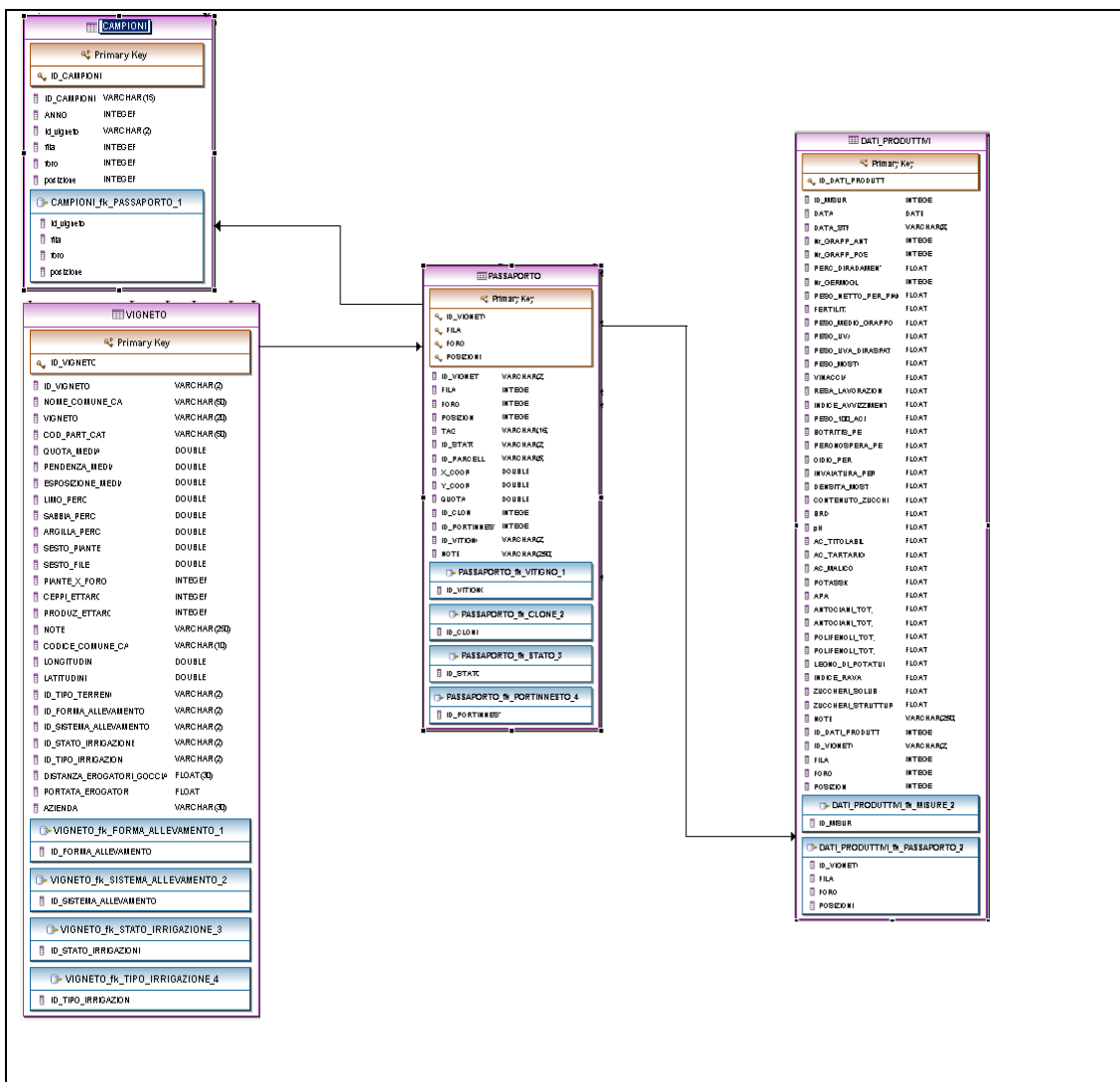


Fig. 3b – Porzione dello schema del GeoDB.

La tabella CAMPIONI, attraverso la composizione dei quattro campi ID_VIGNETO, FILA, FORO, POSIZIONE (che compongono in pratica in maniera univoca l'identificativo della posizione della pianta ID_POSIZIONE) è collegata alla tabella PASSAPORTO; lo stesso tipo di collegamento si ha fra la tabella PASSAPORTO e DATI_PRODUTTIVI; di conseguenza la tabella CAMPIONI e DATI_PRODUTTIVI sono collegate attraverso il campo quattro campi che identificano la posizione.

Mettiamo caso che siamo interessati a estrarre i dati produttivi per tutti i campioni del vigneto di Brolio per l'anno 2007. La query che scriveremo su l'interfaccia pgAdmin III di PostgreSQL è visibile in fig.3c. Chiaramente essendo ancora vuoto il database il riquadro degli output riporta solo la descrizione dei parametri selezionati ma non il loro contenuto.

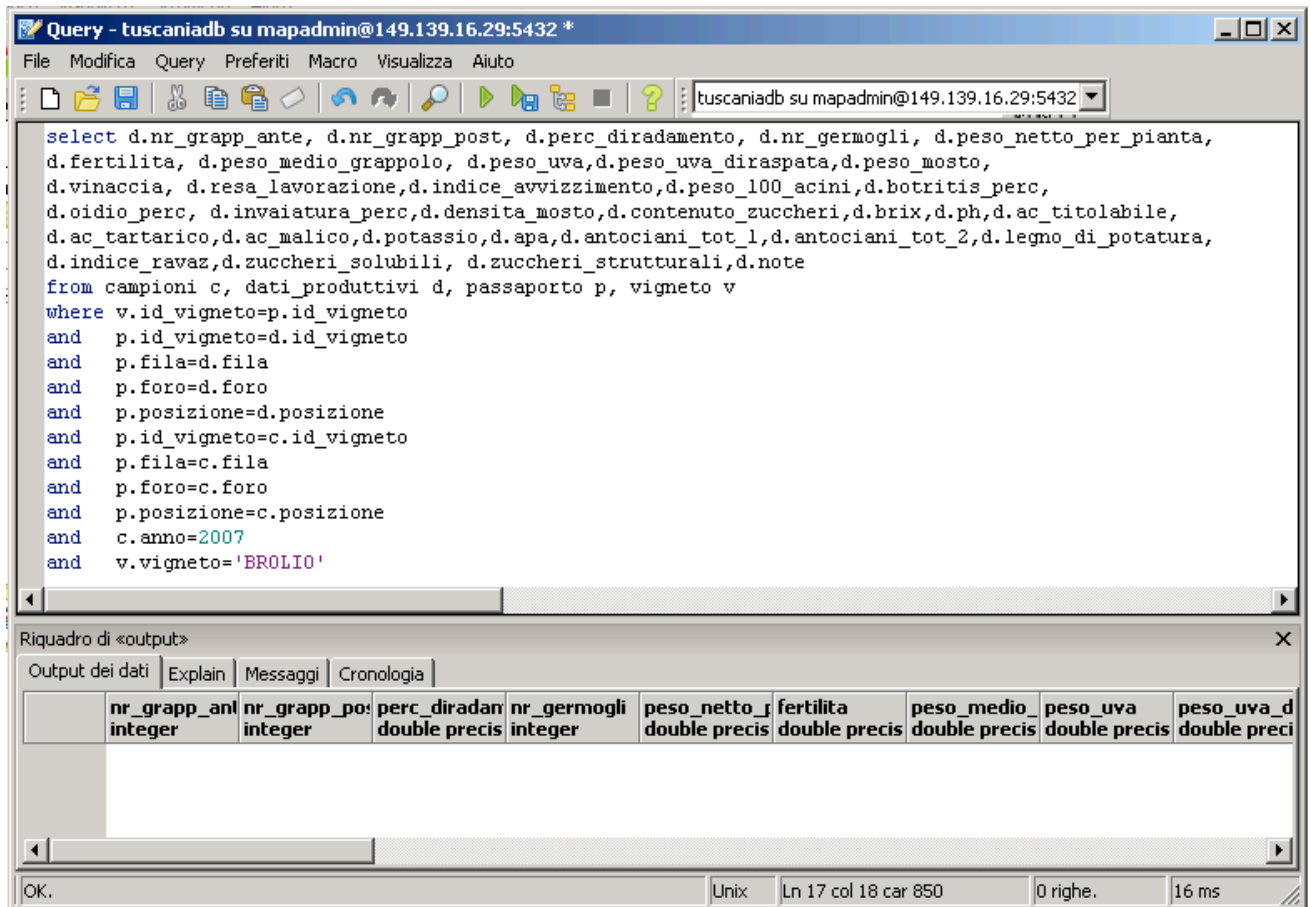


Fig. 3c – Esempio di query complessa.

Superata la fase di test del GeoDB, per agevolare l'accesso ai dati verranno fornite agli utenti del GeoDB una serie di query "preconfezionate" in linguaggio SQL per

l'interrogazione e la estrazione dei dati. Queste permetterà in potenziali sviluppi futuri l'accesso al GeoDB non in maniera diretta ma attraverso degli appositi programmi applicativi, sviluppati sempre lato server, che funzionano, per così dire, da interfaccia. Ovviamente sarà poi possibile creare anche query "ad hoc" per le specifiche necessità dei vari user.

Inserire leelenco query fornite per le analisi dei dati

Allegato 3.5 - Versione sperimentale del Thesaurus

Parametro	Descrizione
Defogliazione	Intervento svolto nel periodo vegetativo la cui importanza risiede nel miglioramento dell'esposizione, ai raggi del sole, dei grappoli attraverso l'asportazione delle foglie più vecchie.
Diradamento	Questa operazione consiste nel togliere parte dei germogli produttivi quando questi hanno raggiunto 10-20 cm di lunghezza con l'obiettivo di regolare la produzione e distribuirla in maniera omogenea ottimizzando l'equilibrio vegeto-produttivo.
Clone	Il clone in viticoltura deriva dalla moltiplicazione per via agamica di un individuo (ceppo) isolato all'interno della varietà: i discendenti del clone hanno lo stesso corredo genetico (tutti gli individui sono identici).
Portinnesto	Parte inferiore di una pianta moltiplicata con la tecnica dell'innesto. fornisce l'apparato radicale e una parte più o meno sviluppata dell'apparato caulinare.
Vitigno	Particolare varietà di uva. Si distinguono per differenti forme e colori dei chicchi di uva, del grappolo e delle foglie, oltre che per differenti periodi di maturazione e soprattutto per le diverse caratteristiche organolettiche dei vini da essi ottenuti.
NDVI	Indice usato per stimare la quantità, qualità, e sviluppo della vegetazione in base alla misura, per mezzo di sensori comunemente installati su una piattaforma spaziale, della intensità della radiazione di specifiche bande dello spettro magnetico che la vegetazione emette o riflette.
LAI	Il Leaf area index è il rapporto fra la superficie fogliare superiore della vegetazione e l'area della superficie su cui la vegetazione cresce. E' un valore adimensionale, che varia da 0 per il suolo nudo a 6 per fitte foreste.
Potenziale Idrico	Misurazione usata per valutare il generale stato idrico di una pianta. Il valore 0 indica l'assenza di stress idrico, mentre valori crescenti negativi ne certificano la presenza.
Fenologia	Osservazione della variazione delle fasi fenologiche in comunità vegetali ed animali, da porsi in relazione con i cambiamenti climatici.