



Past Experience Supports Future Choices for Cropping Systems Management:



American Society of Agronom rop Science Society of Americ Soil Science Society of Americ

The Italian Long-Term Agro-Ecosystem Experiments (LTAE) through the IC-FAR Project and the MaGGnet International Network

I. Iocola¹, D. Antichi², B. Basso³, A. Berti⁴, L. Doro⁵, C. Grignani⁶, A. R. Kemanian⁷, M. Mazzoncini², L. Mula¹, R. Orsini⁸, G. Seddaiu¹, F. Tei⁹, D. Ventrella¹⁰ and **P. P. Roggero**¹

(1) University of Sassari, Italy, email: *pproggero@uniss.it* (2) University of Pisa, Italy (3) Michigan State University, MI (4) University of Padova, Italy (5) Texas Agrilife Research-Blackland Center, Temple, TX (6) University of Turin, Italy (7) Pennsylvania State University, University park, PA (8) University of Marche, Italy (9), University of Perugia, Italy (10 CREA-SCA, Bari, Italy

Introduction

MAGGnet (Managing Agricultural Greenhouse Gases Network), formed under the GRA Croplands Research Group (Liebig et al, 2016; *Carbon Manag*), is a global network which serves as a foundation for inventory and analysis of GHG mitigation and soil C sequestration experimental research. Currently, **315 experimental studies are included in MAGGnet** (Table 1).



Table 1 – current status and duration of studies included in MAGGnet

Country		Status		Duration		
	No. of studies	Completed	Ongoing	1-3 yr	3-10 yr	> 10 yr
Argentina	10	4	6	8	1	1
Australia	16	16	0	16	0	0
Brazil	8	8	0	8	0	0
Canada	12	11	1	11	0	1
CostaRica	1	1	0	1	0	0
Denmark	5	2	3	0	0	5
Finland	12	12	0	8	4	0
France	104	90	14	95	8	1
Germany	15	15	0	14	1	0
Indonesia	2	2	0	2	0	0
Ireland	7	7	0	2	5	0
Italy	19	0	19	1	4	14
Japan	9	4	5	2	2	5
Korea	1	0	1	0	1	0
New Zealand	2	1	1	1	1	0
Spain	12	3	9	3	9	0
Switzerland	10	6	4	0	3	7
United Kingdom	36	36	0	35	0	1
United States	30	15	15	6	15	9
Uruguay	4	0	4	1	0	3
Total	315	233	82	214	54	47

The Italian IC-FAR project network (www.icfar.it) is contributing to MAGGnet with a partnership managing now 20 long-term agro-ecosystem experiments (LTAEs) in 10 different locations across the Italian peninsula and islands (Figure 1). Most of the LTAEs were started many decades ago but are still useful to support the agronomic research challenges posed by climate change (CC) (Roggero, 2016; Eur J Agron). The hypothesis underlying this effort is that such experimental networks provide a valuable source of field experimental evidence to be effectively used to support effective options for CC mitigation or adaptation through crop modelling tools.

Materials and Methods



The LTE datasets were used to understand the interactions of the long-term interactions between crop management, climate and the agroecosystem. As an example, we illustrate in this poster the methods applied to a dataset from the AN site (Seddaiu et al 2016 Eur J Agron) characterized by a rainfed two-year durum wheat -maize rotation:

Recursive Partition Analysis (RPA). The regression tree function *ctree* available in the R package was used to explore the interannual variation of crop yield as influenced by a range of meteorological and management variables. RPA explores the structure of a dataset, developing decision rules for predicting a variable. Regression trees are constructed by recursively splitting the response variable (e.g. grain yield) into two groups on the basis of the explanatory variables so as to minimize variability within a group and maximizing variability between groups. At the end, the terminal nodes (leaves) are characterized by the mean values of the response variable, each followed by a P value indicating the significance of its splitting;
Crop modeling. The LTE database was used to calibrate several simulation models (DSSAT, EPIC, SALUS, CropSyst, CSS, Cycles, RothC) to simulate the impact of CC on soil organic C (SOC).
Climate scenarios. Climate scenarios were generated applying a statistical model, based on Canonical Correlation Analysis to the predictors of CMCC-CM global model to obtain temperature and precipitation at local scale over the period 1971-2000 (ContrRun) and 2021-2050 (RCP45 and RCP85 emission scenarios). The impacts of climate change on both SOC and

crop yields were assessed using the Multi-Model (MM) Mean obtained by the ensemble of the calibrated models.

Results 0-20 cm M1_Tmean p < 0.001 \odot {N1,N2} 1.2 p < 0.001 RMSE=0.6 t/ha´M1_Tmean Carbon p < 0.001 MAE = 0.5 t/ha≤ 6.5 > 6.5 0.0 > 6.5 ≤ 6.5 nic 0.8 M5_Tmean M4_Rain n=21 **0.6** p < 0.001 p < 0.001 p < 0.001 y=1.02 0.6 % ≤ 79 > 79 ≤ 18.3 > 18.3 N2 0.4 4 M4_ET0 ContrRun-C n=18 n=24 n=21 n=18 RCP45-CT M4_ET0 20-40 cm Ó. RCP85-CT p < 0.001 y=2.35 y=2.78 y=1.67 y=1.54 p < 0.001 ContrRun-N RCP45-N1 arbon *** RCP85-NT ≤ 91.1 > 91.1 ≤ 90.5 > 90.5 0.2 0.8 n=21 n=15 n=18 n=24 **rganic** 0.0 y=4.54 y=3.20 y=3.53 0.0 **0**.4 **Figure 2** – Regression tree showing the emerging drivers of the durum wheat yield 2000 8000 6000 4000 Jul-10 Oct-96 Jul-99 Apr-02 Oct-07 Jan-94 Jan-05 **interannual variation**: meteorological variables (Tmean = mean monthly Yield (kg ha⁻¹ dry wt) temperature, RAIN = monthly precipitation, ETO = cumulate monthly reference -NT + weeds -NT = CT - obs = NT - obsevapotranspiration; M[n] = month from 1=January to 5=May) and N fertilization rate **Figure 4 – Cumulative probability distributions** of **MZ yields** in the AN site LTE obtained using the (NO -0, N1-90, N2- 180 N kg ha-1). n = number of observations and y = mean grainFigure 3 – Observed and simulated SOC using DSSAT model in the 0-20 and 20-40 **MM Mean** (DSSAT, EPIC, SALUS) under the future yield, t ha⁻¹) in each terminal node. cm soil layers as influenced by crop management with 90 kg N ha⁻¹. Points denote climate change scenarios (RCP45 and RCP85) for The most relevant management factor influencing wheat yield was **N fertilization**. measured data and bars the standard error. **CT = 40 cm inversion tillage**, **NT= No Till**, 2021-2050 compared to the 1971-2000 baseline The interannual wheat yield variability was constrained by the low temperatures in **NT + weeds = NT including the contribution of weed plants** during wheat-maize period (ContrRun). No Tillage (NT) is shown by red the early growth stages, in relation to the beneficial effects of low T on tillering, and intercropping, simulated by DSSAT as bahia grass. by the water stress constraining grain numbers and weight in spring. lines, Conventional tillage (CT) by black lines.

Conclusions

We reported an example of how LTAE data can contribute to a robust assessment of the long term effects of cropping systems on crop yield and soil fertility. The MAGGnet and ICFAR LTAE network data are an invaluable source of information for site-specific assessments of long term soil C dynamics and GHG emissions. LTAE data coupled with crop models can be contribute to develop adaptive responses and mitigation options on crop management addressing the resilience issues associated to food security and environmental sustainability posed by CC.