



**DIRETTA
STREAMING
15 / 16
DICEMBRE
2021**



MID TERM CONFERENCE

LIFE AGRESTIC

Reduction of Agricultural Greenhouse gases
Emissions Through Innovative Cropping systems



Il progetto LIFE AGRESTIC
ha ricevuto finanziamenti
dal Programma LIFE
dell'Unione Europea

Il consorzio

Coordinatore:

HORT@
— From research to field —

Partner:



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Sant'Anna
Center of Advanced Studies - Pisa



1.

Modelling the GHGs emissions from agricultural soils.

Marco Acutis University of Milano

- What GHGs are involved ?
- Drivers of soil emissions.
- Models for GHGs simulation
- Performances and applicability of the GHG models
- Conclusions

gruppo tecniche nuove
agricole



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What GHGs are involved ?*

Agricultural soil emits:

CO₂ from root respiration and root, anaerobic and aerobic microbial respiration. In average roots emit 50% of CO₂ (range: 10-90%)

CH₄ in soils is produced by methanogenesis under anaerobic conditions and is consumed by methanotrophic microorganisms that use O₂ and CH₄ for their metabolism under aerobic conditions. CO₂ equivalent: 28-30.

N₂O releases are driven by nitrification (oxidation of NH₄⁺ to NO₃⁻ via NO₂⁻) and denitrification (reduction of NO₃⁻ to N₂O and N₂). CO₂ equivalent: 250-300.

NO_x, but they are not GHG



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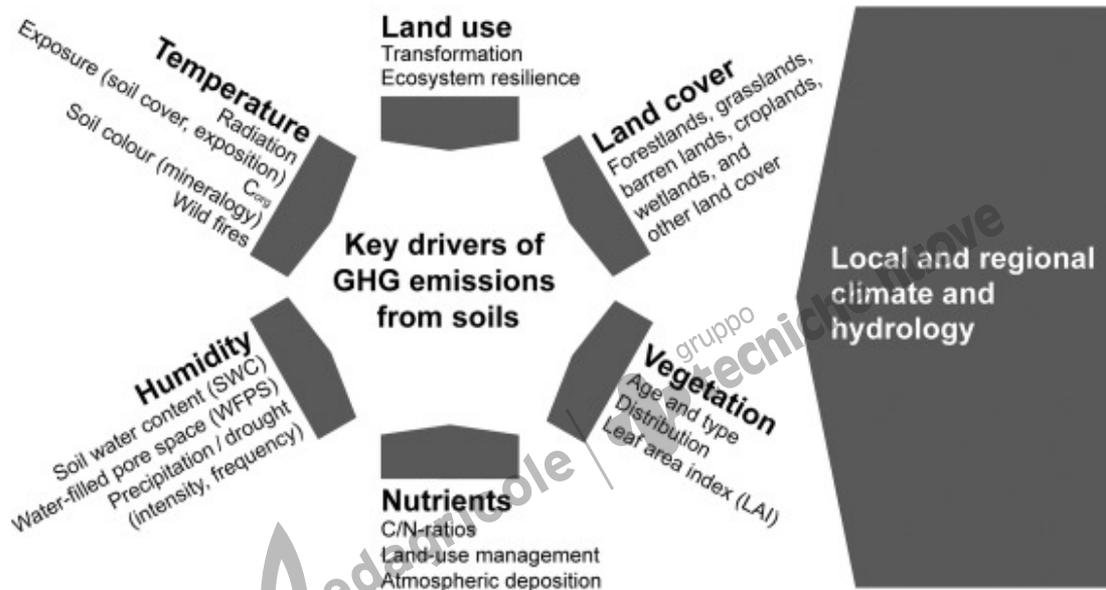
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Drivers of soil emissions



*Oertel, C., Matschullat, J., Zurba, K., Zimmermann, F., Erasmí, S., 2016. Greenhouse gas emissions from soils—A review. *Geochemistry* 76, 327–352. <https://doi.org/10.1016/j.chemer.2016.04.002>



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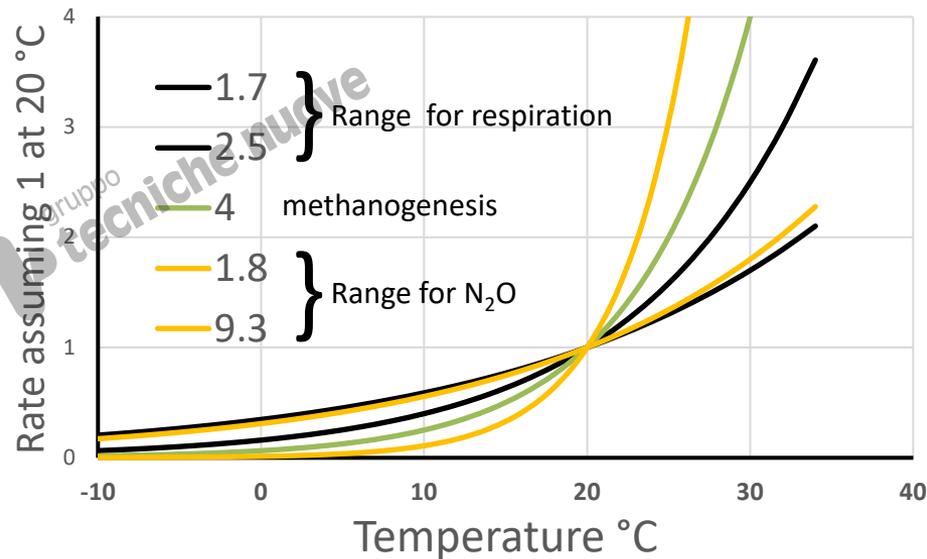
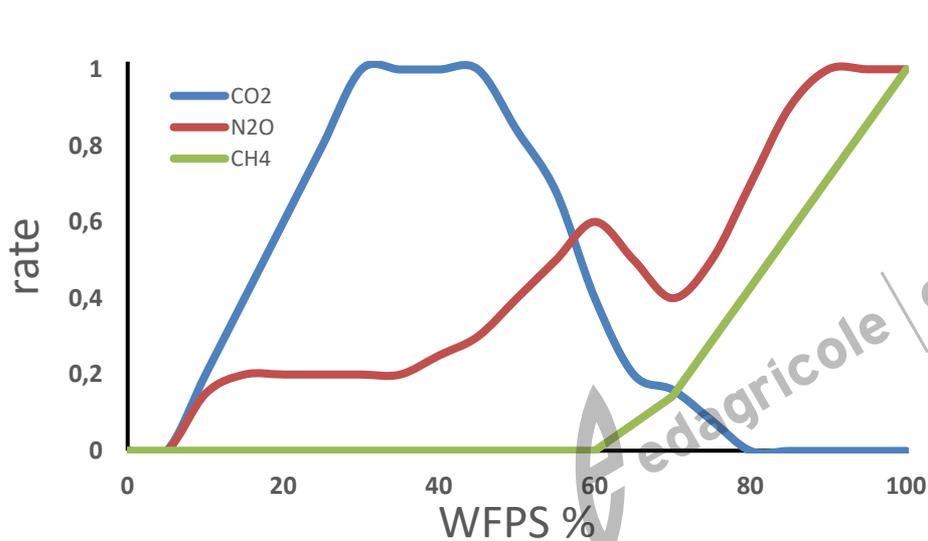


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Drivers of soil emissions (2)

Main drivers are the Soil Water content (expressed in term of Water Filled Pore Space) and temperature



Drivers of soil emissions (3)

Other drivers are absolutely important:

Fertilization (organic and mineral)

Amount of soil organic matter and its change at short and long time

C/N ratio

Tillage (conservation agriculture)

Residues (mulch)

Soil pH

Liming

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Models for GHGs simulation

The IPCC approach (based on EF, it is not really a model)

TABLE 11.1 (UPDATED)
DEFAULT EMISSION FACTORS TO ESTIMATE DIRECT N₂O EMISSIONS FROM MANAGED SOILS

Emission factor	Aggregated		Disaggregated		
	Default value	Uncertainty range	Disaggregation ⁴	Default value	Uncertainty range
EF ₁ for N additions from synthetic fertilisers, organic amendments and crop residues, and N mineralised from mineral soil as a result of loss of soil carbon ¹ [kg N ₂ O-N (kg N) ⁻¹]	0.010	0.001 – 0.018	Synthetic fertiliser inputs ⁵ in wet climates	0.016	0.013 – 0.019
			Other N inputs ⁶ in wet climates	0.006	0.001 – 0.011
			All N inputs in dry climates	0.005	0.000 – 0.011

A large set of coefficients are reported in table to estimate yield, N contents in plant residues, etc..

Models for GHGs simulation

Process based model approach

- process-based models predict daily to weekly fluxes, important for designing and management interventions to mitigate emissions.
- however, they are complex and heavily rely on site and version-specific parameterizations that are sometimes ad hoc tunings.
- algorithms of popular biogeochemical models (DayCent, DNDC, EPIC) are usually derived from laboratory-based responses to individual environmental factors.



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Models for GHGs simulation

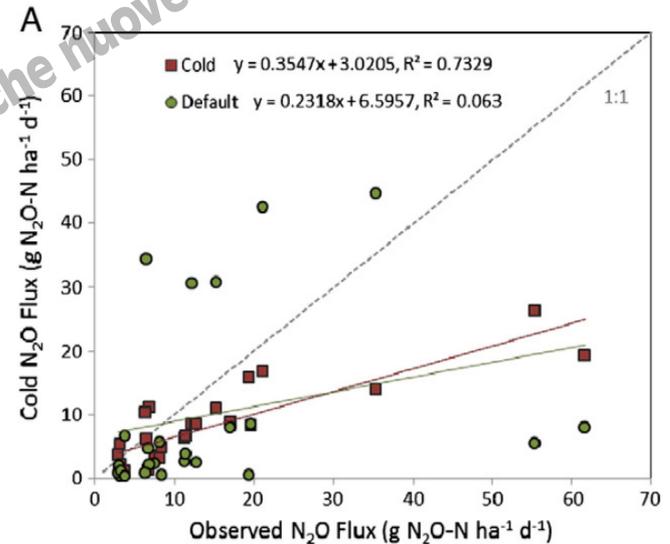
What models are available able to simulate at the same time CO₂, N₂O and CH₄ in agricultural systems ?

- DayCent: "weak" in water dynamic simulation
- DNDC: "weak" in water dynamic simulation, crop simulation, tillage.
- EPIC: "weak" in water dynamic simulation.
- Animo: "weak" in usability (3 model have to run together), tillage, excellent in hydrology
- DSSAT: modular model that can include models for GHG
- A lot of other models can simulate 2 out of 3 GHG e.g.: Stics, Salus, Apsim, Armosa, water and nitrogen management model (WNMM) from China, RZWQM2 and at least 30 others

Performances and applicability of the GHG models

All the actual GHG models need strong calibration on high quality data.

- CO₂ is the GHG easier to measure, and models offer good reliability (but CO₂ emissions from the soil are only a part of the carbon balance).
- When calibrated, all the models offer reasonably good simulations, at least useful to compare cropping systems and for regional scale, also in future climate
- Simulation performance are variable even after calibration, in different experiments
- Correct prediction of WFPS is the key point



- Default = uncalibrated
- Cold = after 47 parameters calibration

Models for GHGs simulation. Statistical model – machine learning

- When a large dataset of measured data is available, few predictors can give good estimate of N₂O emissions when a ML algorithm is used.
- It is possible also to couple a process based model to obtain some variables used after in the machine learning process.

Input	Full ML model	Coupled model	Possible to estimate
Water filled pore space	*	*	model
Ammonium N content		model	model
Nitrate N content		model	model
Clay	*	*	*
Soil organic carbon	*	*	model
Days after fertilization	*	not needed	not needed
N application rate	*	*	*
Mean daily temperature	*	*	*
cumulative rain in the last 2 days	*	*	*

*= measured data

D.Saha, B. Basso, G. P. Robertson (2021) Machine learning improves predictions of agricultural nitrous oxide (N₂O) emissions from intensively managed cropping systems. Environ. Res. Lett. 16



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CONCLUSIONS

We have the need to estimate the emission from agricultural soil.

- When we project new cropping systems, the risk is to store more carbon in soil but to emit more N_2O .
- Models are the best resources that we have to planning cropping systems.
- The quality level of the simulation of the available models just sufficient when there is the support of excellent experimental data.
- Emission factor based prediction are totally inadequate (good only for large scale inventories).
- Machine learning is really promising, but there are few applications.
- There is the need of more research, based on shared approach with stakeholders