

Personalizing Sustainable Agriculture with Causal Machine Learning

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Climate Change AI

Agriculture: Modern Challenges

Ever-increasing demand for
agricultural products



Climate change,
Environmental degradation



“Sustainable Intensification”

Agriculture: Modern Potential

Crops naturally capture CO₂
from the atmosphere and
store it in soil



Maximizing soil carbon
sequestration can enlarge the
global carbon sink



Carbon Farming

A diverse world

Differences in climate,
soil, land use



Different environmental
responses to interventions
carried out by farmers



Effect Heterogeneity

Diverse policy measures (EU)

The new common agricultural policy: 2023-27

The new common agricultural policy will be key to securing the future of agriculture and forestry, as well as achieving the objectives of the European Green Deal.



On 2 December, 2021, the agreement on reform of the common agricultural policy (CAP) was formally adopted. The new legislation, which is due to begin in 2023, paves the way for a fairer, greener and more performance-based CAP.

It will seek to ensure a sustainable future for European farmers, provide more targeted support to smaller farms, and allow greater flexibility for EU countries to adapt measures to local conditions.



Targeted support



Flexibility to adapt measures to local conditions



Geospatial “personalization”

What we need?

“Sustainably Intense” Agricultural Planning

Farming context



Causal Machine Learning

Personalized advice on sustainable,
most impactful policy measures

Ecosystem Services
GHG emissions

Soil Organic Carbon (SOC)



Crop rotation
No tillage

Eco-friendly practices



Farming context

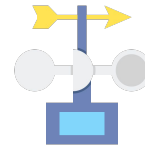


What data do we have for this problem?

Data

Earth Observation data are constantly improving in terms of quality and volume

Coupled with other geospatial data sources they provide useful agro-environmental info



Soil
Climate
Practices
Reanalysis Data
Farmer Declarations

Conditional Average Treatment Effects (CATEs)

What is the impact of a **treatment** for a **unit** with particular **characteristics**?

Eco-Friendly Practice



Field



Agro-environmental info



$$\theta(x) = \mathbb{E}[Y(1) - Y(0) | X = x]$$



Potential SOC when
practice is applied



Potential SOC when
practice is not applied

Outcome Y: Soil Organic Carbon Content (field-level)

Treatment T: Eco-friendly practice (field-level)

Characteristics X: Crop, soil, climate info (field-level)*

Double ML (Chernozukov et. al, 2016)

- Flexible framework for CATE estimation
- Robust for spatial data¹

$$Y = \theta(X) \cdot T + g(X) + \varepsilon$$

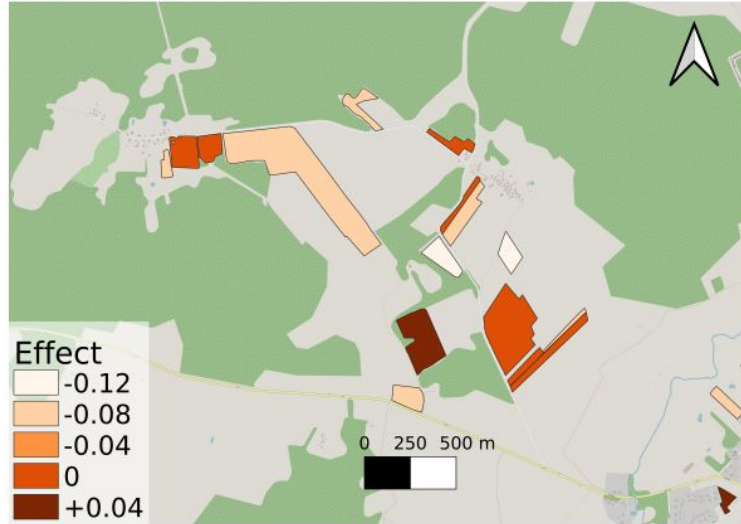
$$T = f(X) + \eta$$

$$\hat{\theta} = \arg \min_{\theta \in \Theta} \mathbb{E}[(\tilde{Y} - \theta(X) \cdot \tilde{T})^2]$$

¹Approaches to spatial confounding in geostatistics, Gilbert et al., 2022

Preliminary Results

(Lithuania, 2017-2021)



Color coded CATE point estimates for selected fields

Conditional Average Treatment Effects

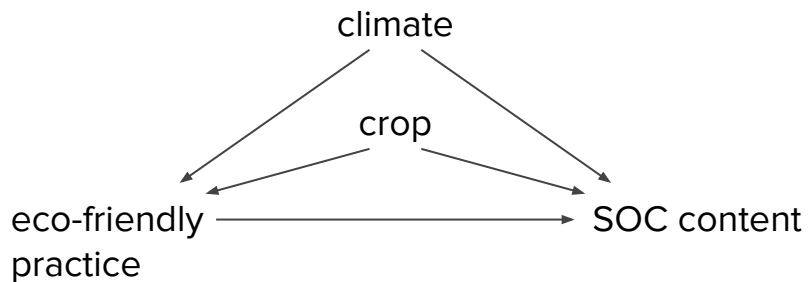
Crops	PP	SP	WW
Point Estimate	0.06	-0.08	-0.09
Standard Error	0.14	0.17	0.28
Z-score	0.41	-0.44	-0.32
P-value	0.69	0.66	0.75
95% CI (Lower)	-0.22	-0.42	-0.63
95% CI (Higher)	0.34	0.26	0.45

CATE point estimates and associated uncertainty

PP = Perennial Pastures, SP = Simple Pastures, WW = Winter Wheat

The impact of eco-friendly practices on SOC content (WIP)

Limitations & Future Work



The causal graph assumed thus far is simplistic, bias in estimates exists

Slowly changing soil properties require the expansion of the time window considered

Did farmers self-select into the treatment?
Interference or unobserved confounding?

Robustness checks and sensitivity analyses should be performed for validating results



Conclusions

Proposing causal machine learning for efficiently uplifting green metrics and personalizing sustainable agriculture

Ever-increasing volume and variety of Earth Observation data enables extraction of actionable insights on large scales

But: be wary of limitations - how to better control sources of bias?

An aerial photograph of a landscape featuring a large, dark, irregularly shaped lake or reservoir in the center. The surrounding land is divided into numerous rectangular and polygonal plots of varying colors, including shades of green, brown, tan, and grey, suggesting agricultural fields or different land uses. The overall pattern resembles a complex mosaic or patchwork. A semi-transparent dark horizontal band runs across the middle of the image, serving as a background for the text.

Thank you!

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