

# A multi-dimensional Sentinel-based **Soil Monitoring Scheme (S2MoS)** for soil clay content estimation

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SSS10.1: Digital soil mapping meets remote sensing for soil monitoring and assessment



EuroGEOSS Showcases: Applications Powered by Europe



## OVERARCHING OBJECTIVE

### **multi-input Sentinel-based Soil Monitoring Scheme (S2MoS)**

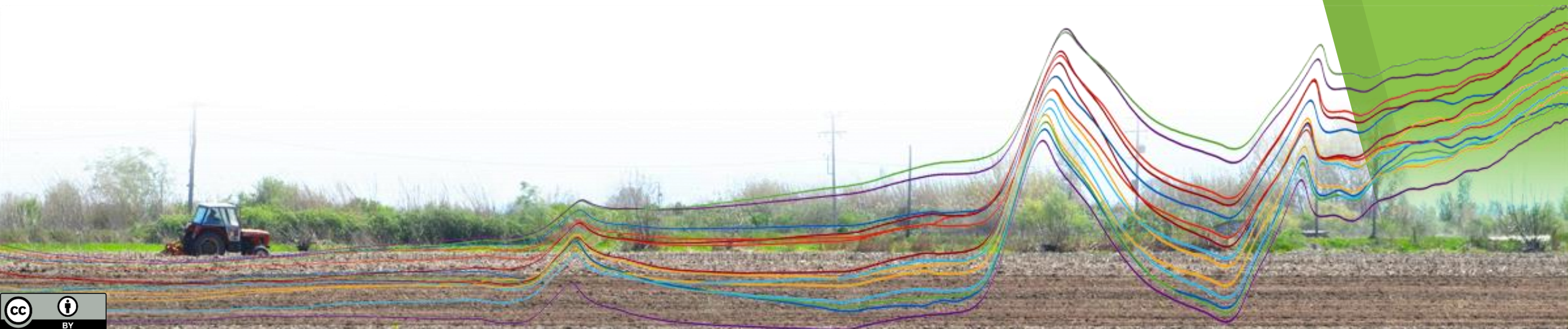
- open-access EO data
- cloud computing

=> for topsoil monitoring



### **Key novelties**

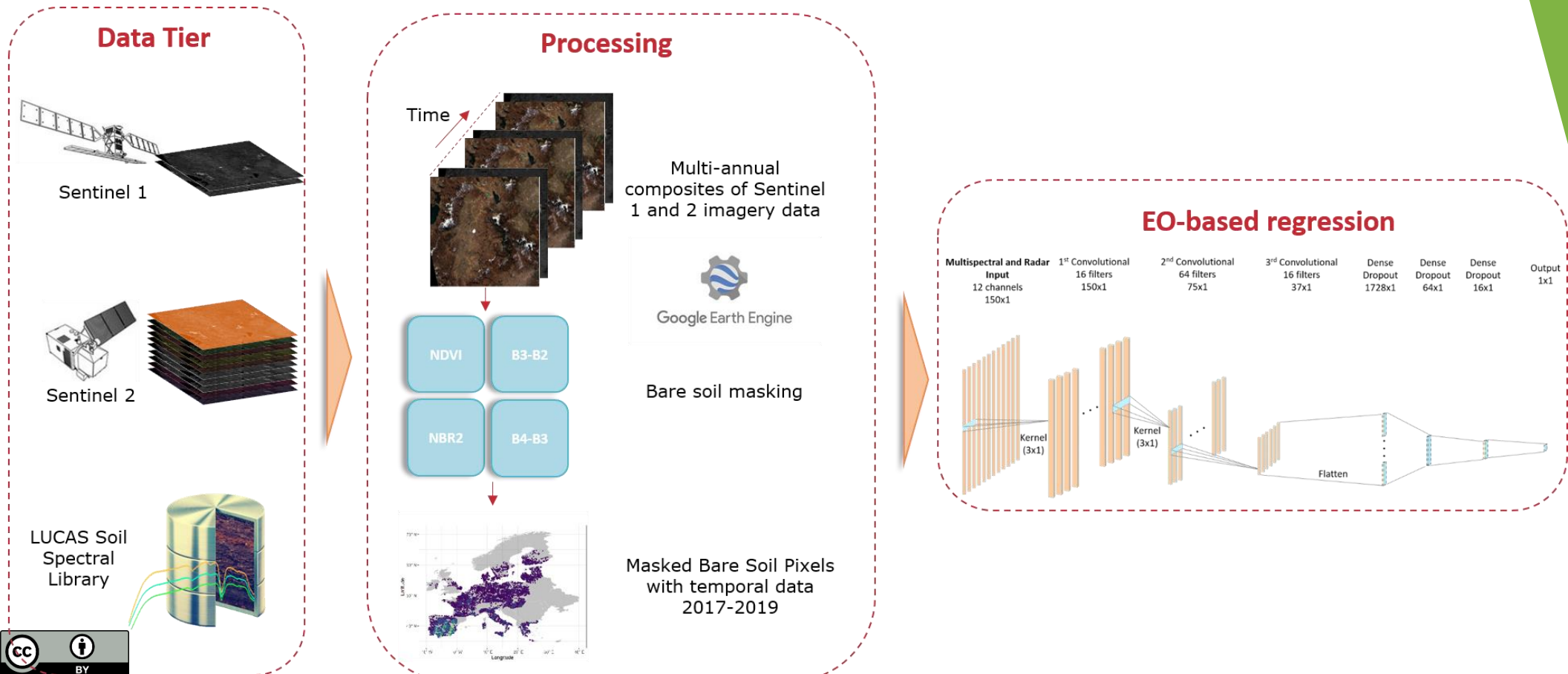
1. Synergistic use of SAR (Sentinel-1) and optical images (Sentinel-2)
2. Multi-temporal analysis (inter- and intra-annual changes)
3. Convolutional Neural Networks (CNNs) to produce fine resolution soil maps



# METHODOLOGICAL APPROACH

**Data Processing** to create multispectral and radar back scattering time – series of masked bare soil pixels

**EO-based regression** analysis supported by Convolutional Neural Networks



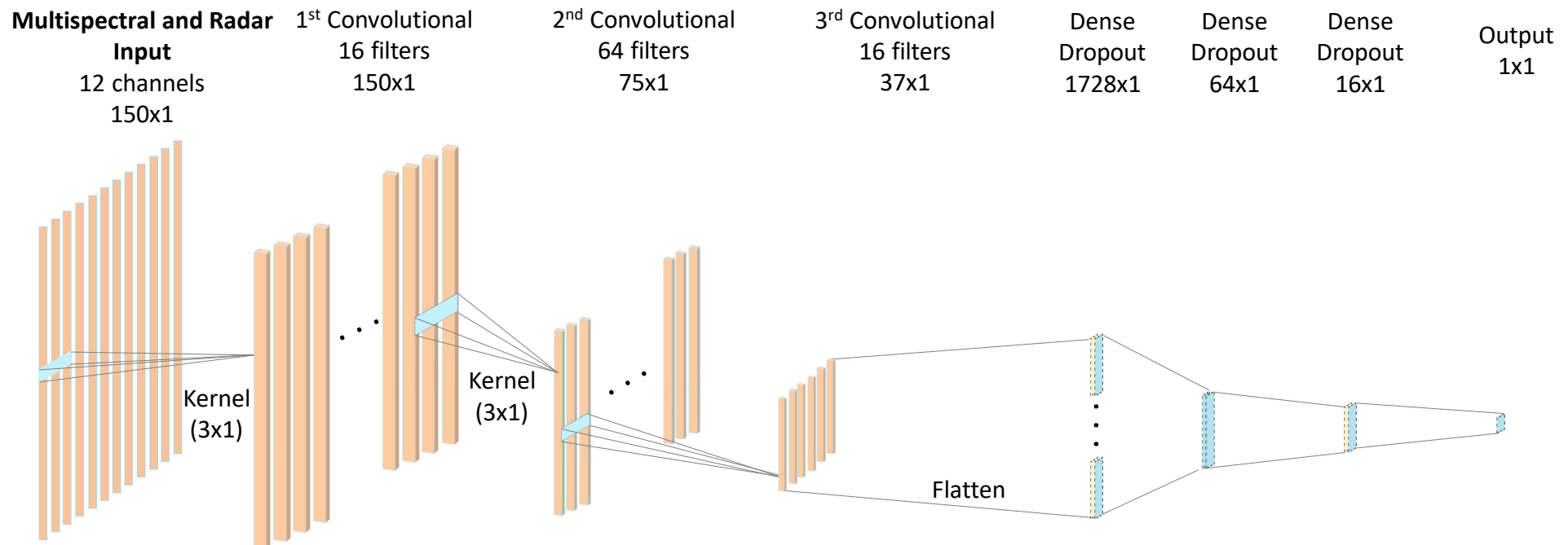
# EARTH OBSERVATION DATA AND REFERENCE SOIL DATABASES

- ~ **42,113** surface reflectance images (2017–2019)
- ~ **79605** VV and VH data; H corresponds to horizontal and V to vertical polarization (2017–2019)
- Agricultural soil samples from the **LUCAS 2009** topsoil database — consistent soil spectral library



## EO- BASED REGRESSION WITH CNNs

- Deep learning: **1-Dimension Convolutional Neural Networks** using as predictors **both** radar and multispectral imagery data of bare soil pixels
- Evaluate the complementary information of inter- and intra-annual spectral variations in the chemometric modeling
- A combination of various layers to filter a given input and extract information from specific spectral regions



# BARE SOIL MASKING RESULTS ON CLOUD

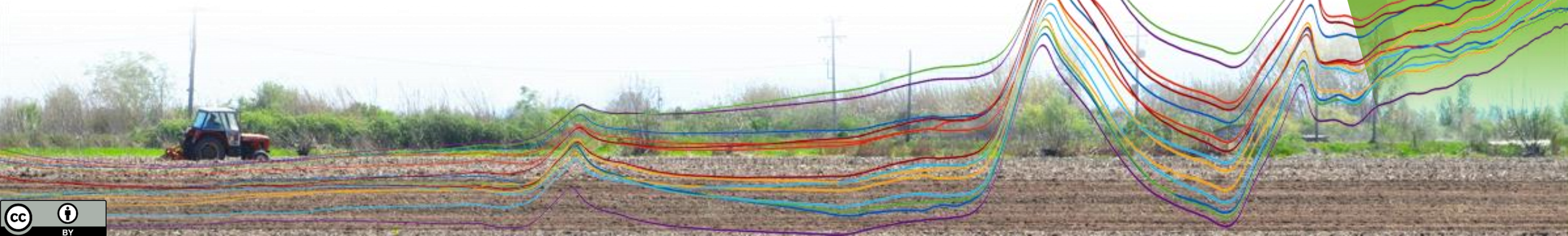
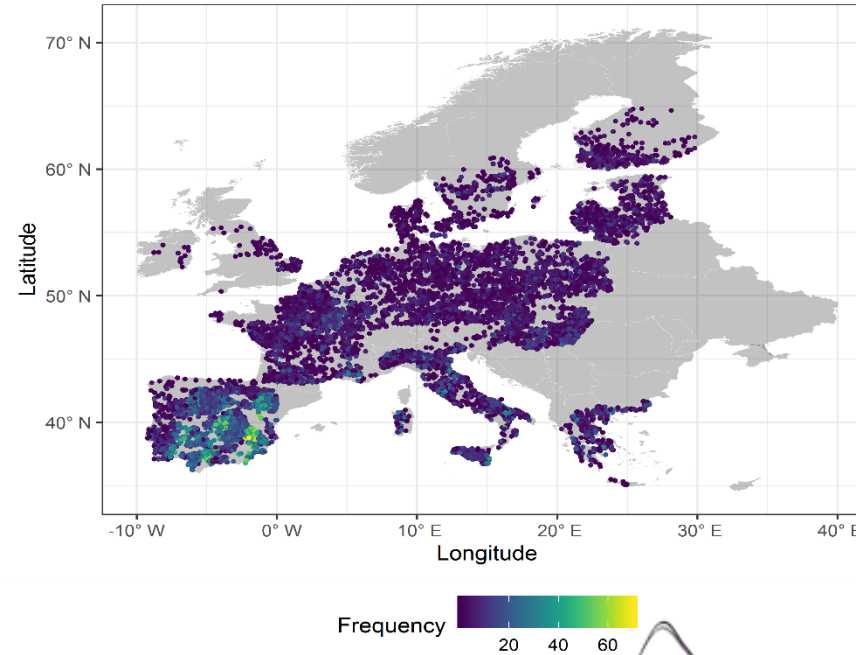


Google Earth Engine

Google Earth Engine provides a **cloud-based contribution to Earth Observation for bare soil masking**

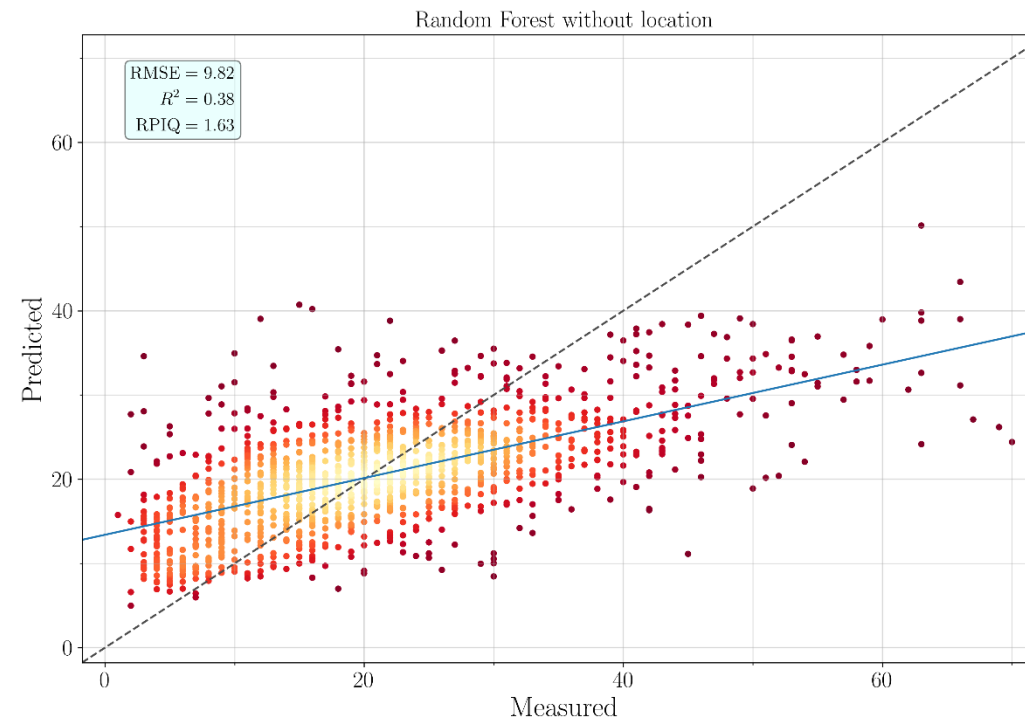
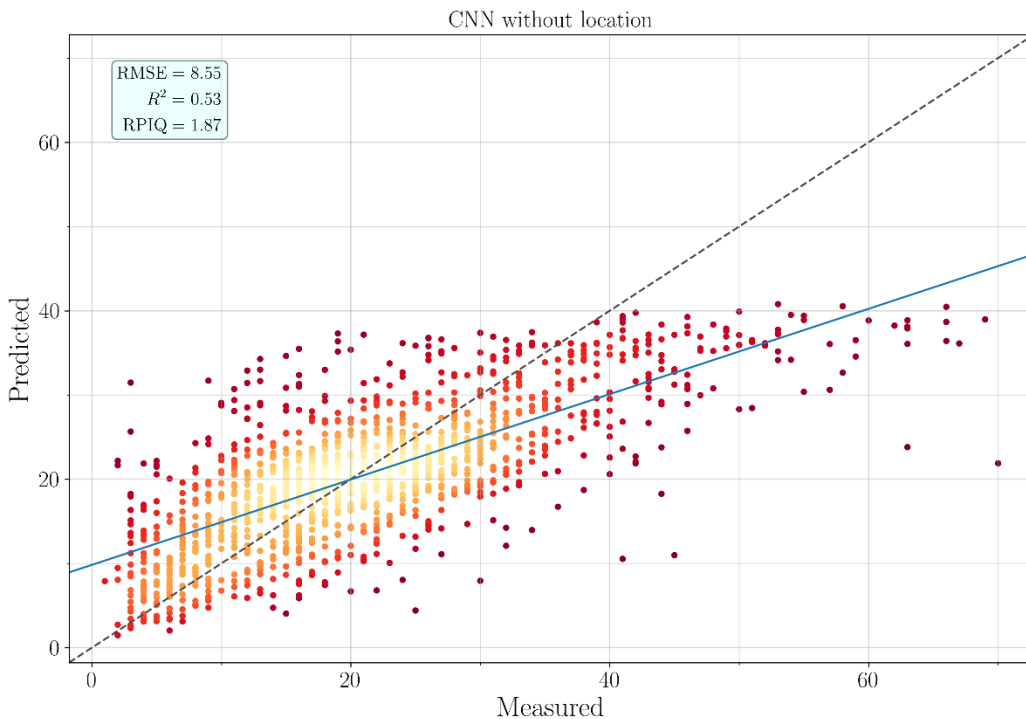
**Bare soil masking** based on normalized difference vegetation index (NDVI), normalized burn ratio index (NBR2) as proposed by [Dematte et.al 2018](#)

Selected Bare soil pixels for 2019



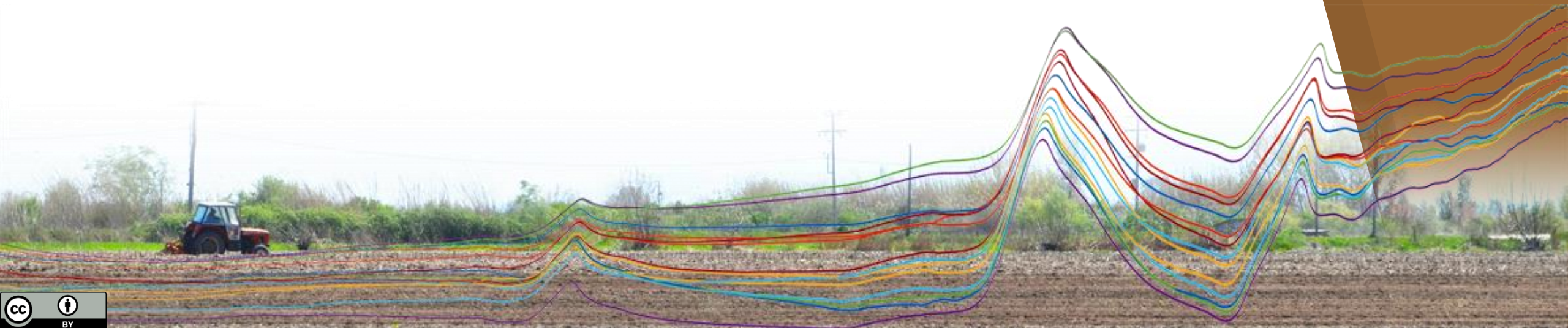
## PRELIMINARY RESULTS AND COMPARISON WITH CURRENT STATE OF THE ART

A higher prediction performance ( **$R^2=0.53$** ) was achieved by the inclusion of both types (SAR and optical) of observations using the convolutional neural network model, demonstrating a significant improvement of in overall accuracy compared to the RF using the multi-year median optical composite



## CONCLUSIONS AND FUTURE STEPS

- Evaluate a new a multi-input architecture's potential to **combine effectively** the complimentary information contained in the pool of both **optical and radar spectral** information and those from auxiliary variables ([Tziolas et.al. 2020](#), MDPI)
- Utilize upcoming release of **LUCAS version of 2015** as well as new efforts in Africa and **Latin America** based on agreed set of harmonization principles that will allow us to have better chance of chemical attributes estimation within an intra-annual calibration of models



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**This project has received funding from the European Union's Horizon 2020 research and innovation programme under Grant Agreement No 820852**