Theme 2 | Advances in soil mapping and monitoring

Mapping different organic carbon fractions in soils affected by recurrent forest fires through an upscaling multispectral UAV - satellite imagery.

Salgado, L<sup>1,2\*</sup>, Rodríguez-Gallego, J L<sup>2</sup>, Colina, A<sup>2, 3</sup>, Forján, R<sup>2,4</sup>, López-Sánchez, C A<sup>1</sup>





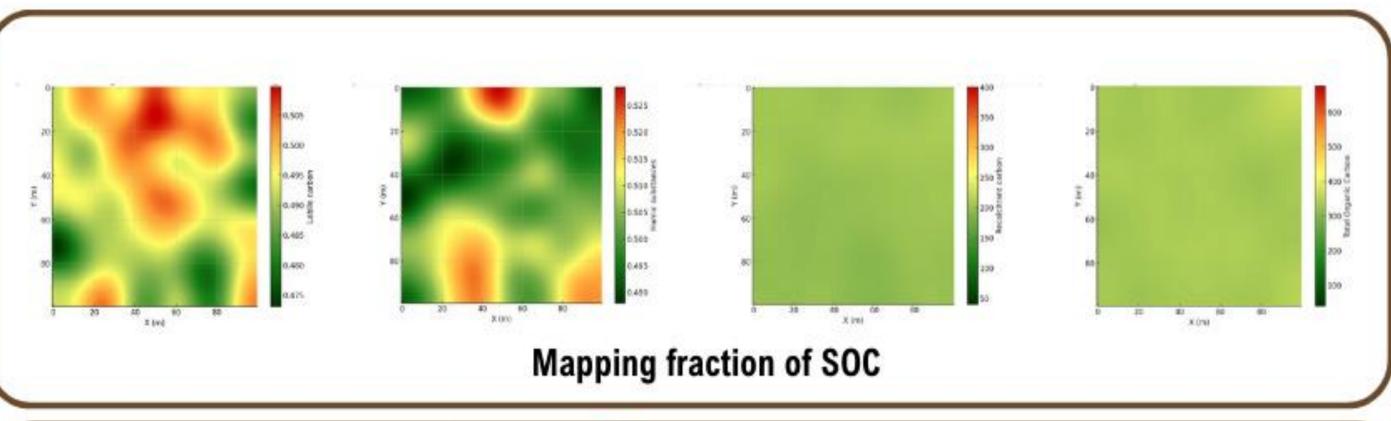
<sup>1</sup> SMartForest Group, Department of Organisms and Systems Biology, Polytechnic School of Mieres, University of Oviedo, 33600 Mieres, Spain. <sup>2</sup> Biogeochemistry & Raw Materials Group and Institute of Natural Resources and Territorial Planning, Campus of Mieres, University of Oviedo, 33600 Mieres, Spain. 3 Department of Geography, Campus del Milán, University of Oviedo, 33011 Oviedo, Spain. 4 Plant Production Area, Department of Biology of Organisms and Systems Biology, University of Oviedo, 33600 Mieres, Spain.

## L. INTRODUCTION

The impact of the recurrence of wildfires on soil organic carbon (SOC) dynamics cause the need for rapid large-scale territorial mapping to determine SOC. Traditional geochemical methods, while accurate, are costly and time-consuming. In contrast, UAVs equipped with advanced sensors capture highresolution spatial data, providing greater flexibility and temporal resolution. These data are crucial for refining remote sensing algorithms, which are then applied to broader scales using satellite data. This hierarchical approach enhances the accuracy and scalability of SOC assessments, offering a robust framework for soil carbon monitoring and climate change mitigation. The study utilized the P4 Multispectral DJI UAV and GEOSAT-2 satellite, both employing passive remote sensing in the VIS-NIR range, to analyze different carbon fractions within the soil.

# 2. METHODOLOGY

Fig. 1 illustrates the methodology, with step 1 involving the modeling between the collected geochemical data and the spectral data derived from UAV followed by mapping, and step 2 involving the modeling based on the mapping obtained with UAV and the spectral data from GEOSAT-2.



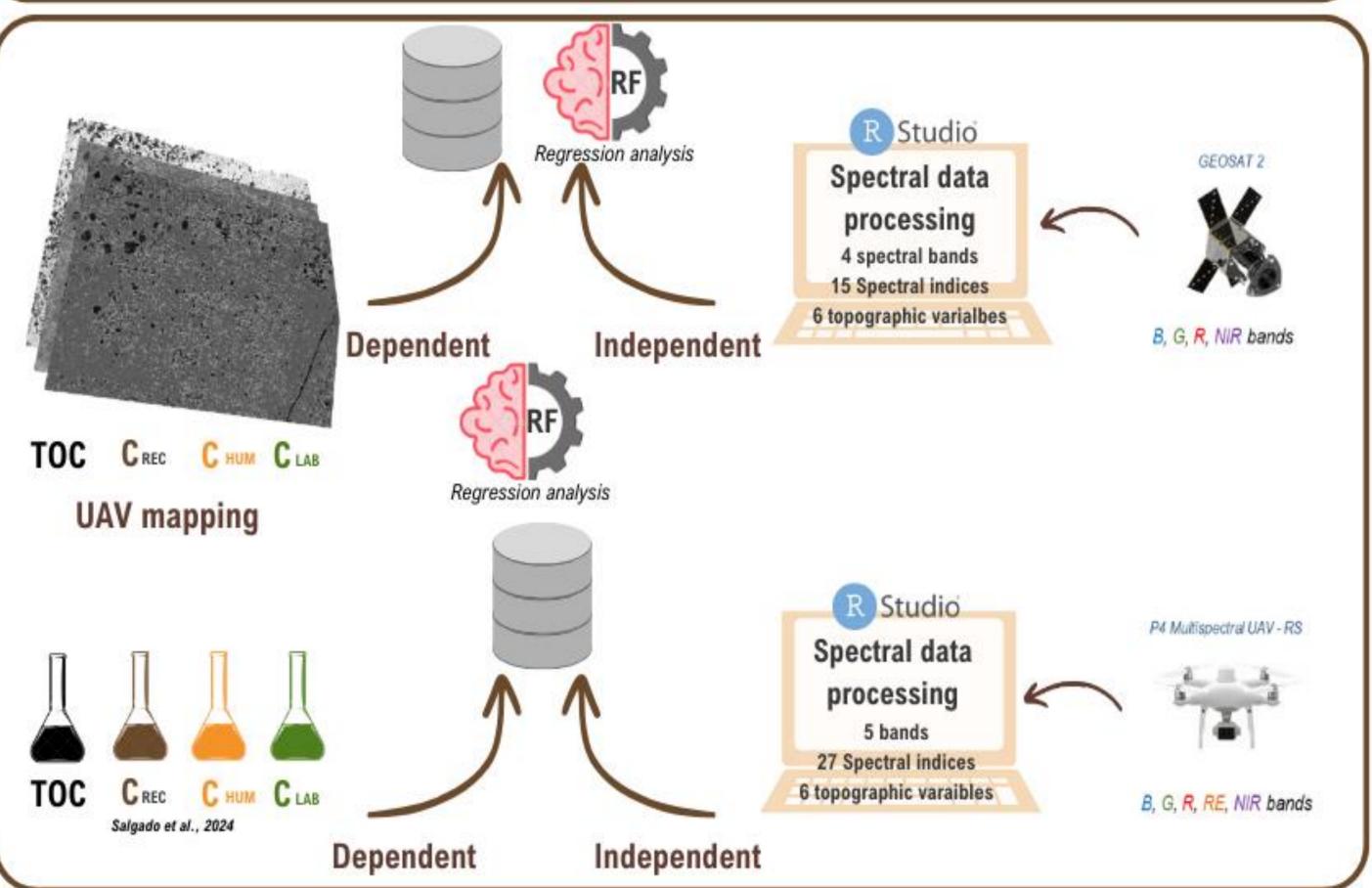




Fig 1. Methodology scheme

# 3. RESULTS

### 3.1 Adjustment of predictive models.

Table 1. Goodness-of-fit statistics given by ML for CLAR, CHIMA, CREE and TOC

		UAV			GEOSAT 2		
		$R^2$	rRMSE	RPQ	$R^2$	rRMSE	RPQ
	C <sub>LAB</sub>	0.30	116.39%	0.56	0.83	65.04%	0.33
	C <sub>HUM</sub>	0.33	58.98%	0.69	0.20	28.62%	0.49
	C <sub>REC</sub>	0.52	28.46%	1.42	0.79	28.14%	0.66
	TOC	0.53	25.67%	1.45	0.91	33.18%	0.33

- The **best** results are obtained for **TOC**, both in the geochemical-UAV estimation and in the upscaling process from UAV mapping to GEOSAT 2. - Among the organic carbon fractions, the **best results** are obtained for the recalcitrant fraction, being the only fraction that presents an optimal goodness-of-fit for upscaling (Table 1).

### 3.2 Mapping.

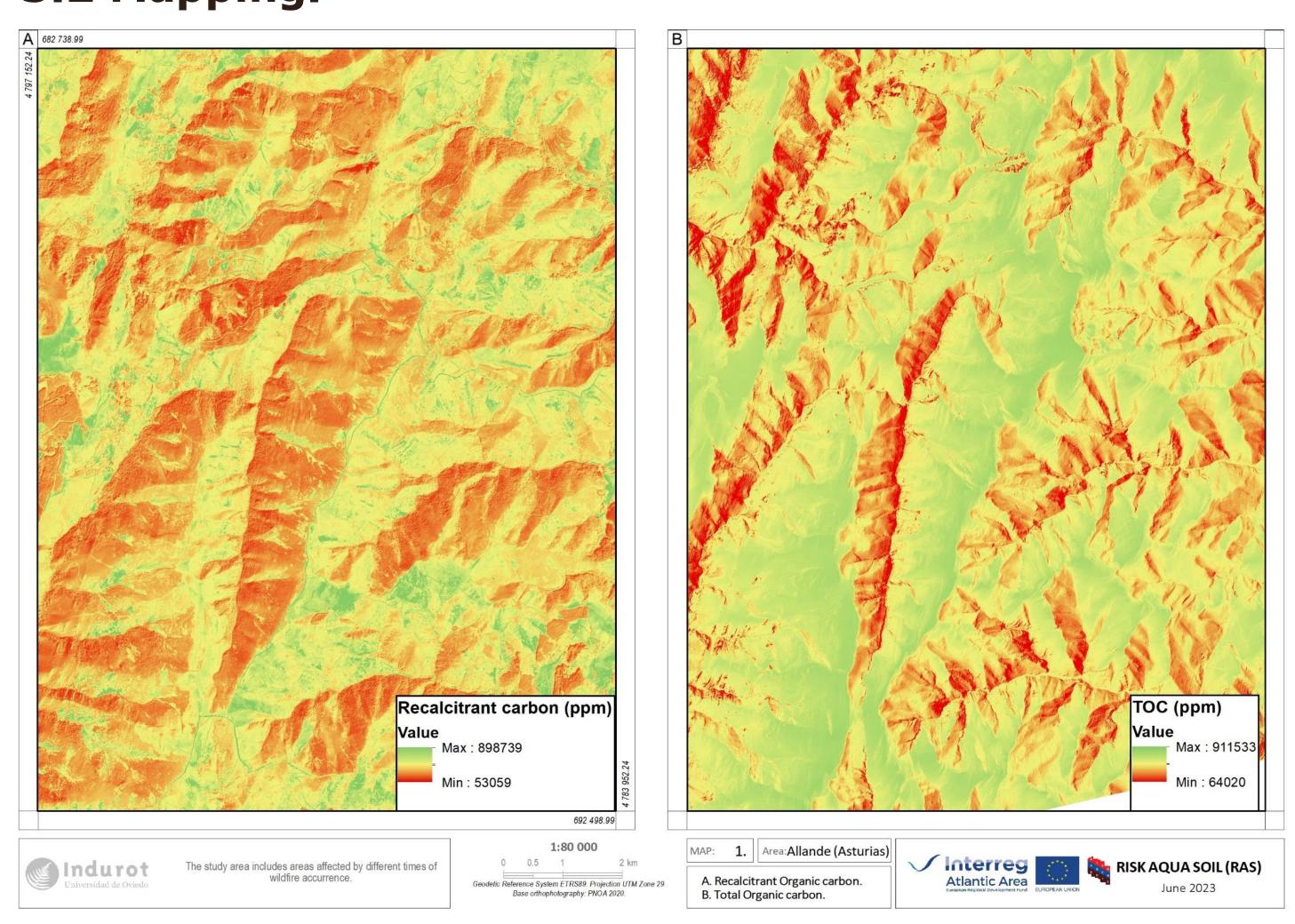


Fig 2. Mapping organic carbon in soil: A. Recalcitrant and B. Total

# 4. CONCLUSIONS

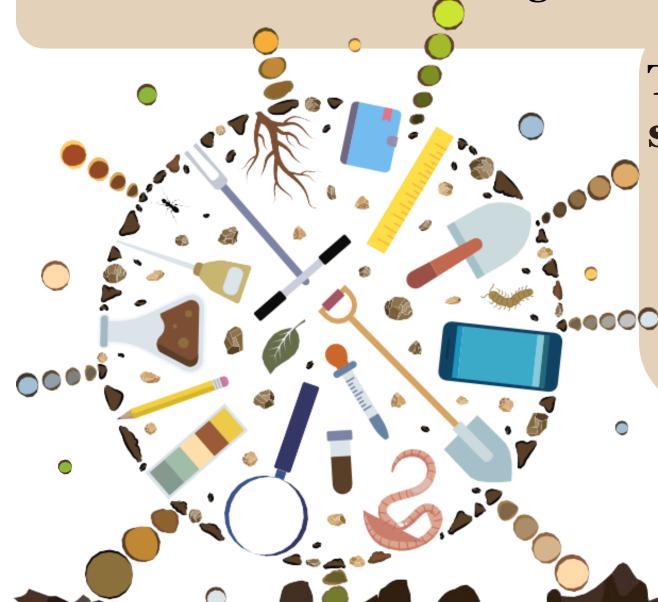
The use of **Remote Sensing** technologies, such as satellite as UAV, to **address organic carbon soil** are complementary, **reliable**, and **rapid** alternatives to classical geochemical methods.

0

aling

Study

are



The use of **UAV** techniques result on **spatial detailed** organic carbon **distribution**. The use of **satellite** data result on spatial distribution of organic carbon of large areas.

According to the results obtained, the proposed methodology, based on UAV-satellite upscaling, yields acceptable results for the recalcitrant fraction and TOC. However, for the less stable fractions, such as labile carbon and humic substances, the results obtained from the initial UAV-derived model are poor, which is consistent with the results later obtained from the satellite data.

#### REFERENCES

Salgado, L., Alvarez, M. G., Díaz, A. M., Gallego, J. R., & Forján, R. (2024). Impact of wildfire recurrence on soil properties and organic carbon fractions. Journal of Environmental Management, 354, 120293.