









EO for Africa Symposium 2024

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Topsoil Organic Carbon Retrieval Using PRISMA Hyperspectral Imagery and Machine Learning Techniques

<u>Saham Mirzaei</u>¹, Simone Pascucci¹, Alessia Tricomi², Raffaele Casa³, Chiara Pratola², Roberta Bruno², Stefano Pignatti¹, Francesca Fratarcangeli², Riccardo Ungaro²

¹ Institute of Methodologies for Environmental Analysis (IMAA)- Italian National Research Council (CNR), C. da S.Loja, 85050 Tito Scalo, Italy ² e-GEOS S.p.A., Via Tiburtina, 965 - 00156 Rome – Italy;

³ Department of Agriculture and Forestry Sciences (DAFNE), University of Tuscia, Via San Camillo de Lellis, 01100 Viterbo (Italy)

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PRISMA 4 AFRICA - Overview

 Development and validation of analysis techniques (developed as open-source) exploiting satellite data as hyperspectral (e.g PRISMA) and thermal (e.g. ECOSTRESS) data for food security in Gabon, Mozambique and South Africa.

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High-level workflow

Problem Statement

• Limitation of First-Hand In-Situ Data

Goal

• Leverage the Global Free Available data along with PRISMA imaging and few Local data to produce high accuracy soil SOC map.

EnM

Hyperspectral data



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Sentinel-2

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Image Acquisition for Study Area



<complex-block>

Gabon GABON Boma: sugarcane area

Boma, Gabon

SASRI, South Africa



Site	Xinavane, Mozambique		Boma, Gabon		SASRI, South Africa	
Quality	Free Cloud	Cloudy	Free Cloud	Cloudy	Free Cloud	Cloudy
PRISMA	14	6	-	4	5	5
EnMAP	9	7	-	15	3	-
DESIS	2	3	-	1	-	-

Global SOC and PRISMA Dataset





Sampling depth: 0-20 cm



Global SOC Dataset



SOC Sample per-PRISMA Image



South Africa SOC and PRISMA Dataset

Sampling depth: 0-20 cm

SOC Measurement Method:

Walkley-Black – dry oxidation after carbonates removal

- International Soil Reference and Information Centre (ISRIC)
- Gamtoos Irrigation Board (GIB)
- Department of Environmental Affairs (DEA)
- Agricultural Research Institute (ARC)





Methods for SOC Retrieval Using EO Data







Deep Transfer Learning for Soc Retrieval





Apply Tuned CNN Model to Xinavane SOC Retrieval





Primarily SOC Map of Xinavane using Already developed Indices











Apply the Tuned CNN Model to SASRI Test Areas SOC Retrieval





Clouds, Water & Built-up Mask & NDVI <0.35 & nCAI<0.03









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Conclusion

 The developed CNN is working properly in the test areas pointing out that the data set is close to being wide enough. At the same time, validation and testing should be improved.

Challenging

- × Availability of a sufficient large independent data set for the test.
- × Explore the CNN transferability robustness in different soil ecosystems.

Future Work

- Provide local SOC samples to validate tuned CNN model at local scale.
- Increasing the number of SOC samples to improve the accuracy of transferred CNN model by covering more diverse soil samples in this region, including a wider range of SOC.
- Investigate the capability of other Hyperspectral sensors (e.g. EnMAP and DESIS) for SOC retrieval.
- The ongoing work, applying the button-up approach, is devoted to up-scaling this methodology to the African agricultural PRISMA4AFRICA test sites by the Gabon, Mozambique, and South Africa Early Adopters

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