



Satellite-Based Remote Sensing for Monitoring Soil Carbon Sequestration in Agroforestry Systems

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Agroforestry systems play a vital role in climate change mitigation by enhancing carbon sequestration both above and below ground. Monitoring soil organic carbon (SOC) dynamics is critical to evaluating the long-term sustainability and carbon sequestration potential of such systems. Traditional soil sampling techniques, while accurate at small scales, are labor-intensive, time-consuming, and spatially limited. Satellite-based remote sensing offers an efficient and scalable alternative for monitoring SOC changes across diverse agroforestry landscapes. This article explores the principles, methodologies, satellite platforms, data integration techniques, and limitations of remote sensing in assessing soil carbon sequestration in agroforestry systems.

Introduction

Agroforestry integrates trees with crops and/or livestock, enhancing biodiversity, improving soil health, and increasing carbon sequestration. Soil carbon sequestration, in particular, refers to the process of transferring atmospheric carbon dioxide (CO₂) into the soil organic matter pool, where it can remain for decades or centuries. Reliable monitoring of SOC is essential to quantify the climate mitigation benefits of agroforestry systems under initiatives like REDD+, the Paris Agreement, and carbon trading markets. Remote sensing, particularly from satellites, provides consistent, repeatable, and wide-area observations that are well-suited to monitoring land cover, biomass, and soil-related parameters. Recent advancements in satellite technologies, sensor resolution, and data fusion techniques enable the estimation and monitoring of SOC with increasing precision.

Principles of Remote Sensing in Soil Carbon Estimation

Remote sensing detects electromagnetic radiation reflected or emitted by the Earth's surface. While SOC itself is not directly detectable, it influences several biophysical and biochemical parameters that can be captured using various spectral bands. These include:

- **Vegetation indices (NDVI, EVI):** Indirect indicators of biomass productivity and organic matter inputs.
- **Soil reflectance:** SOC-rich soils tend to be darker, affecting visible and near-infrared reflectance.
- **Thermal properties:** Soil moisture and texture, often correlated with SOC, influence thermal emissivity.

- **Microwave backscatter:** Sensitive to soil structure, moisture, and texture, which relate to SOC dynamics.

Key Mechanisms:

- **Spectral reflectance modeling**
- **Vegetation and biomass proxy estimation**
- **Land use/land cover (LULC) classification**
- **Change detection over time**

Satellite Platforms and Sensors for SOC Monitoring

Several satellite missions provide data relevant to soil carbon monitoring in agroforestry systems:

Satellite	Sensor	Resolution	Key Parameters
Landsat 8/9	OLI/TIRS	30 m	LULC, NDVI, surface reflectance
Sentinel-2	MSI	10–20 m	Multispectral indices, vegetation cover
MODIS	MOD09, MOD13	250–1000 m	Temporal dynamics, biomass proxies
PRISMA	Hyperspectral	30 m	Detailed soil/vegetation spectra
SMAP	L-band Radiometer	9–36 km	Soil moisture
PlanetScope	RGB/NIR	3–5 m	High-resolution vegetation monitoring
GEDI (on ISS)	LiDAR	Footprint (~25 m)	Vertical forest structure, biomass

Hyperspectral Imaging

Hyperspectral sensors offer the potential to detect SOC directly by resolving subtle differences in soil spectral signatures across 100s of narrow bands, particularly in the VNIR-SWIR range (400–2500 nm).

Methodological Framework

A generalized workflow for satellite-based SOC estimation in agroforestry includes:

Step 1: Data Collection

- Acquire multi-temporal satellite imagery
- Obtain ancillary data (DEM, land cover maps, soil databases)

Step 2: Preprocessing

- Radiometric and geometric correction
- Atmospheric correction
- Cloud and shadow masking

Step 3: Feature Extraction

- Derive vegetation indices (e.g., NDVI, SAVI)
- Analyze soil reflectance bands
- Extract texture and topographical features

Step 4: Ground Truthing

- Conduct field surveys for SOC sampling
- Use GPS for geolocation of sample points

Step 5: Model Development

- Use machine learning models (e.g., Random Forest, SVM, PLSR)
- Train SOC estimation models using spectral and ancillary features

Step 6: Validation and Uncertainty Analysis

- Cross-validate model with independent datasets
- Evaluate model performance (R^2 , RMSE, bias)

Step 7: Spatial Mapping and Time-Series Analysis

- Generate SOC distribution maps

- Analyze trends and detect spatial-temporal changes

Applications in Agroforestry Systems

In agroforestry, remote sensing supports:

- Baseline SOC stock estimation** for carbon accounting
- Monitoring temporal SOC changes** due to afforestation, land management, or degradation
- Evaluating tree-crop interactions** and their impact on biomass and SOC input
- Supporting carbon credit schemes** and sustainable land management practices

Agroforestry systems often include complex heterogeneity in vegetation and soil types. Therefore, combining optical, thermal, and radar data can improve monitoring accuracy.

Challenges and Limitations

While satellite-based SOC monitoring offers great potential, several limitations remain:

Challenge	Description
Indirect Estimation	SOC is inferred from proxies, not directly measured.
Sensor Limitations	Coarse spatial resolution in many platforms limits plot-level monitoring.
Soil Surface Conditions	Vegetation cover, moisture, and residues can obscure soil signal.
Cloud Cover	In tropical agroforestry zones, persistent cloud limits optical data availability.
Model Transferability	Machine learning models may not generalize across regions.
Validation Needs	Field data are still essential for calibration and validation.

Future Directions

Emerging technologies and integration approaches are expected to enhance remote sensing capabilities for SOC monitoring:

- Synergy of satellite + UAV (drone) + proximal sensing
- Advanced machine learning and AI for big data analysis
- Integration of LiDAR and radar for 3D biomass-carbon relationships
- Development of global soil carbon monitoring platforms (e.g., SoilGrids, OpenLandMap)
- Open data and cloud computing platforms (Google Earth Engine, NASA Earthdata) to enable large-scale analysis

Conclusion

Satellite-based remote sensing is a promising and scalable approach to monitor soil carbon sequestration in agroforestry systems. While not a substitute for in-situ measurements, remote sensing can provide cost-effective, repeatable, and spatially comprehensive insights when combined with field data and advanced modeling. Continued advancements in sensor technology, data integration, and AI-driven analytics will further unlock its potential to support sustainable agroforestry practices and global climate mitigation strategies.

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