Airborne Hyperspectral Measurements and Superficial Soil Organic Matter

OSTC

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OUTLINE Research Goals Study Area Field And Laboratory Measurements Data Analysis And Methodology Results Conclusion & Perspectives





RESEARCH GOALS

Validate Soil Organic Matter (SOM) Prediction Model by Means of Hyperspectral Images

Study the impact of disturbing factors on SOM-Hyperspectral signal relationship



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STUDY AREA : SOUTHERN BELGIUM

Study area located in the Province of Luxemburg, Southern Belgium



- 14 agricultural parcels
- 135 soil samples
- Sandy to clayey soils



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ATTERT (2003)
Area : 50 km²
10 agricultural parcels
100 soil samples
Sandy to clayey soils



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MEASUREMENTS

Casi airborne campaign : Pixel resolution : 2.5x2.5 m

Casi 96 spectral bands : 400-950 nm

ial image

Hyperspect

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Hyper-Cube

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Pixel Spectra

Wavelength

MEASUREMENTS

Casi airborne campaign :
Pixel resolution : 2.5x2.5 m
Casi 96 spectral bands : 400-950 nm

Soil properties
Soil samples
Soil moisture

Surface roughness
 (SR)

SR = St.Dev Relative height of 100 nails.





MEASUREMENTS

Casi airborne campaign :
Pixel resolution : 2.5x2.5 m
Casi 96 spectral bands : 400-950 nm

Soil properties:

Soil samples

- Surface roughness
- Soil moisture
- ASD measurement 400-2500 nm:
 Laboratory measurements
- Field acquisition







- Soil chemical analysis
- Spectral Signature Analysis
- Statistical analysis :
- Defining Best Correlated Bands by Stepwise Procedure
- Multi-regression : $SOM_p = A_0 + A_1R_{\lambda 1} + A_2R_{\lambda 2} + ... A_nR_{\lambda n}$
- Calibration & Validation
 1/3 Samples for validation
- Prediction evaluation

$$RMSE_and_PRMSE=$$



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 $\sum_{i=1}^{n} \left(V_{fi} - V_{pi} \right)$

n-1







RESULTS : SOM and Soil Moisture



RESULTS : Spectra pre-proceesing

Noise and atmospheric effects are reduced by smoothing and 1st derivative algorhitms.

RESULTS : Calibration phase 1

Tintigny 734 563 Attert

Best models for Tintigny (9 bands) and Attert (11 bands)

The models were applied to Tintigny and Attert site respectively:

	Tintigny-site (p-77)		Attentister (n=63)	
	Field SOM	Predicted SOM	Field SOM	Predicted SOM
Mean	2.28	2.28	1.99	1.99
St. deviation	0.78	0.74	0.97	0.89
R ²	0.88		0.85	
ME(%)	-0.003		-0.001	
RMSE	0.266		0.373	

RESULTS : Calibration phase 2

Good fit of field data in the Prediction intervals

Bigger Prediction intervals for Attert site

Validation with independent set of data

Better accuracy for Tintigny site with PRMSE = 0.46 vs. 0.76 for Attert site
 Validation from one site to another did not work well
 Relationships between SOM and reflectance are site dependent

RESULTS : Disturbing factors

Influence of disturbing factors is difficult to quantify:

- Soil Moisture
- Vegetated debris
- Soil Roughness

CONCLUSIONS

Hyperspectral Remote Sensing is a useful tool to derive SOM

Predictive equations are site dependent

Disturbing factors should be taken into account

The method has the advantage of using small samples to determine SOM on regional scale

PERSPECTIVES

Impacts of disturbing factors are under study:

- to better understand their effects
- to quantify their impacts

Using of unmixed models to separate disruptive elements from SOM-Reflectance relationship:

- Endmembers spectral mixture model (Roberts et al., 1998; Galvao et al., 2001)

- Rayleigh Criterion (Ogily, 1991; Matthias et al., 2000)

