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Mapping agricultural phenology using repetitive optical remote sensing over a peri-urban region

Framework : map soil organic carbon content, using remote sensing, ground and laboratory measurements, modelling. Project SURFAC_EGC_SENTINEL_PLEIADES_CO, financed by CNES (French National Space) Agency).

Objectives of this work : determine typology of crop rotations as a further input to modelling. Remote sensing data used : Landsat-8 surface reflectance, PROBA-V, SPOT4-TAKE5, Pleiades.



Data presentation and preparation



Figure 2 : temporal variability based colour composite *image*, based on NDVI min, max and amplitude within 3 years. Indicative range : Yellow : [0.2-0.8]. Red : [-0.6-0.1]. Dark blue : [0.2-0.3]. Cyan : [0.6-0.8].



Figure 3 : crop mask, selected as NDVI_{min} <0.3 and *NDVI_{max}>0.6. Morphological reconstruction dilatation filter* was applied to remove isolated pixels. Vector layer : stands from national land parcel registry subdivised by individual crops based on high resolution imagery (Vaudour et al. 2015).

Figure 1 : NDVI from Landsat-8 surface reflectance (L8SR), subsetted and masked for clouds and strong aerosols loads.

Figure 4 : Comparative dataset, Vaudour et al. (2015). See area in Fig.3. Crop map obtained from the SVM classifier implemented on the bi-temporal 3 and 24 April 2013 Pléiades images.

Vaudour, E., Gilliot, J.M., Bel, L., Lefevre, J., Chehdi, K., (2016a). Regional prediction of soil organic carbon content over temperate croplands using visible near-infrared airborne hyperspectral imagery and synchronous field spectra. International Journal of Applied Earth Observation and Geoinformation, 49, 24-38. DOI: 10.1016/j.jag.2016.01.005 Vaudour, E., Noirot-Cosson, P.E., Membrive, O., (2015). Early-season mapping of crops and cultural operations using very high spatial resolution Pléiades images. International Journal of Applied Earth Observation and Geoinformation, 42, 128-141. DOI:10.1016/j.jag.2015.06.003

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Typology of NDVI temporal variations in NDVI for agricultural areas.

Constraint : incomplete time series (clours, aerosols). Methods for 2013 :

1/Missing data reduction \rightarrow Maximum value composite in July 2013 (DOY 191-200), December 2013 (344-360), September 2014 (251-258).

2/ Kmeans classification on pixels with only valid data (2013 $_{DOY104}$, 2013 $_{July}$, 2013 $_{DOY248}$, 2013 $_{DOY280}$, 2013 $_{Dec}$, 2014 $_{DOY75}$, 2014 $_{DOY107}$, 2014 $_{DOY107}$, 2014 $_{DOY139}$, 2014 $_{Sept}$, 2014 $_{DOY315}$, 2015 $_{DOY110}$, 2014 $_{DOY270}$). 60 classes \rightarrow 60 temporal profiles. 2/ Assignation of all pixels (including pixels with missing values) to a temporal profile. 3/ Clustering temporal profiles according to their mutual correlations, and grouping the classified pixels on this basis. **Results** :



Figure 6 : comparison of the temporal classes (Fig. 5) with classes obtained from multi-spectral classification in April 2013 (Fig. 4)

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Pixel aggregation based on NDVI temporal variations across 3 years (cultural cycle of 4 years in the study area).

Some agreement with multipectral classification in spring 2013 : e.g. NDVI variation type 3 matches well to spring cultures.

mapped.







Conclusion

Some disagreement : e.g. NDVI variation type 4 is not specific to any crop

Perspective 1 : Densifying time series

Figure 7: NDVI temporal variations, stand-wise. Landat, PROBA-V (100m central camera), SPOT4-TAKE5. 9 stands from Fig. 3.



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