



EOStat - Agriculture Poland: Services for Earth Observation-based statistical information for agriculture

Jędrzej S. Bojanowski¹, Edyta Woźniak², Sebastian Aleksandrowicz², Zbigniew Bochenek¹, Katarzyna Dąbrowska-Zielińska¹, Martyna Gatkowska¹, Stanisław Lewiński², Jan Musiał¹, Karol Paradowski¹, Marcin Rybicki², Dariusz Ziółkowski¹, Magdalena Mleczko³, Tomasz Milewski³, Przemysław Slesiński³, and Artur Łączyński³

¹ Remote Sensing Centre, Institute of Geodesy and Cartography, ² Space Research Centre, Polish Academy of Sciences, ³ Statistics Poland (all from Warsaw, Poland)

Project aim

- The EOStat project aims at developing an EO-based system to support operational activities of Statistics Poland related to gathering statistical information on agriculture. The ultimate objectives are:
- to create a system of EO-based support for acquisition of statistical information for agriculture production in Poland; to assist and support the system implementation at Statistics Poland

Project tasks

(1) Development of operational in-season crop type classification routines based on:

- multi-temporal S-2 and Landsat composites with Sen2Agri
- multi-temporal S-1 data
- using S-1 and S-2 data fusion

(2) Development of an approach for operational crop growth monitoring and crop yield forecasting (at LAU-1 level) using apparent Sentinel-3 OLCI data and the long term averages from NOAA/AVHRR data.

- to deliver a suite of evaluated and documented tools for automatic processing of satellite data and derivation of information relevant for agricultural production assessment (i.e., crop maps and crop yield forecasts).
- to provide an assessment of economic viability of the system application in the operational work of Statistics Poland

(3) Assessment of applicability of field-specific S-2 vegetation indices for crop growth monitoring and early warning system.

Preliminary results of the tasks marked in **bold** are presented below.

Classified crops Winter wheat Spring wheat Winter triticale Spring triticale Winter barley Spring barley Oat Rye Buckwheat Rapeseed Green maize Grain maize

Area and period of interest

Two testing NUTS-2 areas (Wielkopolskie, Warmińsko-Mazurskie) in 2017-2018





In-situ data

- Field campaigns of Statistics Poland for 2017 and 2018 (> 500 fields for each region)
- Field campaigns of IGIK for 2018 at the Wielkopolska JECAM site (60 fields)
- Farmers' declarations (thousands of samples)
- In-situ controls of the Paying Agency (ARMA)

For training/validation fields, to the field geometry from ARMA (declarations, LPIS), we assigned a crop type attribute from in-situ data. Then we applied a 20m inner buffer to reduce the noise from not agricultural pixels.

Towards S2-based crop growth monitoring at a field level

Crop type classification with Sen2Agri

We classified crop types for two seasons 2017-2018 and two NUTS-2 regions using two types of training data:

Training: 1000 in-situ samples, Statistics Poland, most reliable

Validation: 5-fold cross-validation: insitu data were randomly allocated into 5 groups, classification repeated 5 times using one group as validation data (i.e. 20% data for validation).

Training: 11000 fields of farmers' declarations, subject to error

Validation: 3000 fields of farmers declarations (i.e. 20% for validation)

We applied Sen2Agri system to derive NDVI time series for 2018 growing season for the AOI. The figure below presents the random examples of NDVI mean for individual fields (blue dots). The red curves (smoothed by loess) are added to ease the comparison. The signal seems to have a negligible noise (e.g. from underestimated cloud cover), allows for spring and winter crops separation, and reveals the agricultural practices (i.e. catch crops and grasslands cutting). Based on these findings, we are going to build a database of NDVI time series for all agricultural fields in Poland and to use it for cropspecific yield forecasting at LAU-1 level.



Confusion matrix:



Confusion matrix:

Overall accuracy: 0.55, Kappa: 0.5

Overall accuracy: 0.76, Kappa: 0.74

	Precision (PA)	Recall (UA)	F1	Precision (PA)	Recall (UA)	F1
Buckwheat	0.78	0.75	0.77	0.99	0.98	0.98
Maize	0.82	0.93	0.87	0.95	0.99	0.97
Oat	0.34	0.26	0.29	0.53	0.66	0.59
Pastures	0.48	0.72	0.58	-	-	-
Perm. grasslands	0.71	0.46	0.56	0.99	1.00	0.99
Rye	0.62	0.53	0.57	0.78	0.84	0.81
Spring barley	0.40	0.49	0.44	0.65	0.71	0.68
Spring rapeseed	0.82	0.55	0.66	0.95	0.68	0.79
Spring triticale	0.28	0.15	0.20	0.67	0.36	0.47
Spring wheat	0.31	0.30	0.30	0.63	0.62	0.63
Winter barley	0.65	0.65	0.65	0.93	0.86	0.89
Winter rapeseed	0.77	0.83	0.80	0.78	0.94	0.85
Winter triticale	0.46	0.43	0.44	0.55	0.68	0.61
Winter wheat	0.44	0.58	0.50	0.79	0.70	0.74



Conclusions & outlook

- (1) Random forest classifier (used by Sen2Agri) trained with a large sample of farmers' declarations (subject to errors!) significantly outperforms the classifier trained with 10 times smaller sample of in-situ data
- Except a distinction between spring crops, all crops classified with F1-score (2) above 0.6 (> 0.8 for non-cereals)
- Sen2Agri classifications will be compared with S1-based classifications that (3)are under development within the project
- Mean NDVI estimates reveal a clear signal of crop development, and (4) therefore can potentially be used for field-specific crop growth monitoring and detection of agricultural practices.

Project funded by the European Space Agency