

Sentinel-2 for Agriculture Project

Take 5 Data Set to Anticipate Sentinel-2 Exploitation for Agriculture Monitoring Applications

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European Space Agency

Sentinel-2 for Agriculture Launched by ESA in Feb. 2014



Objective:

- Preparation for national to regional agricultural monitoring based on Sentinel-2
- Consolidate best practices for EO agricultural monitoring
- Strengthening national capacity for agricultural monitoring
- 3-phases project over 3 years



Toolbox for 4 S2-based products in line with the GEOGLAM core products



A unique dataset made of Sentinel-2 like EO time series .



12 sites globally distributed to develop algorithms where EO dataset mainly made of SPOT4- Take 5

(Bontemps et al., RS2015)



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8 sites to repeat the experience and test the robustness of the selected algorithms with dataset mainly made of **SPOT5-Take 5**



... and globally spread in-situ data set

Site	2013	2015		
Argentina	Crop (108) – No crop (39)	Х		
Belgium	Crop (31244)-No crop (78156) GAI, PAI, FAPAR, FCOVER	Crop (41) – No crop (68594) GAI		
Burkina Faso	Crop (496) – No crop (101)	Crop (611) – No crop (141)		
China	Crop (54) - No crop (22)	Crop (142) – No crop (20)		
France	Crop (1500) - No crop (7659)	Crop (985) – No crop (809)		
	GAI, PAI, FAPAR, FCOVER	GAI, PAI, FAPAR, FCOVER		
Mali	Х	Crop (497) – No crop (80)		
Morocco	Crop (636) - No crop (500) LAI	Х		
Pakistan	Crop (228) – No crop (54)	Х		
Russia	Crop (205) - No crop (56)	Crop (85) – No crop (44) GAI, PAI, FAPAR, FCOVER		
South Africa	Crop (120) - No crop (44)	Crop (182) – No crop (599)		
Ukraine	Crop (221) - No crop (56) GAI, PAI, FAPAR, FCOVER	Crop (70) – No crop (69)		
Maricopa (USA)	Crop (4907) – No crop (11351)	Х		





Successful development thanks to JECAM collaborative network

Wide range of agricultural systems



Morocco

Russia

South Africa

Maricopa

Ukraine

ESBIC





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Madagascar

Benchmarking for a transparent and objective algorithms selection using SPOT4Take5 data





Repeating and testing the robustness with SPOT5-Take5 dataset



















ESBIO





Repeating and testing the robustness with SPOT5-Take5 dataset



Monthly cloud-free composite



Benchmarking conclusions

- Weighted average approach
- Compositing period can vary between 30 to 50 days window
- Implements a directional correction for seamless composites
- Recurrent implementation : L3A product updated with each new L2A product

To limit the data volume to keep on-line in Sen2Agri system





Glimpse of 2015 results on Mali site

Remaining cloud or cloud shadow

Site : Ourikela-Mali

SPOT5(Take5)

Apr

From 2015-04-10 to 2015-09-12 5 monthly composites Compositing period: 50 days

- Artefacts and higher proportion of remaining gaps on the edges due to SPOT5 footprint instability

- Even with a temporal period of 50 days for compositing, there are still gaps in rainfall season. (~25 % of soil is not visible from 07-20 to 09-10)



30/04/2015

ND PIX

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Composite period w/r to the data availability



Dynamic cropland mask



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(Valero S. et al., RS2016)

Benchmarking conclusions

- 2 chains implemented to deal with presence/absence of in-situ data
- RF supervised algorithm
- Trimming to clean an reference map
- A posteriori smoothing based on a per-object approach





Exemples on few SPOT4 sites



2015 Crop mask – Using in situ data Auch-France

SPOT5 30/05/2015 - False colors composite



Crop mask legend No crop

Crop

Validation data

Crop









2015 Crop mask – Using in situ data Auch-France

Crop mask pixel



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2015 Crop mask – Using in situ data Auch-France

Crop mask pixel

Majority vote with segmentation

	9 month			
	Fscore_noCrop	Fscore_Crop	Kappa	OverallAccuracy
Mean	0.873	0.939	0.812	0.918
CI-95	0.007	0.003	0.007	0.003

9 month with post-processing

	Fscore_noCrop	Fscore_Crop	Kappa	OverallAccuracy
Mean	0.885	0.945	0.829	0.925
CI-95	0.005	0.003	0.006	0.003

2015 Crop mask – Using in situ data Mali - Ourekila

SPOT5 30/07/2015 - False colors composite



Validation data

No crop
Crop









2015 Crop mask – Using in situ data Mali - Ourekila

Crop mask pixel

Majority vote with segmentation

	6 month			
	Fscore_noCrop	Fscore_Crop	Карра	OverallAccuracy
Mean	0.834	0.922	0.756	0.894
CI-95	0.011	0.009	0.018	0.010

6 month with post-processing

	Fscore_noCrop	Fscore_Crop	Карра	OverallAccuracy
Mean	0.869	0.940	0.809	0.918
CI-95	0.010	0.007	0.016	0.009



Efficient annual cropland mapping along the season



Approach using in situ better than without in situ one.

Without in situ method reachs the OA targets => also interesting as many situations where field campaigns are not ensured (e.g. 1st year of activity, food insecure countries, political instabilities, etc.)

Crop type map



(Inglada et al., RS2015)

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Benchmarking conclusions

- Based on the crop mask previously generated
- Random forest classifier
- Classifier applied on features :
 - Surface reflectance
 - NDVI
 - NDWI
 - Brightness





2015 Crop type over Toulouse region (France)









UCL





sentinel-2

Crop type (Auch - France)



* Overall Accuracy : 0,8512
* 4 main crops F-Score > 0,80
* Barley : confusion with wheat











Argentina





France

Wheat Maize Barley Rapeseed Sunflower Other annual crops





Belgium



Russia

Winter wheat Spring barley Rapeseed Sunflower Other annual crops



3 LAI products

Mono-date LAI estimation



• Multi-date and fitted LAI reprocessing



Benchmarking conclusions

- Non-linear regression model
- Reflectance are simuated using the ProSail model
- 2 reprocessing options :
 - Weighted average using the n last LAI value
 - Fitting a phenological model on the full time series









2016 Growing season



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20-04-2015

10-04-2015

LAI value





La force de l'innovation



2016 Growing season









Nov



Ground data

LAI SPOT5









2015 monodate LAI profiles over France



— Ground data

LAI SPOT5









Conclusions

- SPOT4 (Take5) enabled to test methods in conditions similar to Sentinel-2
- Repeated experiment with SPOT5 (Take5) allows obtaining the same encouraging results
- With both experiments the project is now able to provide a strong scientific contribution to the JECAM network and GEOGLAM initiatives
- Contribute to fill the gap between operational systems and state-of-theart
- Ready for demonstration phase allowing to test Sentinel-2 data in real conditions
- Many thanks to CNES and ESA for making this unique dataset available
- Successful development thanks to JECAM collaborative network









Thank you for your attention



Early results from Sen2-Agri system using S-2 over Czech Republic











