



Agriculture and
Agri-Food Canada

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Climate Change Canada

Environnement et
Changement climatique Canada



Mapping Crops and their Biophysical Characteristics with Polarimetric Synthetic Aperture Radar and Optical Remote Sensing

Part 3: Sen4Stat Open-Source Toolbox for Large Scale Crop Mapping (Theory and Practical)

Pr. Pierre Defourny (UCLouvain)

26 April, 2022

Training Outline

12 April

Part 1: SAR
Polarimetry for
Agriculture (Theory
and Practice)

19 April

Part 2: Practical SAR
Polarimetry with
Sentinel-1, RCM, &
SAOCOM Imagery for
Agriculture

April 26

**Part 3 - Sen4Stat Open-
Source Toolbox (Theory
and Practical)**

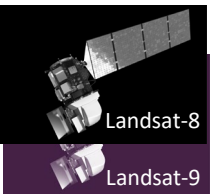
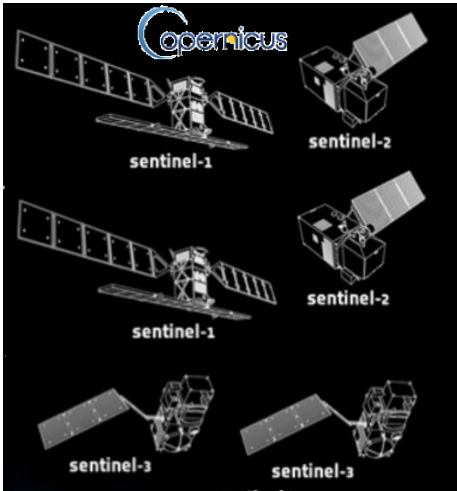
May 3

Part 4: Crop-Specific
Time Series Analysis for
Growth Monitoring

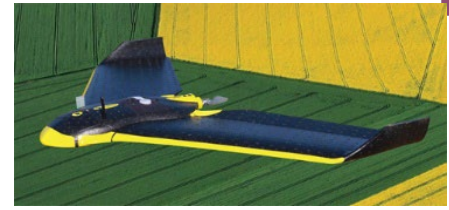
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- Section 2: Large scale crop mapping using Sen4Stat toolbox
- Section 3: Practical using Sen4Stat toolbox for crop type mapping
- Section 4: Q&A session

New satellite missions and IT (r)evolutions have changed the game for agriculture



Server farms with full EO data archives
Exploitation at low/no cost



Free, open and long term data policy (US, EU,...)



Sen2-Agri

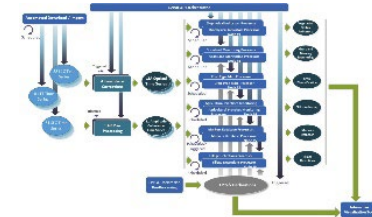


sen4cap
common agricultural policy



Sen4Stat
Statistics for Agriculture Statistics

Open source processing tools using ML & AI algo.



CEOS
Coordination for EO systems redundancy & harmonization

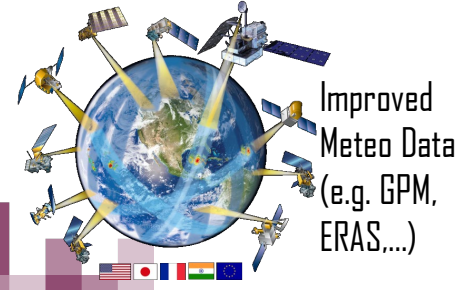
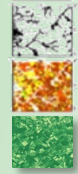
JECAM
Joint Experiment for Crop Assessment and Monitoring
Global network with cross site experim.



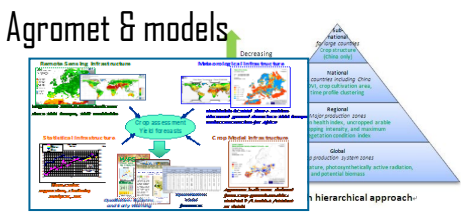
Best practices & standards

EO ag. products => (intra-) field level

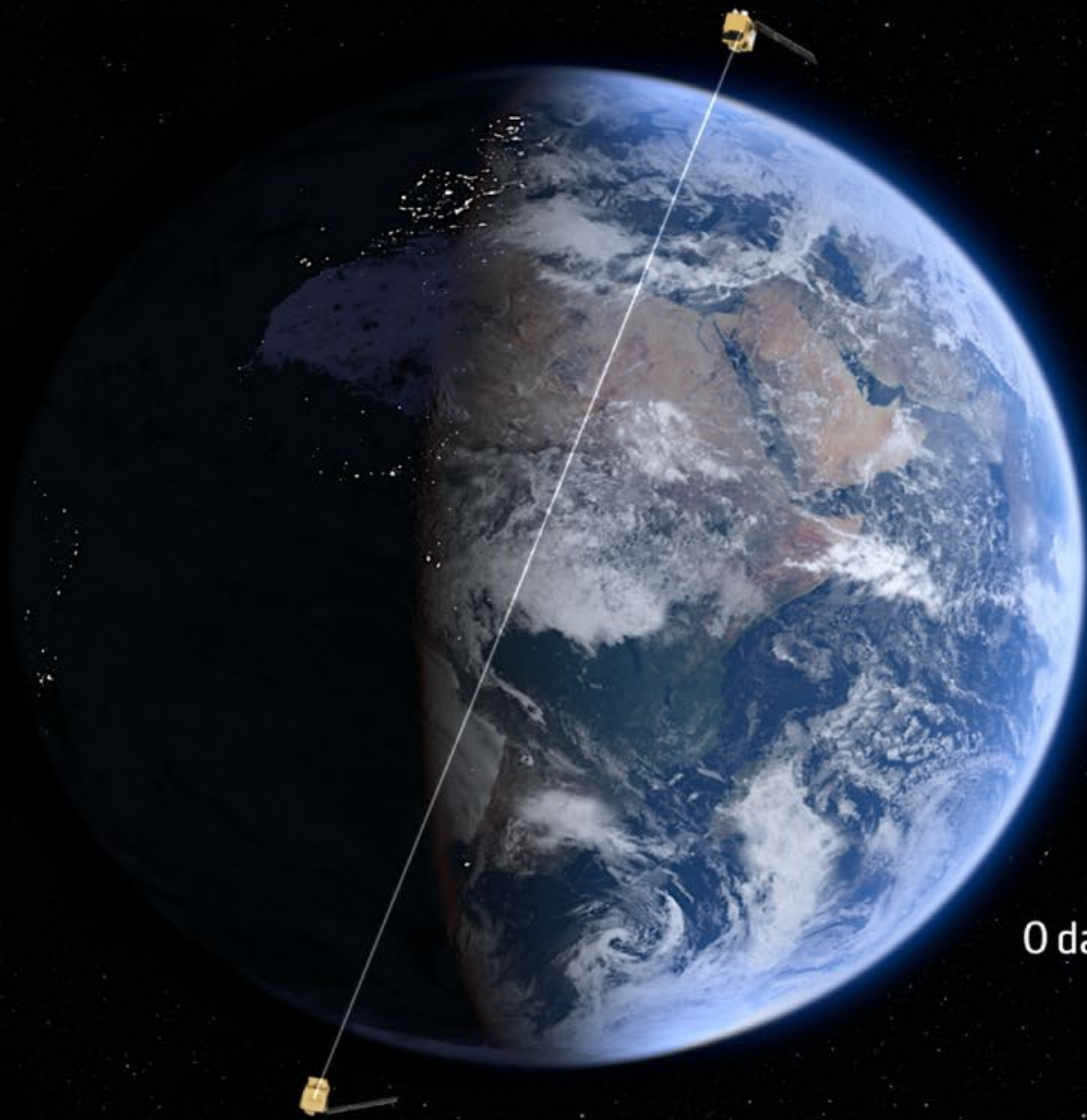
1. Cropland mapping
2. Crop type mapping
3. Biophys. variables (fCover, LAI, fAPAR,...)
4. Crop condition monitoring
5. Ag. practices monitoring



Improved Meteo Data (e.g. GPM, ERAS,...)

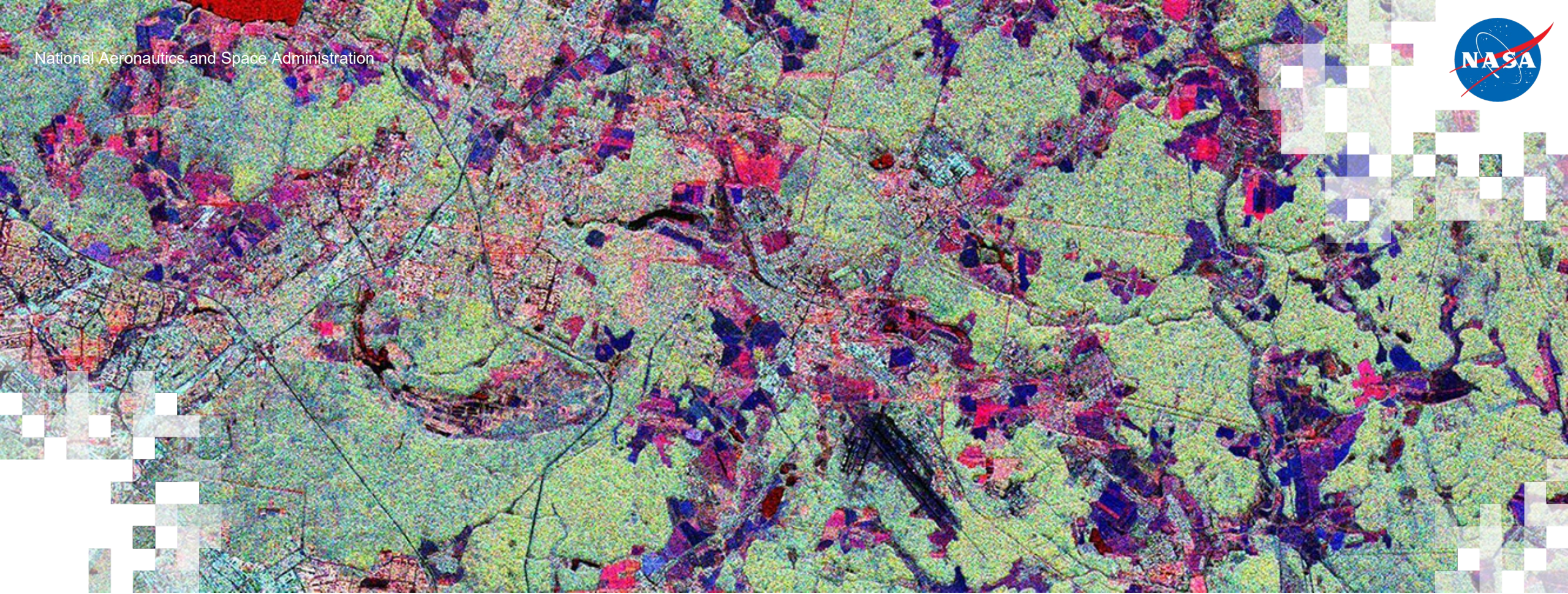


How to take advantage of Copernicus Sentinel-1 and Sentinel-2 data for crop mapping at national scale?



0 days 00 hours 00 minutes
Sentinel-2 constellation:
summer solstice





Section 1: Challenges for large scale crop mapping at high spatial resolution

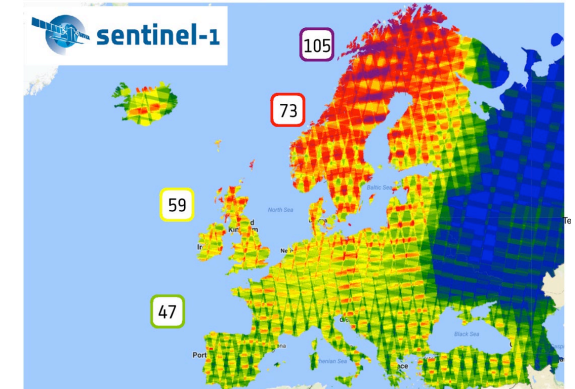
Challenges for large scale mapping

In addition to large data volume and the IT requirements



- Observing satellite systems at high resolution do not cover large area at once
 - Sentinel-2 instrument swath: 290 km
 - Sentinel-1 interferometric wide swath mode: 250 km
 - Overlap between images acquired from adjacent orbit
 - Increasing overlap due to converging orbit to North Pole

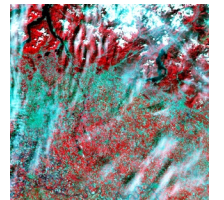
⇒ **Heterogeneity of observation density** across the area



Nbr of observation dates by S-1A & -1B (July-Sept)

- Cloud coverage frequency quite variable (optical systems) and nonsystematic acquisition plan over space and time (optical and SAR systems)

⇒ **Heterogeneity of valid observation density** across the area



- Significant **agro-climatic gradients** introducing large variability in crop calendars and vegetation development across the area

Challenges for large scale crop mapping – why is it so special?

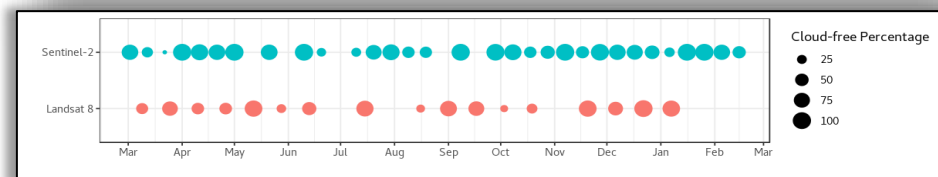
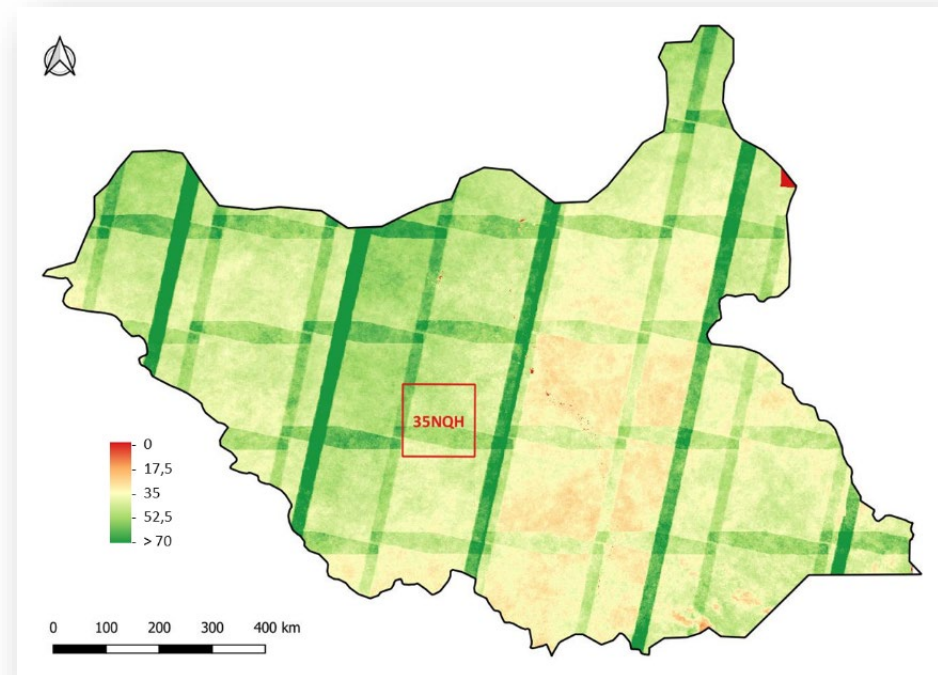
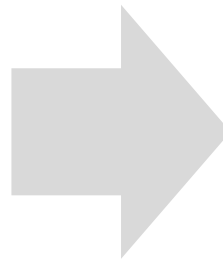
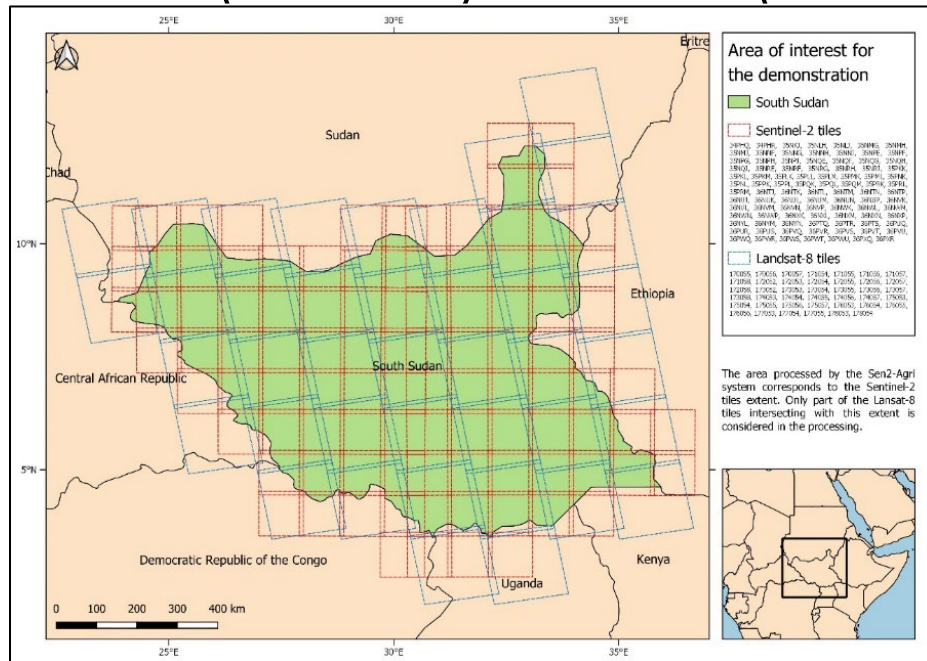
3 challenges already identified

temporal resolution critical for crop discrimination

Example of South Sudan (619.745 km²)

Nbr of granules for satellite data:
scenes (Landsat) and tiles (Sentinels)

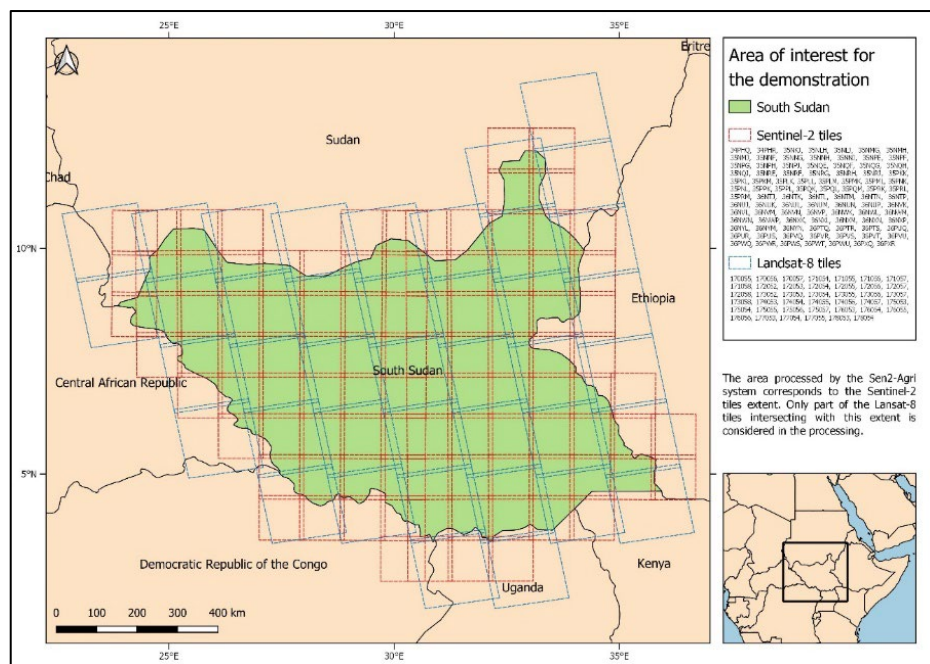
Nbr of valid images (cloud cover < 10%)
from March 2017 to Feb. 2018



Challenges for large scale crop mapping

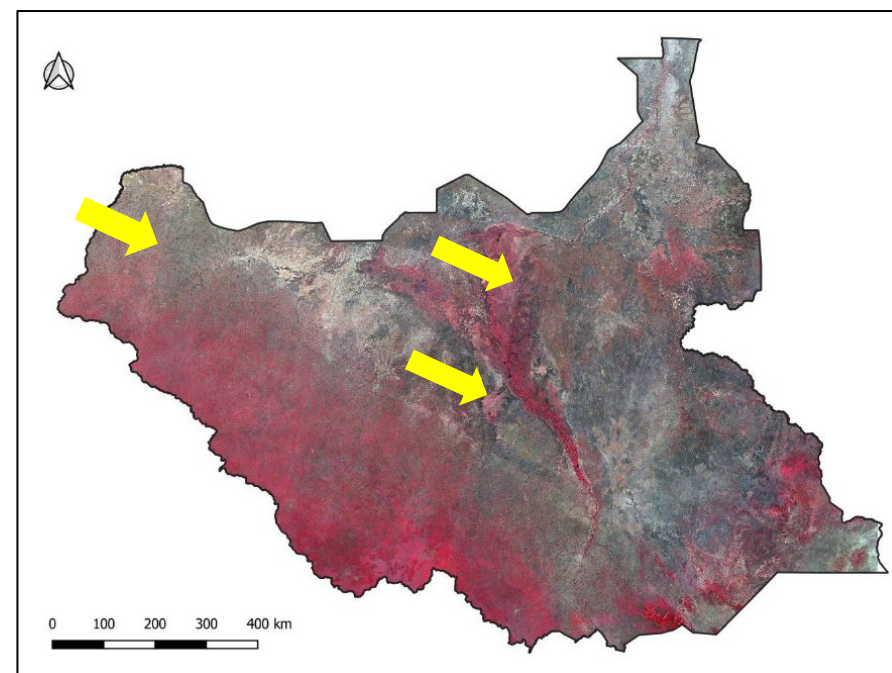
Example of South Sudan – **first solution to cope with heterogeneous valid observ. density**

Nbr of granules for satellite data:
scene (Landsat) and tile (Sentinels)



Monthly synthesis (June 2017)

by averaging all cloud free reflectances

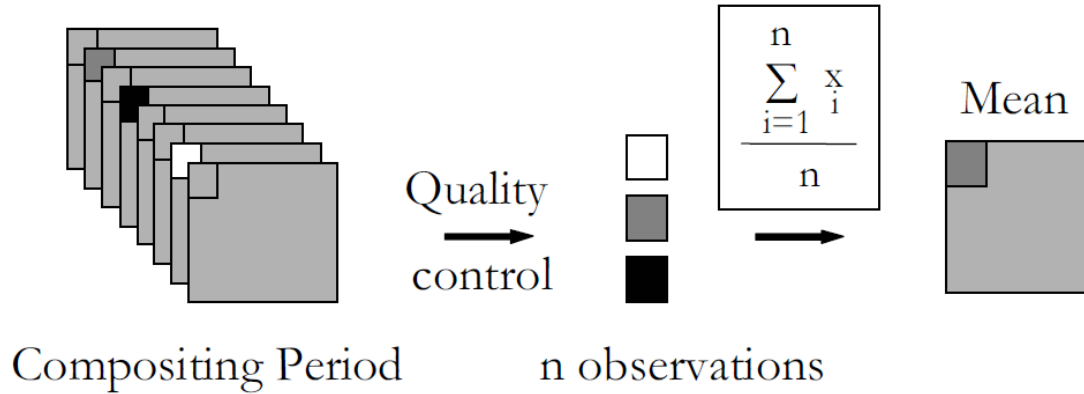


=> **artefacts** due to the variability of input data

=> **reduction of temporal resolution** to monthly

Challenges for large scale crop mapping

Temporal synthesis methods averaging all cloud free reflectances acquired during the period



Sen4Stat: Weighted Average Synthesis Processor

- weight as function of date (central date distance)
- weight as function of Aerosol Optical Thickness
- weight as function of distance to clouds
- weight as function of sensor (S2>L8)

=> reduction of the compositing artefacts

=> monthly temporal resolution not really suitable for crop discrimination

Comparison of temporal synthesis methods

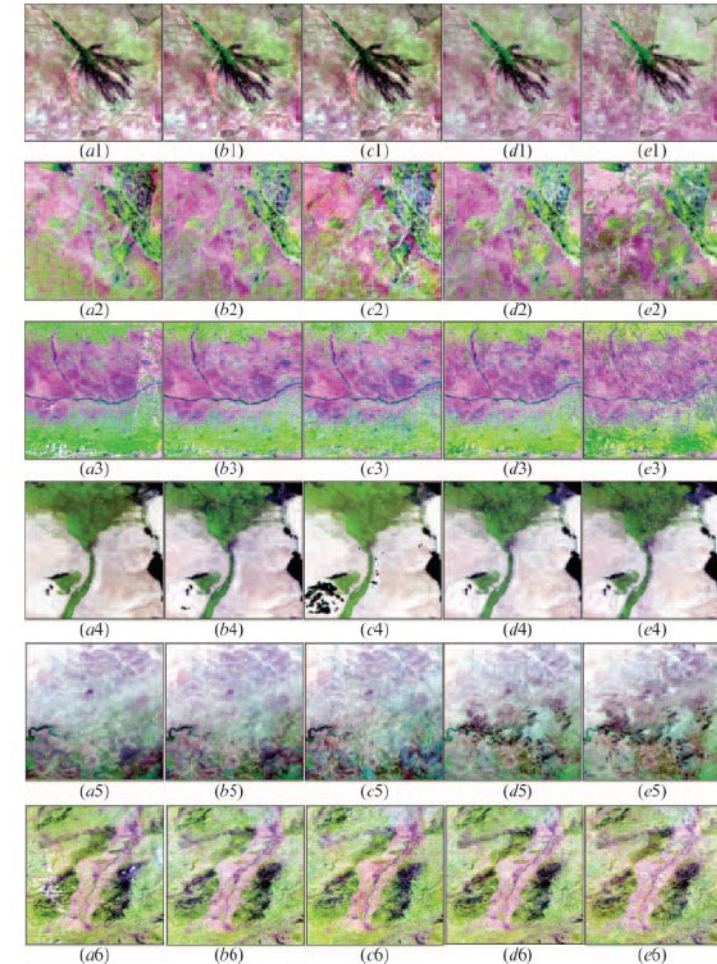


Figure 3. Visual comparison between the best cloud-free image of the decade (a), the MC (b), the BDC (c), the AVG (d) and the MVC (e) colour composites (SWIR, NIR, red), for 6 samples: (1) Botswana (1st decade of June 2002), (2) Sudan (1st decade of November 2002), (3) Romania (2nd decade of June 2002), (4) Egypt (1st decade of November 2002), (5) Senegal (1st decade of June 2002), and (6) Germany (2nd decade of June 2002). Available in colour online.

(Vancutsem et al., JRS 2008)

Western Cape Province monitored by Sentinel 2 monthly composites

2016 winter grain production region (South Africa)

June

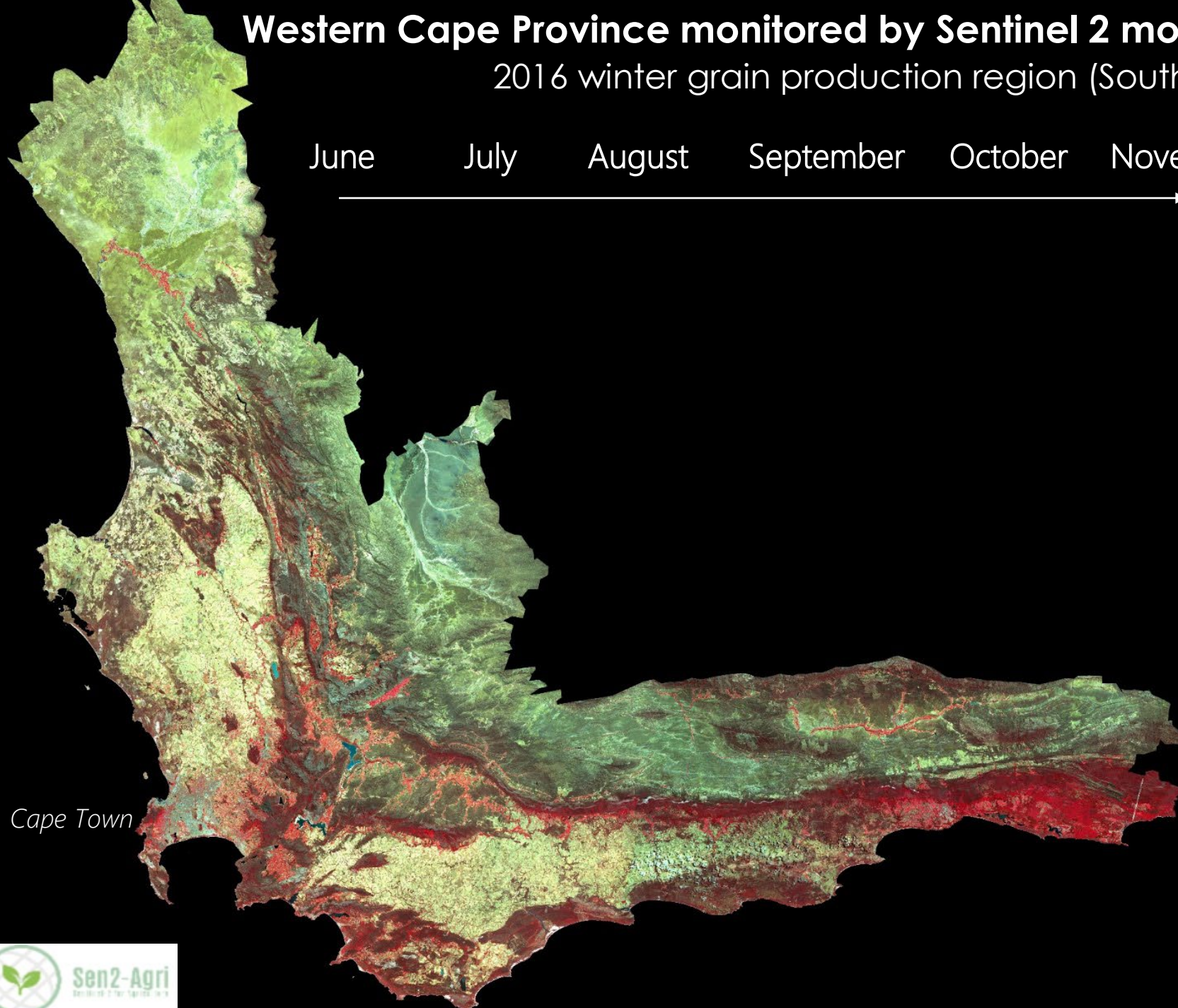
July

August

September

October

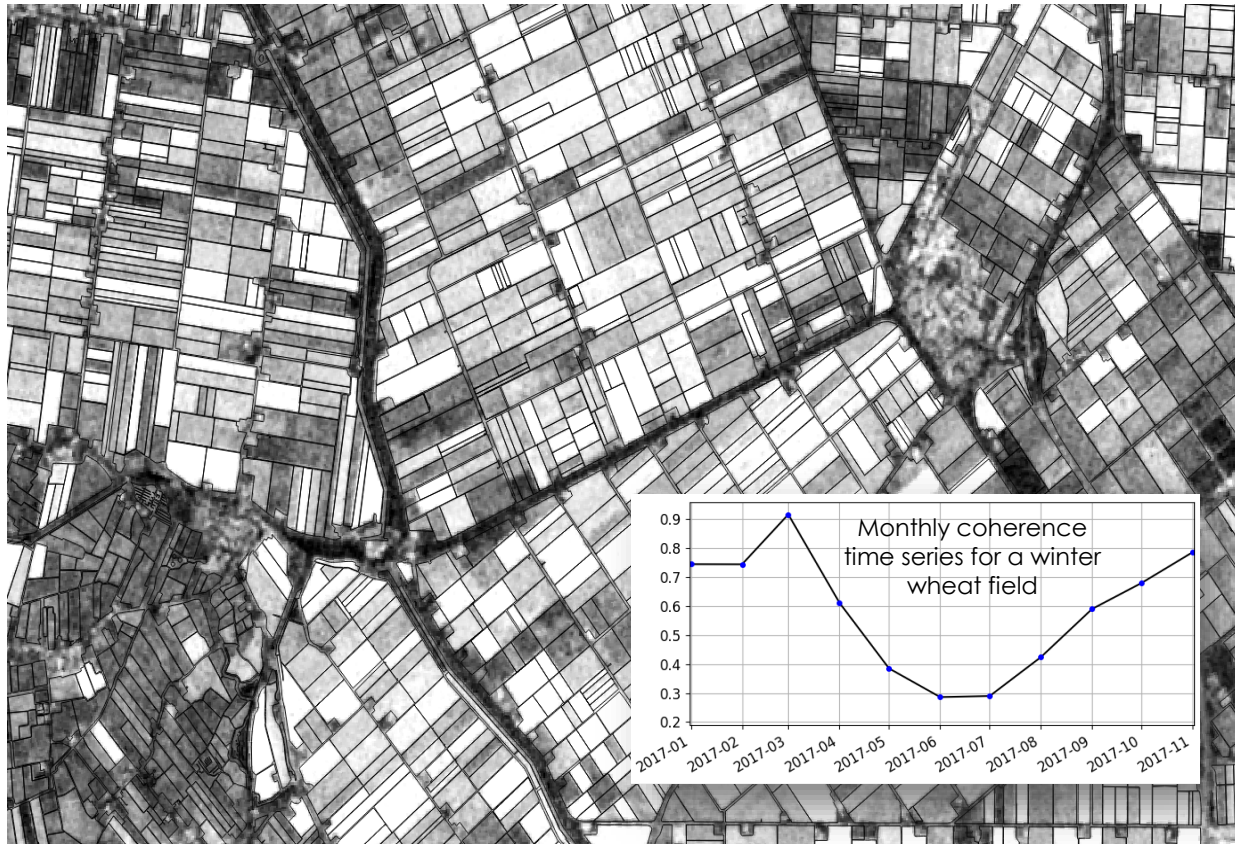
November



Cape Town

Challenges for large scale crop mapping

SAR temporal synthesis by averaging all SAR observations over the period for noise reduction



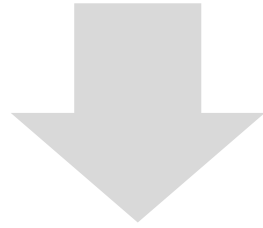
Mean of the coherence values as derived from Sentinel-1 interferometric pairs for March 2017 over Netherlands



Sentinel-1 color composite (blue: mean of the July coherence; green: mean of the March coherence; red: seasonal standard deviation)

Challenges for large scale crop mapping

Temporal information very critical for crop discrimination
Second solution to cope with heterogeneous obs. density over large area

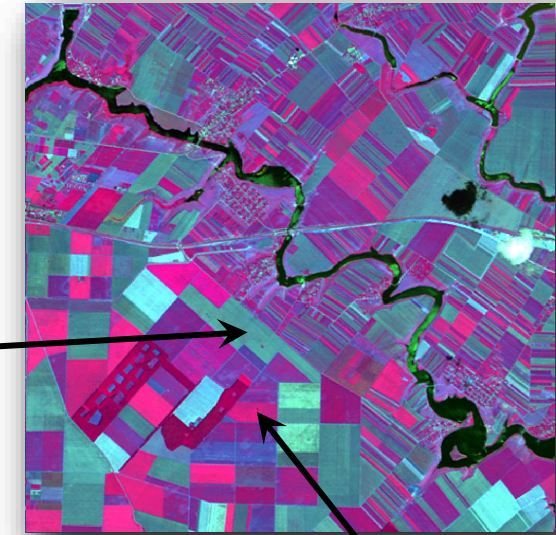


Production of an **optical or SAR time series** which is:

- **gap-filled** using a linear interpolation method with respect to missing data (clouds, cloud shadows, no data, ...)
- **temporally resampled** at regular interval

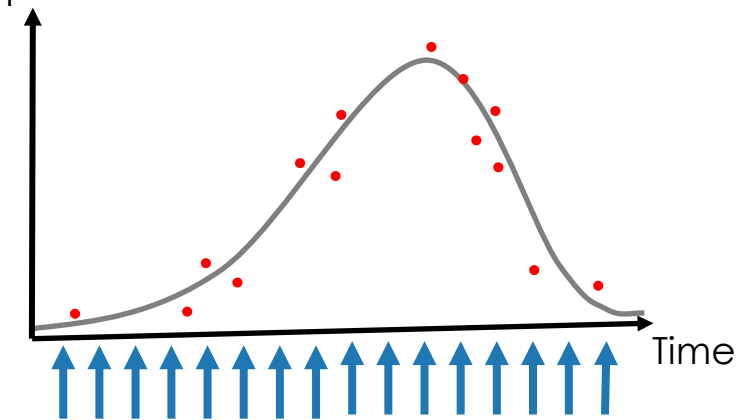
⇒ This approach provides the same “observation” dates over the whole area independently from the sensor orbits and cloud cover which is very critical to run most of the classification methods.

Winter crop



Summer crop

Reflectance
NDVI
SAR
...



Challenges for large scale crop mapping

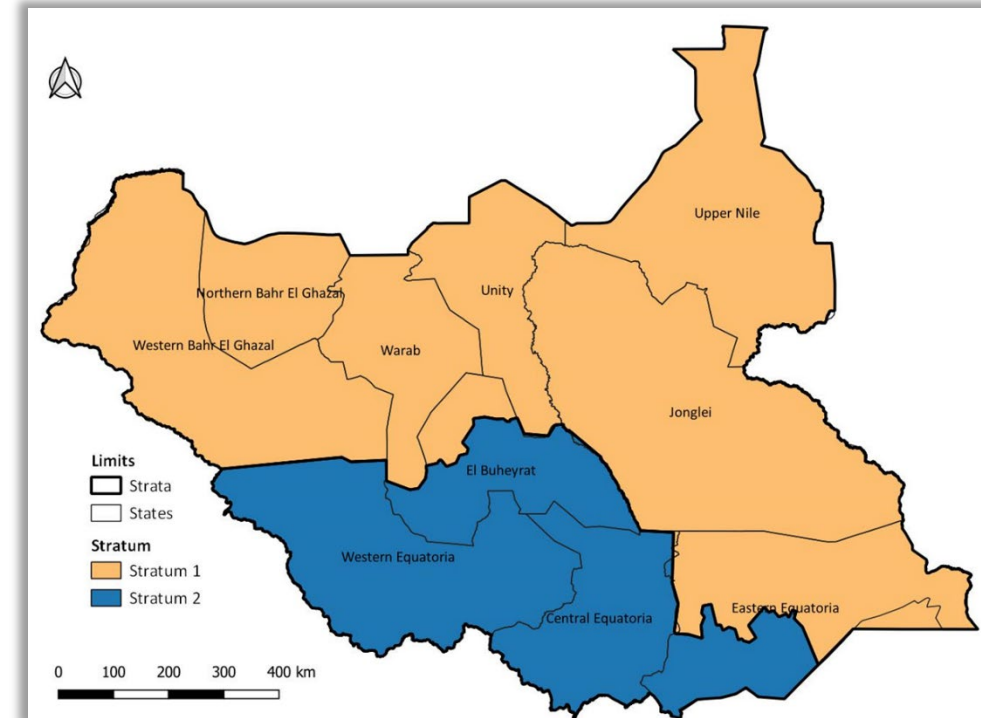
Example of South Sudan

Major rainfall gradient across the country defining two different crop calendars



	Jan	Feb	Mar	Apr	May	June	July	Aug	Sept	Oct	Nov	Dec
Rainfall in the north	Dry season				Wet season				Dry season			
Rainfall in the south	Dry season				Wet season				Dry season			
Unimodal production	Main crop				Land preparation and planting	Growing season			Harvest			
	Long-cycle crop	Harvest			Land preparation and planting	Growing season					Harvest	
Bimodal production (only in the south)	First crop			Land preparation and planting	Growing season			Harvest				
	Second crop	Harvest						Land preparation and planting	Growing season		Harvest	

Stratification in zones to be classified separately



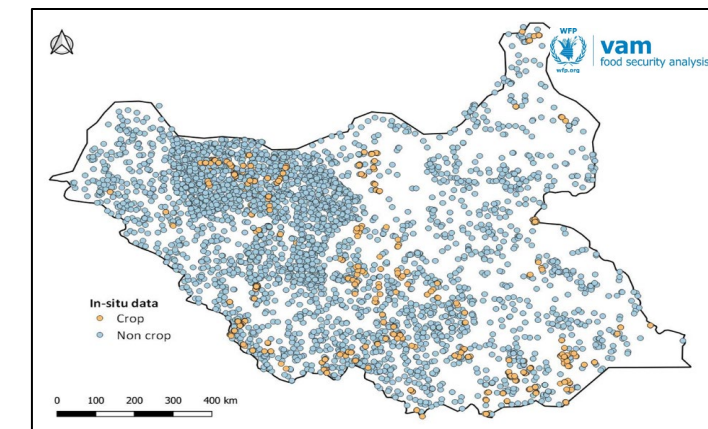
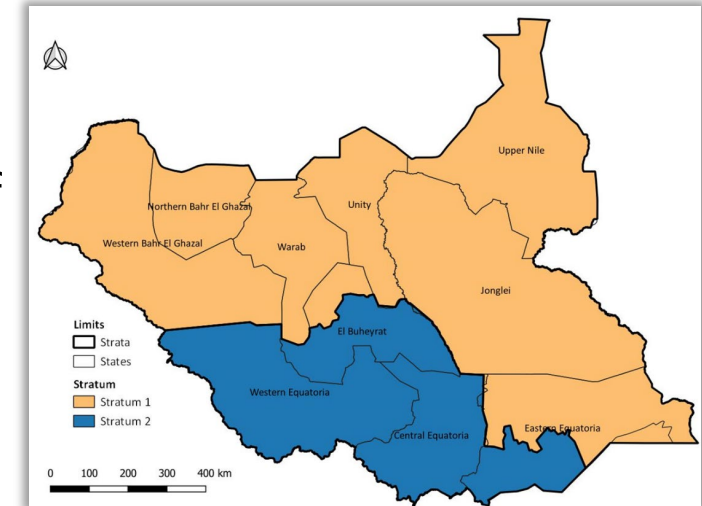
Challenges for large scale crop mapping

Example of South Sudan – stratification

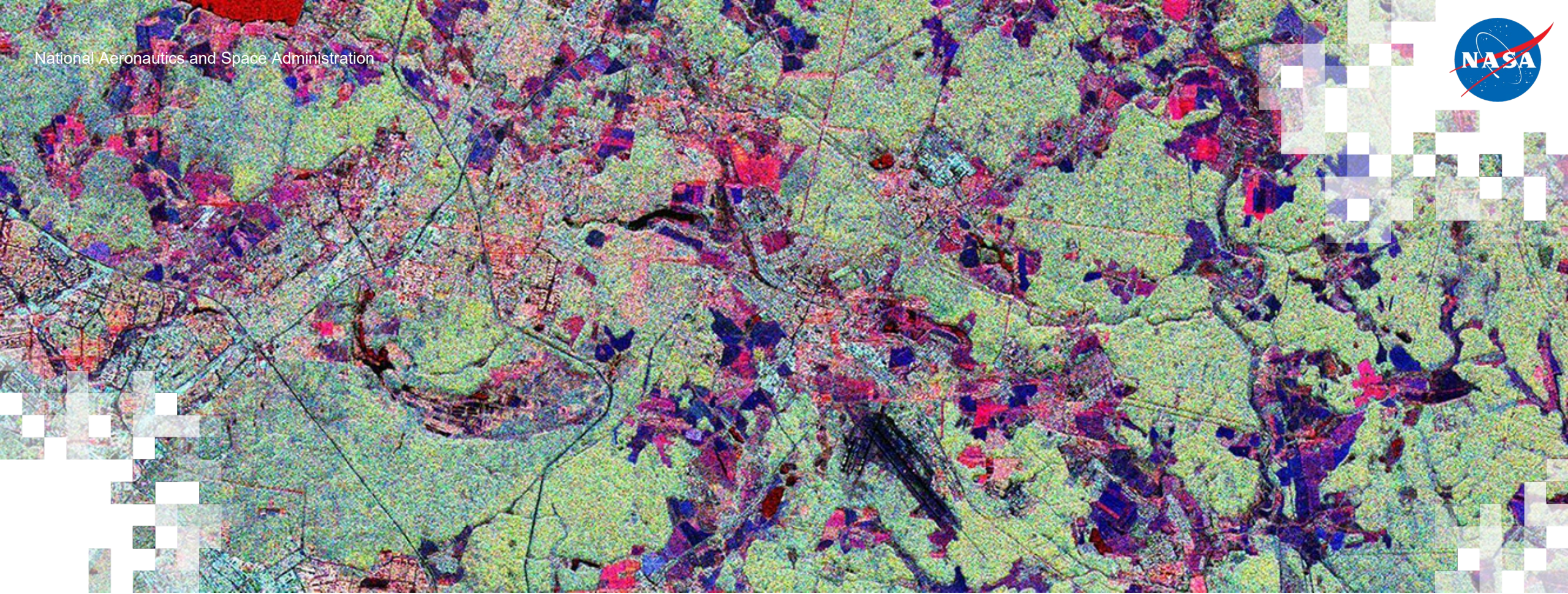
A model to calibrate for each stratum requires collection of enough ground samples for each zone covering all crop types but also all non-cropland classes.

Crop / Non crop	Land cover class	Nr of samples	Proportion in crop / no crop categories	Proportion in total
Non crop	Bush	1504	44%	33%
	Forests	566	17%	12%
	Grasslands and meadows	565	16%	12%
	Shrubland	408	12%	9%
	Wetland	93	3%	2%
	Water bodies	78	2%	2%
	Fallow	76	2%	2%
	Built-up surface	71	2%	2%
	Bare soil	64	2%	1%

Crop type	Nr of samples	Proportion in crop / no crop categories	Proportion in total
Early Sorghum	369	31%	8%
Late Sorghum	244	21%	5%
Maize	122	10%	3%
Groundnut	116	10%	3%
Early Sorghum Mixed	105	9%	2%
Cassava	79	7%	2%
Late Sorghum Mixed	55	5%	1%
Maize Mixed	48	4%	1%
Sesame	13	1%	0%
Rice	10	1%	0%
Bulrush Millet	9	1%	0%
Beans	7	1%	0%



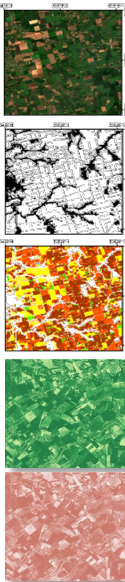
=> 4602 samples well distributed



Section 2: Large Scale Crop Mapping Using Sen4Stat Toolbox

ESA open-source SAR and optical toolbox for operational crop type mapping and monitoring over very large areas

- Process Sentinel-1, Sentinel-2 and Landsat-8&9 time series according to the state-of-the-art including advanced SAR products (coherence, gamma naught,...)
- Deliver automatically or on request **5 types of products** (*processors*) in near real-time along the satellite data acquisition or off-line:
 1. **10m optical cloud free temporal synthesis and SAR temporal synthesis**
 2. **Time series of spectral indices** (NDVI, coherence,...) and **biophysical variables** (LAI, fCover, fAPAR)
 3. **10m crop type maps** along the season based on in situ dataset and stratification
 4. **Large set of crop growth conditions metrics** (including meteorological data)
 5. **Crop yield estimation** at various aggregation levels (national, regional, ...)





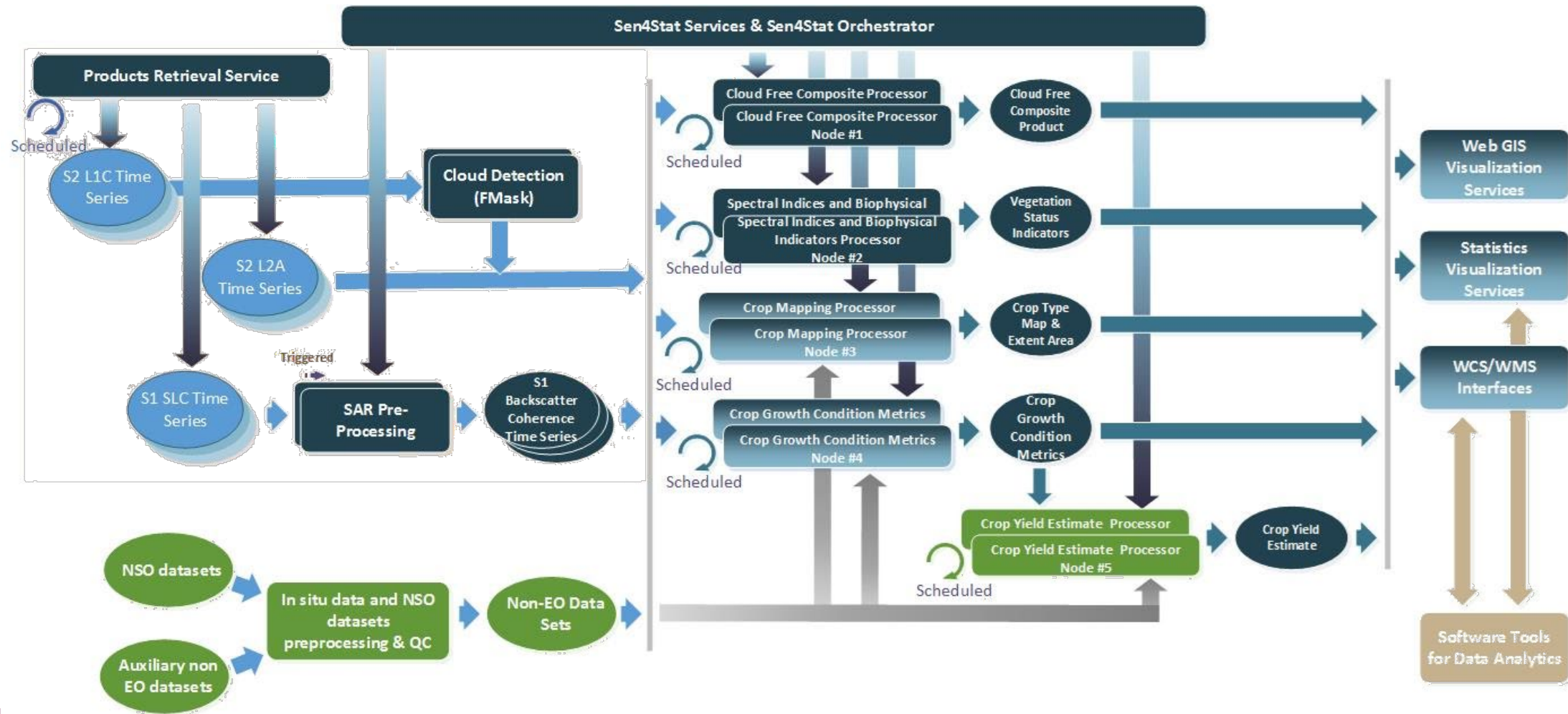
ESA open-source SAR and optical toolbox for operational crop mapping and monitoring over very large areas

Data access /download

Pre-processing In situ quality control

Processors of products

Analytics and visualization



ESA open-source SAR and optical toolbox with a graphical user interface to configure the system to launch the production and monitor the processes

System configuration

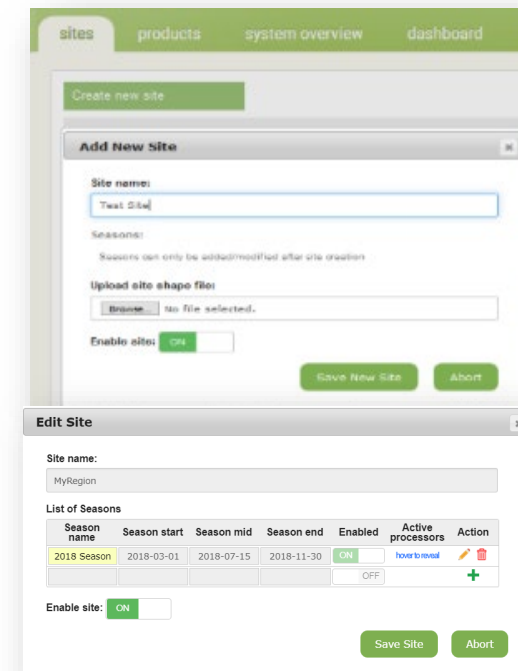
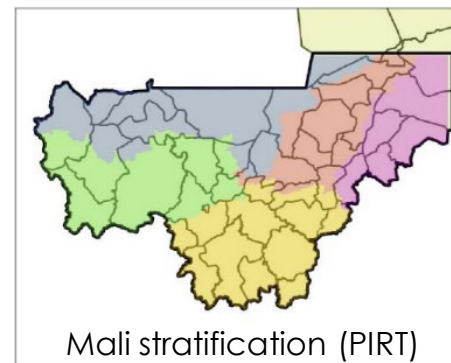


Start of the season

End of the season...

Sen4Stat: parameters settings

Area of Interest	Shapefile to be uploaded
Monitoring period	Start and end dates to be defined
S1 - S2 - L8&9	To be selected
Data sources	ESA&USGS – AWS – DIAS – local storage...



Sen4Stat: field campaign

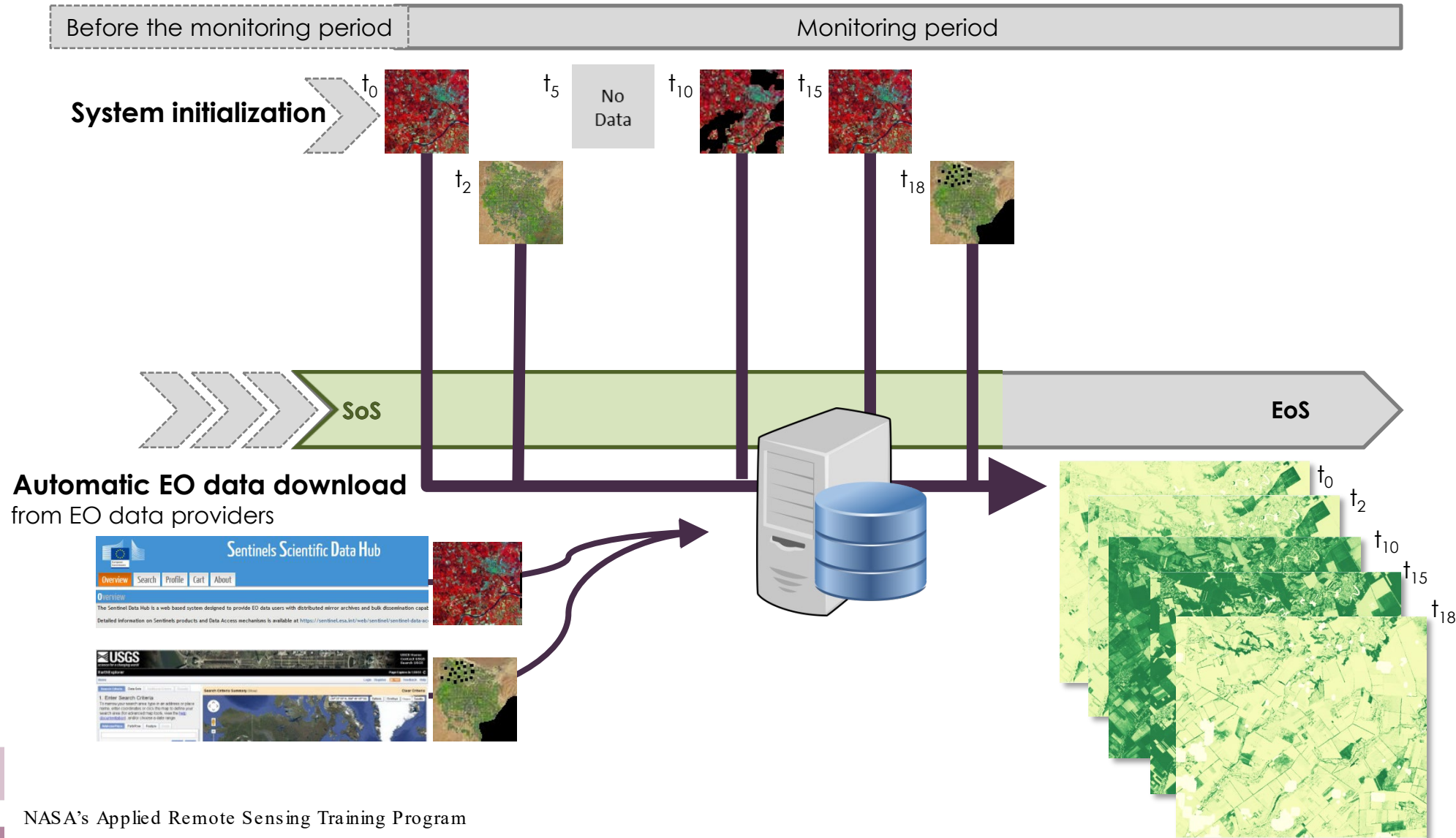
Sampling design	Stratification and sampling
Field visit	In situ data collection – early survey
	In situ data collection – mid-season survey
Data upload	Field data quality control and formatting

File	Shape	ID	CROP	LC	CODE	IRRIGATION
1	Polygon	1.1	Spring Wheat		112	0
1	Polygon	2.1	Spring Wheat		112	0
2	Polygon	3.1	Spring Wheat		112	0
3	Polygon	4.1	Spring Wheat		112	0
4	Polygon	5.1	Triticale		1911	0
5	Polygon	6.1	Triticale		1911	0
6	Polygon	7.1	Triticale		1911	0
7	Polygon	8.1	Triticale		1911	0

Cloud Computing

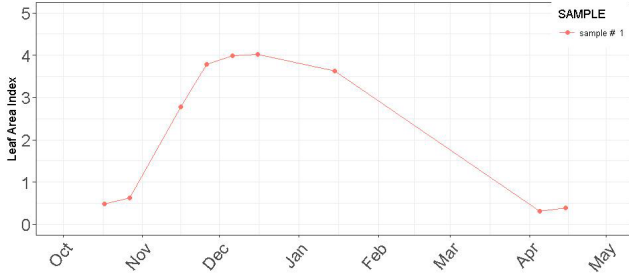


– system operation for crop growth monitoring : Leaf Area Index (LAI) production in near-real time along the season

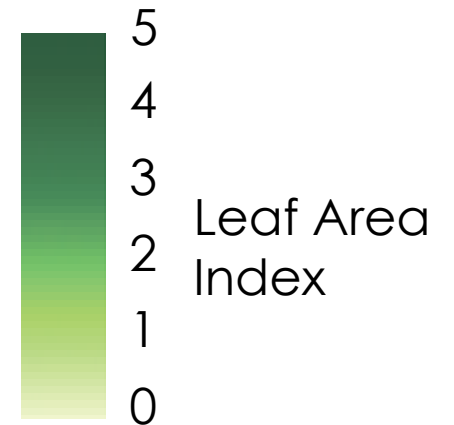
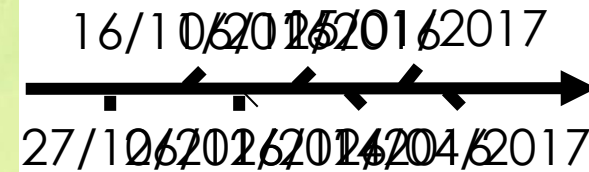
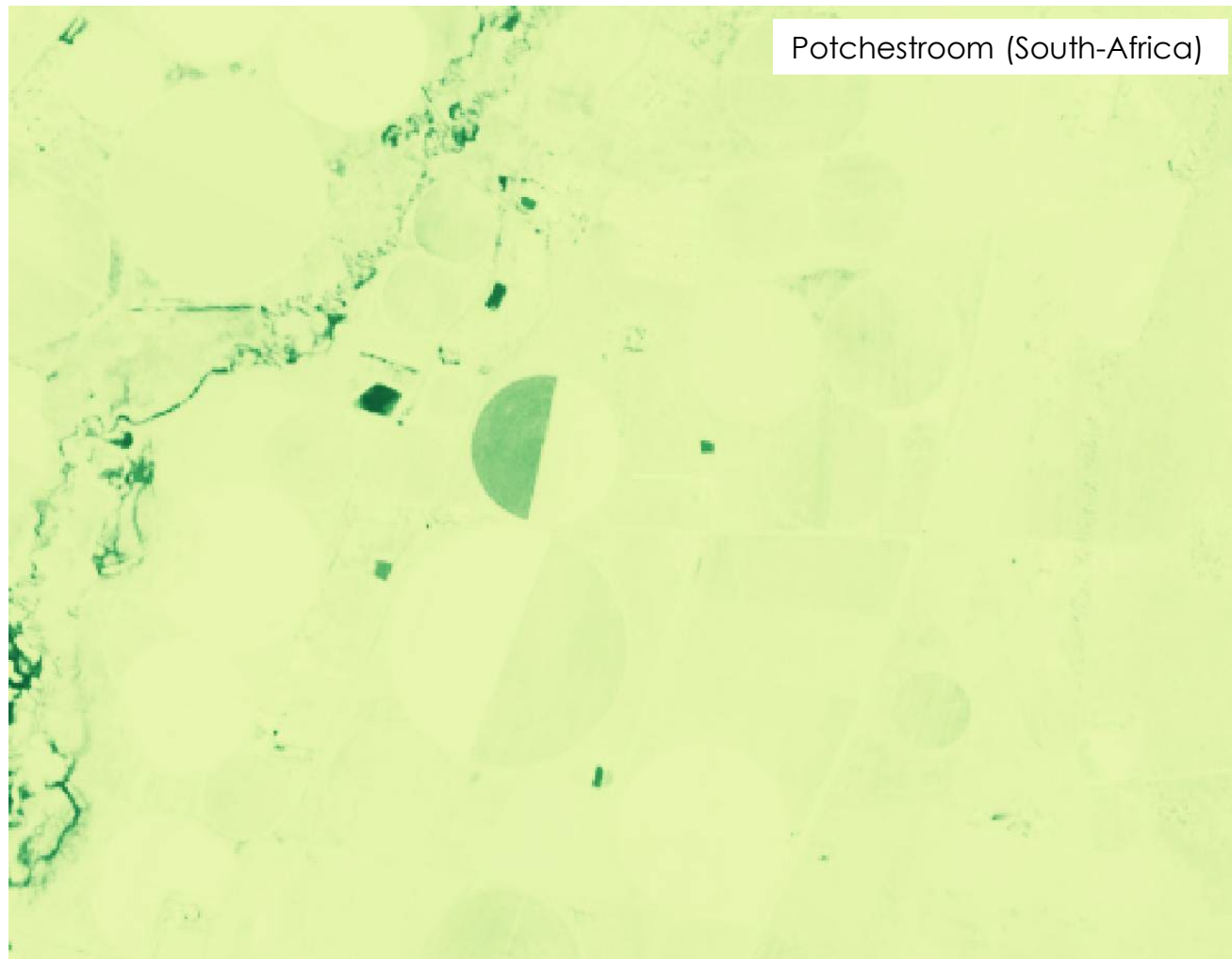


System operation for crop growth monitoring : **Leaf Area Index (LAI)** production in near real-time along the season

LAI profile for a Maize field near Potchefstroom - South Africa



Sen4Stat: 2016 – 2017 Leaf Area Index at 10 m



– System operations running according to different modes:
NRT with orchestrator (fully automated) or on request

Automated mode through the web graphical user interface (GUI)

- a) based on the Orchestrator with by-default parameterization, automatic data download and processing until the end of the season, on-time delivery
=> **operational production in near real time (NRT)**
- b) Processor execution on user request, with by-default parameterization

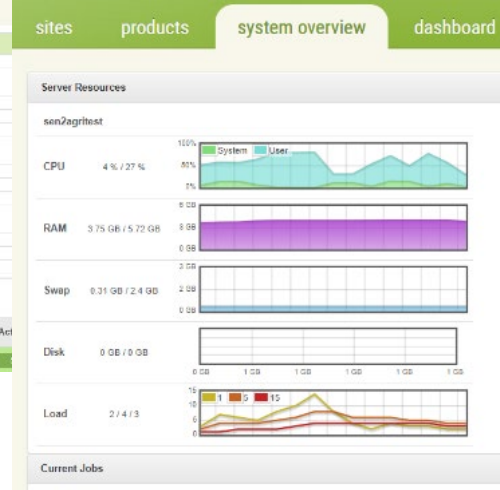
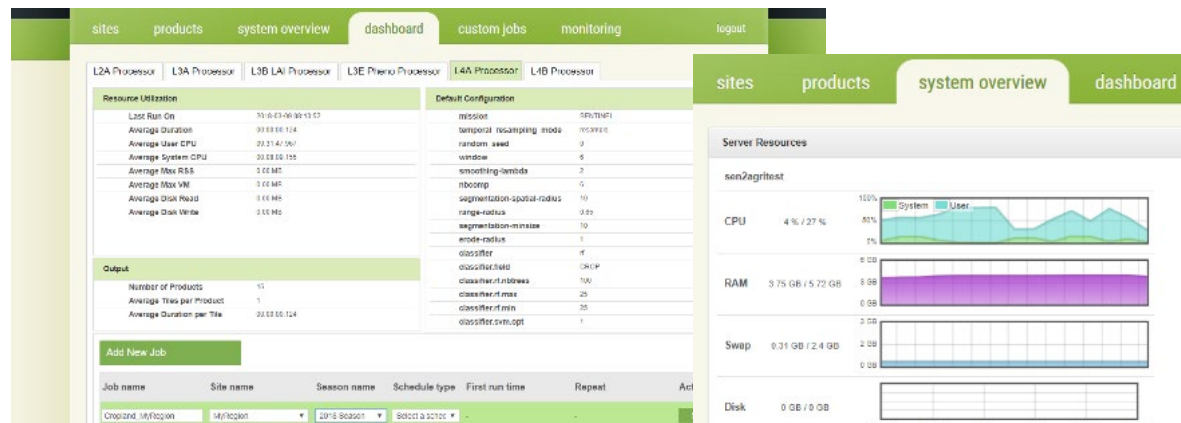


Manual mode: to run processor independently, with custom parameters

- a) through the GUI, with the *Custom job* approach
- b) through SNAP software (only processor of Level 3 and 4)
- c) in command line through a Linux terminal

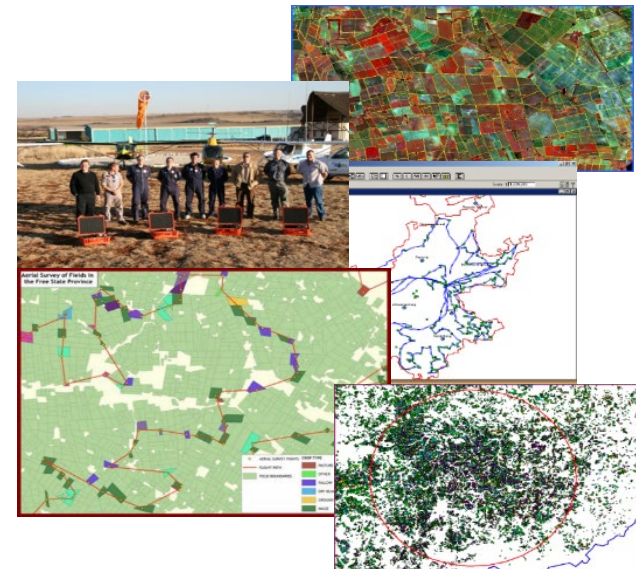
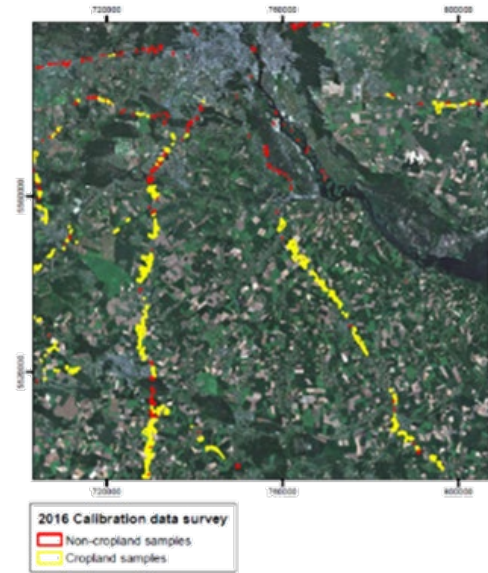
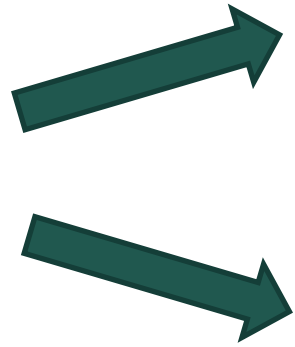
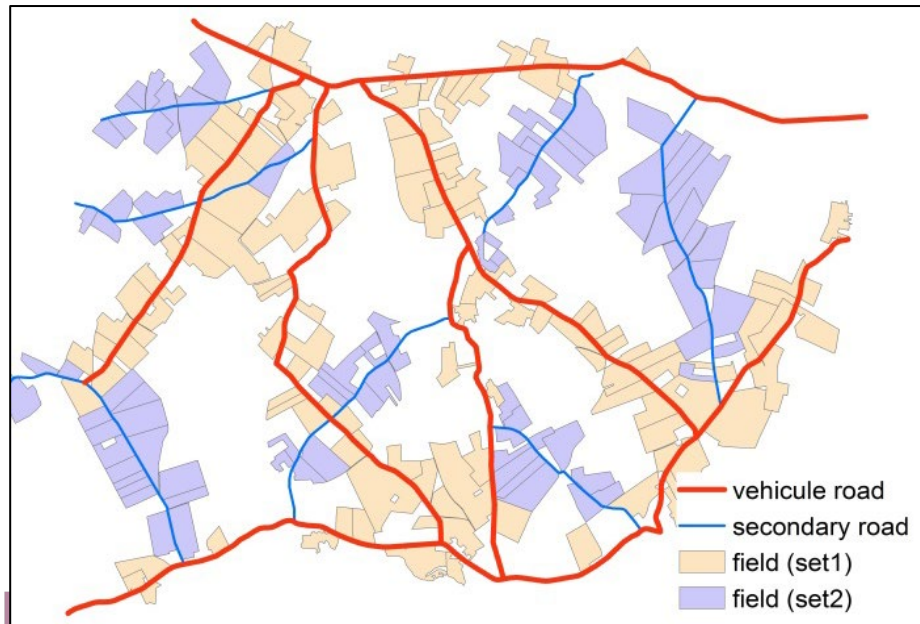
```

lai_retrieve_processing.py --input
/mnt/archive/maccs_def/mali/l2a/S2A_OPER_PRD_MSIL2A_PDMC_20160718T093045_R008_V2016
0717T104833_20160717T104833_SAF/S2A_OPER_SSC_L2VALD_30PVV___20160717.HDR --res 10
--outdir /mnt/archive/temp/test_lai --nsrCFG /usr/share/sen2agri/nsr_cfg.txt --
modelsfolder /mnt/archive/temp/test_lai --generatemodel YES --generatemonodate YES -
-genreprocessedlai NO --genfittedlai NO
    
```





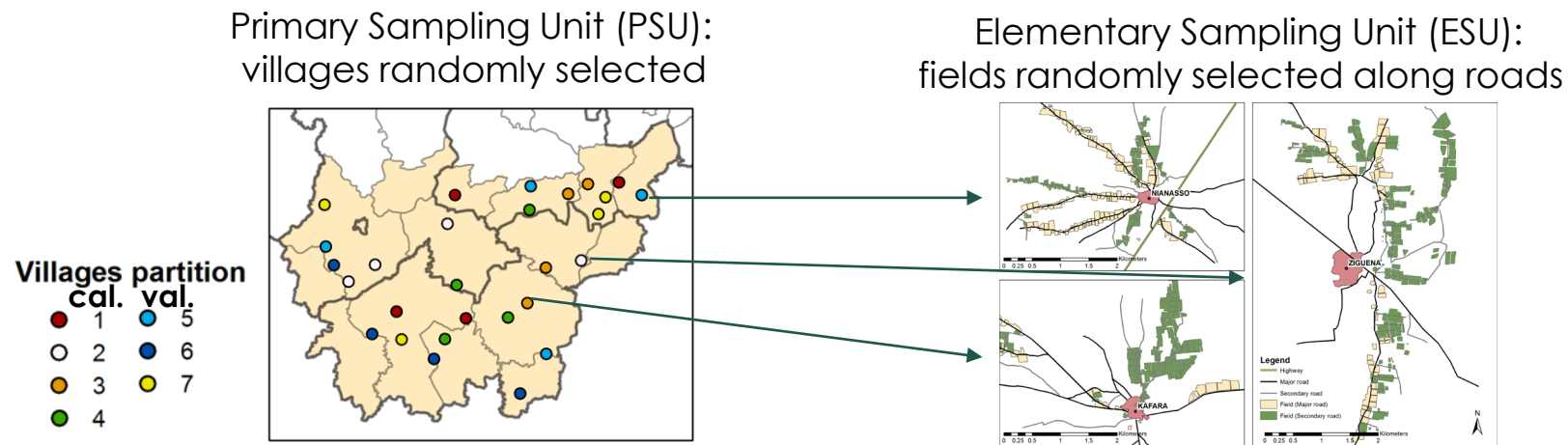
- ❖ Available reference dataset collected according to an area or list sampling frame (national statistical surveys, Ministry of Agriculture,...)
- ❖ JECAM Guidelines – ad hoc “windshield SURVEY” by motorized vehicles (car, motorbike) selecting a set of appropriate roads (set 1) and complemented by regular additional transect (set 2) using secondary roads and tracks to try to reduce the spatial bias brought about by roadside sampling.



VLA used for aerial observations



- ❑ In situ data for **calibration (training)**: sampling to cover the diversity of situations existing in the study area (possibly a national territory) in order **to cover the range of possible signatures for the different elements of interest** (i.e. most crop types and the main non cropland classes).
- ❑ In situ data for **validation** to estimate the products accuracy (with a confidence interval) using a **statistically-sound sampling to be objective and independent**; for logistic reasons, sampling not strictly random but 2-stage sampling (with PSU and ESU) to assess the crop types (one field campaign)



⇒ calibration and validation field campaigns for crop type mapping can be combined but the sampling design should be statistically sound to obtain fully independent dataset.



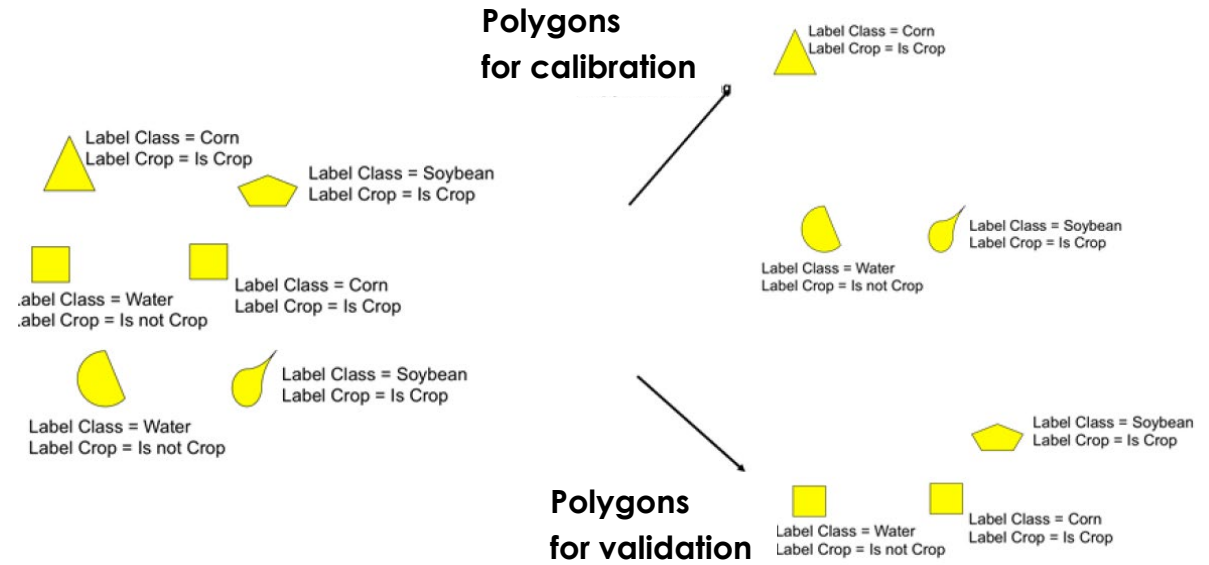
1) High quality reference data collection:

- Crop type acquired from field campaigns
- Non-crop data obtained by visual interpretation of very high spatial resolution imagery available online

2) Quality control: geometry consistency, typology standardisation, homogeneity check...

3) Splitting polygons for calibration (25%) and validation (75%) for each class (alternative strategies might be required in case on poorly balanced in situ dataset)

4) Enhancing the samples number for marginal crop type using Synthetic Minority Over-sampling Technique (SMOTE)





The **Overall Accuracy** (OA) is calculated as the total number of correctly classified pixels (n_{ii} ~ diagonal elements of the confusion matrix) divided by the total number of test pixels:

$$OA = \frac{\sum_{i=1}^r n_{ii}}{\sum_{i=1}^r \sum_{j=1}^r n_{ij}}$$

Precision or **User's Accuracy** (UA) for the class i is the fraction of correctly classified pixels with regard to all pixels classified as this class i in the classified image:

$$UA_i = \frac{n_{ii}}{\sum_{j=1}^r n_{ij}}$$

Recall or **Producer's Accuracy** (PA) for the class i is the fraction of correctly classified pixels with regard to all pixels of that ground truth class i :

$$PA_i = \frac{n_{ii}}{\sum_{j=1}^r n_{ji}}$$

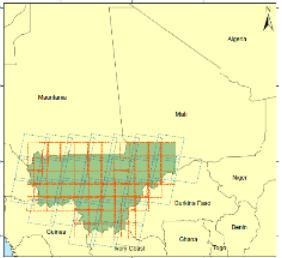
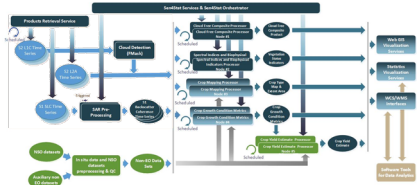
F-Score (also known as F-1 score or F-measure) is the harmonic mean of the Precision and Recall and reaches its best value at 1 and worst score at 0:

$$FScore = 2x \frac{Precision * Recall}{Precision + Recall}$$

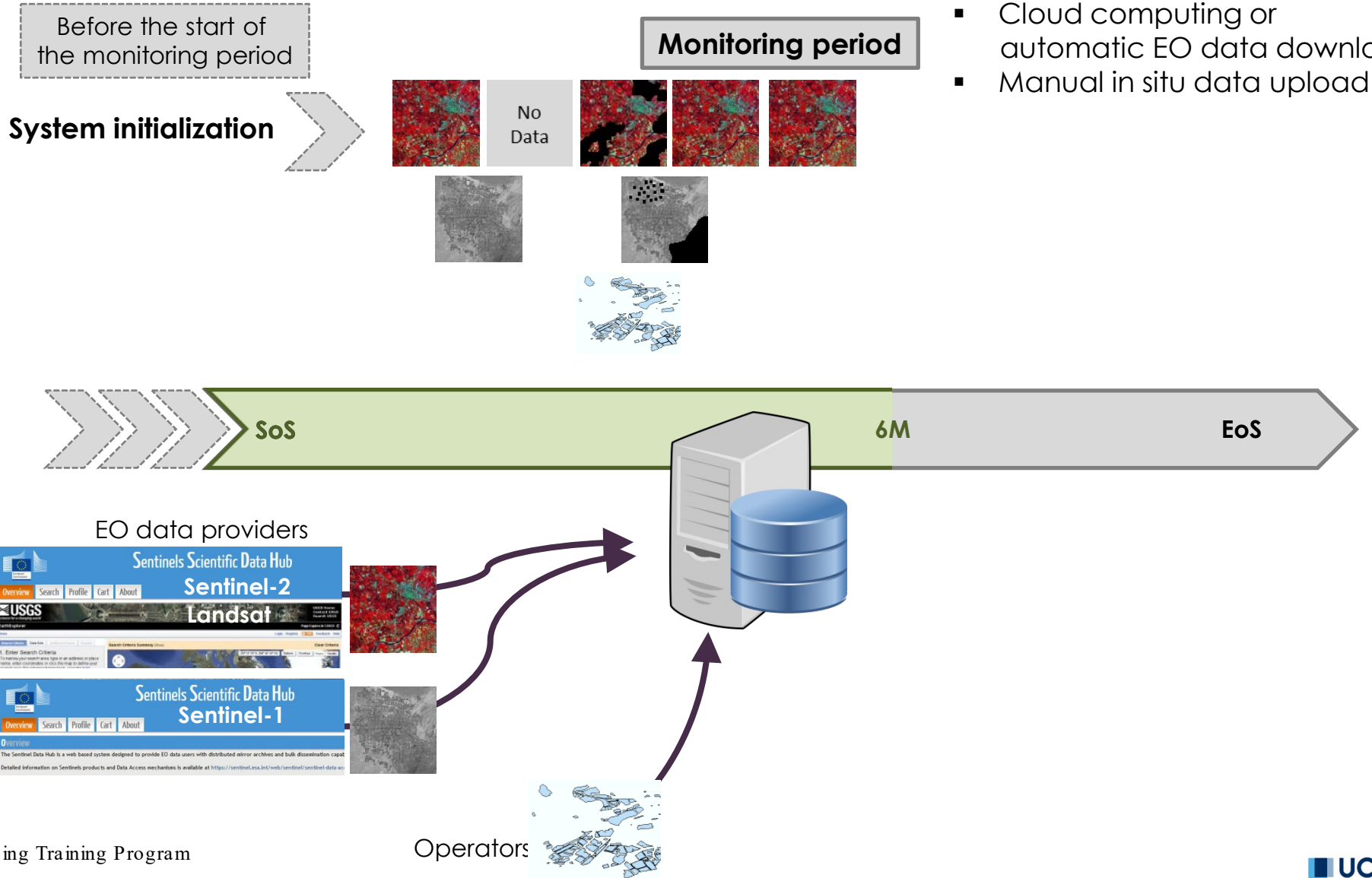
		Reference Data ^a						
		W	S	F	U	C	H	Row Total
Classification data								
W		226	0	0	12	0	1	239
S		0	216	0	92	1	0	309
F		3	0	360	228	3	5	599
U		2	108	2	397	8	4	521
C		1	4	48	132	190	78	453
H		1	0	19	84	36	219	359
	Column total	233	328	429	945	238	307	2480
Producer's Accuracy								User's Accuracy
W = 226/233 = 97%								W = 226/239 = 94%
S = 216/328 = 66%								S = 216/309 = 70%
F = 360/429 = 84%								F = 360/599 = 60%
U = 397/945 = 42%								U = 397/521 = 76%
C = 190/238 = 80%								C = 190/453 = 42%
H = 219/307 = 71%								H = 219/359 = 61%
Overall accuracy = (226 + 216 + 360 + 397 + 190 + 219)/2480 = 65%								

^aW, water; S, sand; F, forest; U, urban; C, corn; H, hay.

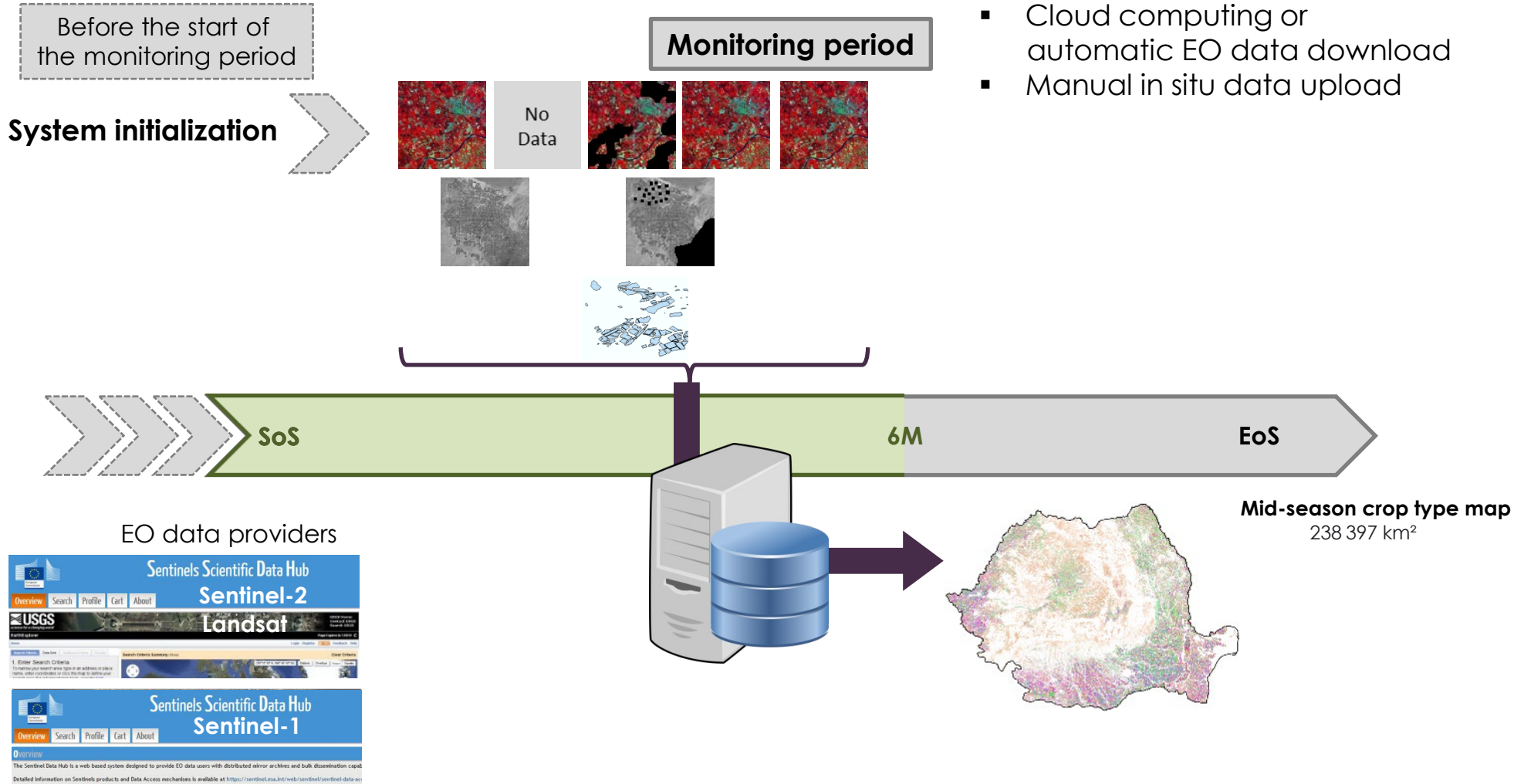
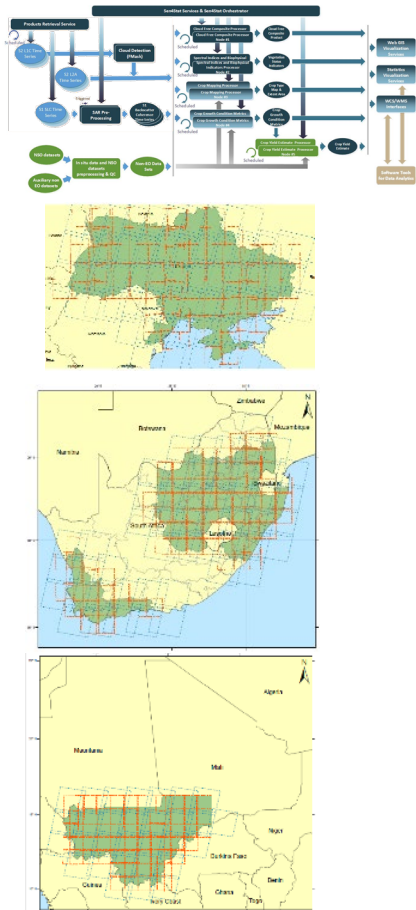
System operation in automated mode to deliver in NRT 10-m crop type maps at national scale



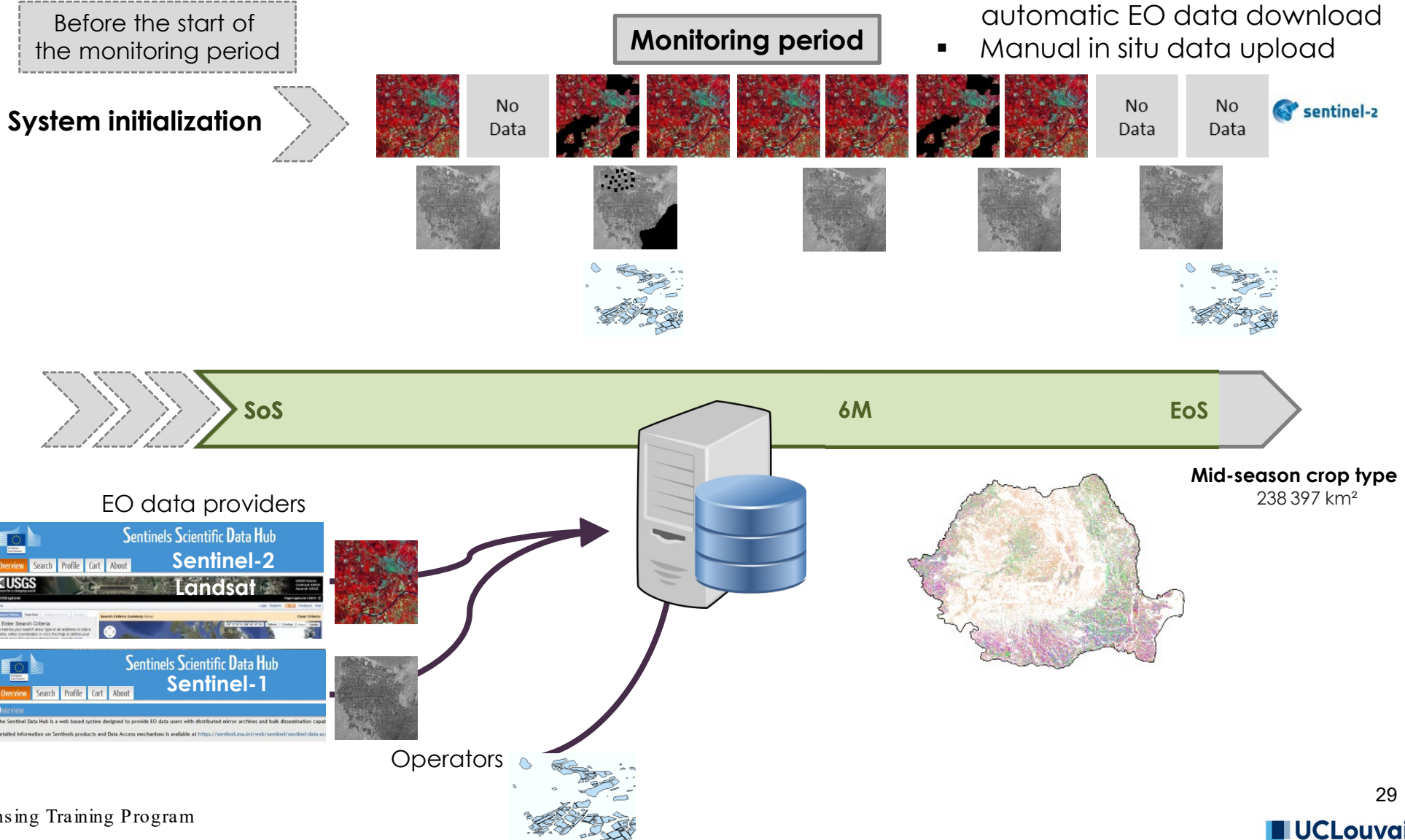
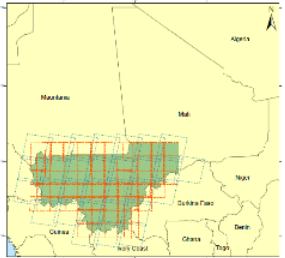
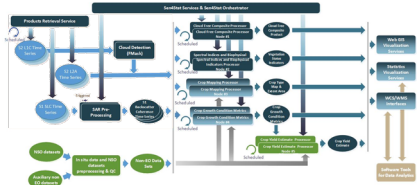
(Defourny et al., RSE2019)



– System operation in automated mode to deliver in NRT 10-m crop type maps at national scale



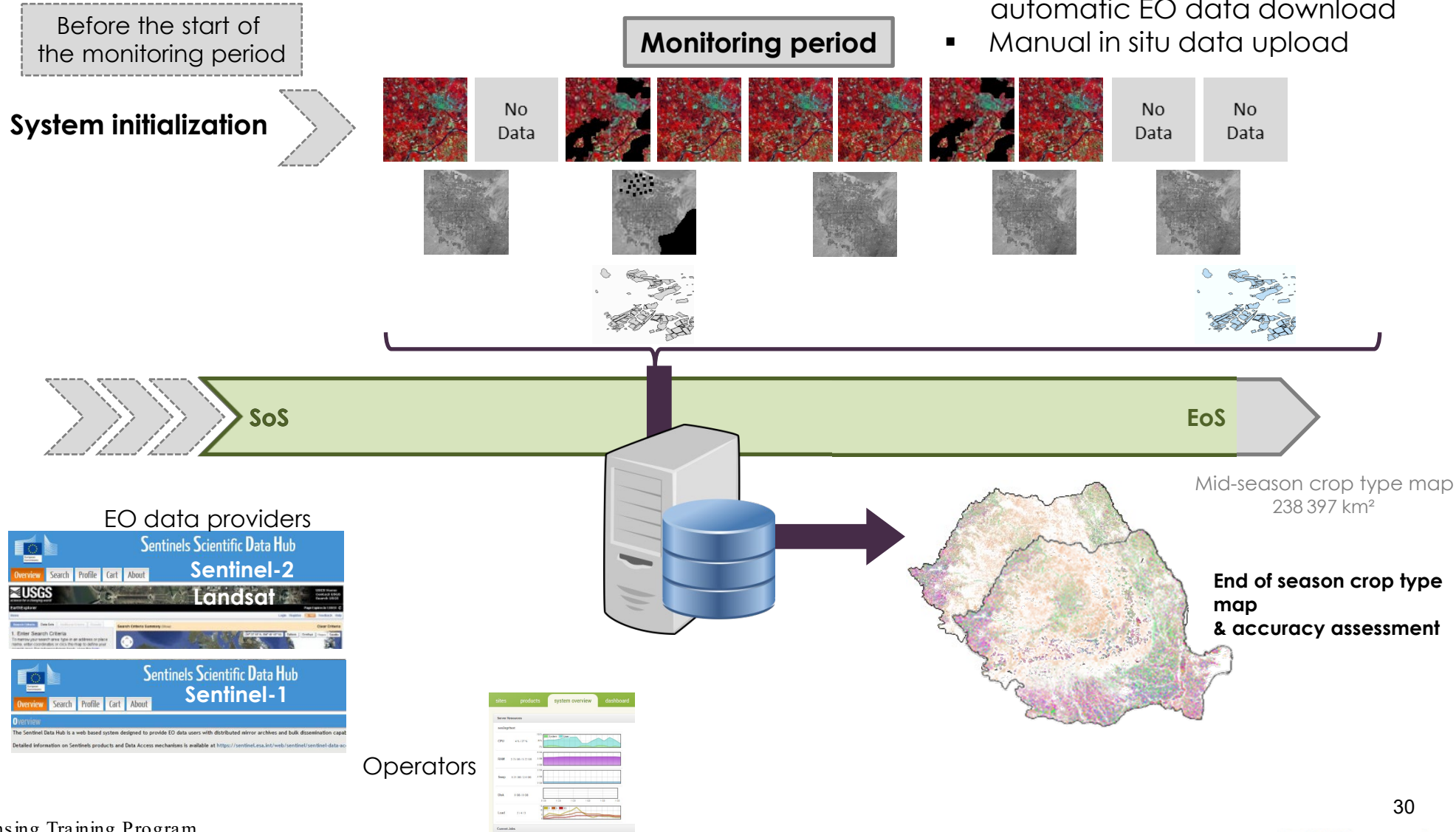
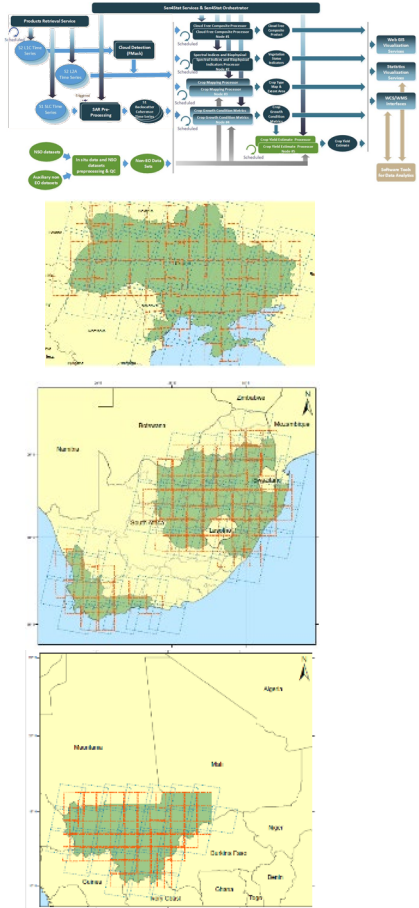
System operation in automated mode to deliver in NRT 10-m crop type maps at national scale



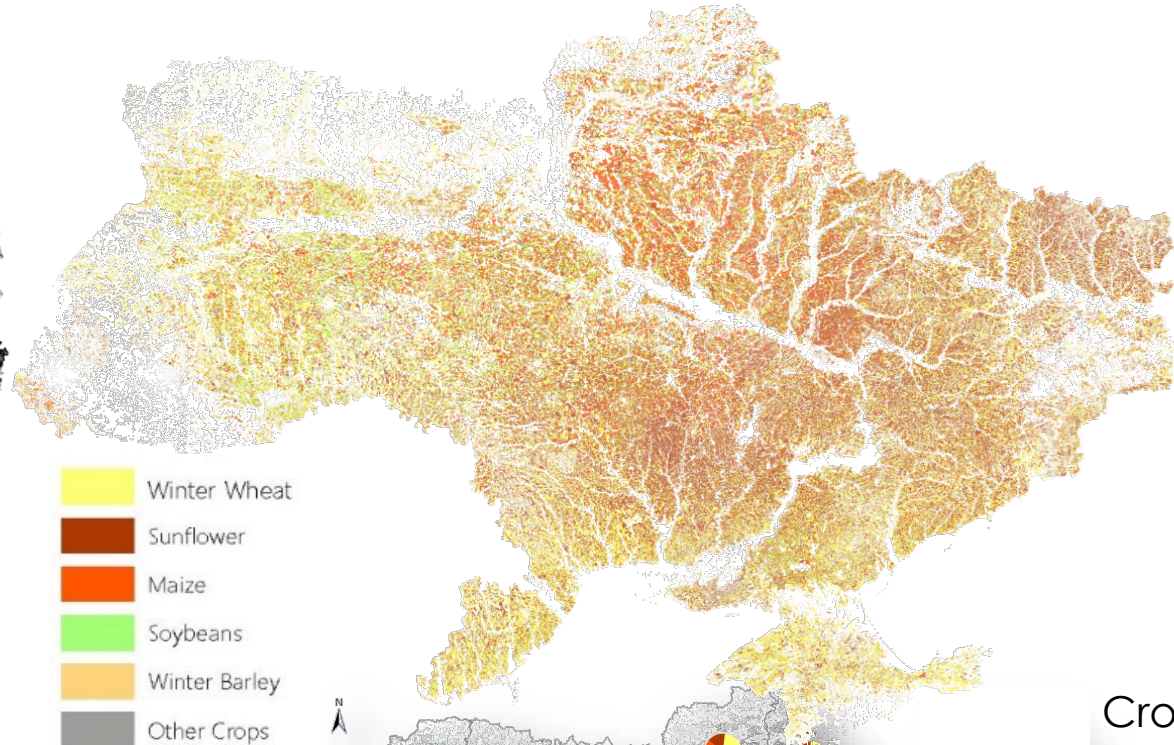
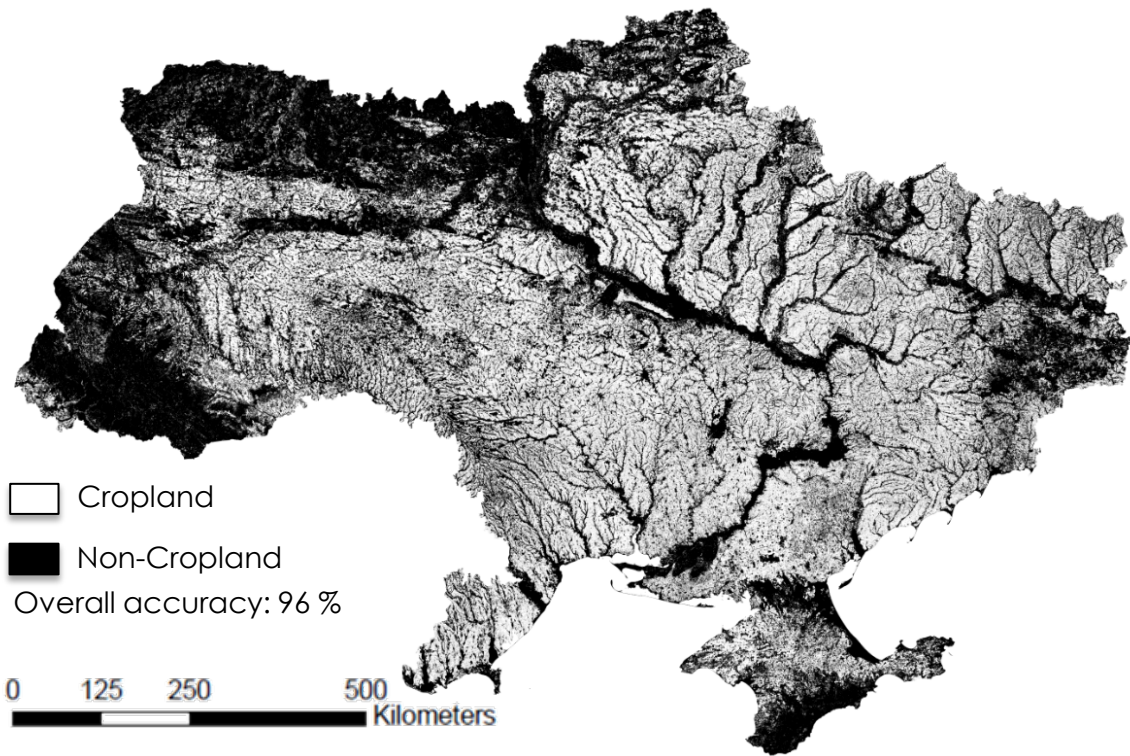
- Cloud computing or automatic EO data download
- Manual in situ data upload

System operation in automated mode to deliver in NRT 10-m crop type maps at national scale

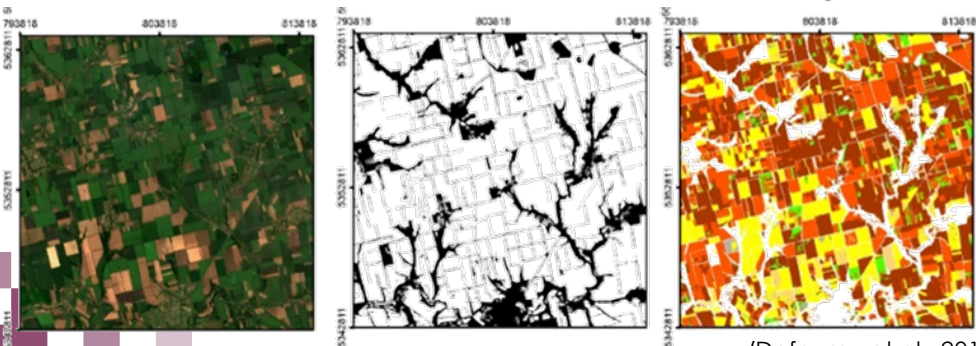
- Cloud computing or automatic EO data download
- Manual in situ data upload



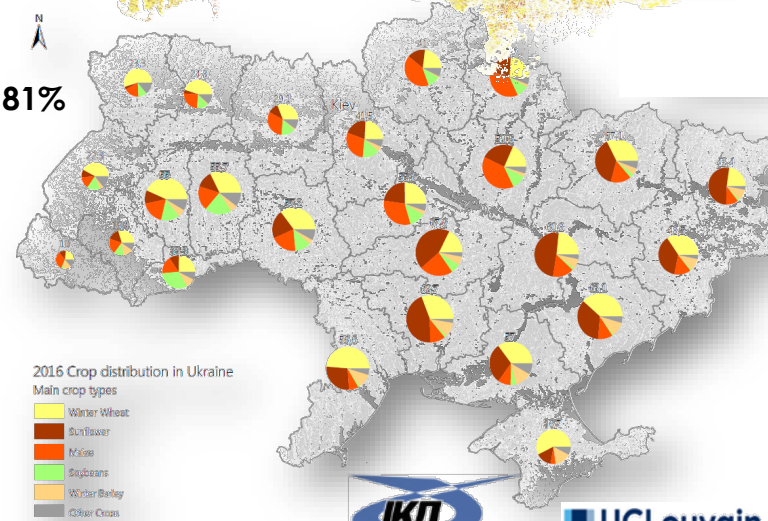
Very first 10-m national cropland and crop type maps based on Sentinel-2a only (Ukraine, July 2016)



Crop area Indicator at mid-season and end of season



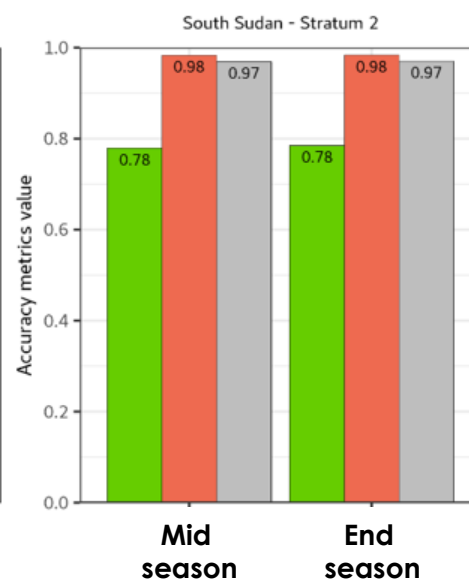
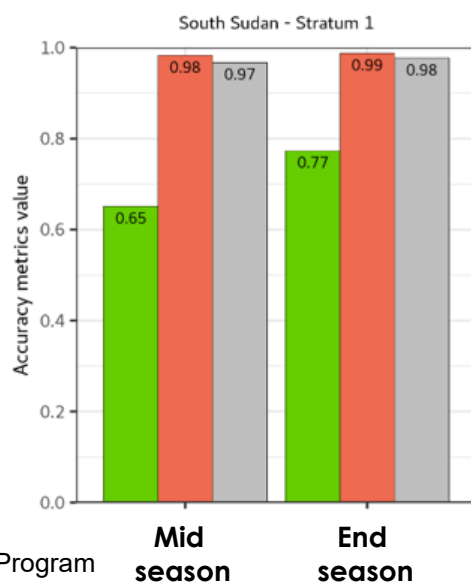
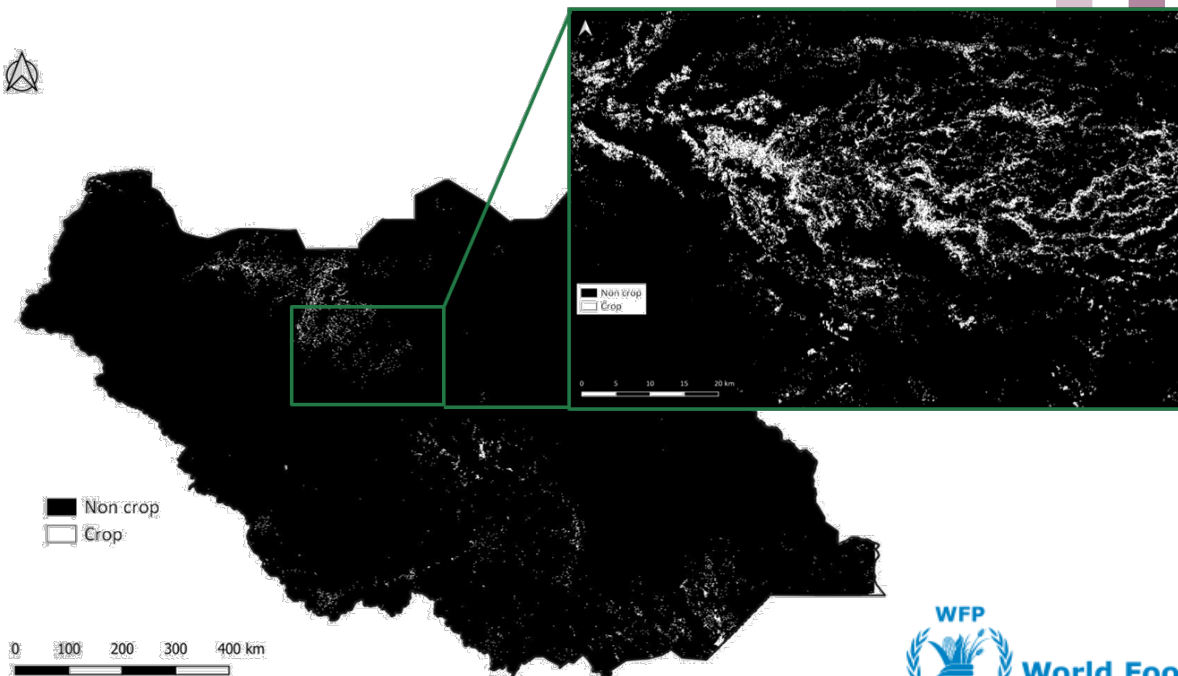
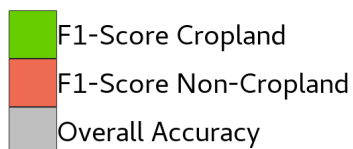
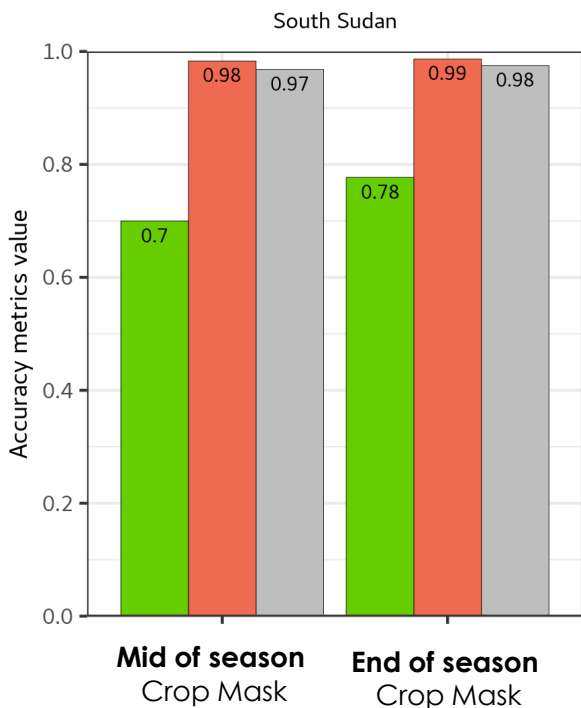
(Defourny et al., 2019)



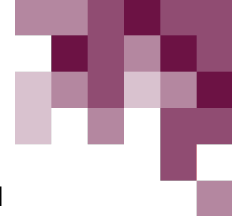


Cropland map at mid and end of the season for the two strata of South Sudan

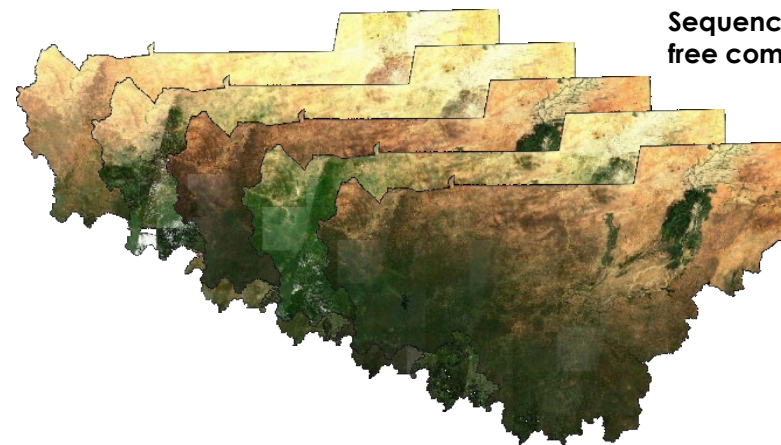
F1-Score of cropland increases with the amount of EO data available along the season



- Crop mapping in Mali (2017)



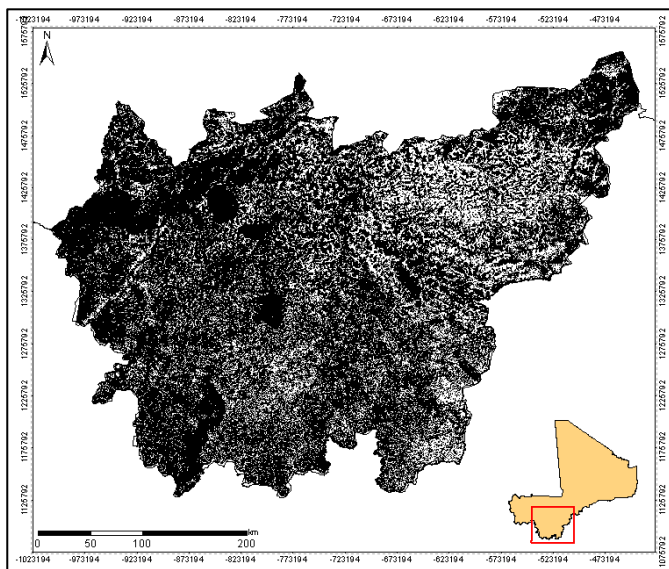
- Successful crop mapping at 10m resolution in collaboration with ICRISAT
- ~4.5 TB of Sentinel-2 data (2152 S2 and 542 L8 images)



Sequence of monthly cloud free composites over Mali

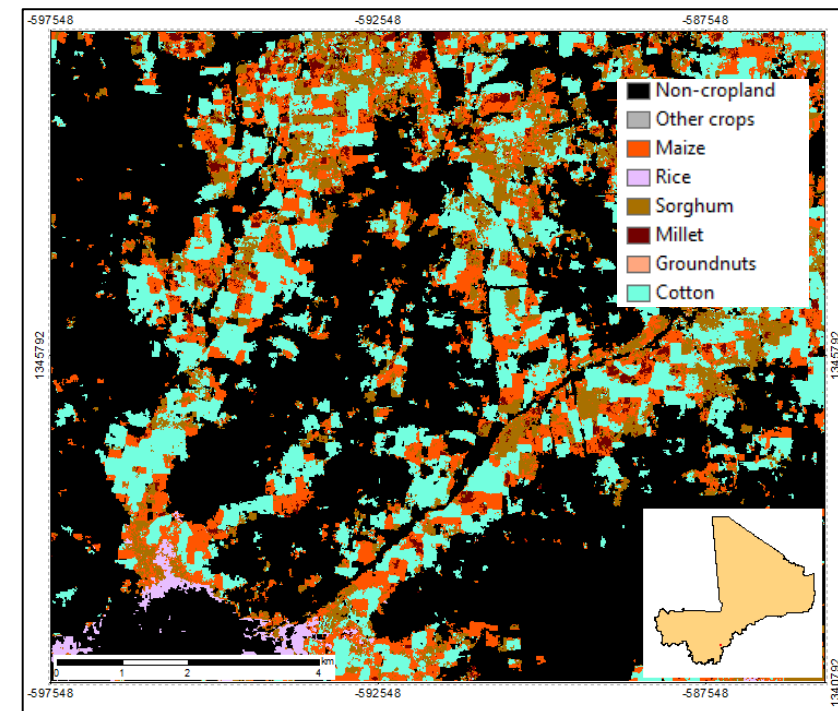
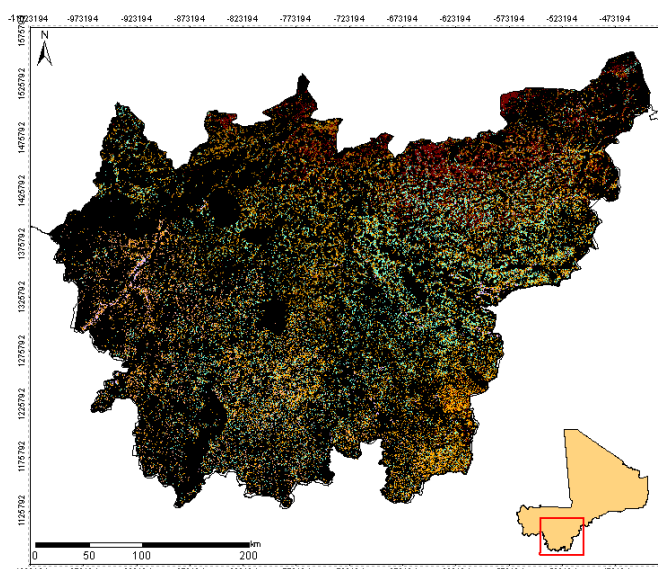
Cropland mask for the 2017 growing season (end-of-season)

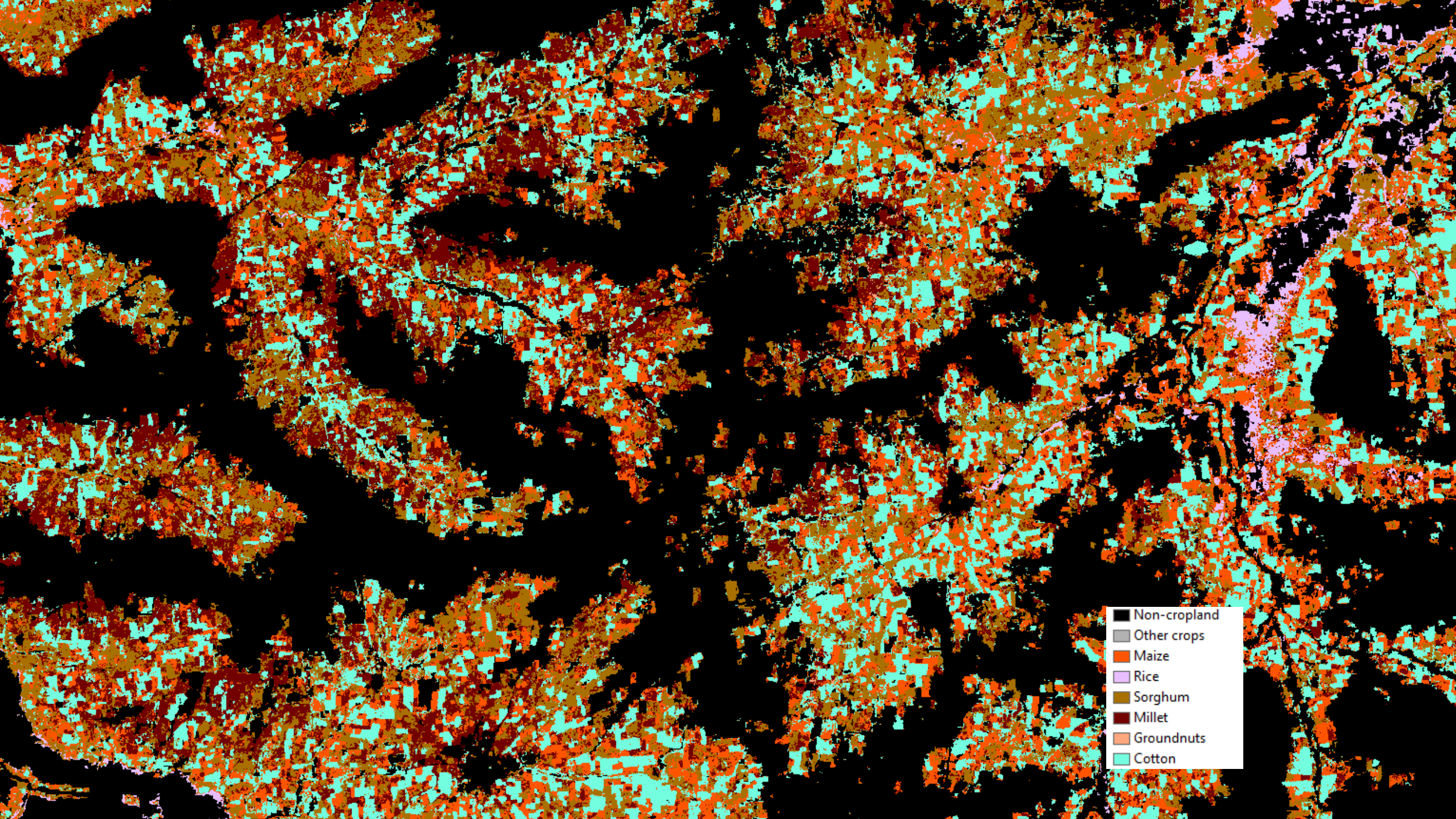
OA = 96%



Crop type map for the 2017 growing season (end-of-season)

OA = 66%

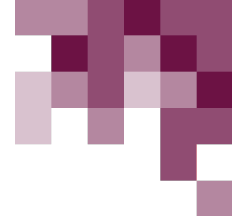




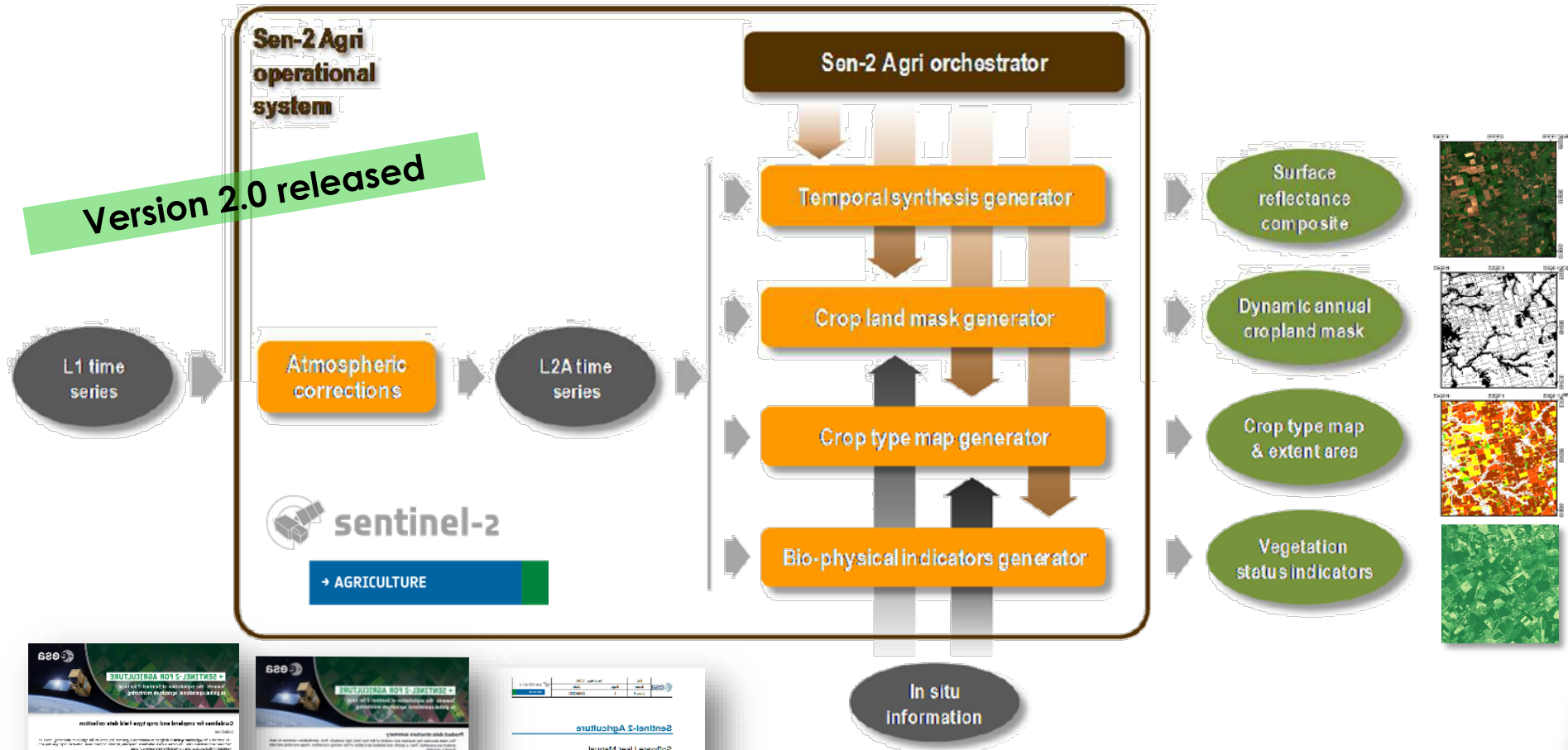
Sen2Agri system: Open source system precursor of Sen4Stat



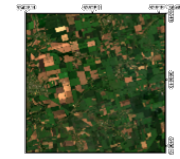
running in NRT or off-line, running locally or in the cloud



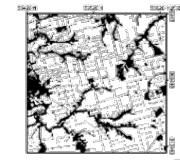
Version 2.0 released



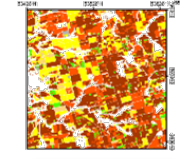
Monthly cloud free surface reflectance composite at 10-20 m



Cropland mask at 10m updated every month



Crop type map at 10 m for the main regional crops

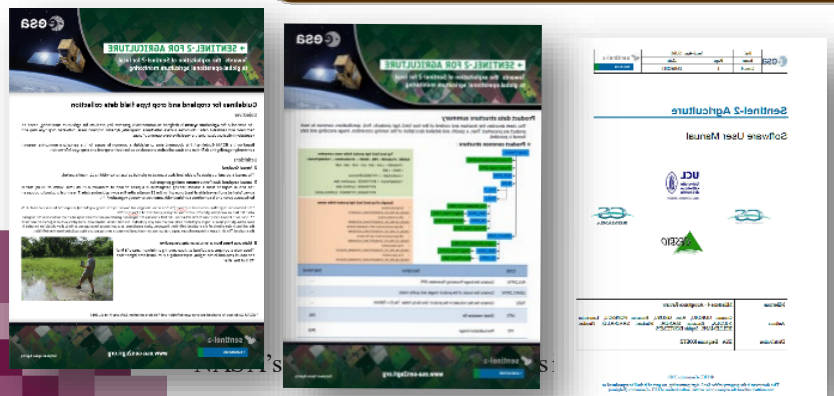


Vegetation status map at 10 m delivered every



week (NDVI, LAI, fCover, fAPAR)

Documented and downloadable at www.esa-sen2agri.org/resources/software/



Sen2Agri as a service on the EODC cloud (e.g. for WFP)



Sen4Stat
Solution for Agricultural Statistics

Open-source operational solution for national processing in near real time and for local applications



Under GNU-GPL License

5 main processors to be possibly combined



Open-source software based on **Orfeo ToolBox**



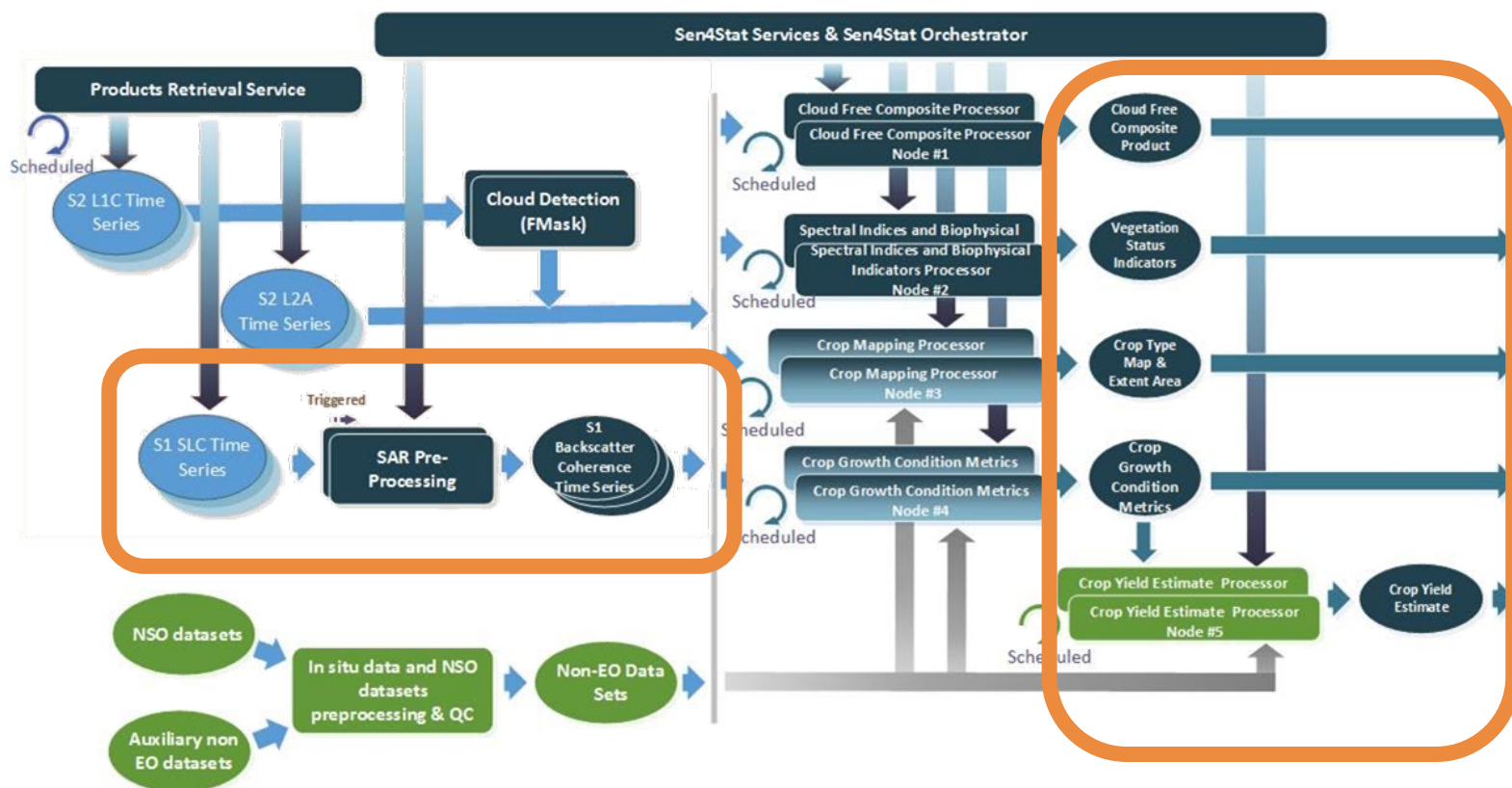
Cluster-ready architecture for distributed processing



Integration with **Sentinel-2 ToolBox**

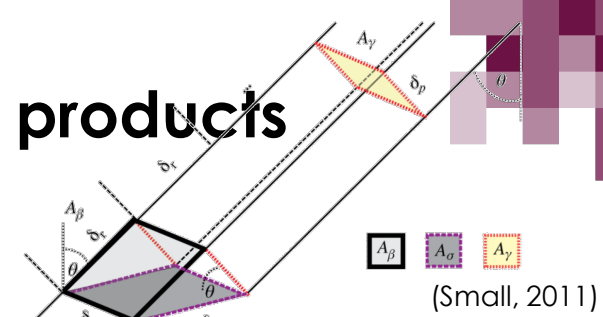


Operational system required: **CentOS 7** (GNU/LINUX)



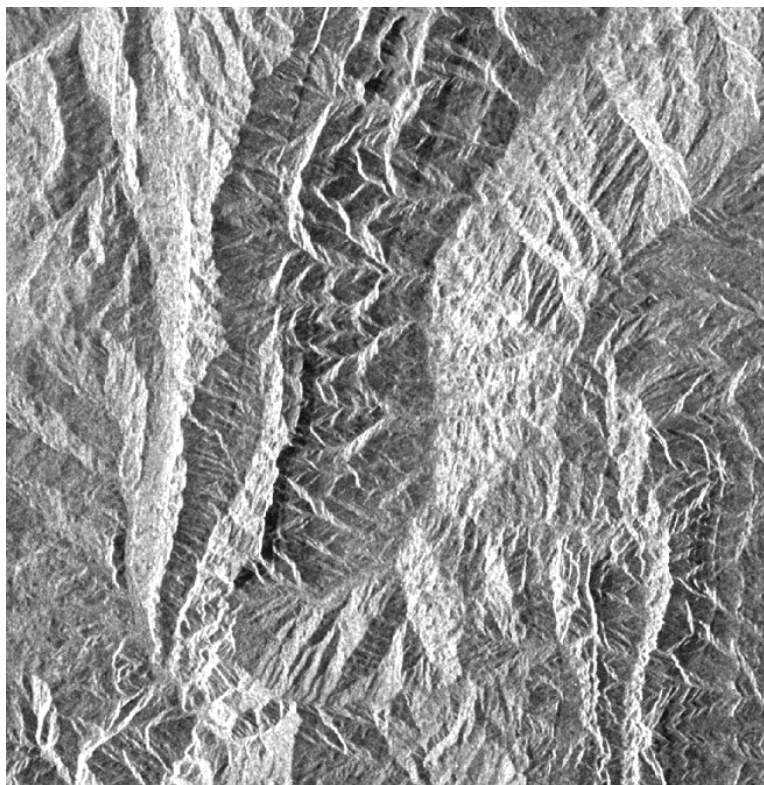
State-of-the-art SAR processing for high level products

State-of-the-art SAR processing for high level products for amplitude and coherence



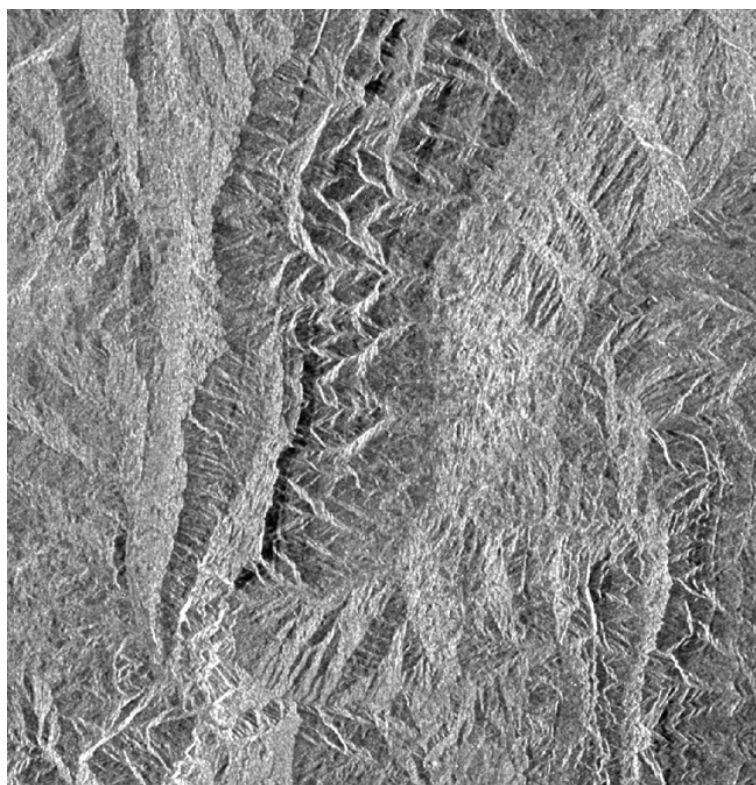
beta nought β_0

(A_β slant range geometry)



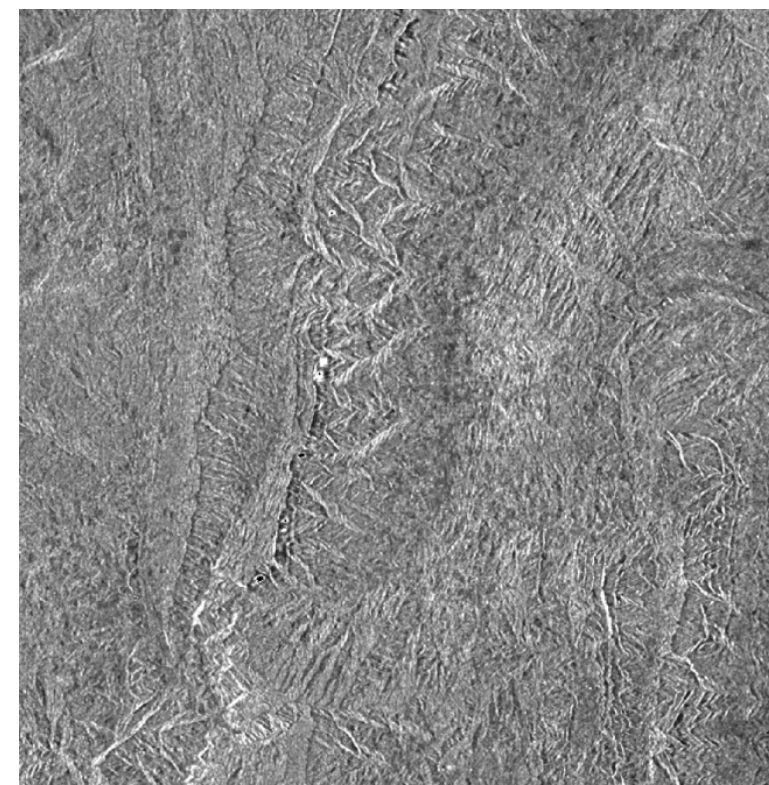
sigma nought σ_0

(A_σ corrected for local inc. angle)



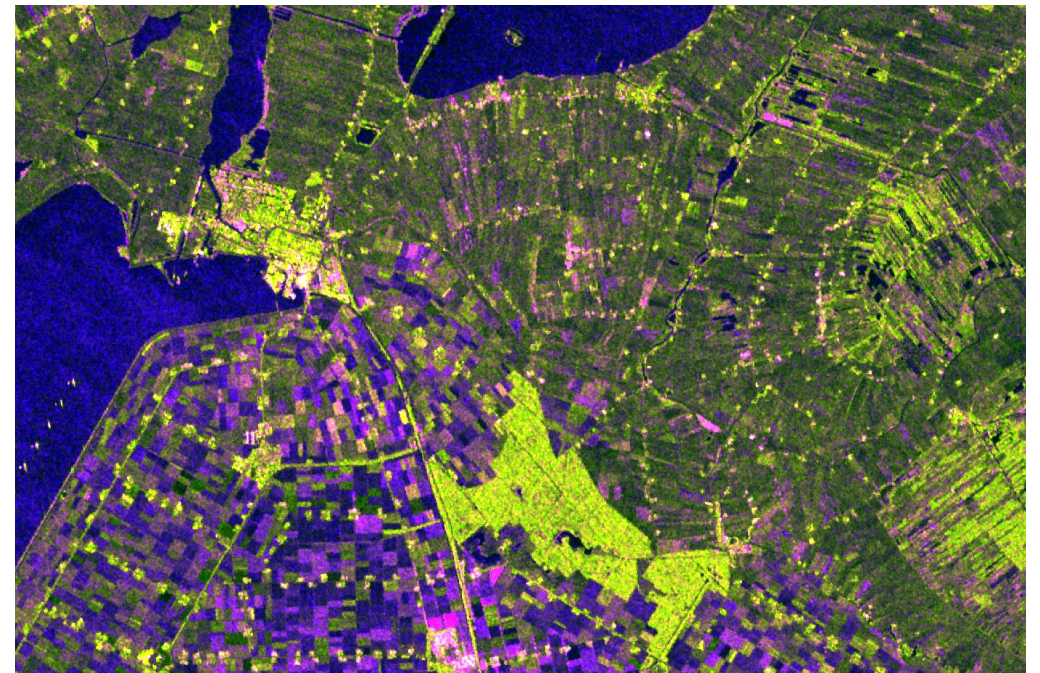
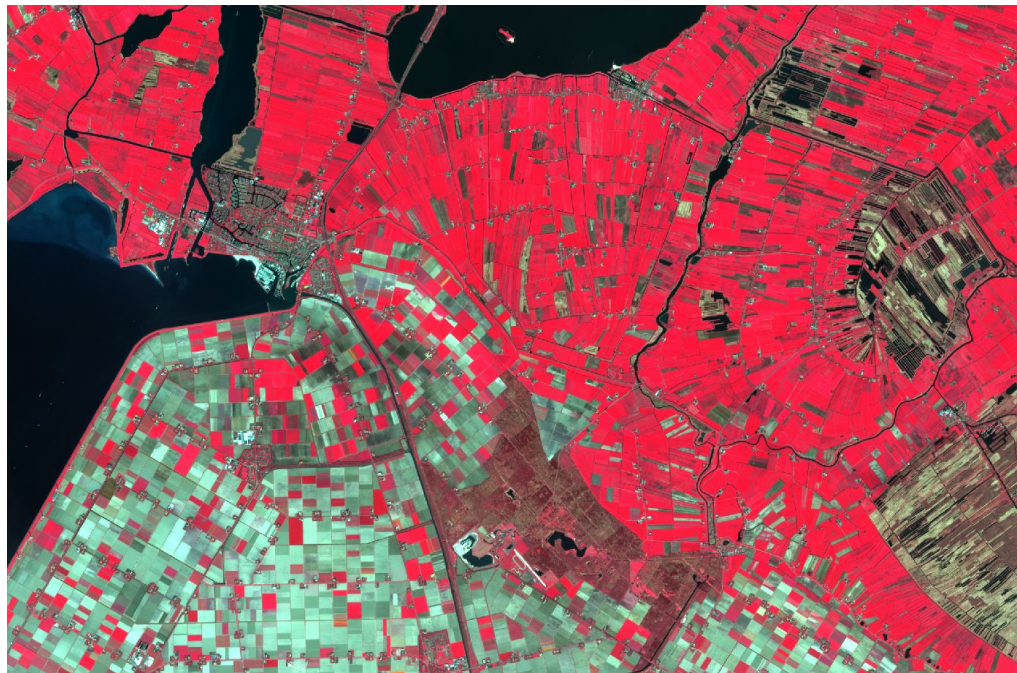
gamma nought γ_0

(A_γ flattened corrected using DEM)





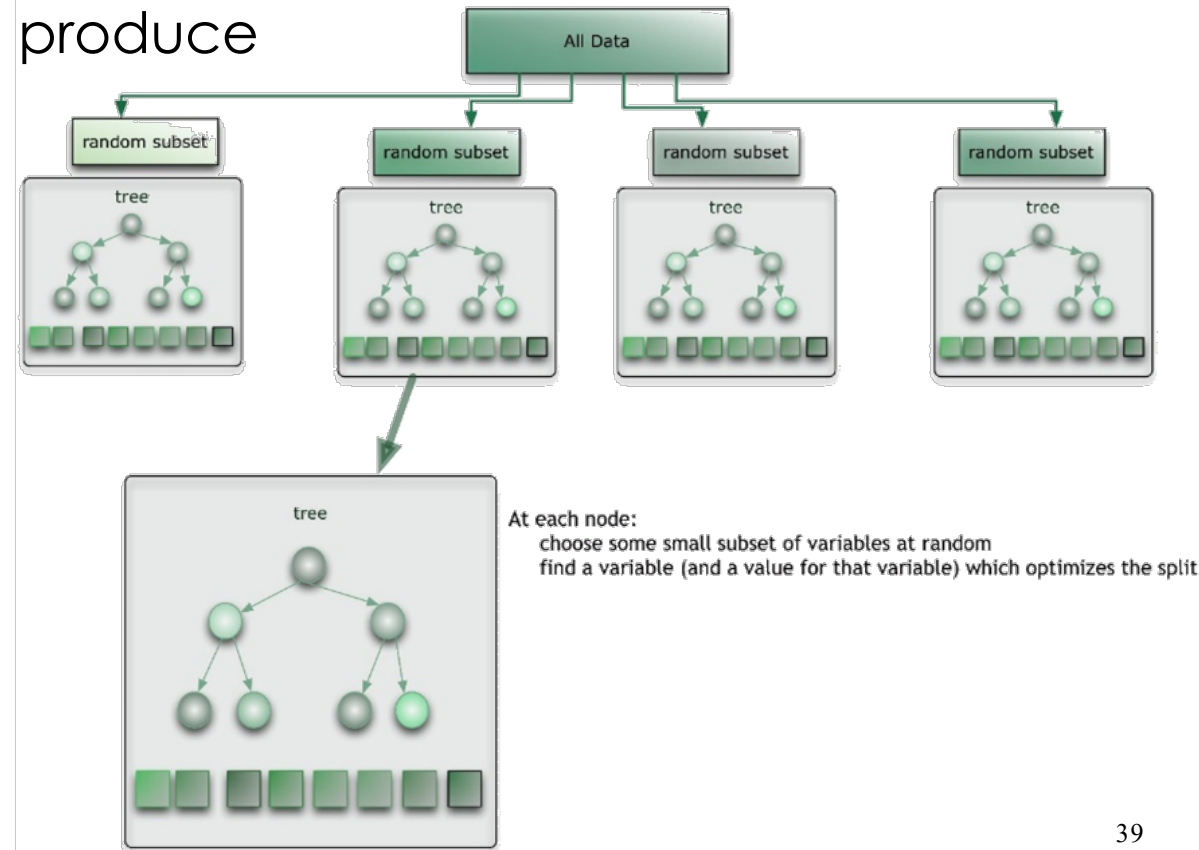
- **Random Forest as classifier by default** (deep learning alternative: Neural Net [Transformer])
- **Specific quality control of in situ dataset to balance the calibration dataset**
- **Large set of features used by the Random Forest model including:**
 - S2 & L8 surface reflectance time series and S1 amplitude and coherence time series
 - NDVI, NDWI, Red Edge Indices, brightness and S1 polarization ratio time series
 - Temporal metrics from above time series and S1 temporal metrics from SAR time series





– Random Forest (RF)

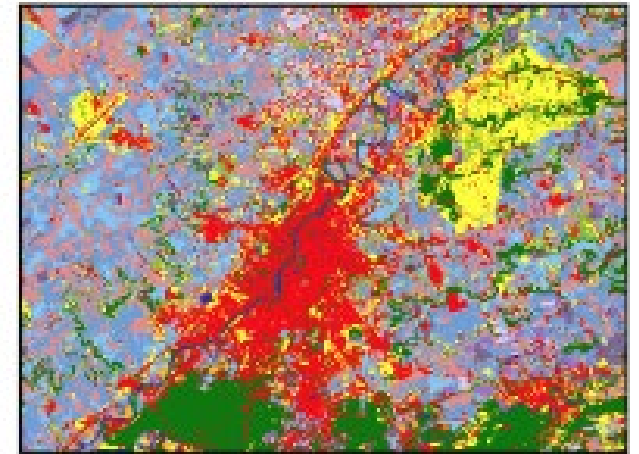
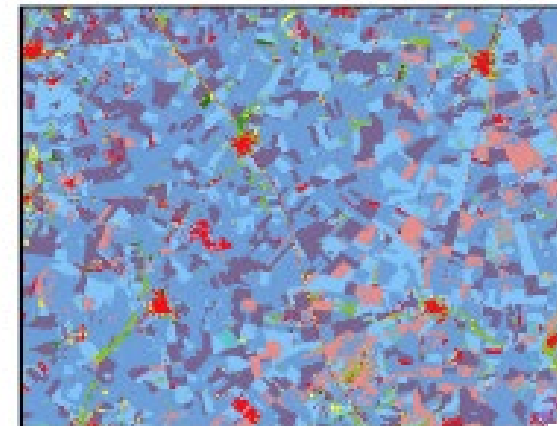
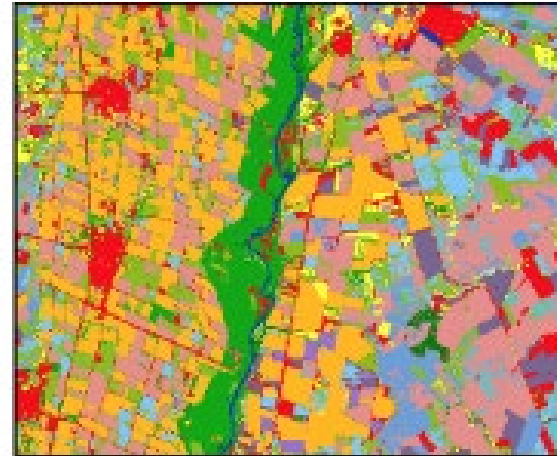
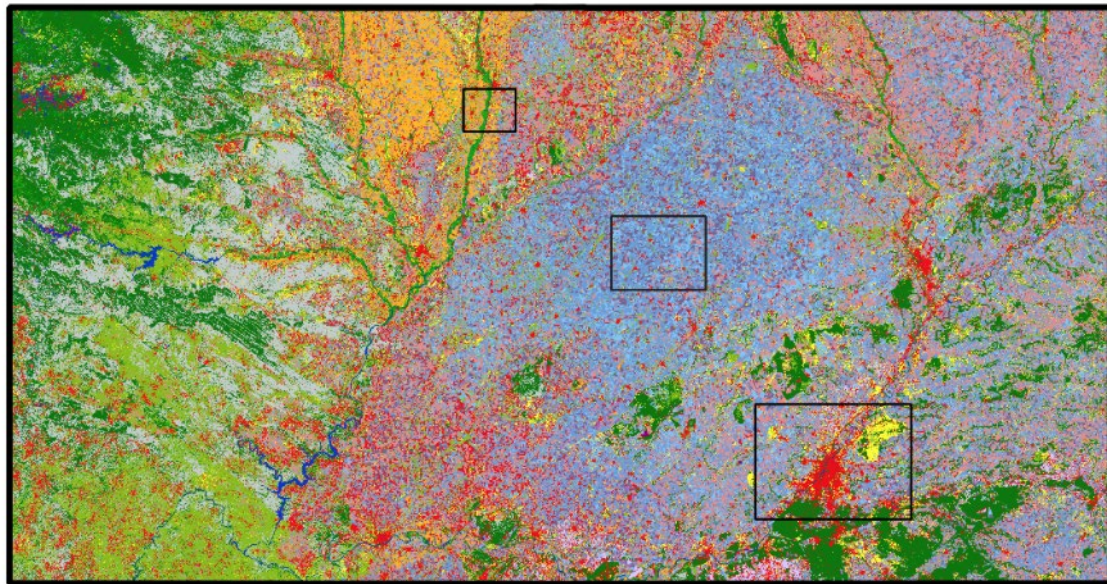
- Decision Trees ~ recursive partitions calibrated on training set to get regions (nodes) increasingly homogeneous with respect to their class variable
- Random Forest ~ ensemble-learning algorithm that combines multiple classifications of the same data to produce higher classification accuracies
- Sequence:
 - **Bagging**: individual trees are grown from different subsets of training data (random subsampling of the original data for growing each tree)
 - **Single majority vote** to identify the most frequent class decided from all trees





Crop map for Spain area (35 crop types)

Very similar accuracies for Random Forest and Neural Net (Transformer)
Overall Accuracy (OA): 80% (crop type) & 88% crop group

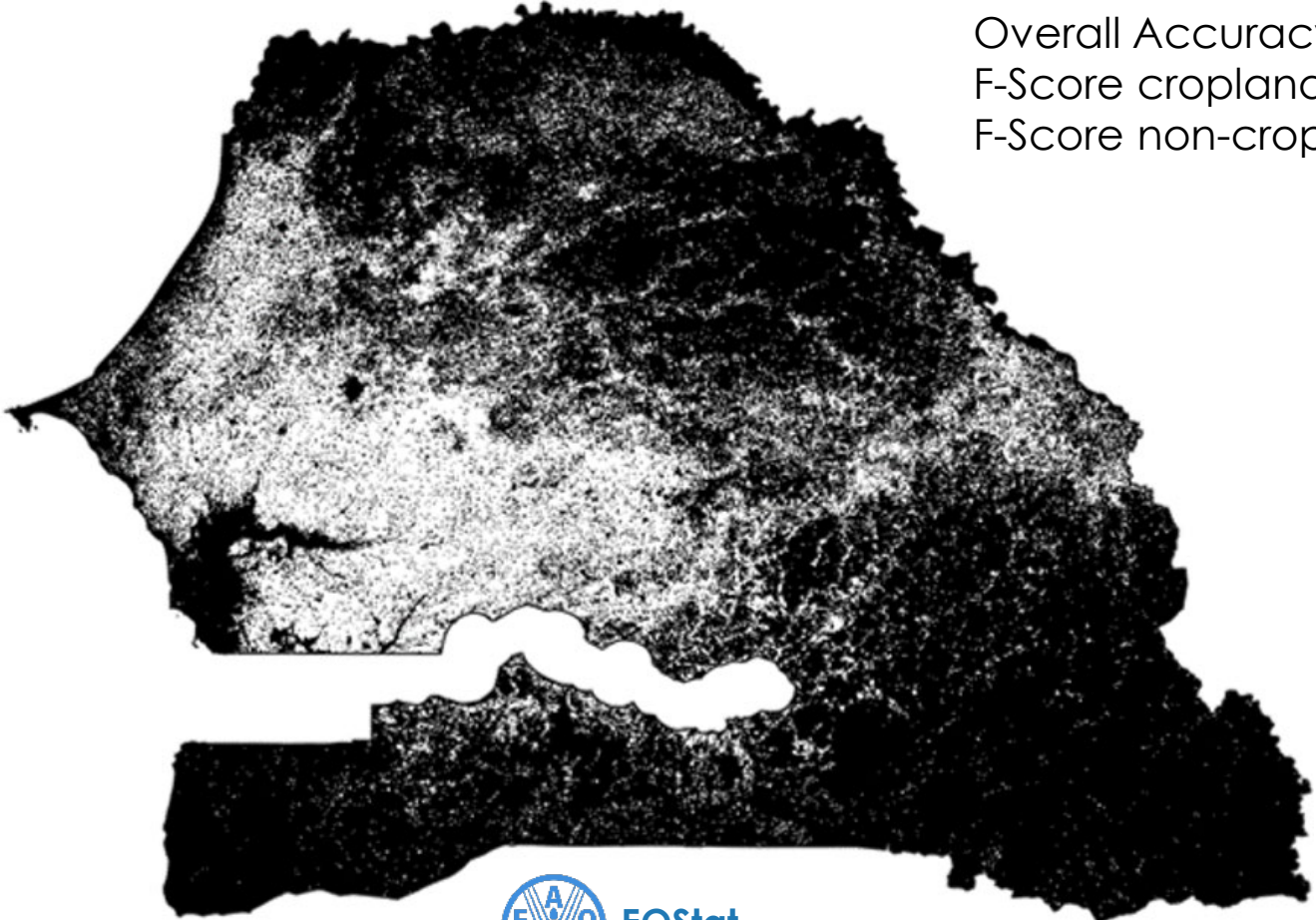


Crop Type			
Wheat	Leafy or stem vegetables	Spice crops	Vineyards
Maize	Fruit-bearing vegetables	Hops	Olive groves
Rice	Root bulb or tuberous vegetables	Leguminous crops	Trees
Sorghum	Mushrooms and truffles	Sugar beet	Succulent plant
Barley	Soya beans	Sugar cane	Shrub land
Rye	Groundnuts	Grassland and meadows	Forest
Oats	Other oilseed crops	Fibre crops	Bare soil
Millets	Potatoes	Medicinal aromatic pesticidal or similar crops	Build-up surface
Quinoa	Sweet potatoes	Flowers crops	Water bodies
	Cassava	Tobacco	
	Yams	Fruits trees	

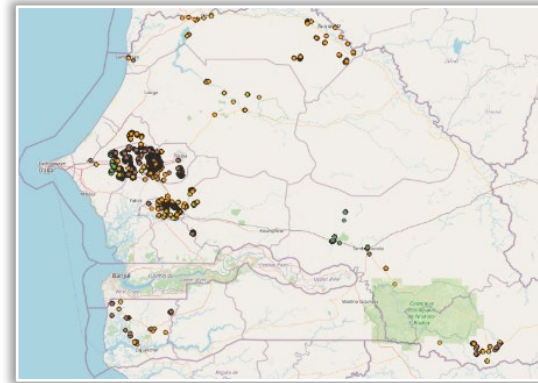


Sen4Stat – National cropland map at 10-m resolution (Senegal, 2020)

Random Forest classification based on S2 & L8 time series only



Overall Accuracy: 96%
F-Score cropland: 0,97
F-Score non-cropland: 0,88



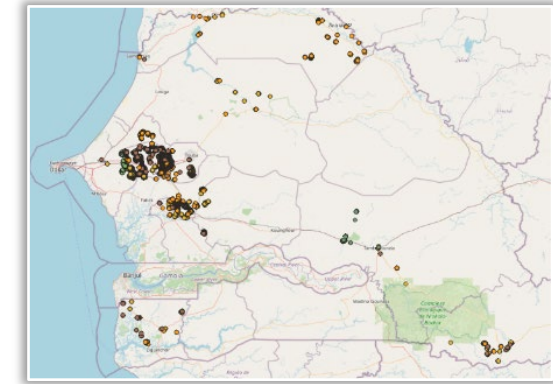
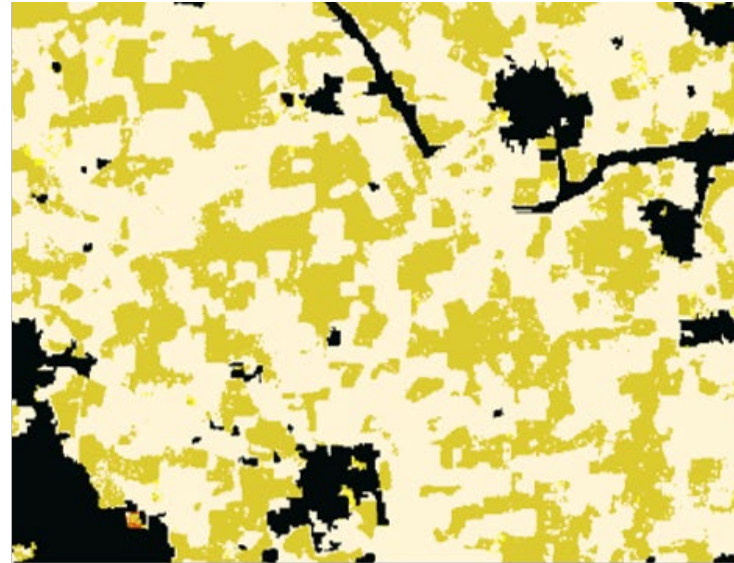
2000 GPS traces for holdings selected through a stratified sampling from the 526.000 holdings in Senegal (~ 0.4%)



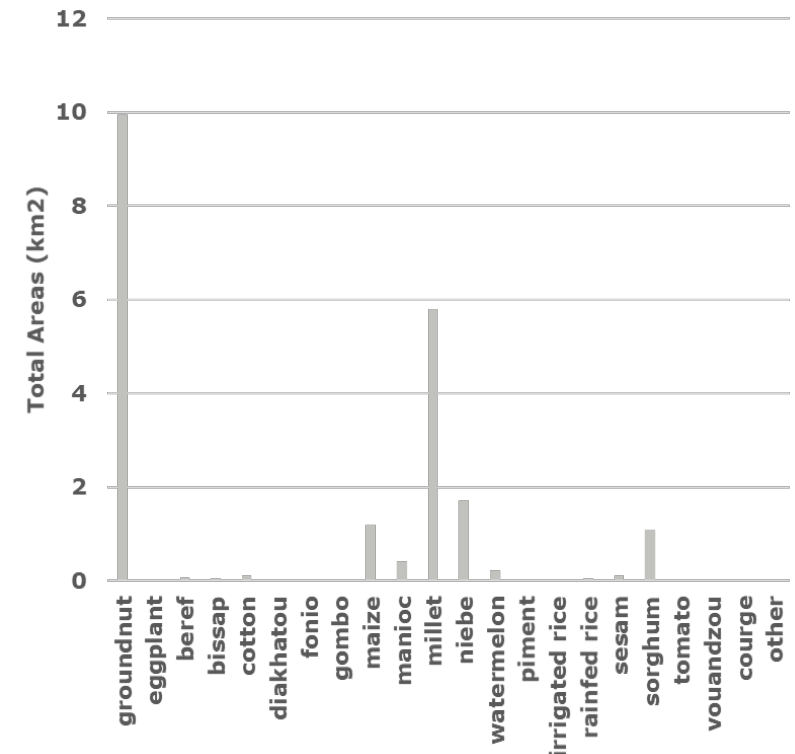


Sen4Stat – Quality of in situ data as major bottleneck today

Poor accuracy of crop type map because of the unbalanced ground observation dataset

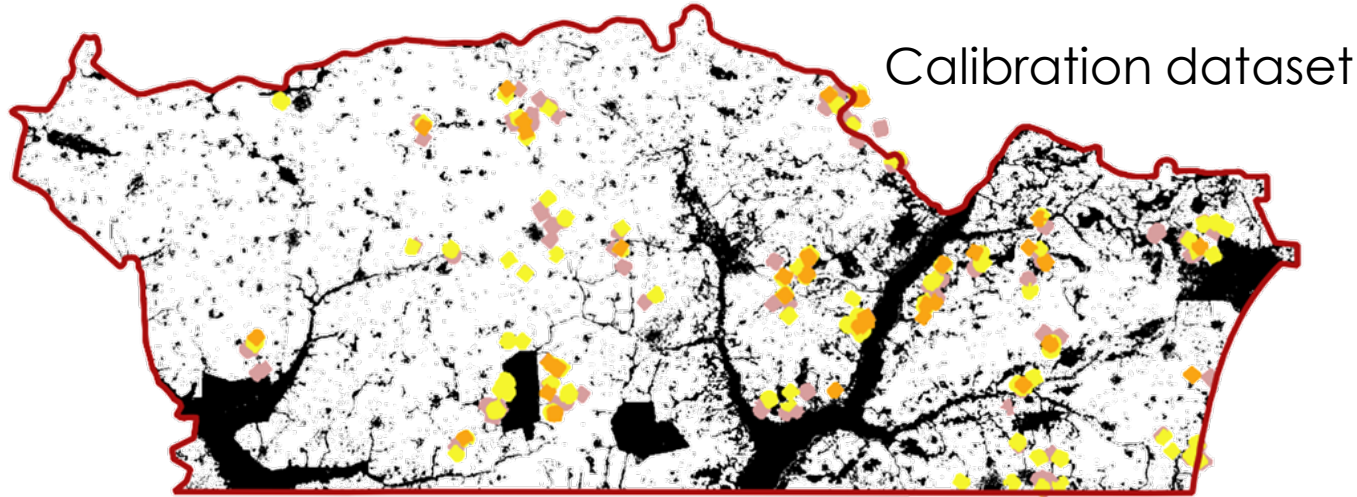


	Arachide	Maïs	Millet	Niébé	Sorgho	Autres	
Arachide	23296	0	6	1101	158	1184	90 %
Maïs	1	0	0	0	0	6	0 %
Millet	178	0	918	93	0	5	77 %
Niébé	3445	0	0	11318	158	1068	71 %
Sorgho	1996	0	0	1200	443	270	11 %
Autres	5037	0	84	2610	191	4256	35 %
	69 %	N/A	91 %	69 %	47 %	63 %	68 %

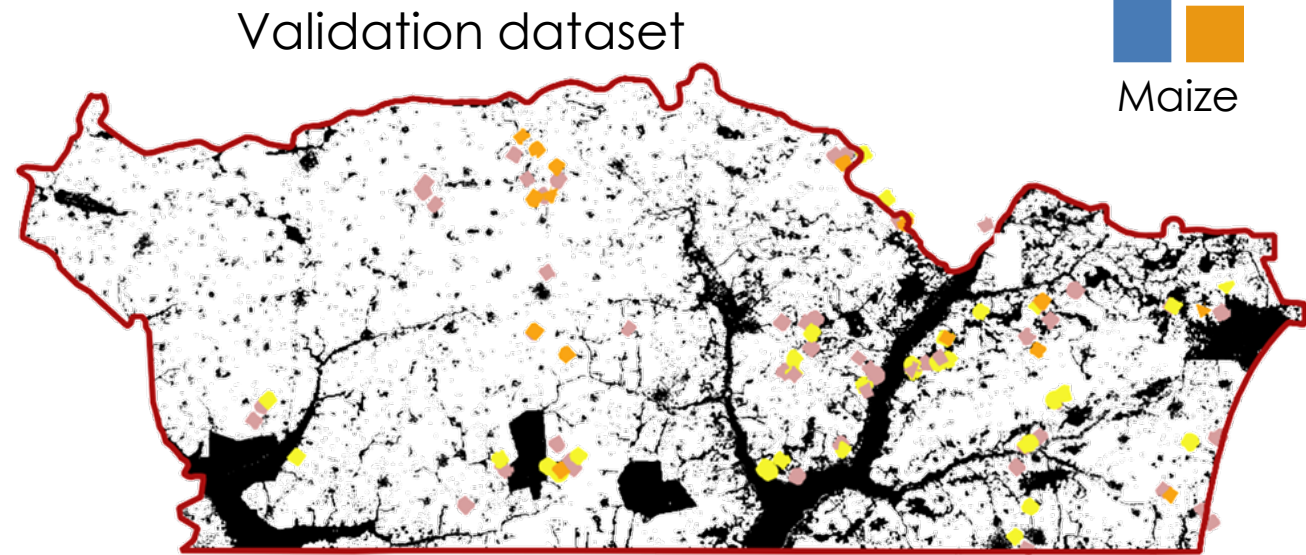




– Specific in situ data collection for department of Nioro (Senegal, 2021)



Calibration dataset



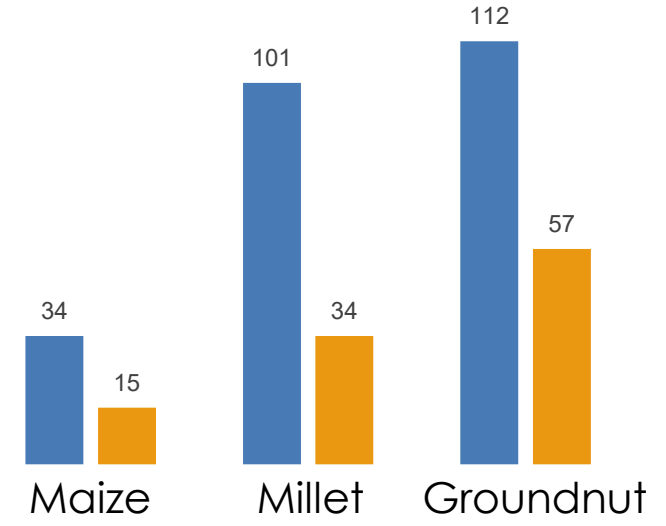
Validation dataset

-  Maize
-  Millet
-  Groundnut

0 10 km



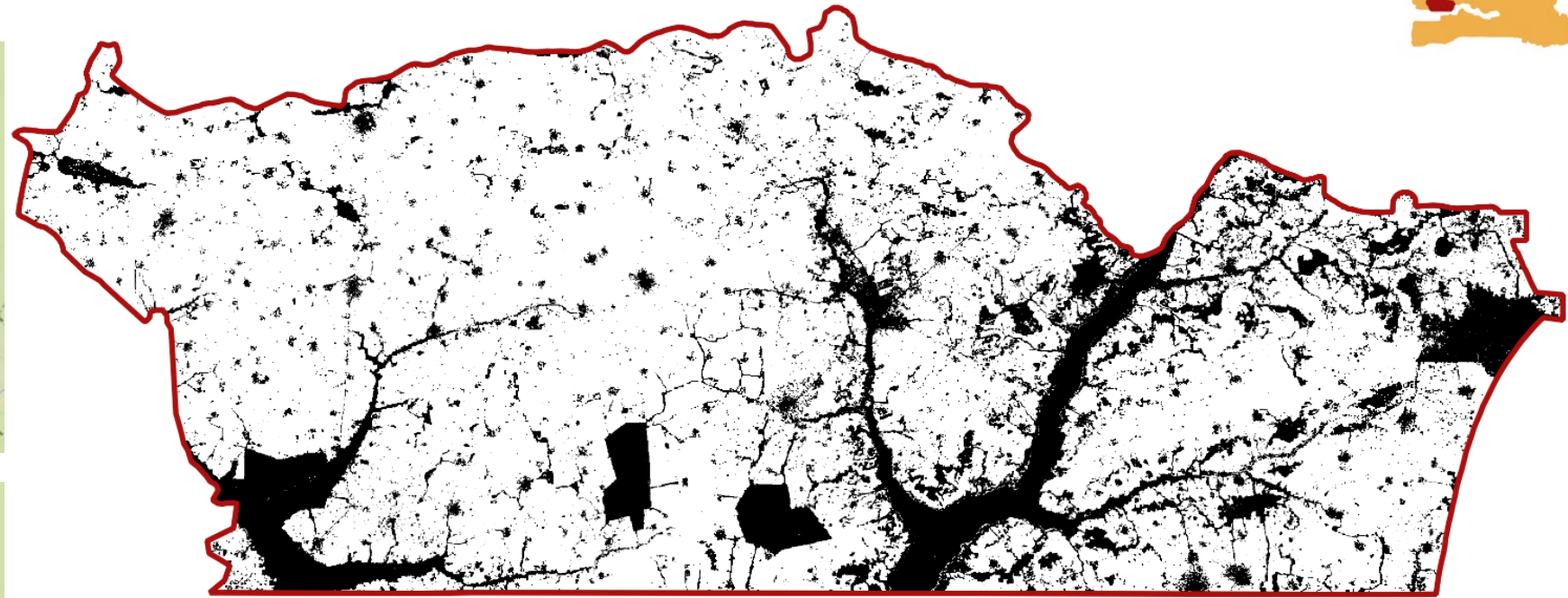
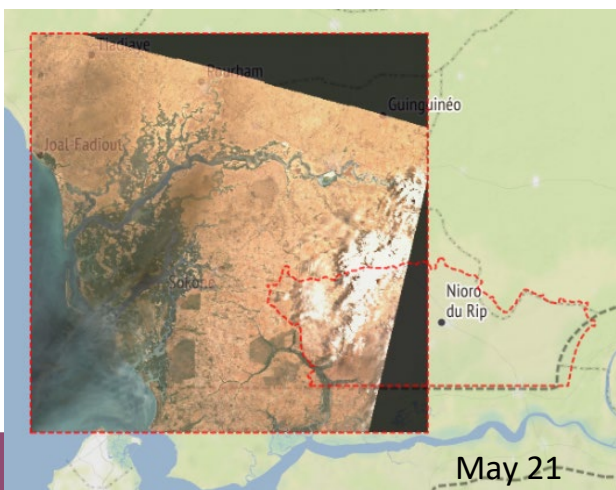
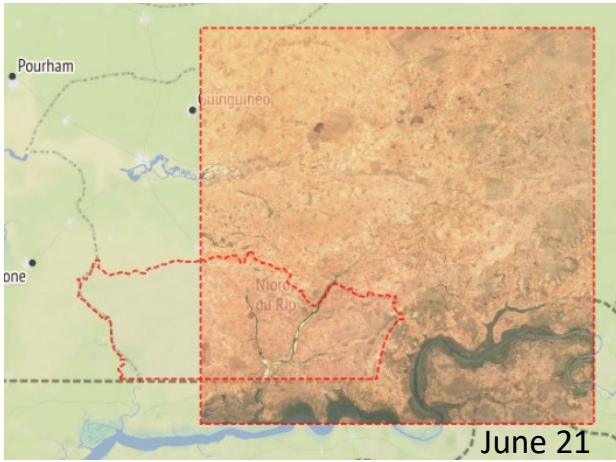
Calibration polygons
Validation polygons





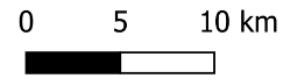
Sen4Stat – 10-m cropland map using balanced in situ data collection for the Department of Nioro (Senegal)

Each S2 image covering only a part of the department



Overall Accuracy: 97,4 %
 F-Score cropland: 0,98
 F-Score non-cropland: 0,95

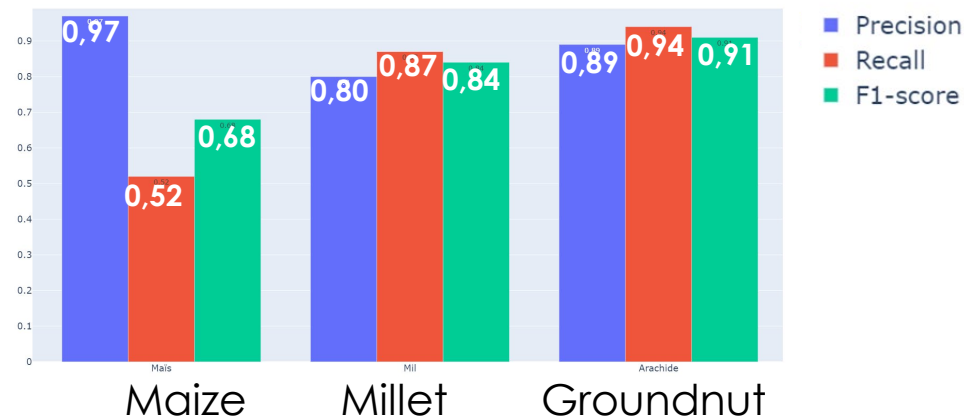
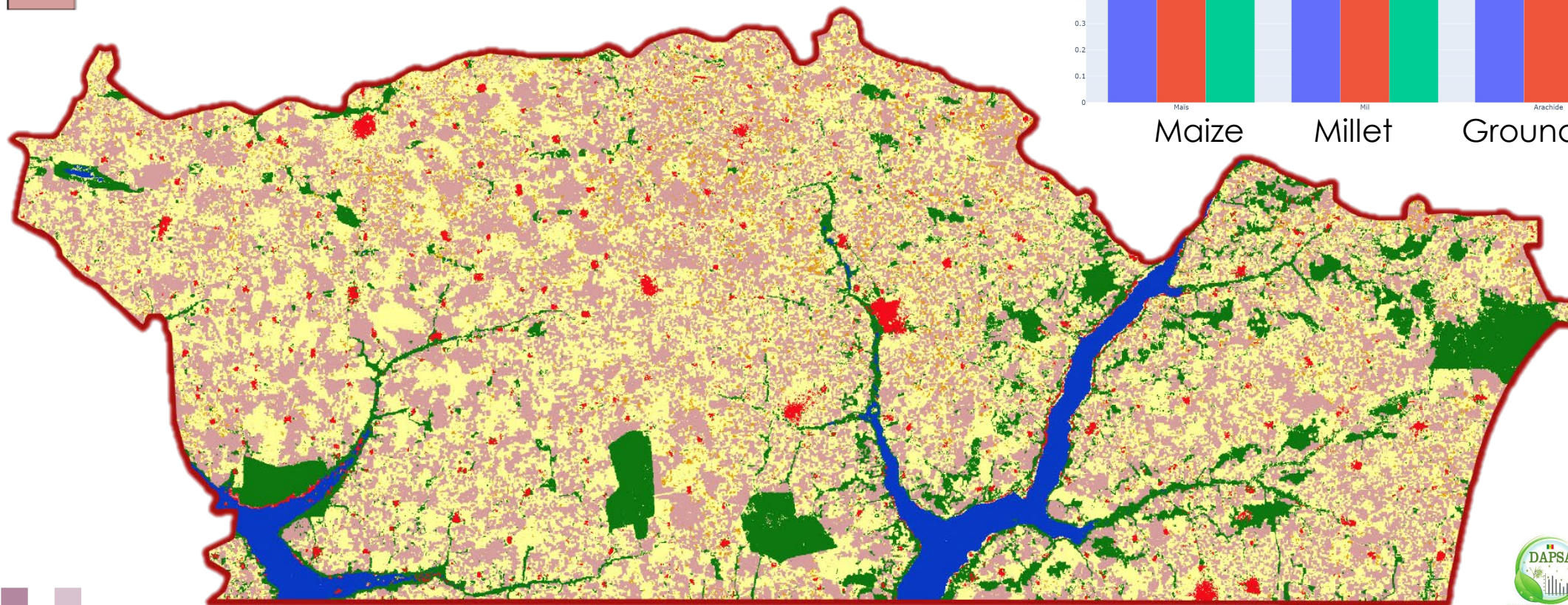
Cropland
 Non-cropland



– 10-m crop type map from Sentinel-1 and Sentinel-2 time series acquired over the season and using balanced in situ dataset

- Maize
- Millet
- Groundnut

Season from 01-05-2021 to 31-12-2021
Overall Accuracy: 84,8 %



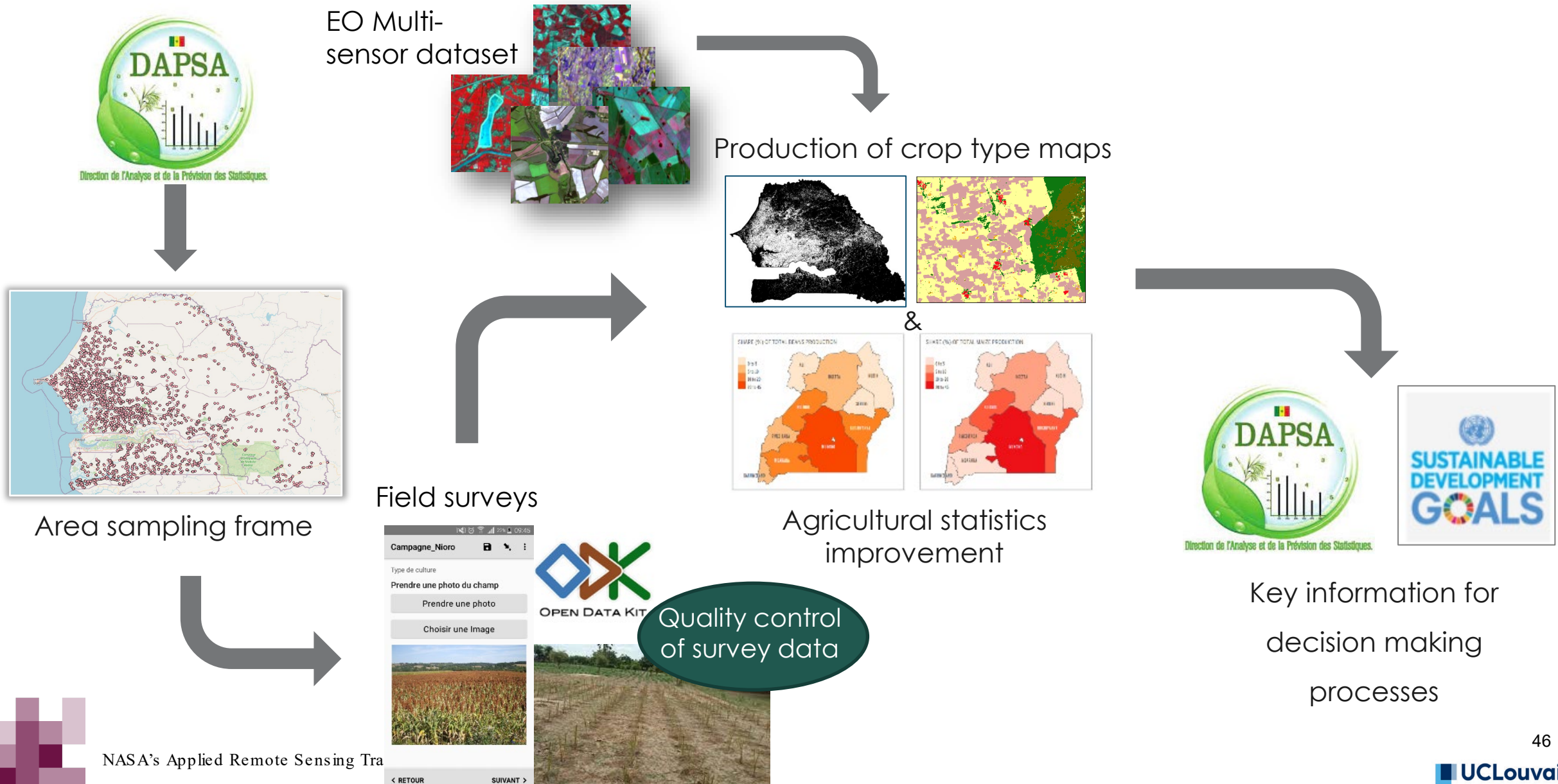
0 5 10 km

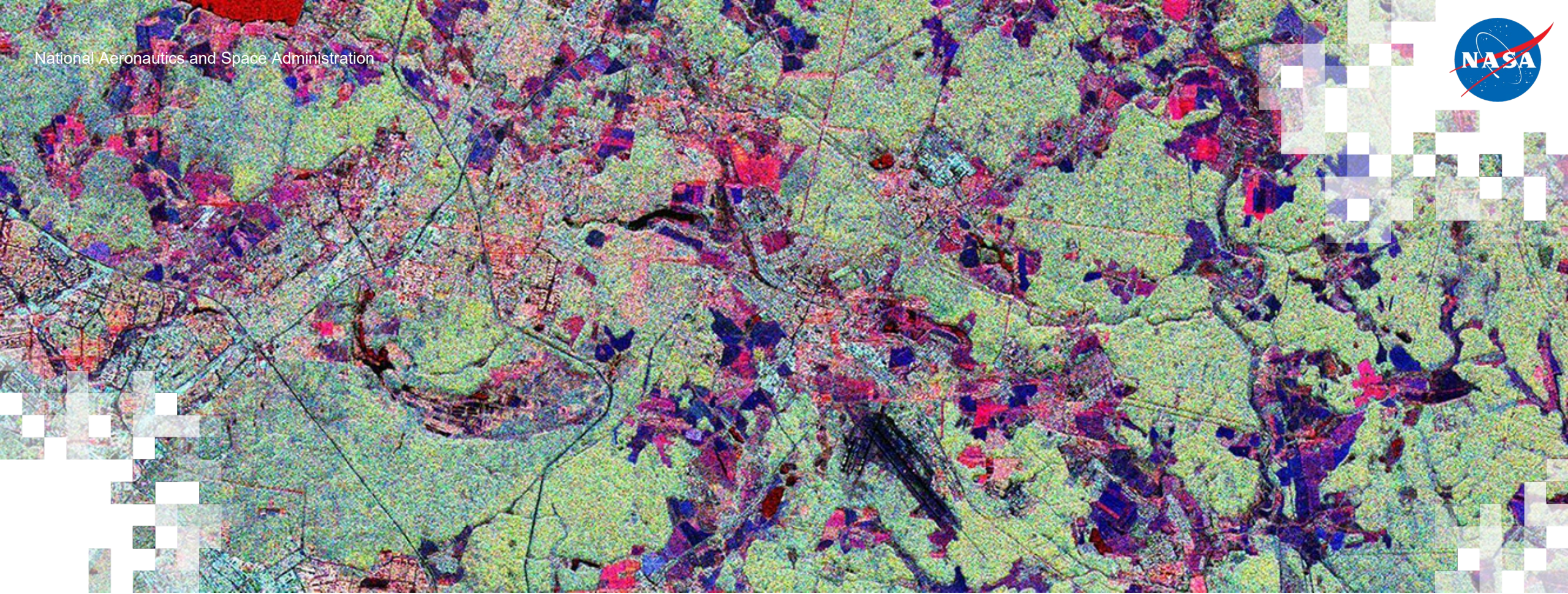


EOStat



Operational exploitation of satellite missions to improve timely information about crop areas, yields and production





Section 3: Practical using Sen4Stat toolbox for crop type mapping

Contacts

- Trainers:
 - Pr. Pierre Defourny pierre.defourny@uclouvain.be
- Training Webpage:
 - <https://appliedsciences.nasa.gov/join-mission/training/english/arset-agricultural-crop-classification-synthetic-aperture-radar-and>
- ESA's EO4Society Website:
 - <https://eo4society.esa.int/training-education/>
- Twitter: [@EOOpenScience](https://twitter.com/EOOpenScience)



Useful links

- **Sen4Stat**

<https://esa-sen4stat.org/>

- **Sen2Agri**

<https://github.com/Sen2Agri/Sen2Agri-System/>

- **SNAP download**

<http://step.esa.int/main/download/snap-download/>

- **Copernicus Open Access Hub (download Sentinel data)**

<https://scihub.copernicus.eu/>

- **NASA ARSET Agricultural Crop Classification with Synthetic Aperture Radar and Optical Remote Sensing Part 5: Biophysical Variable Retrieval using Optical Imagery to Support Agricultural Monitoring Practices**

<https://appliedsciences.nasa.gov/join-mission/training/english/arset-agricultural-crop-classification-synthetic-aperture-radar-and>

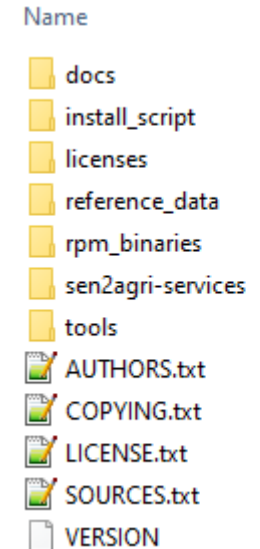
- **ESA 'Advanced Training Course on Land Remote Sensing with the Focus on Agriculture' held in Louvain-la-Neuve, Belgium, on 16-20 September 2019**

<https://eo4society.esa.int/resources/advanced-training-course-on-land-remote-sensing-with-the-focus-on-agriculture/>



Sen4Stat Installation Package (available in June 2022)

- Single installation archive containing all files needed
 - Documentation (docs)
 - Installation scripts
 - Sen4Stat rpm files
 - Sen4Stat Services
 - Other tools and configuration files
- CentOS 7 is required for system installation
- Installation performed through a single script
 - `sudo ./install.sh`
- Two directories with write permissions for all users are needed:
 - `/mnt/archive`
 - `/mnt/upload`
- During installation, all needed third party packages are installed via yum. Needed docker images are installed
- PostgreSQL database is used for storing system configuration, product information etc.
- The regular user `sen2agri-service` is created during installation
- System services are created and launched upon installation:
 - `sen4stat-executor` `sen4stat-orchestrator` `sen4stat-http-listener` `sen4stat-demmaccs`
`sen4stat-demmaccs.timer` `sen4stat-monitor-agent` `sen4stat-scheduler` `sen4stat-services`



Homework and Certificate

- Homework Assignment:
 - Answers must be submitted via Google Form
 - Due Date: May 17, 2022
- A certificate of completion will be awarded to those who:
 - Attend all live webinars
 - Complete the homework assignment by the deadline (access from website)
 - You will receive a certificate approximately two months after the completion of the course from: marines.martins@ssaihq.com

Questions?

- Please enter your questions in the Q&A box. We will answer them in the order they were received.
- We will post the Q&A to the training website following the conclusion of the webinar.



<https://earthobservatory.nasa.gov/images/6034/pothole-lakes-in-siberia>



Thank You!

