

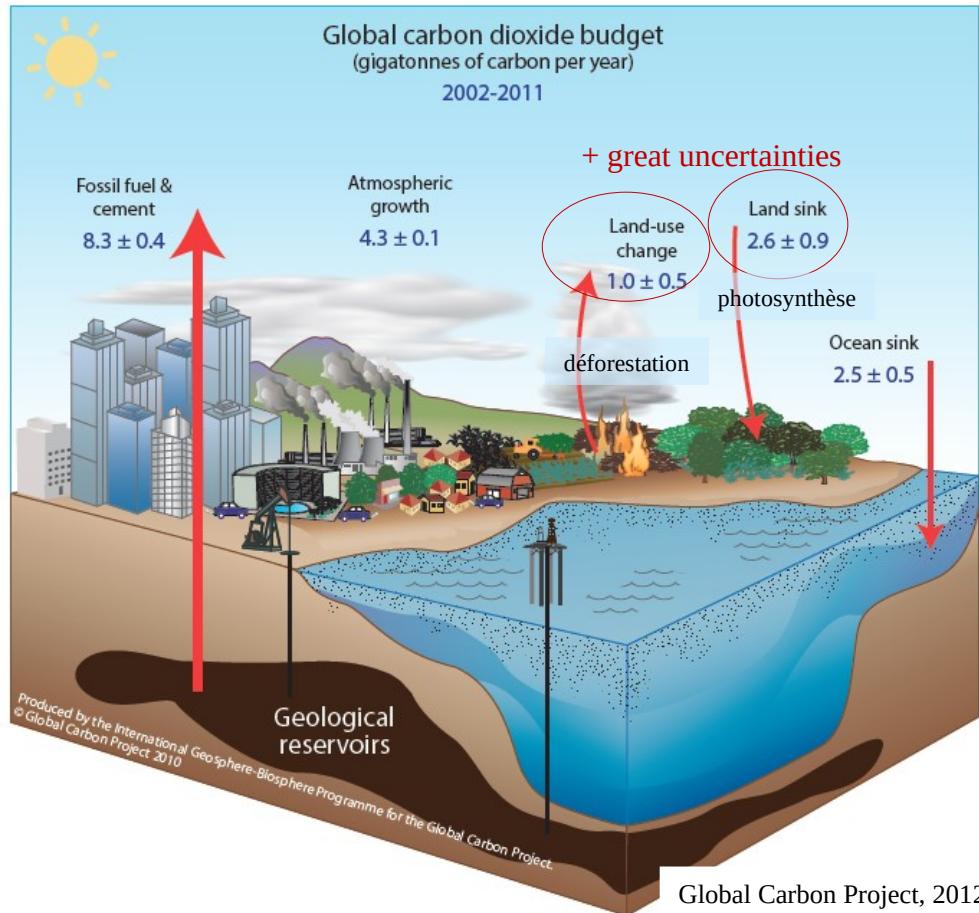
OUTLINE

- I. Radar imaging - Spatial resolution
- II. Polarization - Polarimetry
- III. Radar response sensitivity
- IV. Relief effects
- V. Speckle and Filtering

TROPICAL RAIN FORESTS MONITORING

Tropical Forests monitoring

Role of biomass in the Carbon Global Cycle?



C Stock C = 50% Biomass
Forests: 70-90% aerial biuomass

CO₂ flux with land surfaces :
30% flux anthropiques
great uncertainty

Need for forest biomass estimation improvement:

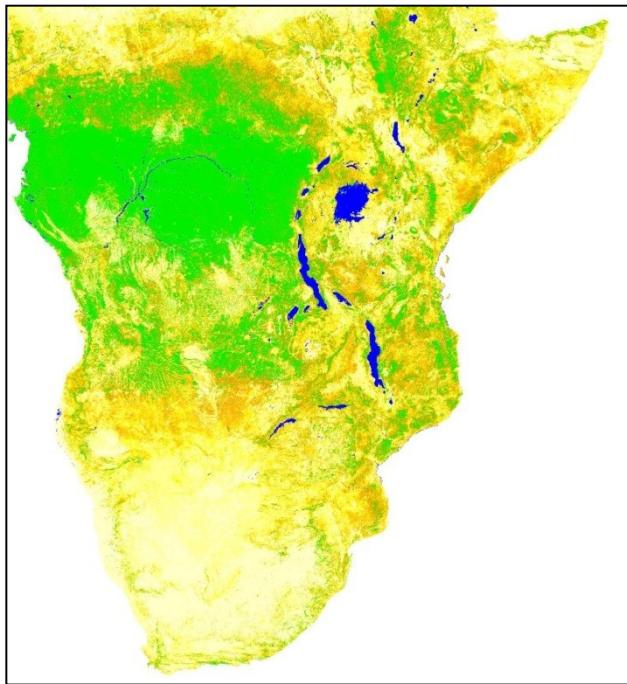
- Spatial distribution
- Stocks
- dynamics

IPCC , UNFCCC

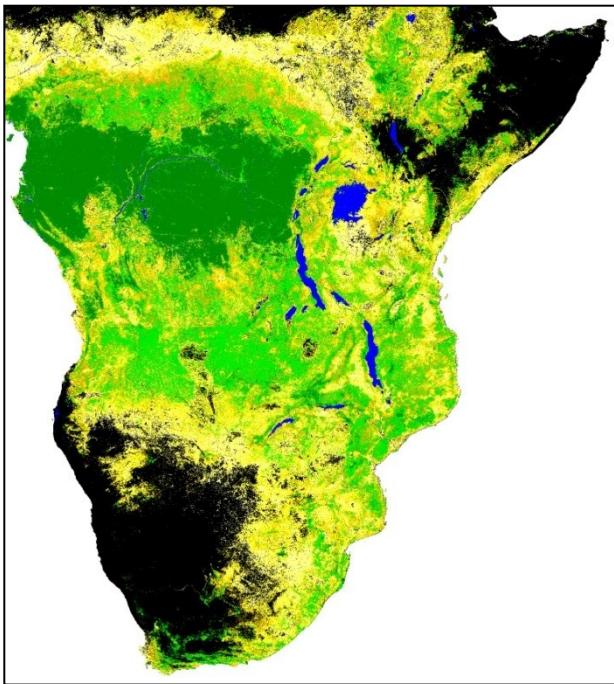
REDD+:

Financial incentive for sustainable forests ==> tropical countries

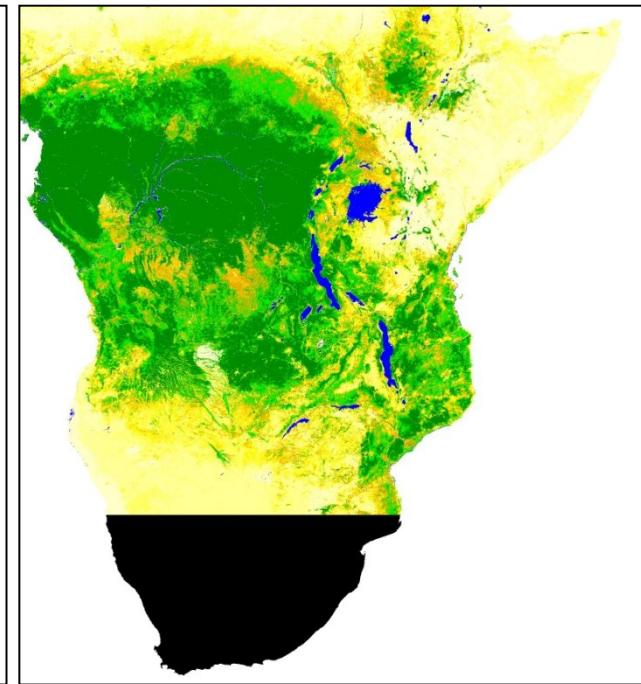
Mermoz et al., 2016



Saatchi et al., 2011



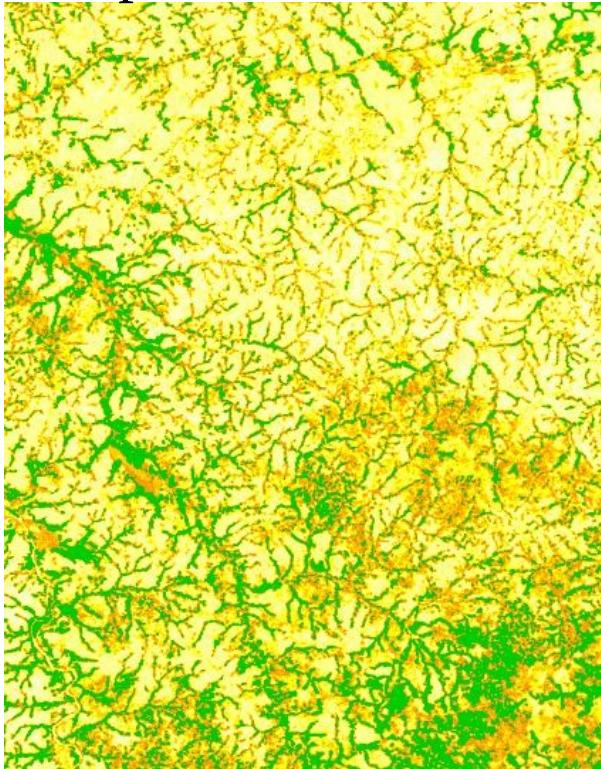
Baccini et al., 2012



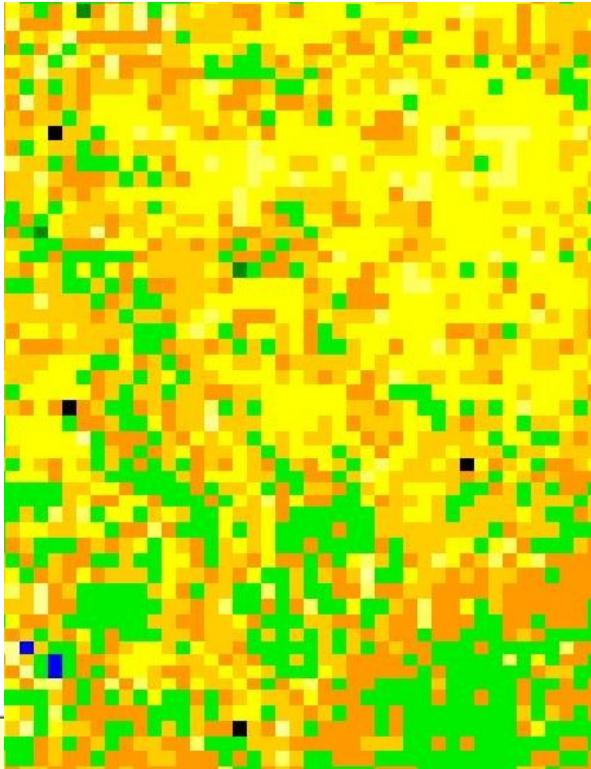
0-10 Mg ha ⁻¹	50-60 Mg ha ⁻¹
10-20 Mg ha ⁻¹	60-100 Mg ha ⁻¹
20-30 Mg ha ⁻¹	> 100 Mg ha ⁻¹
30-40 Mg ha ⁻¹	water
40-50 Mg ha ⁻¹	no data

Produit 'Biomasse forestière' Afrique

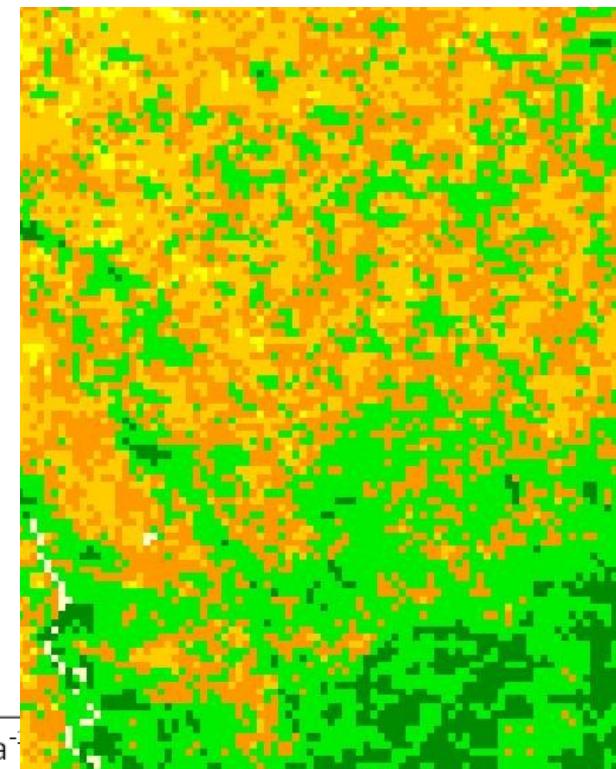
Map from CESBIO



Saatchi et al., 2011



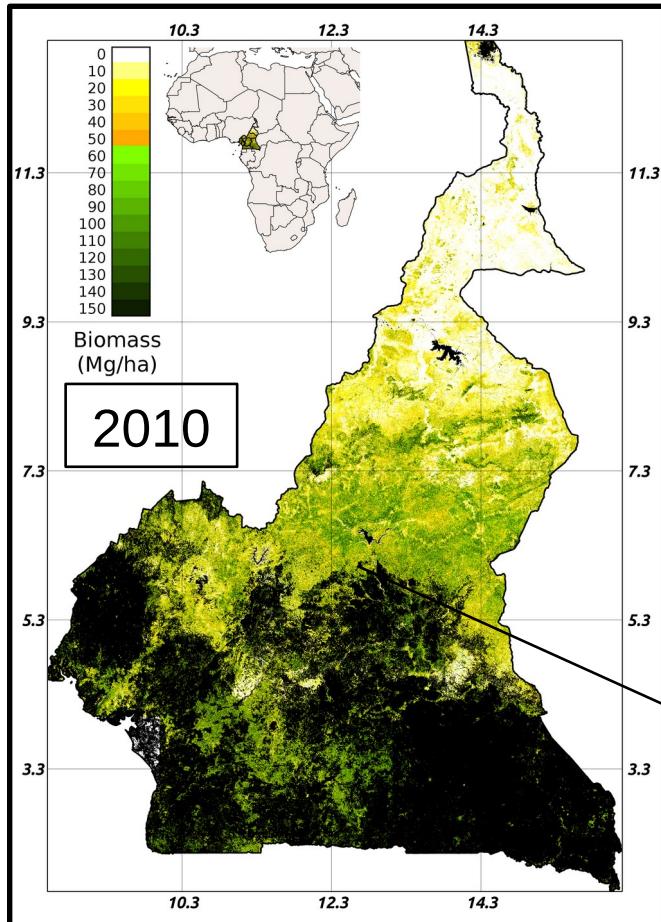
Baccini et al., 2012



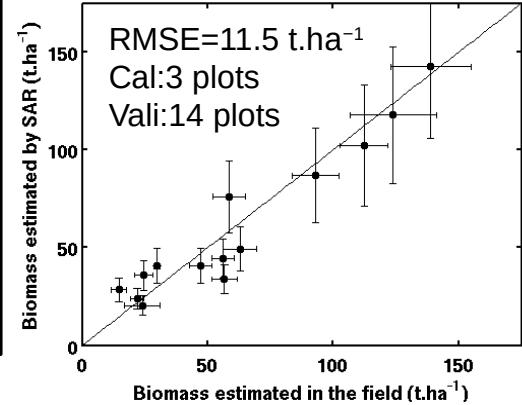
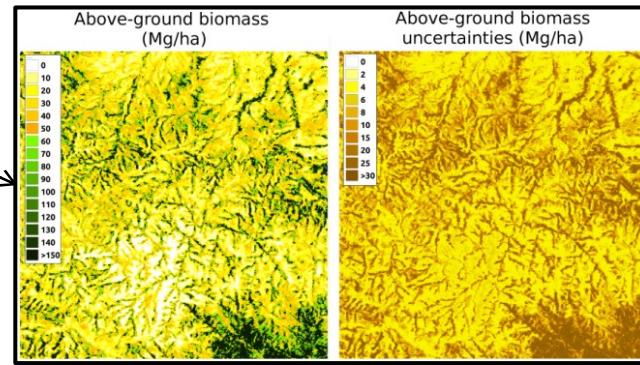
	10-20 Mg ha^{-1}		60-100 Mg ha^{-1}
	20-30 Mg ha^{-1}		> 100 Mg ha^{-1}
	30-40 Mg ha^{-1}		water
	40-50 Mg ha^{-1}		no data

Subset from:
 Latitude: 10°S to 5°S
 Longitude: 20°E to 25°E

Produit 'Biomasse forestière' Cameroon



	Surface area (ha)	Mean AGB (Mg.ha ⁻¹)	AGB (Tg)	Carbon (TgC)
Mosaic forest-croplands	1,811,150	89.5	162.9	81.4
Mosaic forest-savanna	5,187,900	75.6	394.2	197.1
Deciduous woodland	10,352,400	53.3	553.6	276.8
Deciduous shrubland – sparse trees	1,949,000	30.7	59.8	29.9
Others	6,622,340	12.6	83.4	41.7
TOTAL	25,922,790	48.2	1253	626.9

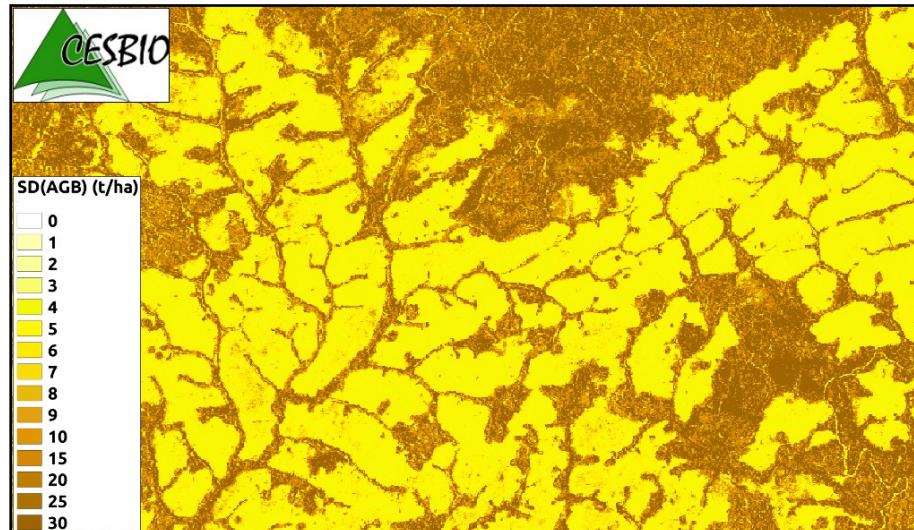
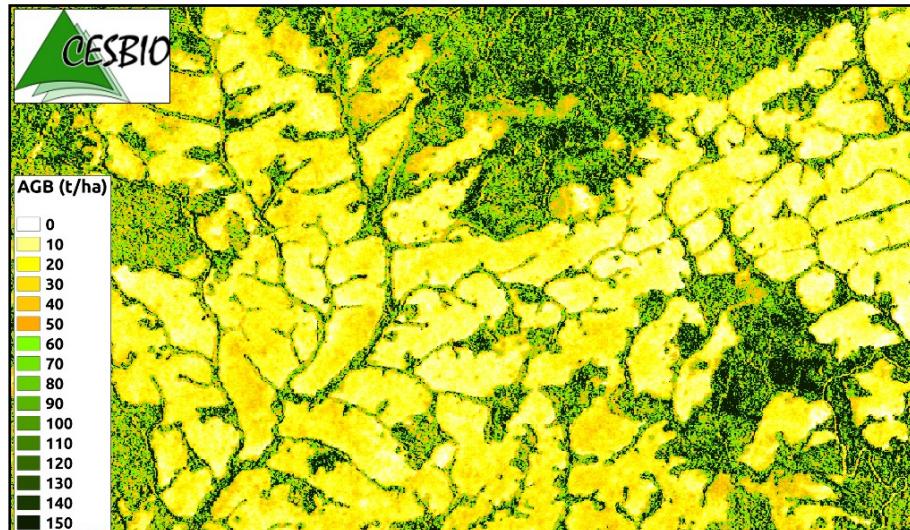
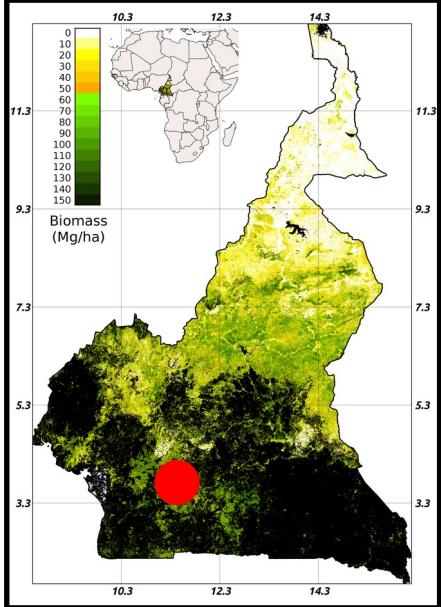


Total aboveground carbon stock:

- This study: 626.9 TgC
- Nasi et al. (2009): 710 TgC



Produit 'Biomasse forestière' Cameroon



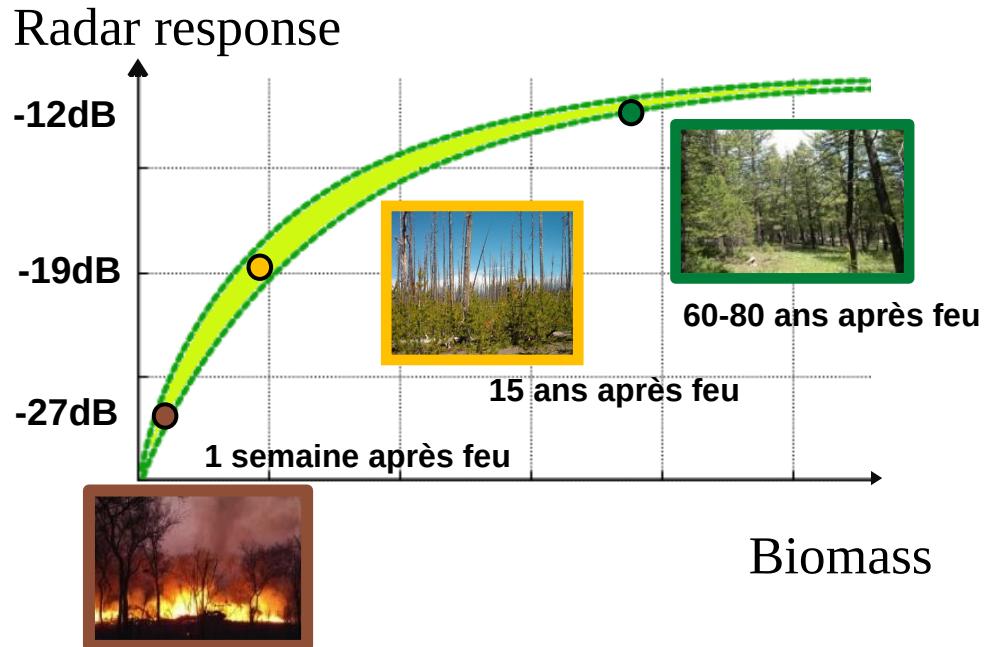
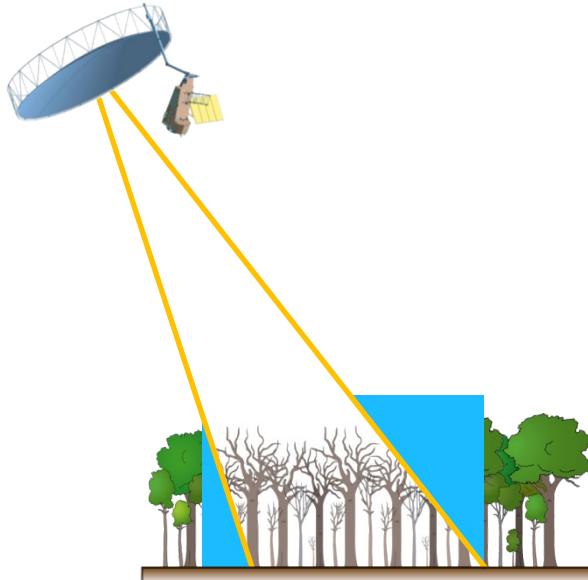
Tropical Forests monitoring

7th ESA Earth Explorer mission: BIOMASS (2022)

Large wavelength for biomass cartography

1. Canopy penetration for every biomes
 2. Interacts with woody elements of vegetation
 3. Forest height estimation ability
- P band band P ($\lambda = 70 \text{ cm}$) *largest available wavelength from space*

P Band Radar
 $(\lambda = 70 \text{ cm})$



Tropical forest monitoring

7th ESA Earth Explorer mission: BIOMASS (2022)

A key mission for a better understanding of Global Carbon Cycle



Aerial biomass (t / ha)

- Resolution: 200 m
- 1 map / 6 mois (4-year period)
- Global cover of forested areas
- **Precision 20%, or 10 t ha⁻¹** for biomass < 50 t ha⁻¹

Canopy height (m)

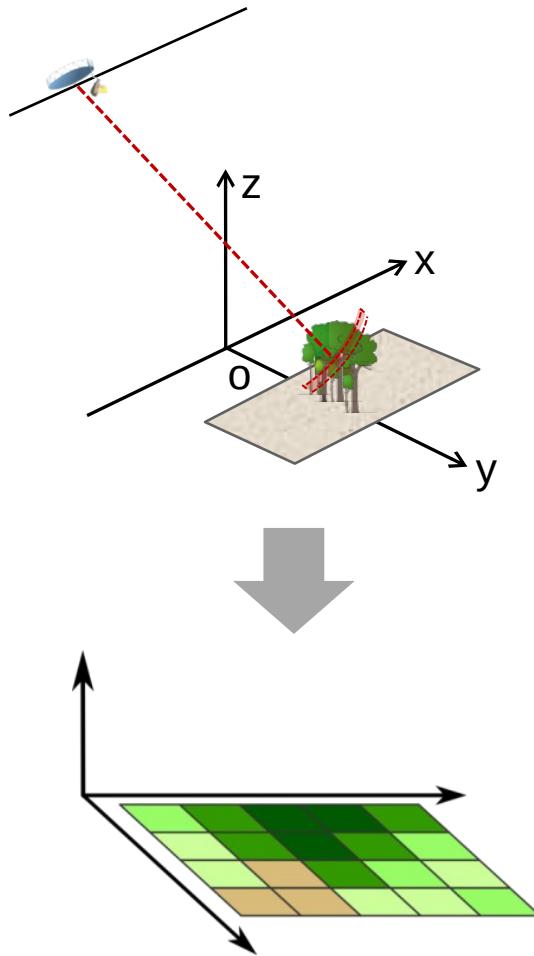
- Resolution: 200 m
- 1 map / 6 months (4-year period)
- Global cover of forested areas
- Precision 20 – 30%

Deforested areas (ha)

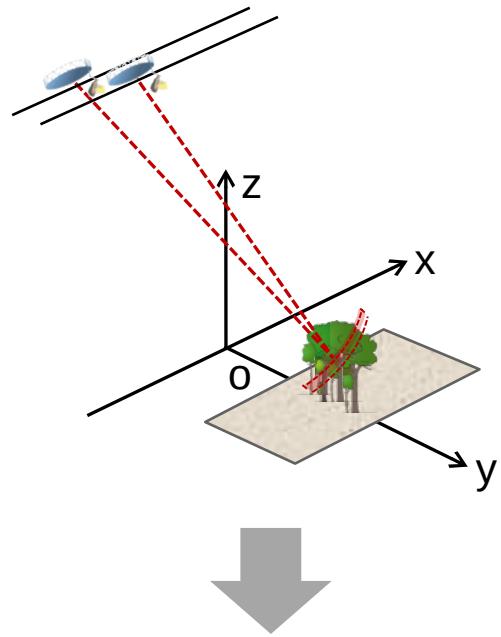
- Resolution: 50m
- 1 map / 6 months (4-year period)
- Global cover of forested areas
- Classification precision: 90%

Biomass Mission

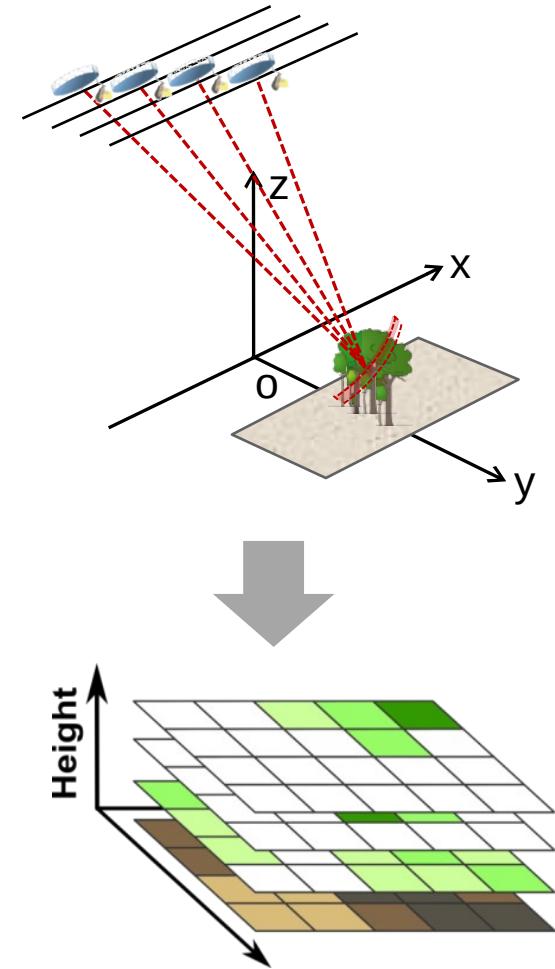
PoISAR
(SAR Polarimetry)



PolInSAR
(Polarimetric SAR Interferometry)



TomoSAR
(SAR Tomography)



Radar response sensitivity

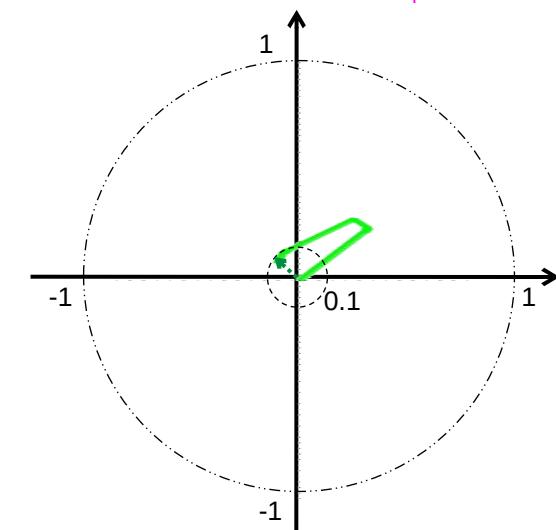
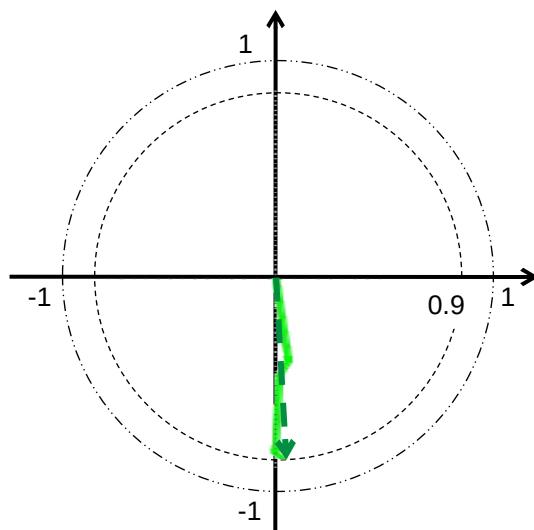
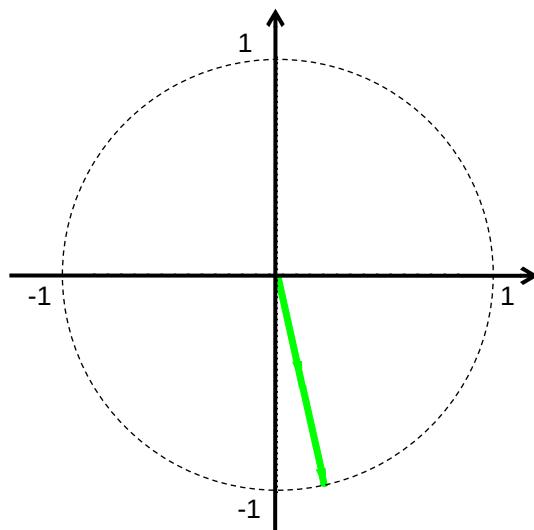
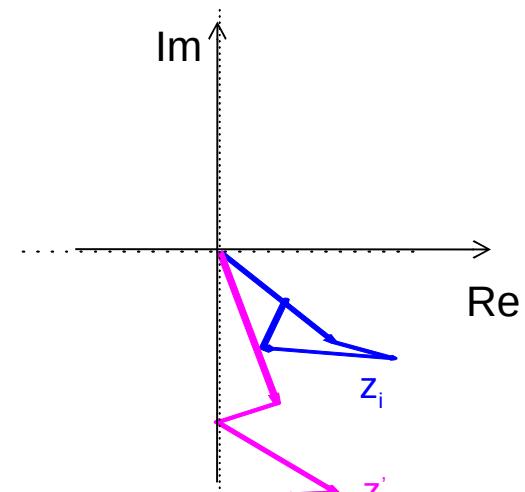
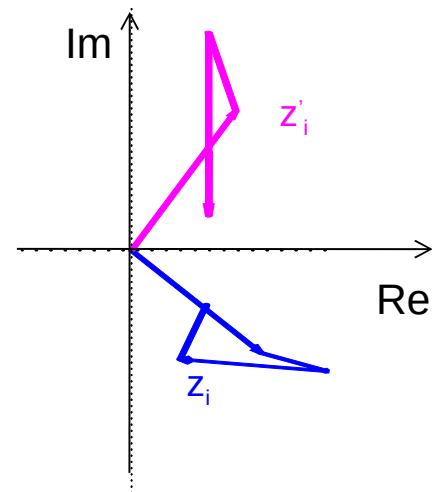
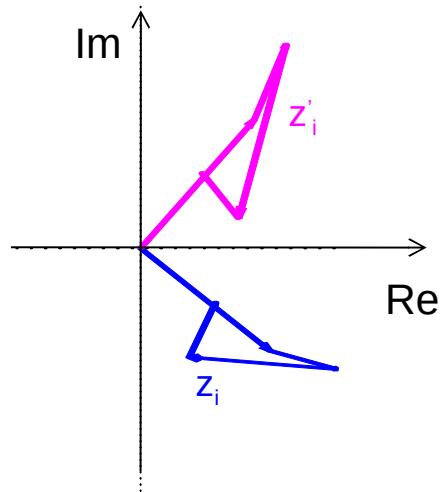
RADAR COHERENCE:

2 radar acquisitions

Temporal geometrical stability ($\leq \lambda$)
of the scatterers within each resolution cell

Complex coherence

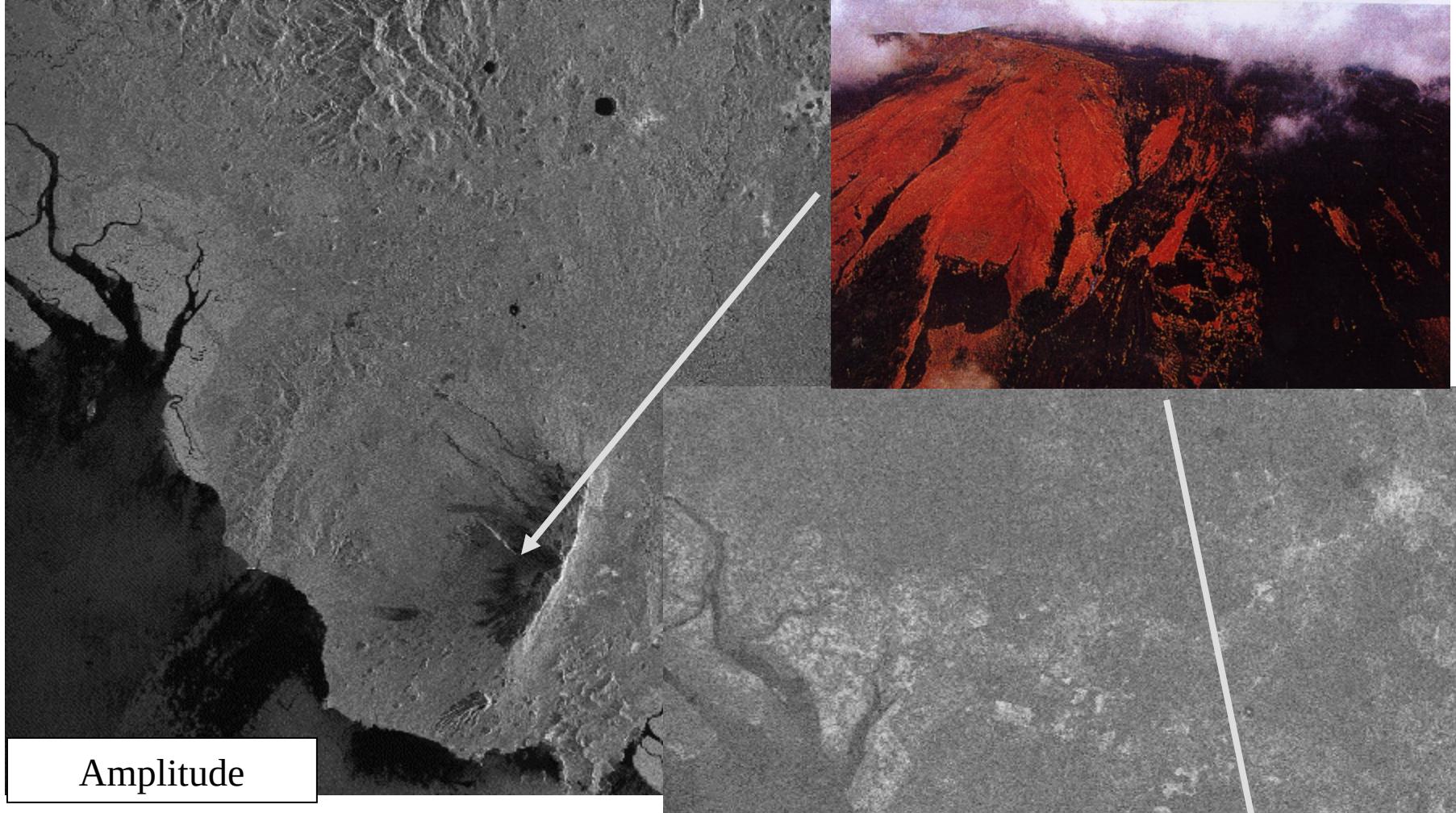
$$\rho = \frac{\langle z_i z'^*_i \rangle}{\sqrt{\langle |z_i|^2 \rangle \langle |z'^*_i|^2 \rangle}}$$



$|\rho| = 1$

$|\rho| = 0.9$

$|\rho| = 0.1$

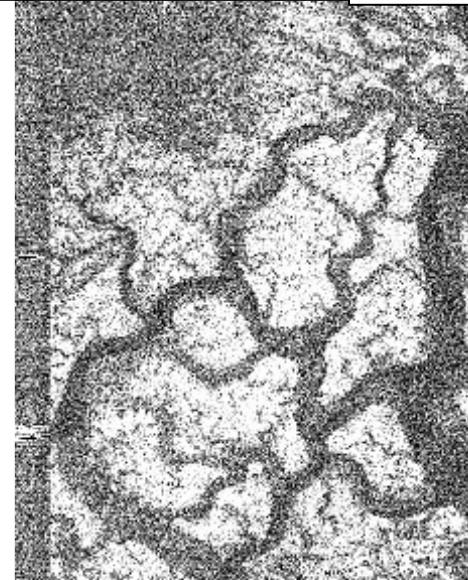
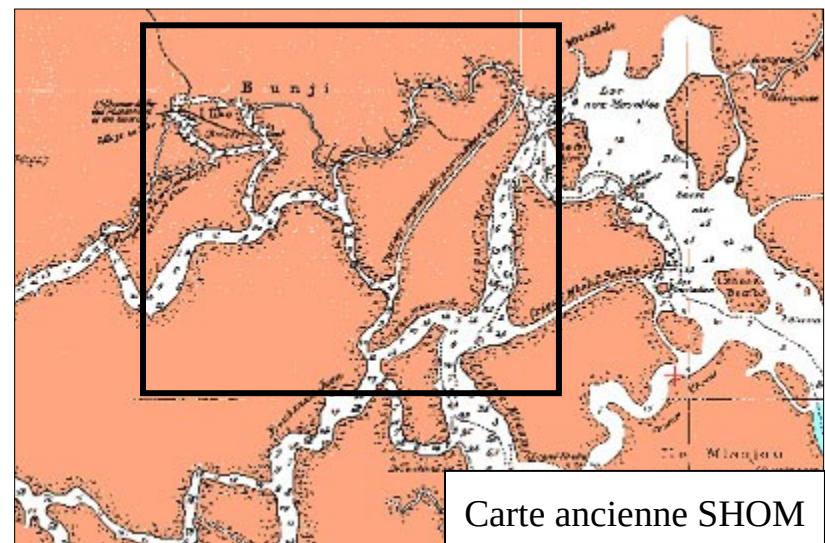
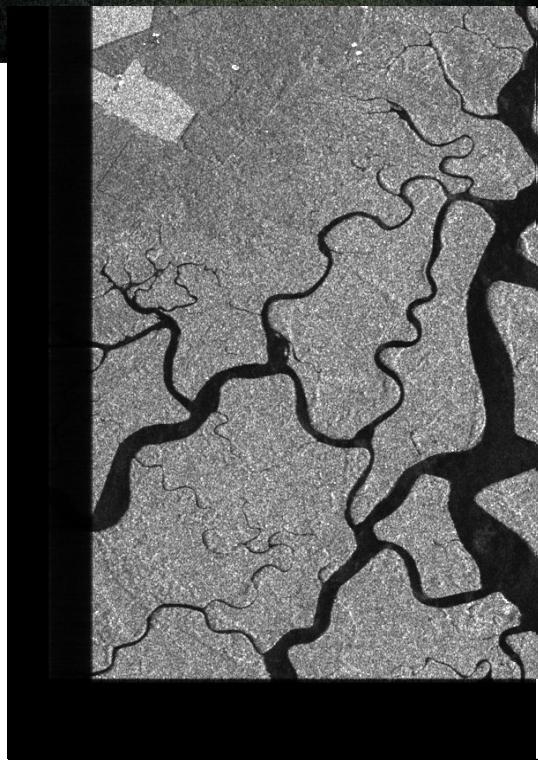


Mount Cameroun, ERS

Coherence

Radar response sensitivity

Mangroves

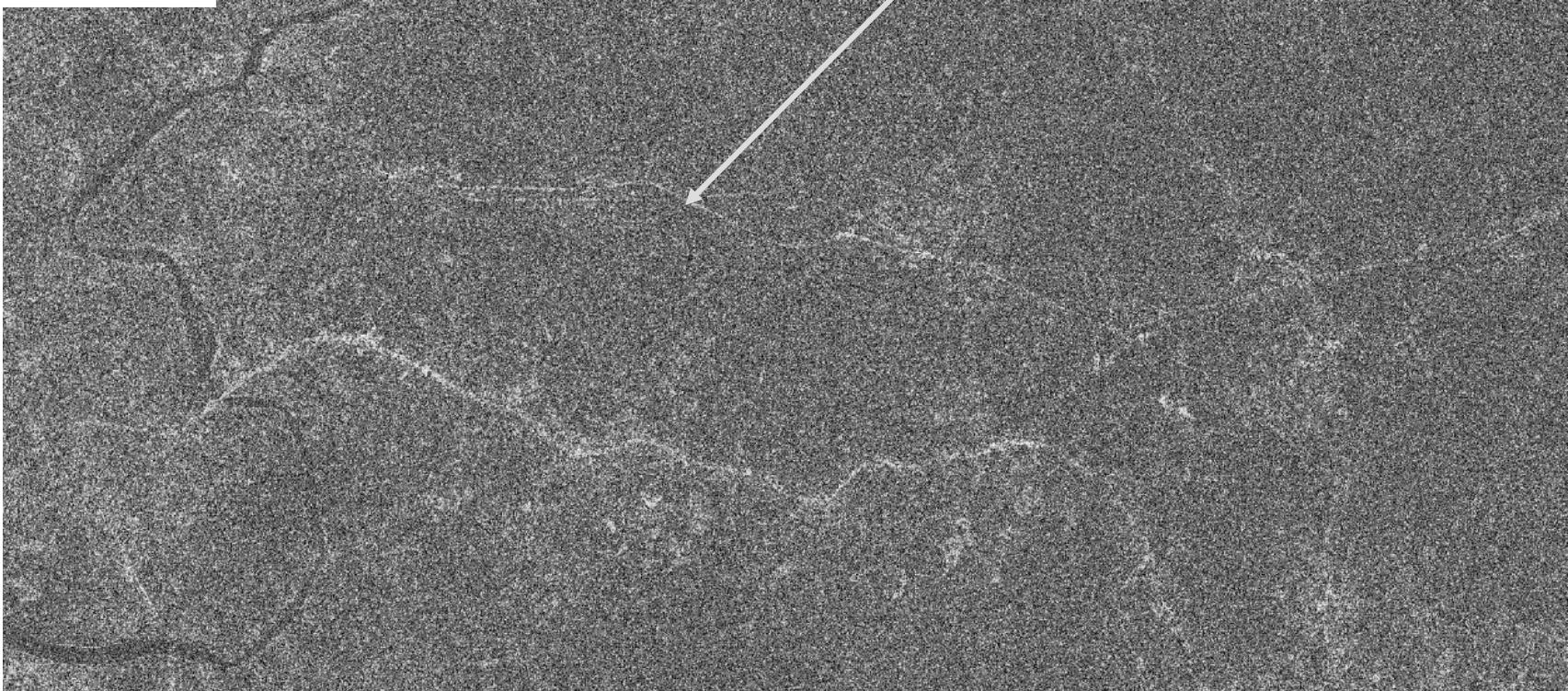


Coherence ERS

Amplitude ERS

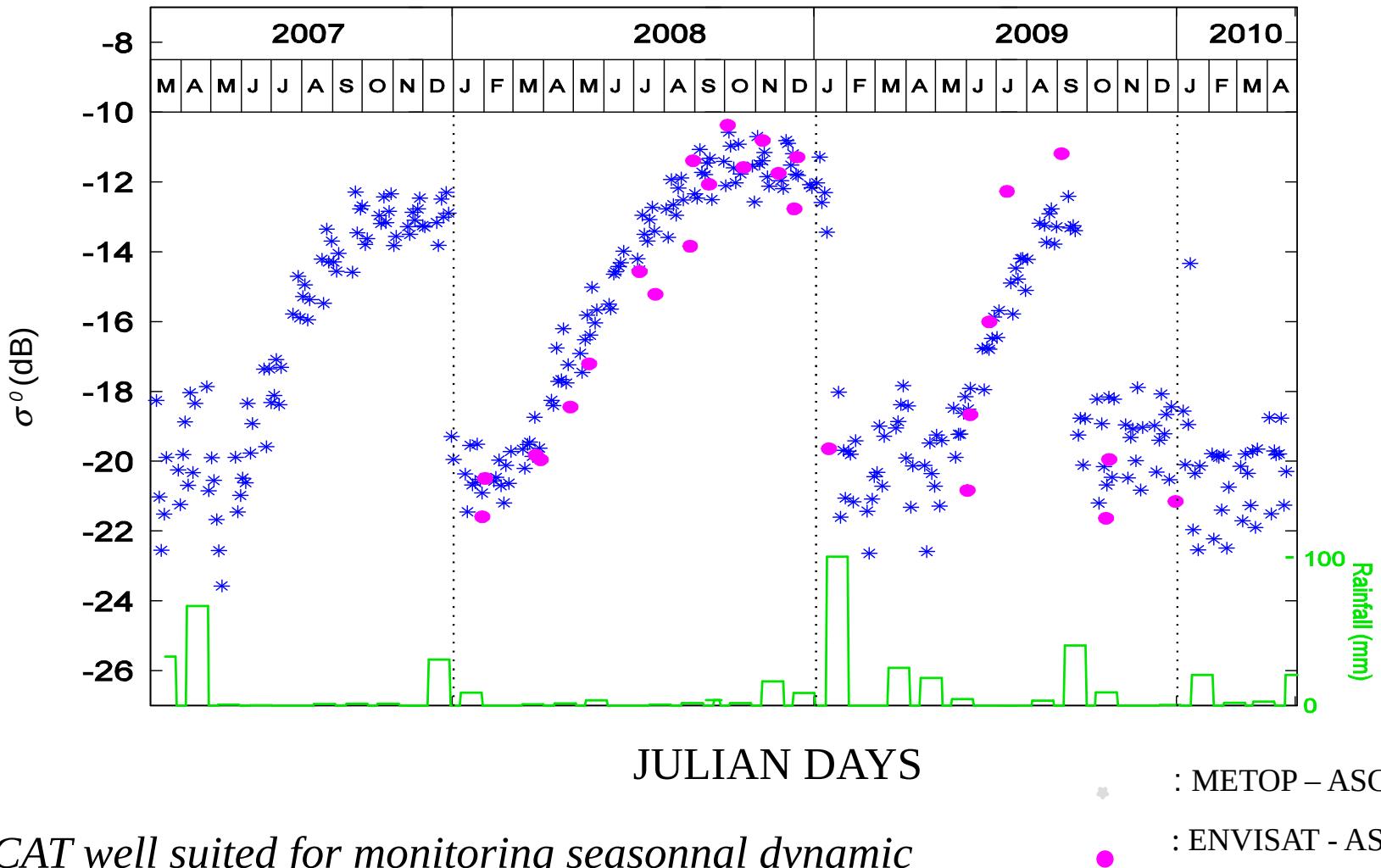


Coherence



ASCAT/ASAR temporal signature over the Chott el Jerid

Incidence angle: 40°



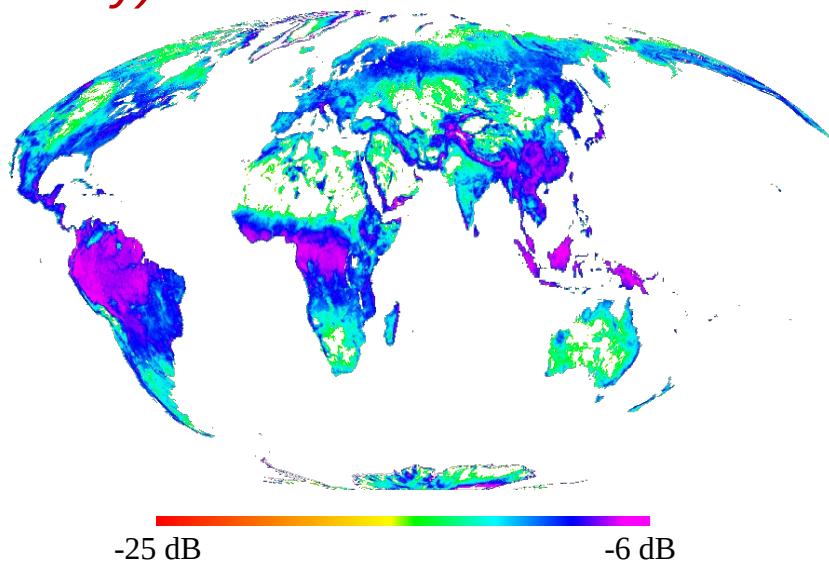
Radar remote sensing for land surfaces monitoring

Side looking radar sensors

Scatterometers

- Radar reflectivity estimation (σ°)

- *low spatial resolution: $\sim 10 - 50 \text{ km}$*
- *high frequency of acquisitions (\sim day)*



Scatt. ERS – May 1992

SAR

- Surface imaging

- *high spatial resolution: $\sim 10 \text{ m}$*
- *low frequency of acquisition (\sim month)*



Sentinel-1
Les Landes – March 2015

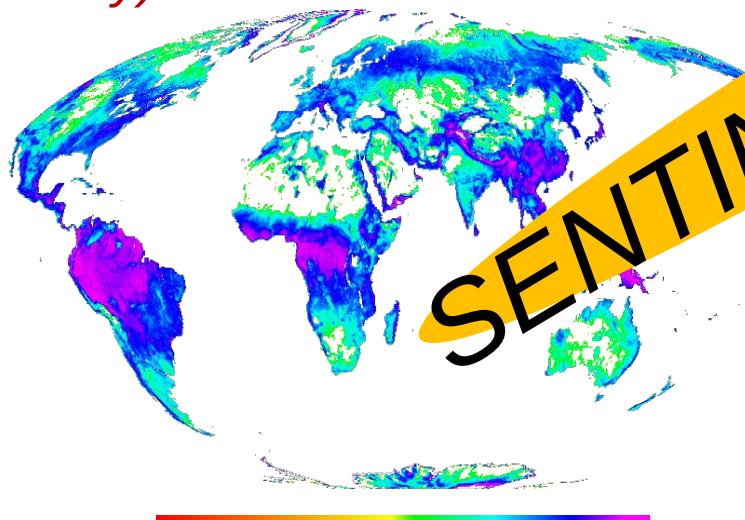
Radar remote sensing for land surfaces monitoring

Side looking radar sensors

Scatterometers

▫ Radar reflectivity estimation (σ°)

- *low spatial resolution*: $\sim 10 - 50 \text{ km}$
- *high frequency of acquisitions* (\sim day)



Scatt. ERS – May 1992

SAR

▫ Surface imaging

- *high spatial resolution*: $\sim 10 \text{ m}$
- *long latency of acquisition* (\sim month)



Sentinel-1
Les Landes – March 2015

The Sentinel-1 missions

Sentinel-1A: launched the 3rd April 2014

==> SAR data from March 2015 Revisit time: 12 days

]} 6 days!!

Sentinel-1B: launched the 22th April 2016 Revisit time: 12 days

==> SAR data from September 2016

- C band
- Spatial resolution: 20 m
- Swath width: 250 km
- Two polarizations over land surfaces: VV and VH

SAR

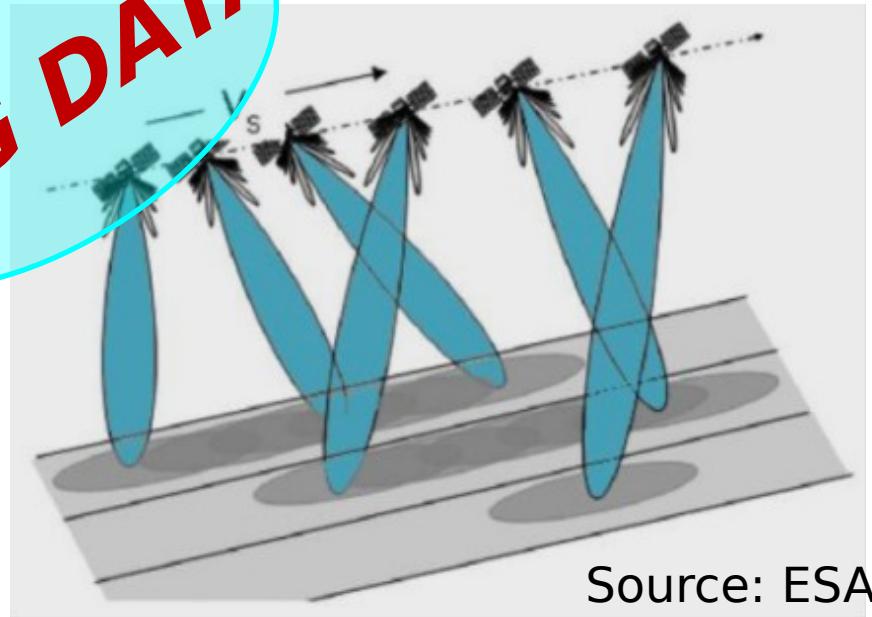
SENTINEL-1

Scatter

Acquisitions period: **12 days** (S1-A) - **6 days** (S1- A+B)

Planned mode over land surfaces: **Interferometric Wide (IW)**

BIG DATA



2 Polarisations: **VV - VH**

Swath: 250 km (3 sub-swaths)

GRD Products :

Spatial resolution: **20 m**

Pixel: 10 m

SLC Products

Spatial resolution: 3 x 20 m;

Pixel: 2 x 14 m (rge x az.)

Temporal monitoring of seasonal variations of land surfaces

Radar Backscattering Coefficient σ^0

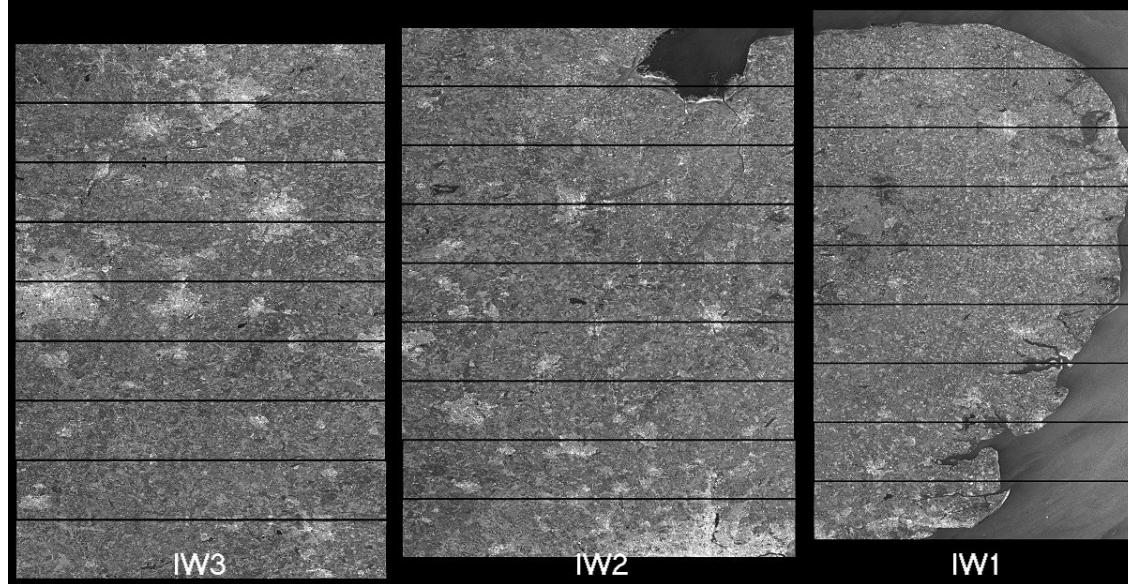
Interferometric Coherence $|p|$

SENTINEL-1 INTERFEROMETRIC WIDE, MODE

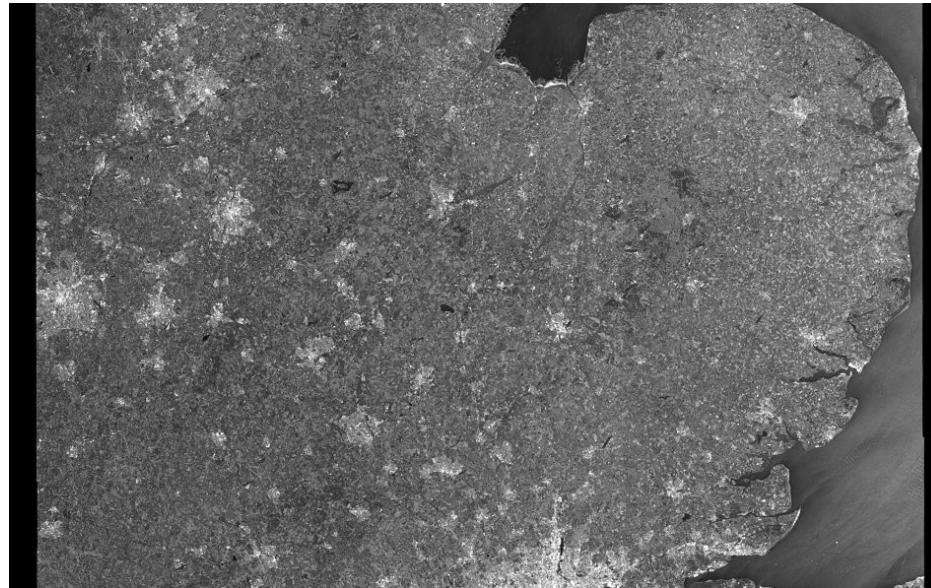
3 subswaths

SLC products

8 bursts

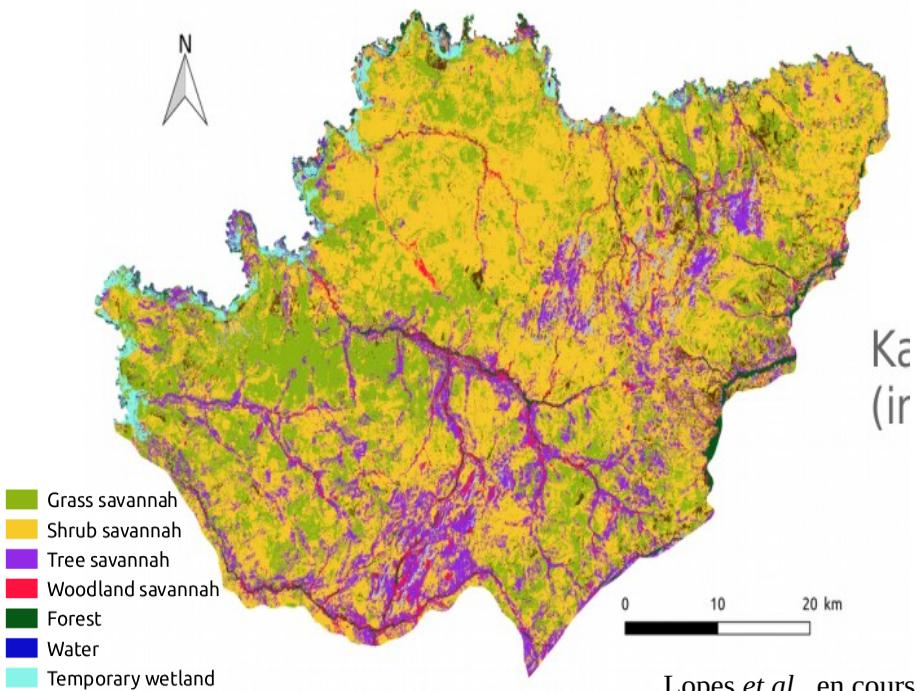


GRD products



Sentinel: Apport des séries temporelles

Formations végétales
Parc de la Pendjari, Bénin



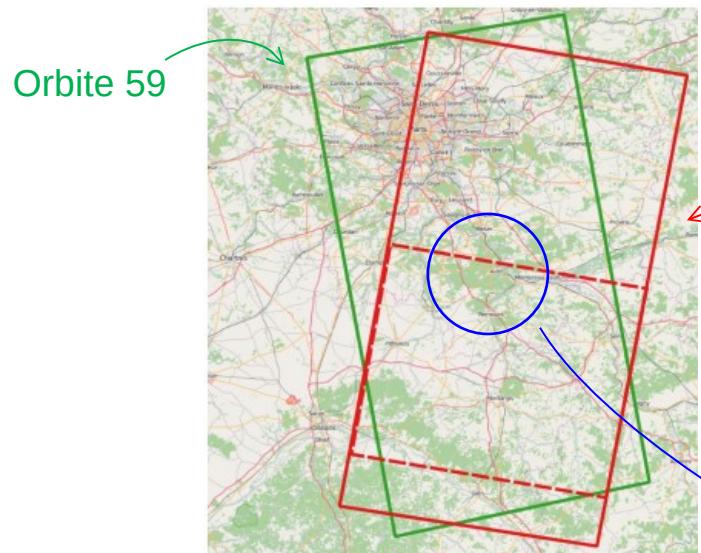
Année 2018
Données Sentinel-1 et -2

Données Sentinel (*Big Data*)

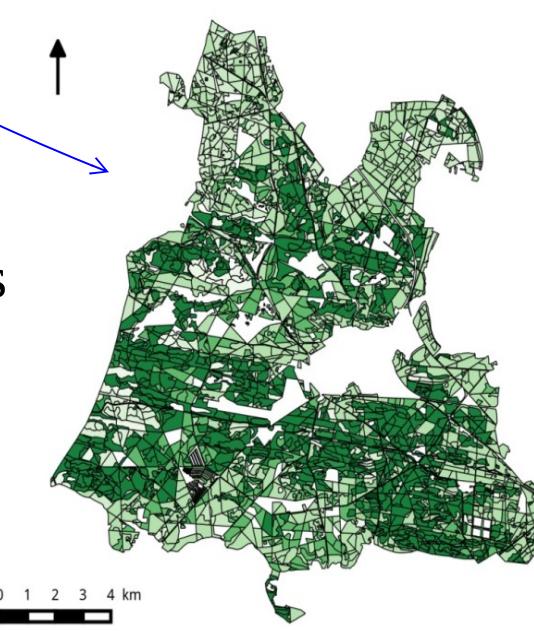
Fouille de données
Intelligence artificielle
Deep learning

Complémentarité optique / radar

Acquisitions over the Paris region



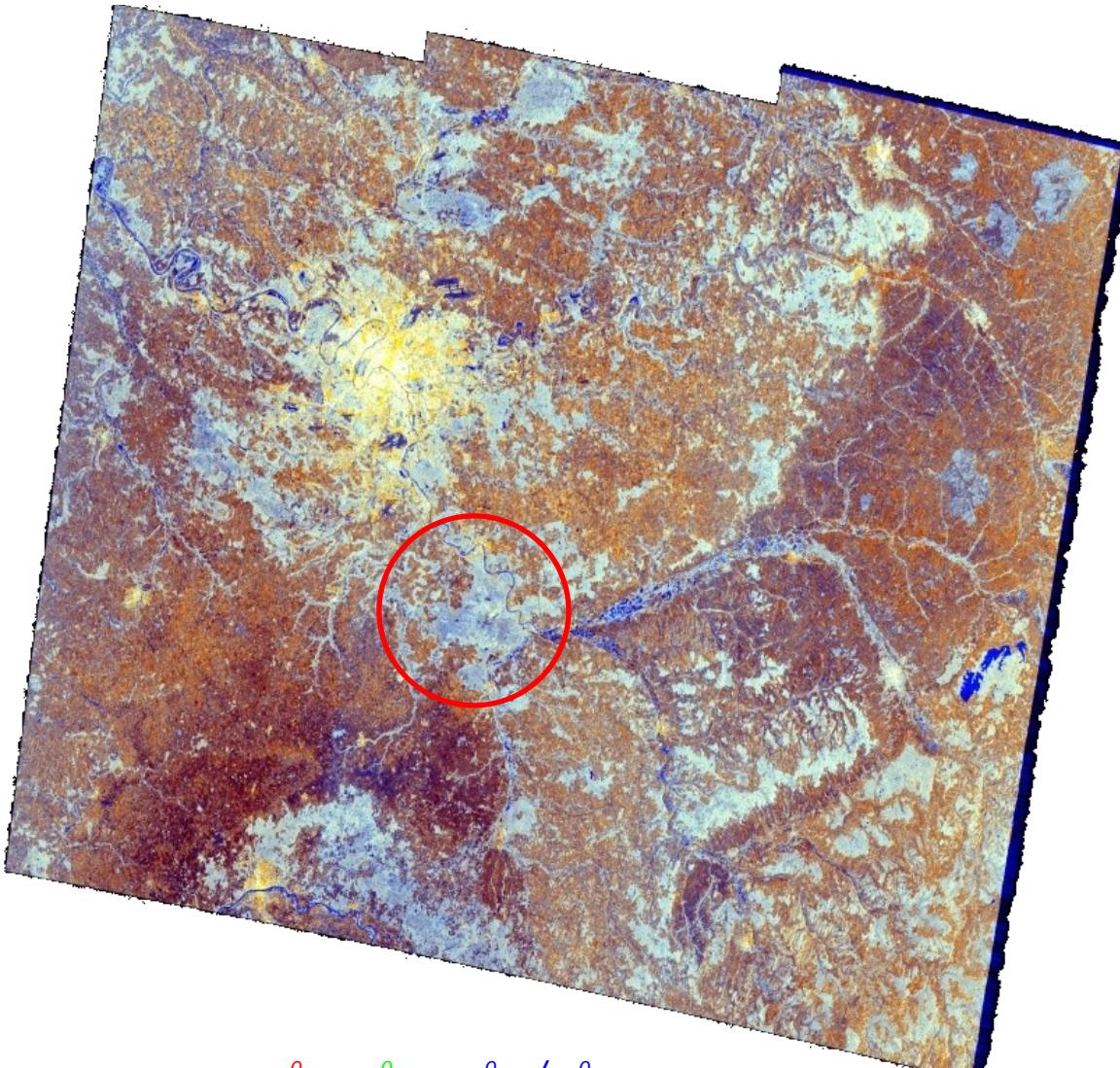
Oaks, pines and beeches stands



Fontainebleau Forest

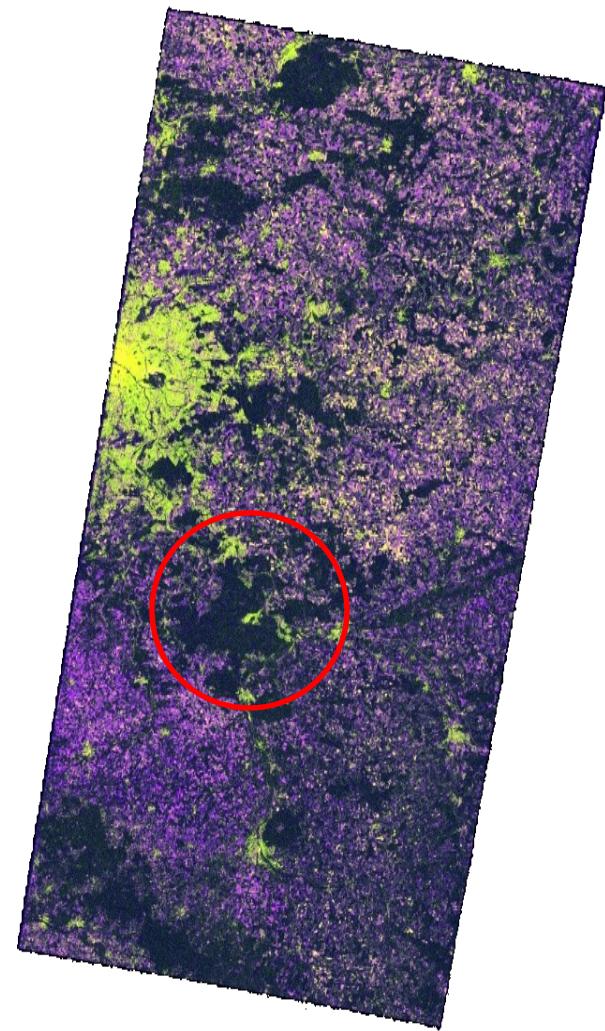
18th March 2015 IW Acquisition

Radar reflectivity (σ^0) image



$$\sigma_{\text{VV}}^0 - \sigma_{\text{VH}}^0 - \sigma_{\text{VH}}^0 / \sigma_{\text{VV}}^0$$

18 – 30 March coherence image

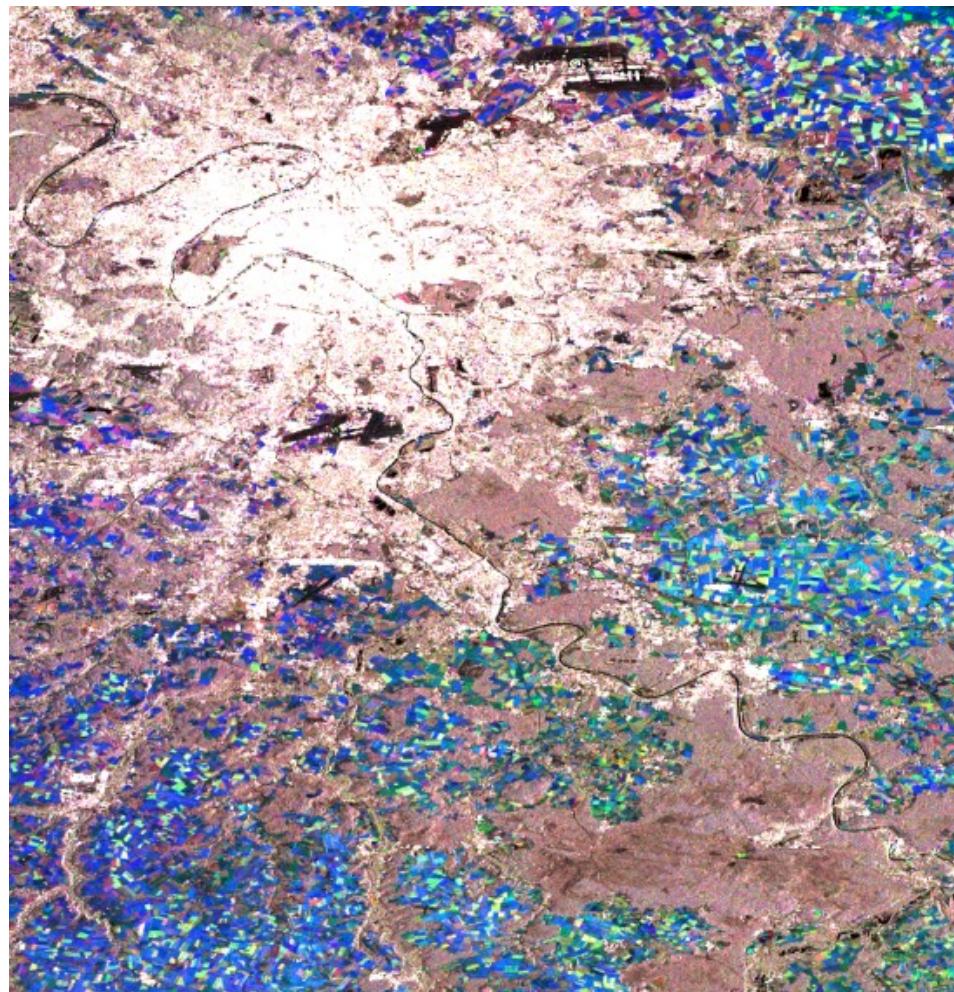


$$|\rho_{\text{VV}}| - |\rho_{\text{VH}}| - |\rho_{\text{VH}}| / |\rho_{\text{VV}}|$$

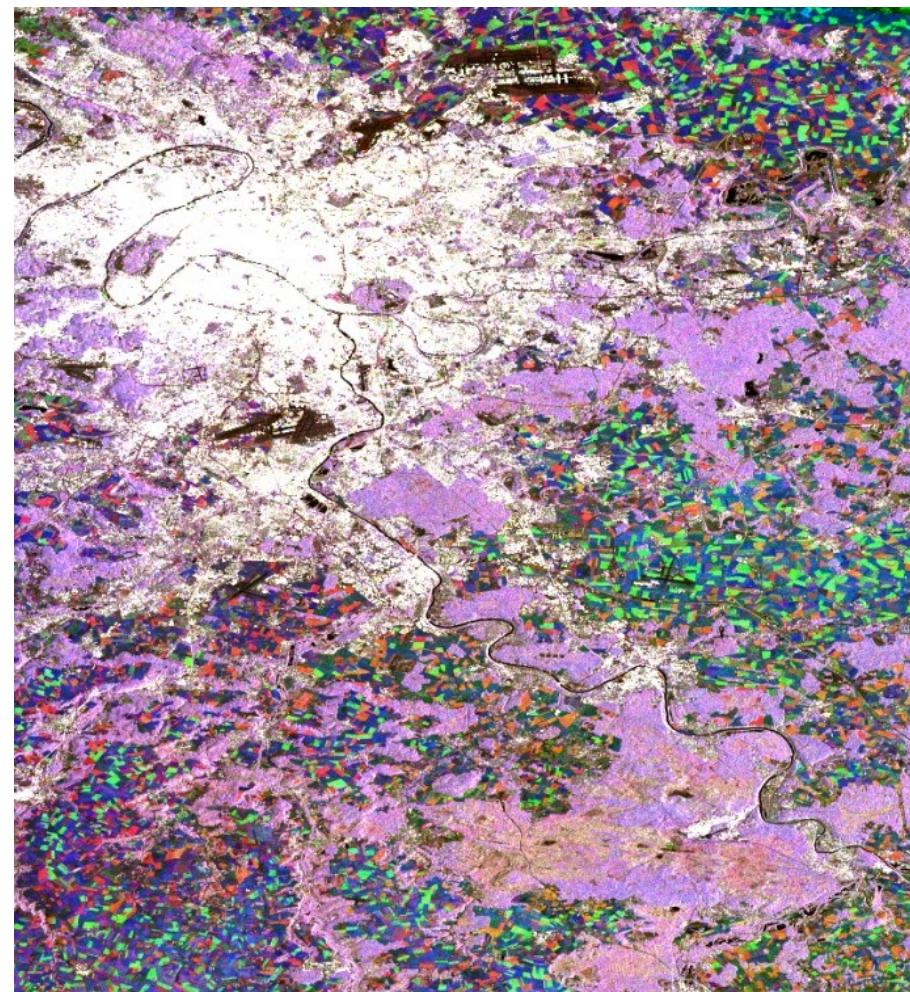
σ^0 Color composite image

5 May - 2 Sept. - 19 Dec. 2015

Polarisation **VV**



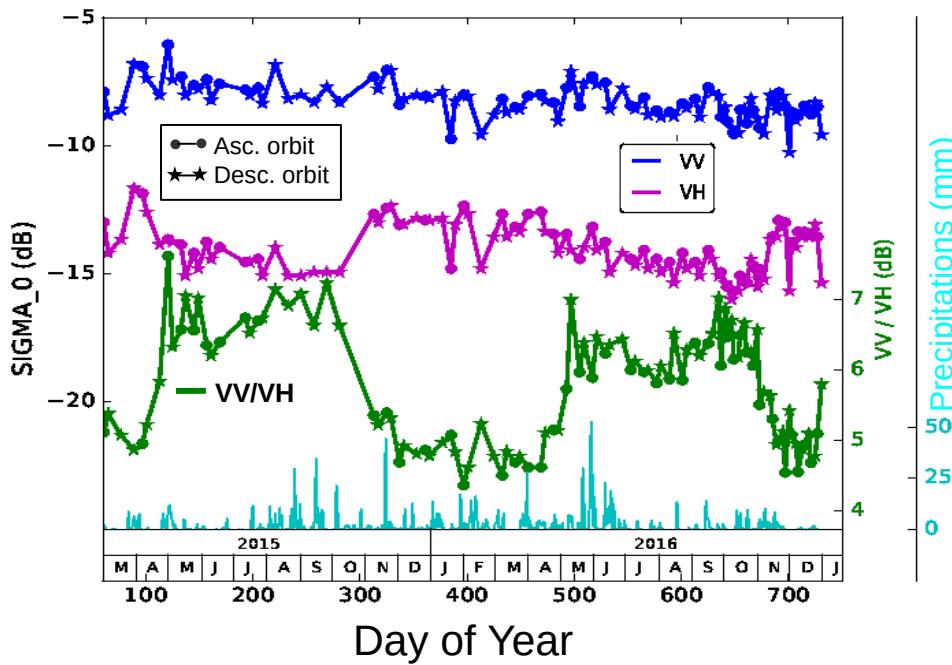
Polarisation **VH**



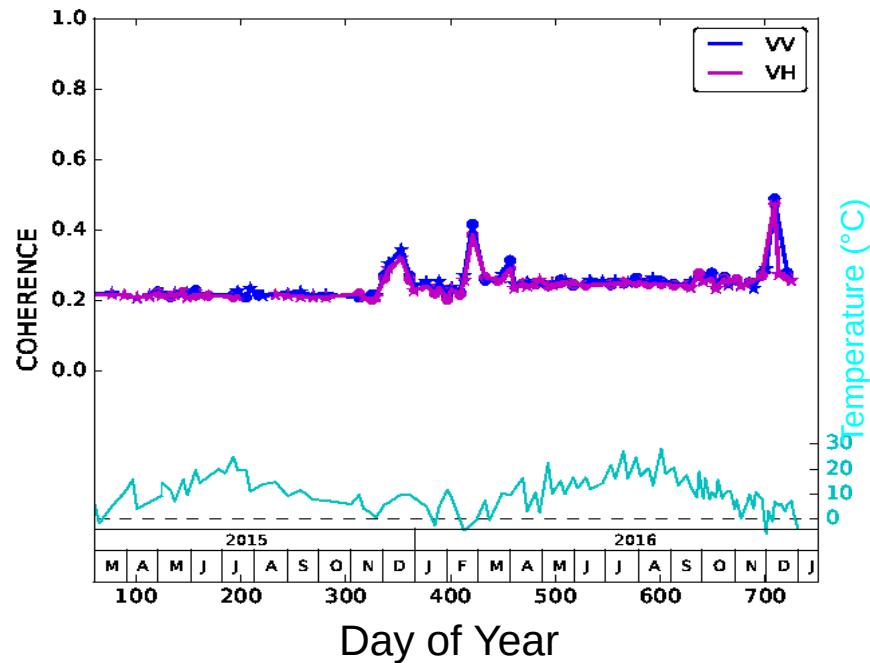
§ High spatio-temporal variability over crop fields

Oaks stand

Radar Backscat. Coeff. σ^0



Coherence $|\rho|$

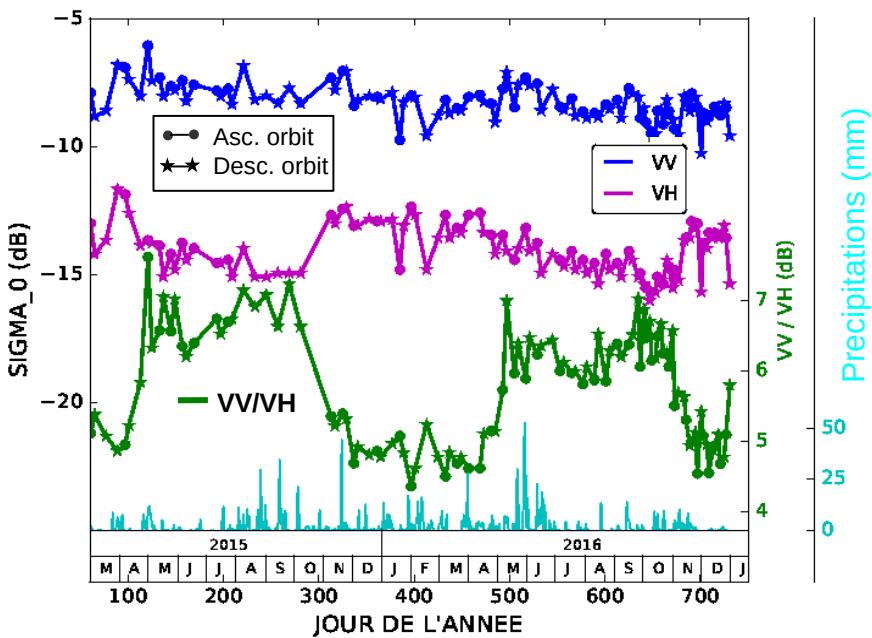


- No seasonal cycle s_{VV}^0
- Seasonal cycles $s_{VH}^0 \Rightarrow \sigma_{VV}^0 / \sigma_{VH}^0$
(*yearly amplitude 3 dB*)

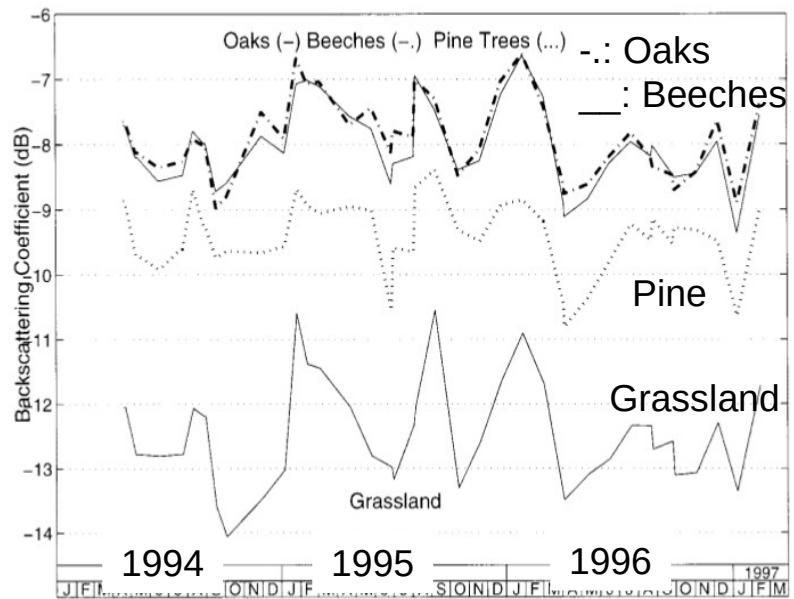
- signal low and constant (Mar. - Nov.)
- $|\rho_{VV}|$ et $|\rho_{VH}|$ Identical
- higher values for low temperatures

Oaks stand

Radar Backscat. Coeff. σ^0



ERS (VV) temporal signature



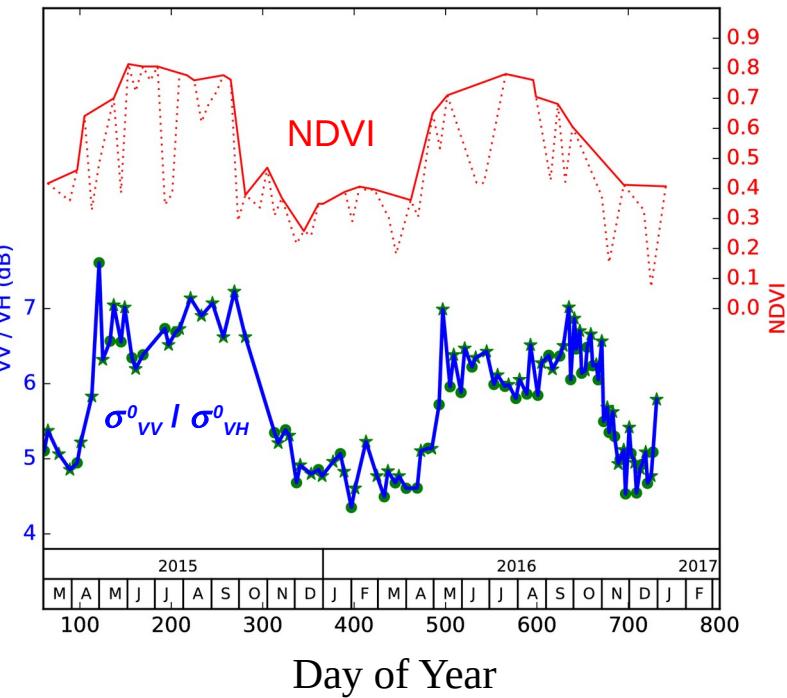
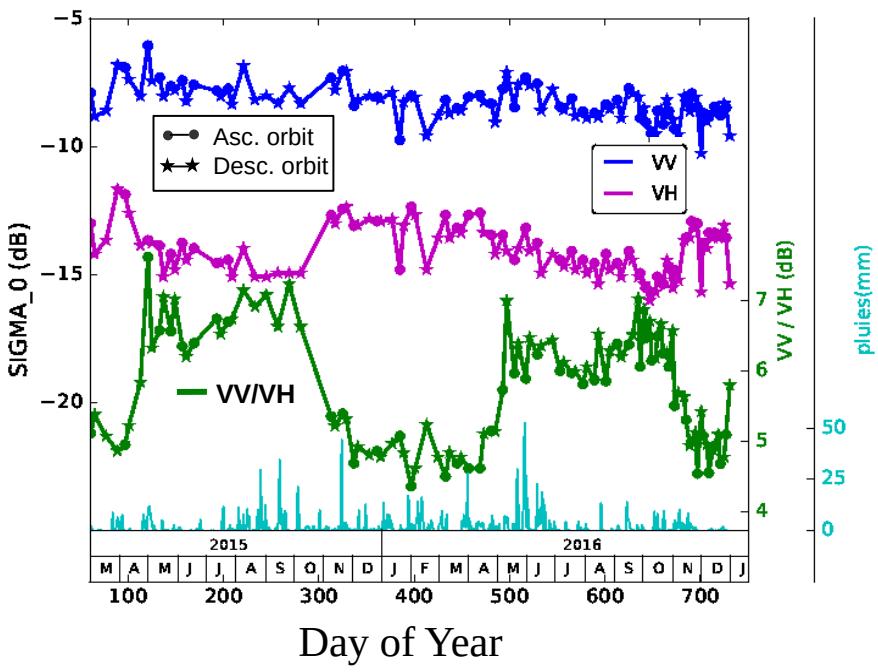
Proisy et al., 1999

no seasonal cycle in VV pol.

Seasonal cycle $\sigma^0_{VH} \Rightarrow \sigma^0_{VV} / \sigma^0_{VH}$

Oaks stand

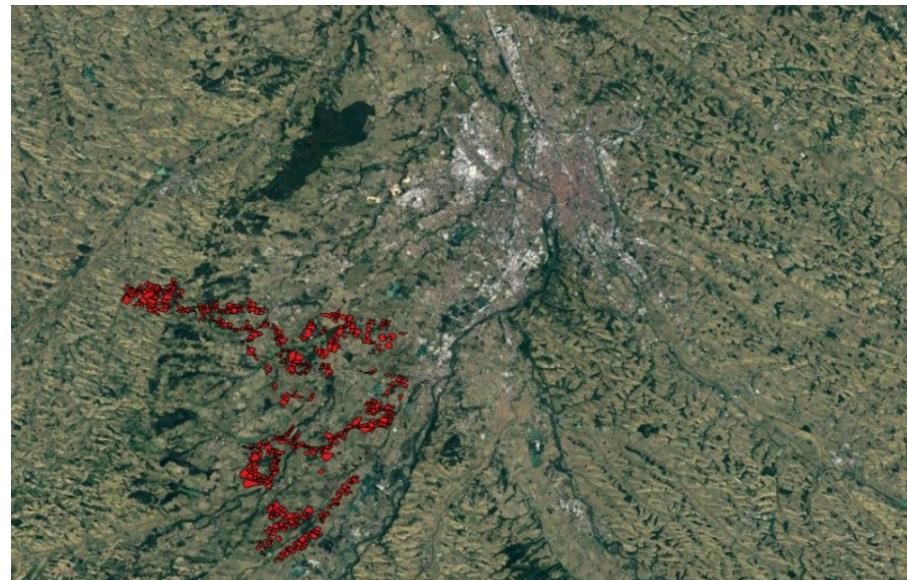
Radar Backscat. Coeff. σ^0



$\sigma^0_{VV} / \sigma^0_{VH}$ and NDVI in phase

⌚ C band sensitive to foliar activity

Crops monitoring – Lamasquère region



in situ survey (CESBIO)

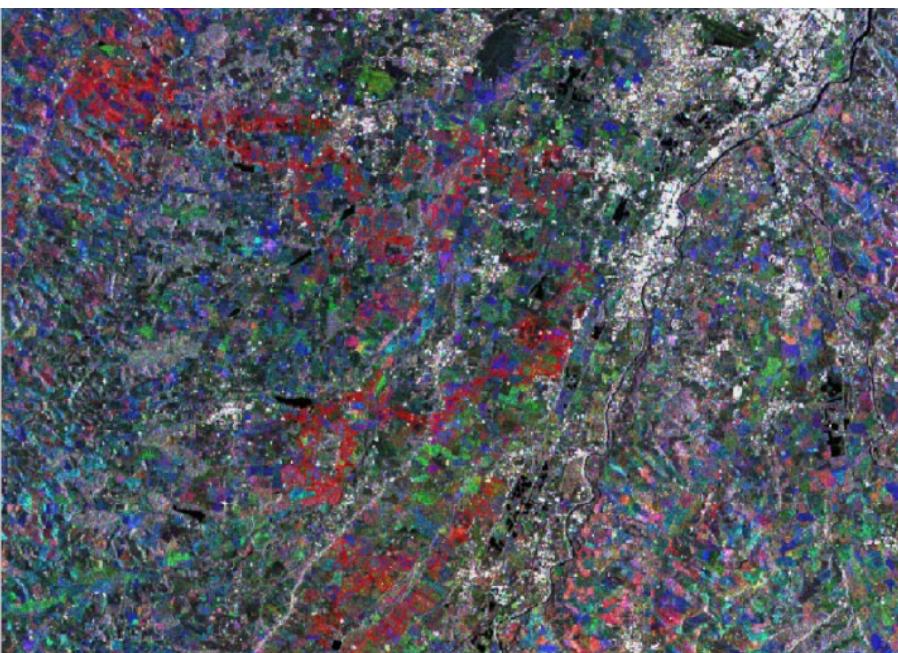
Winter crops: wheat, barley, rapeseed

Summer crops: soybean, sorghum, maize, sunflower

Agricultural area (Lamasquère region)

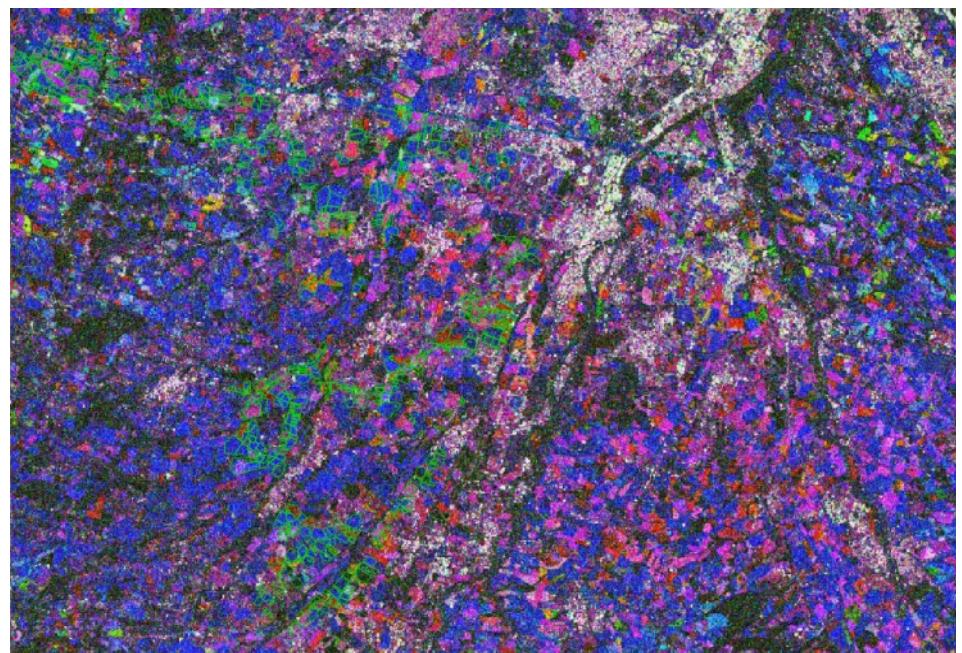
Multi-temporal color-composite images

Radar Backscatterin Coeff.



10 June- 14 Sept. – 7 Dec.

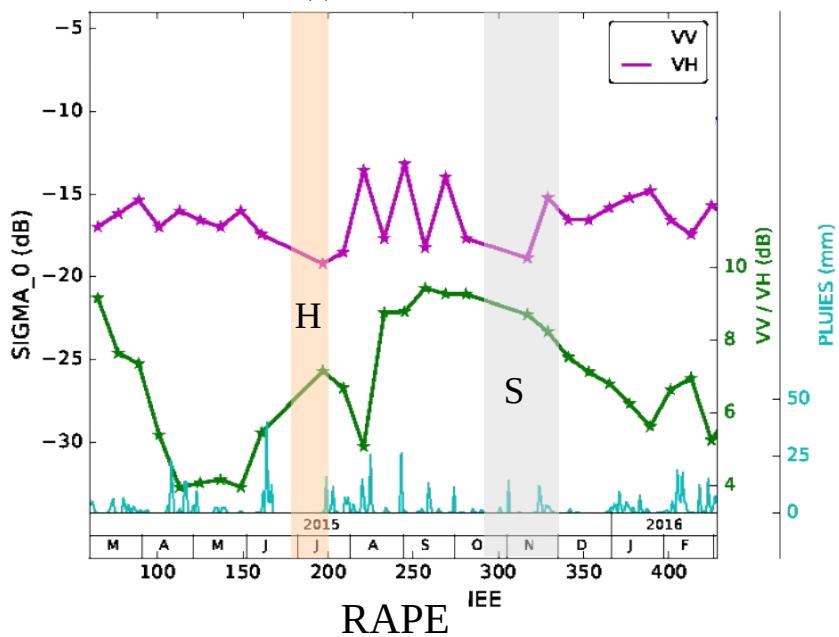
Cohérence



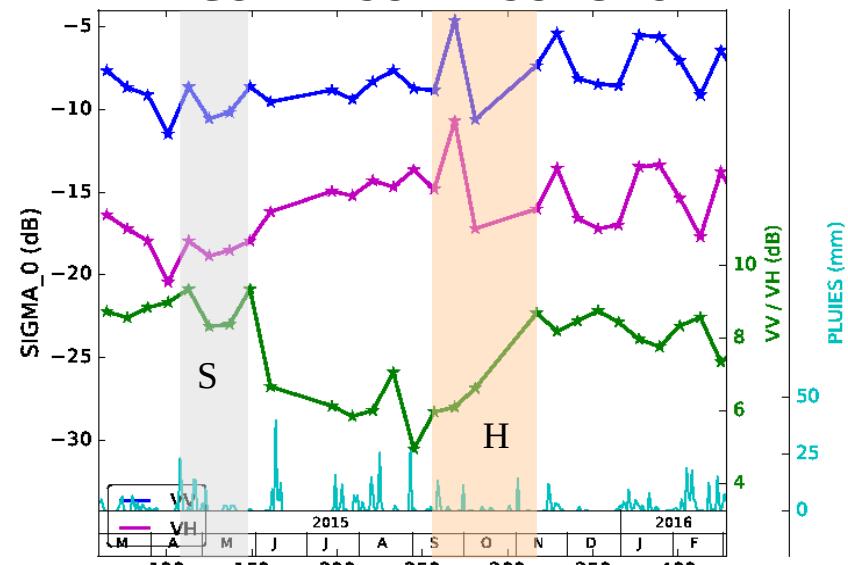
4-16 Jul .- 9-16 Aug. – 7-19 Dec.

CROP FIELDS: Temporal profile σ^0

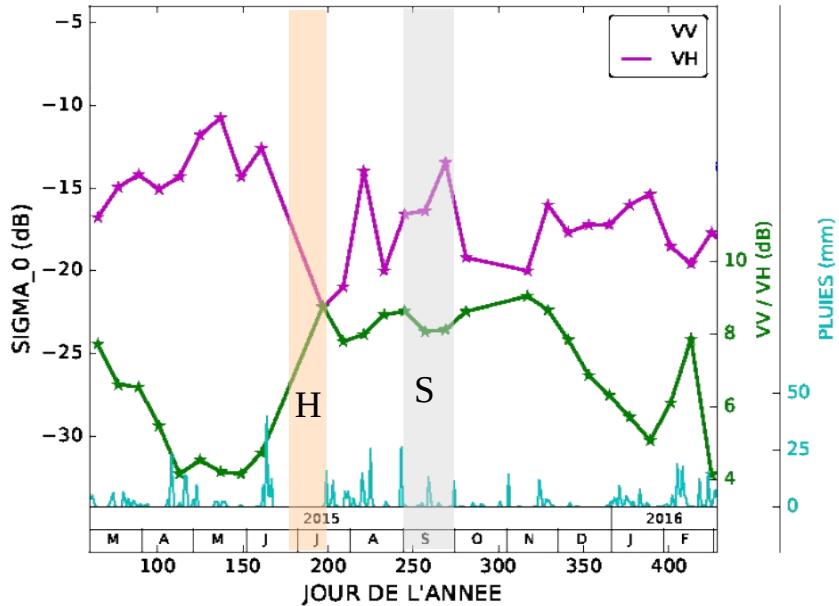
WHEAT



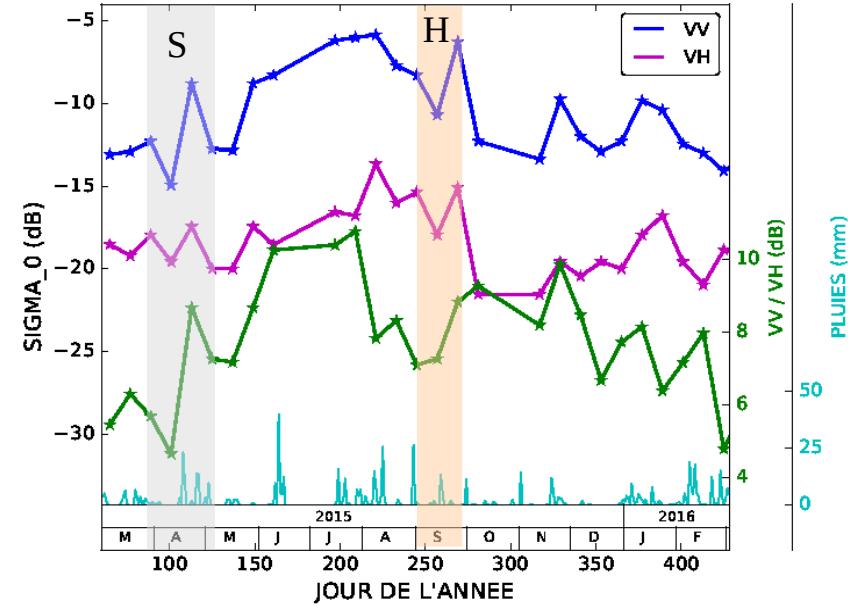
CORN – SOYA - SORGHUM



RAPE

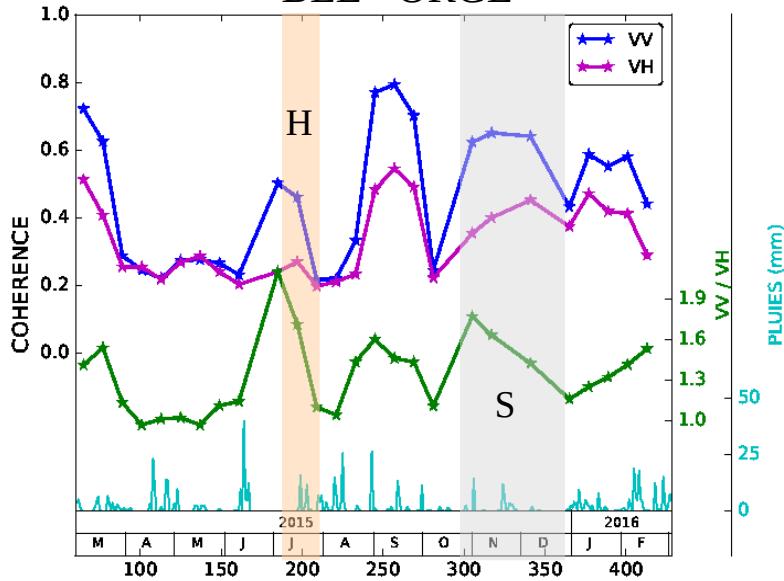


SUNFLOWER

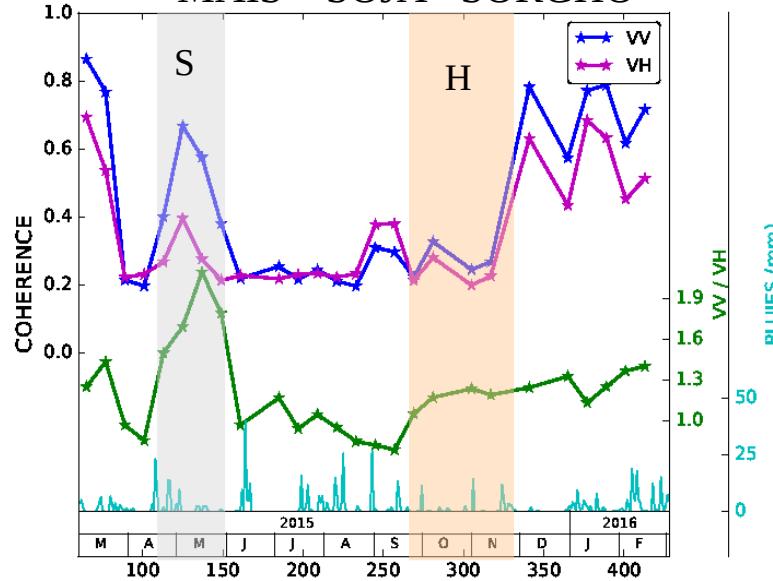


CROP FIELDS: Temporal profiles coherence

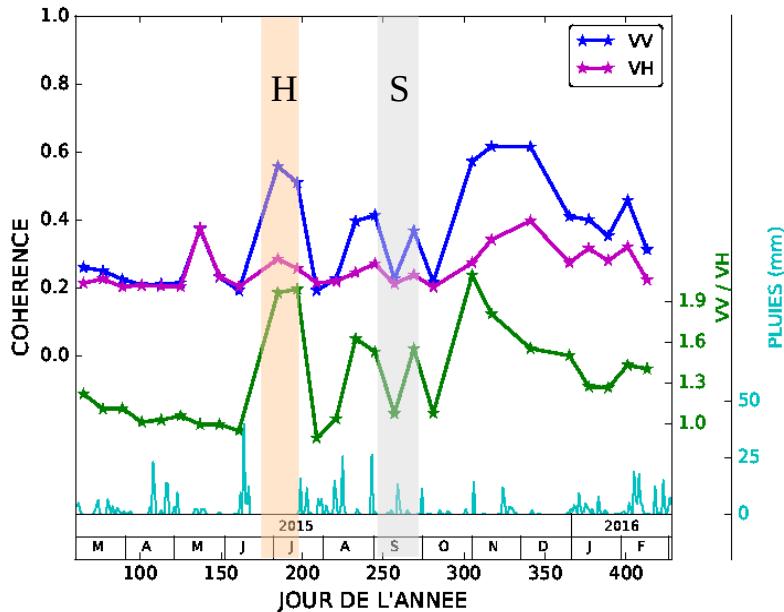
BLE - ORGE



MAIS – SOJA - SORGHO



COLZA



TOURNESOL

