



Monitoring small-scale changes in Forest coverage

Comparing passive and active Sentinel data over Białowieża forest

Margherita Di Leo

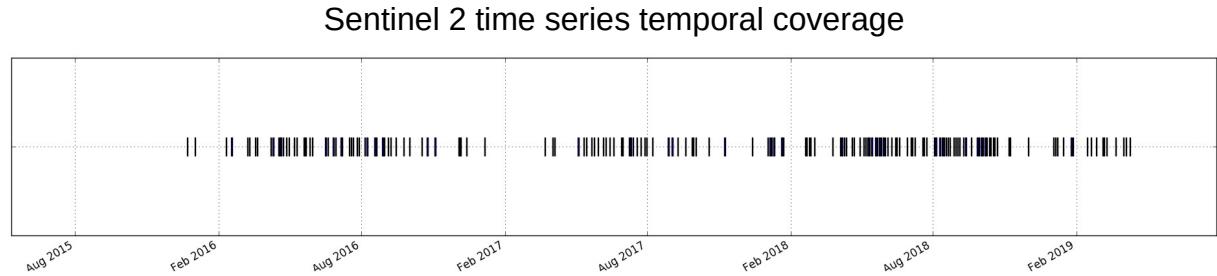
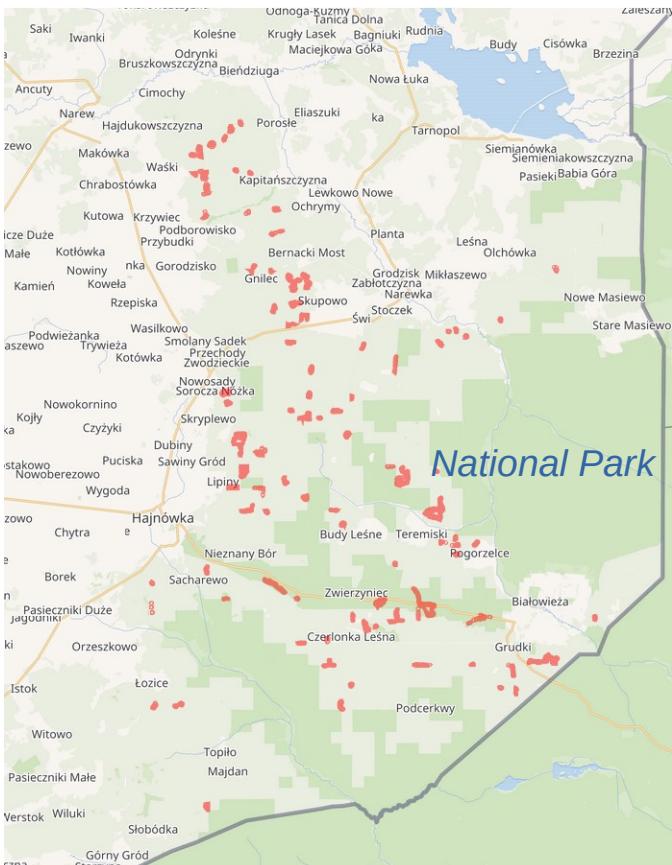


Logging activities in a Natura 2000 site: Białowieża (PL)



Deforestation in the Białowieża Forest. (Photo: Greenpeace Polska, CC BY-ND 2.0)

Białowieża (PL)



The AOI is ~1122 km², of which 105 km² National Park

Elevation

min = 126 m a.s.l.
 max = 208
 range = 82
 mean = 163
 sd = 12.8

Slope:

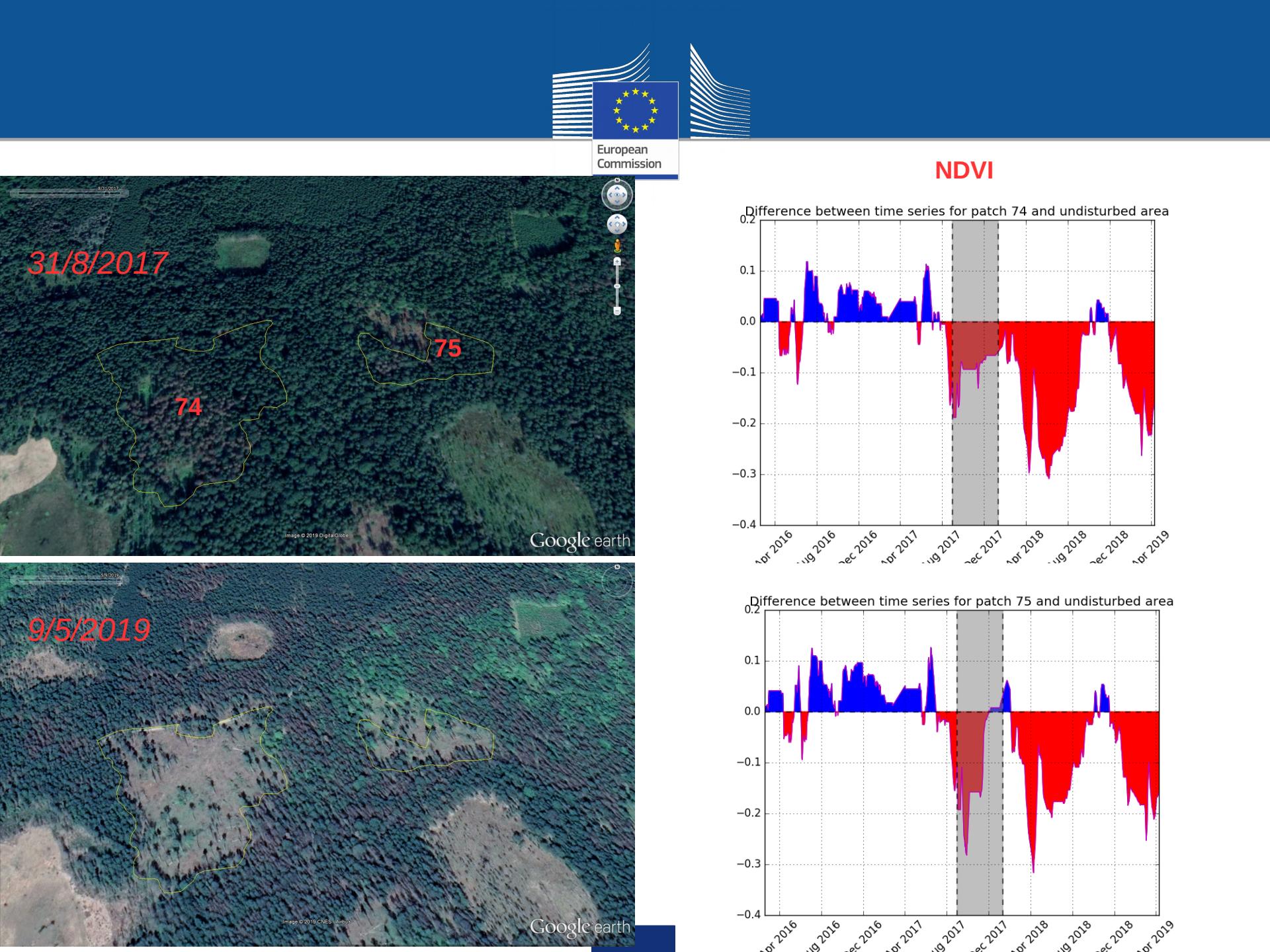
min = 0
 max = 17.95
 mean = 1.46
 sd = 1.1

WH:

- Signal / noise ratio similarity
- Phenological homogeneity

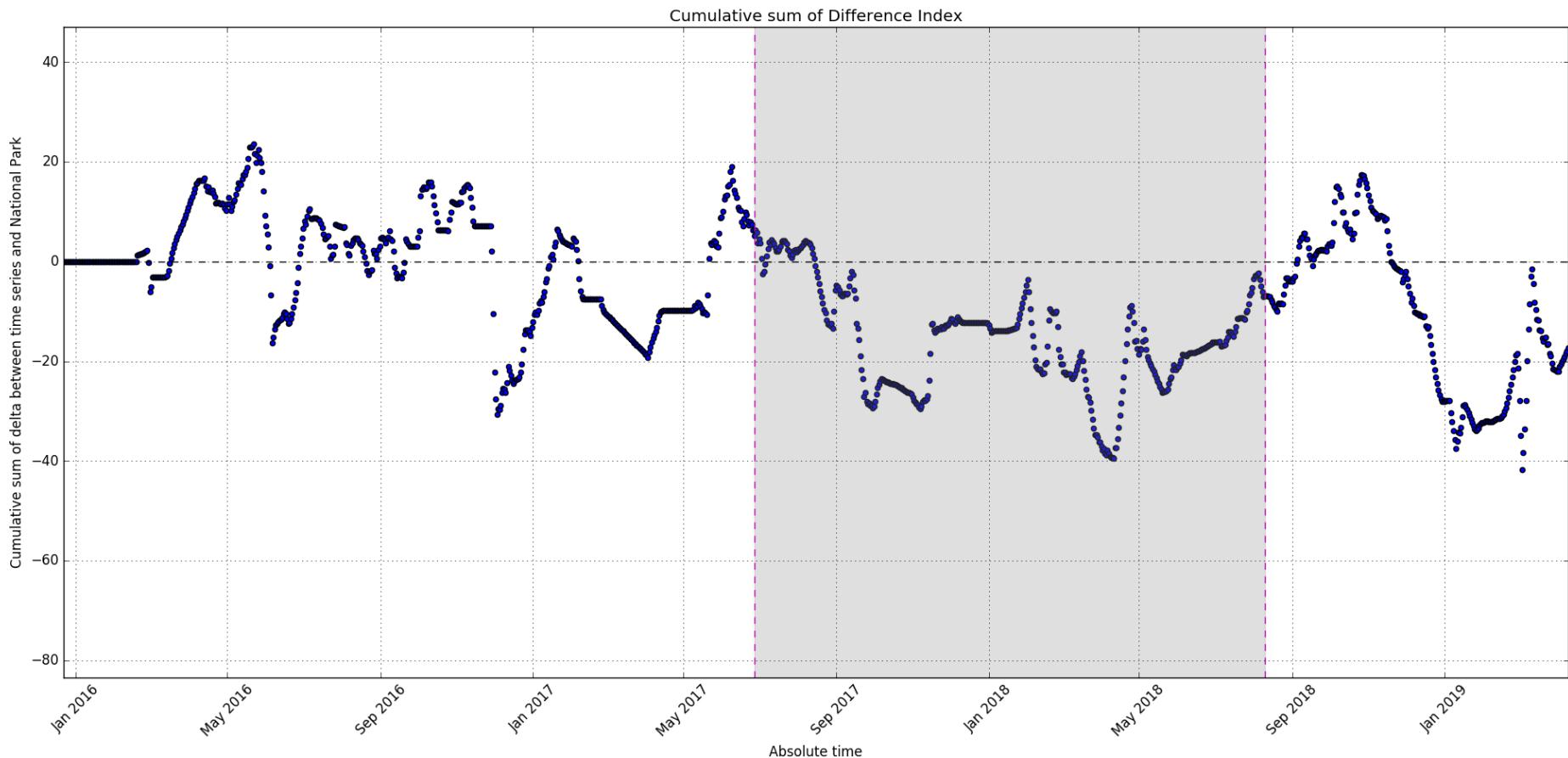
Difference index

$$\Delta_i = TS_i - NP$$





Cumulative sum of difference index





Where to look?



NDVI Anomalies

Departure of NDVI from long-term average, normalized by long-term variability

Generated by subtracting the long-term mean from the current value for that month of the year for each pixel

Indicate if the vegetation **greenness** at a particular location is typical for that period or if the vegetation is more or less green

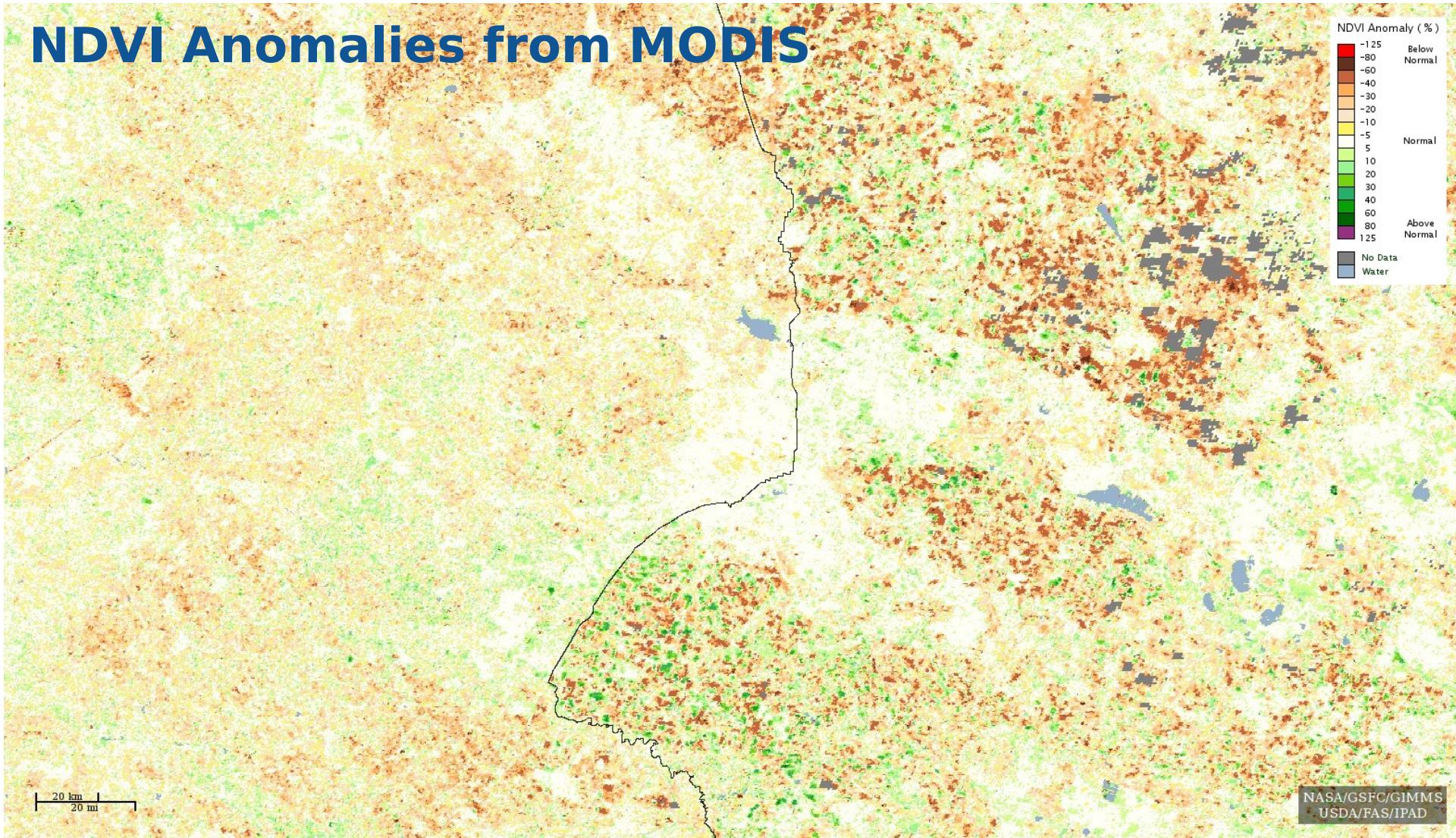
Limitations

Sentinel 2: good resolution but time series too short for stats

MODIS: resolution insufficient to detect small changes in the forest coverage



NDVI Anomalies from MODIS





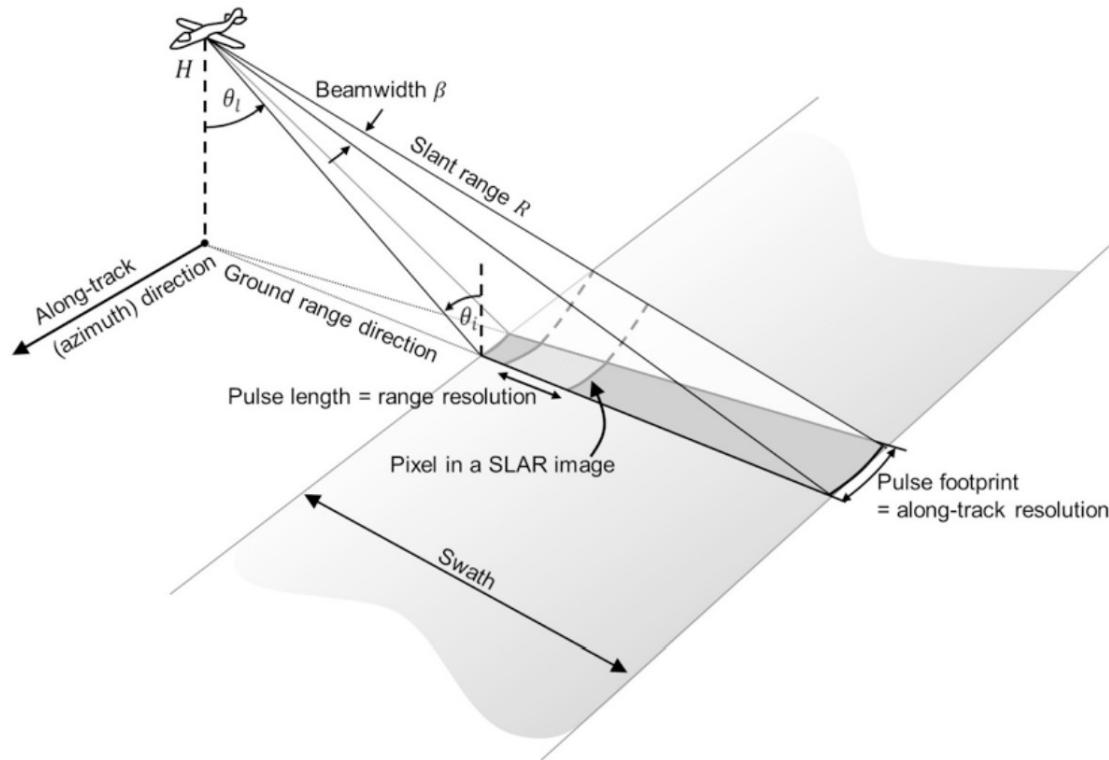
Exploring SAR



Applications of SAR in Forest Monitoring

- Deforestation and forest degradation
- Forest structure estimation
- Biomass estimation
- Mangrove monitoring

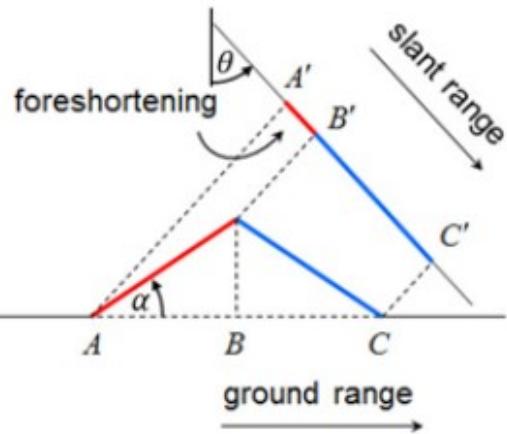
Side-Looking Airborne Radar (SLAR)



Main geometric distortions on SAR images with their dependence on acquisition geometry: (a) foreshortening, (b) layover, and (c) shadow

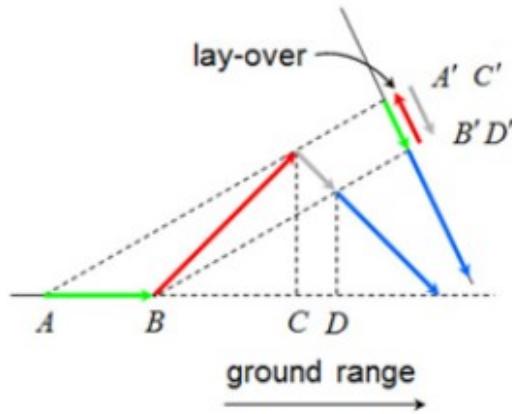
FORESHORTENING

- Sensor-facing slope foreshortened in image
- Foreshortening effects *decrease* with increasing look angle



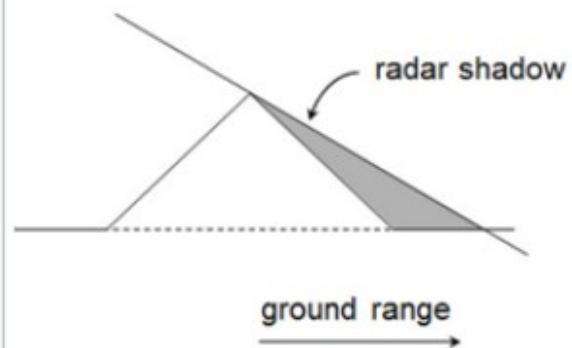
LAYOVER

- Mountain top overlain on ground ahead of mountain
- Layover effects *decrease* with increasing look angle



SHADOW

- Area behind mountain cannot be seen by sensor
- Shadow effects *increase* with increasing look angle



SAR signal penetration by sensor wavelength λ

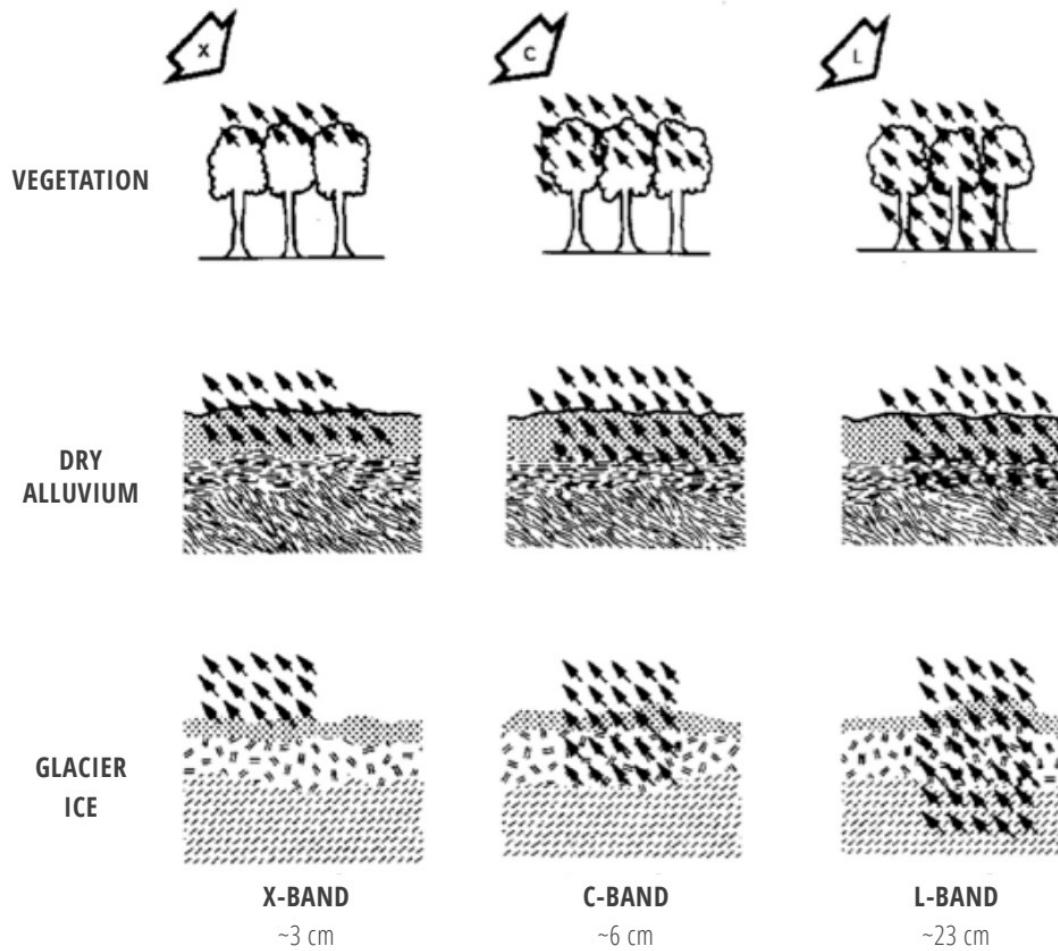


Image after Flores-Anderson, Africa Ixmuca, et al. "The SAR Handbook: Comprehensive Methodologies for Forest Monitoring and Biomass Estimation." (2019).



Designation of microwave bands

BAND	FREQUENCY	WAVELENGTH	TYPICAL APPLICATION
Ka	27 – 40 GHz	1.1 – 0.8 cm	Rarely used for SAR (airport surveillance)
K	18 – 27 GHz	1.7 – 1.1 cm	Rarely used (H_2O absorption)
Ku	12 – 18 GHz	2.4 – 1.7 cm	Rarely used for SAR (satellite altimetry)
X	8 – 12 GHz	3.8 – 2.4 cm	High-resolution SAR (urban monitoring; ice and snow, little penetration into vegetation cover; fast coherence decay in vegetated areas)
C	4 – 8 GHz	7.5 – 3.8 cm	SAR workhorse (global mapping; change detection; monitoring of areas with low to moderate vegetation; improved penetration; higher coherence); Ice, ocean, maritime navigation
S	2 – 4 GHz	15 – 7.5 cm	Little but increasing use for SAR-based Earth observation; agriculture monitoring (NISAR will carry an S-band channel; expands C-band applications to higher vegetation density)
L	1 – 2 GHz	30 – 15 cm	Medium resolution SAR (Geophysical monitoring; biomass and vegetation mapping; high penetration; InSAR)
P	0.3 – 1 GHz	100 – 30 cm	Biomass. First P-band spaceborne SAR will be launched ~2020; vegetation mapping and assessment. Experimental SAR.

Table after Flores-Anderson, Africa Ixmucá, et al. "The SAR Handbook: Comprehensive Methodologies for Forest Monitoring and Biomass Estimation." (2019).

Schematic sketch of the three main scattering types considered for SAR data

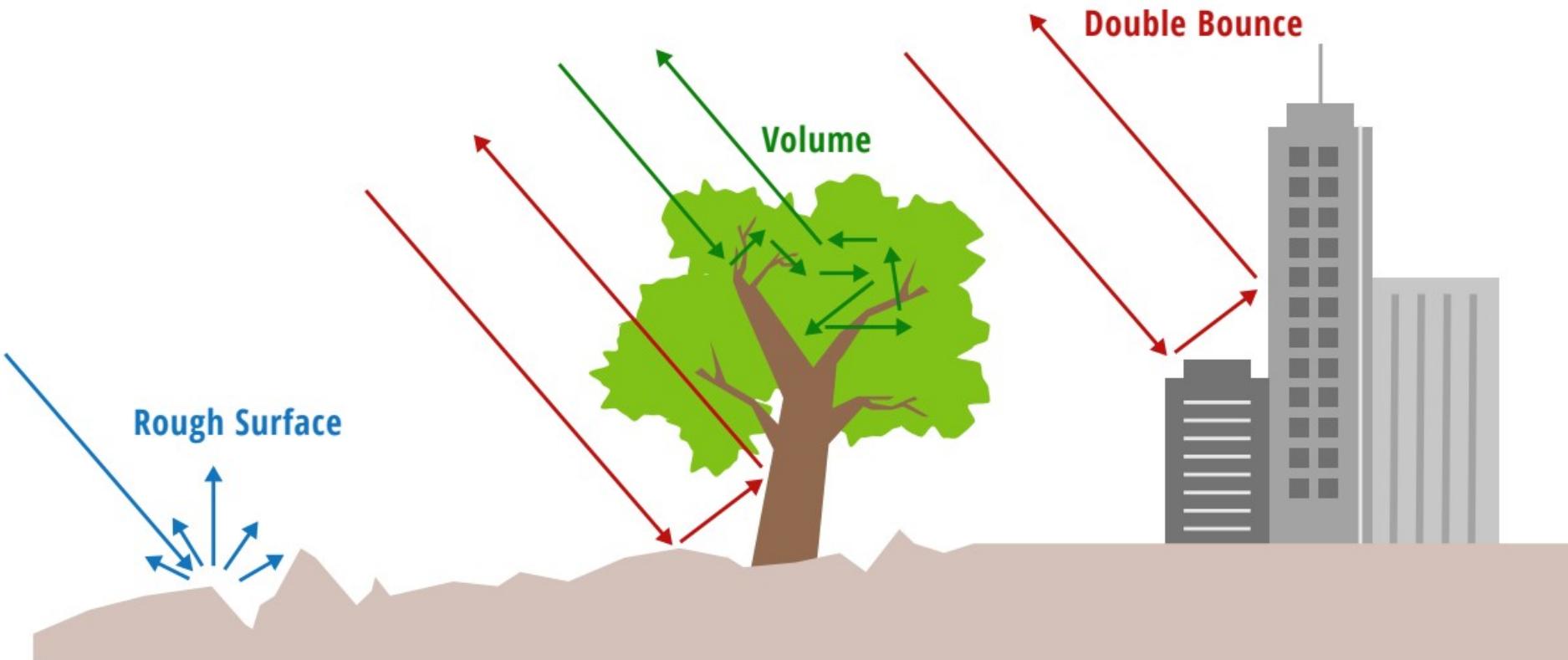


Image after Flores-Anderson, Africa Ixmuca, et al. "The SAR Handbook: Comprehensive Methodologies for Forest Monitoring and Biomass Estimation." (2019).



Each polarimetric channel prefers certain scattering types:

RELATIVE SCATTERING STRENGTH BY POLARIZATION:

Rough Surface Scattering $|S_W| > |S_{HH}| > |S_{HV}| \text{ or } |S_{VH}|$

Double Bounce Scattering $|S_{HH}| > |S_W| > |S_{HV}| \text{ or } |S_{VH}|$

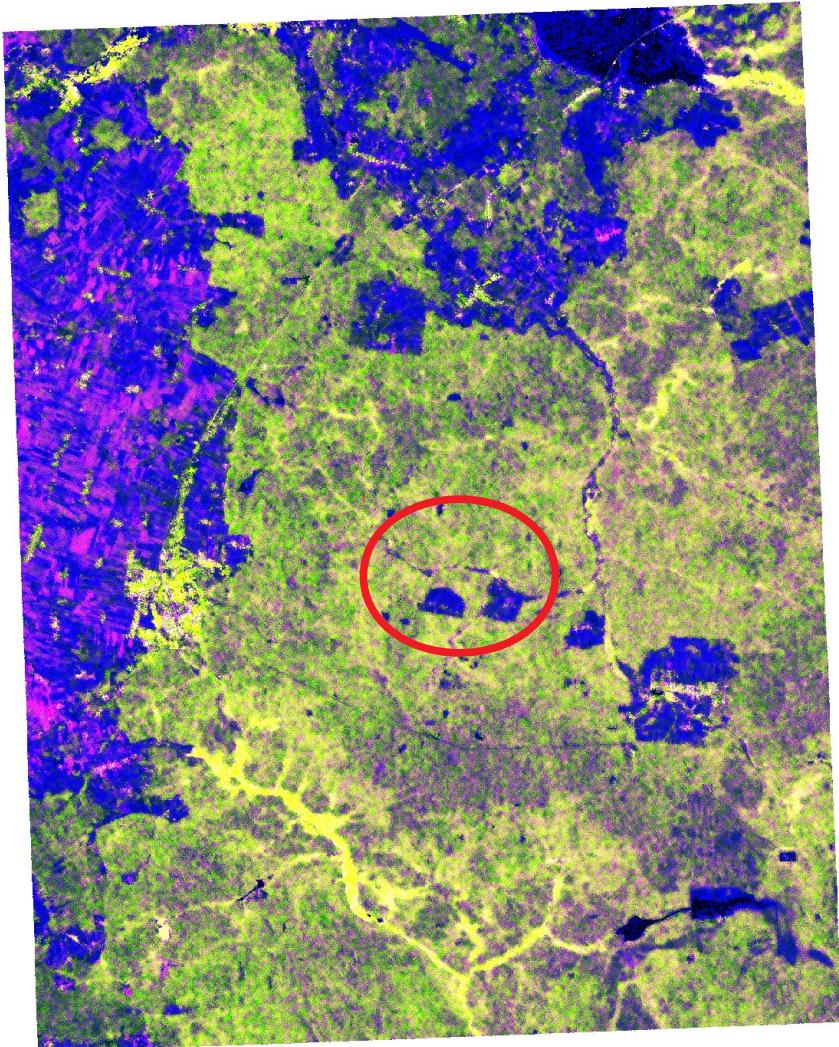
Volume Scattering Main source of $|S_{HV}|$ and $|S_{VH}|$

- Strong scattering in $|S_{HH}|$ indicates a predominance of **double-bounce scattering** (e.g., stemmy vegetation, manmade structures)
- Strong $|S_{VV}|$ relates to **rough surface scattering** (e.g., bare ground, water)
- Spatial variations in $|S_{HV}|$ indicate the distribution of **volume** scatters (e.g., vegetation, high-penetration soil types)



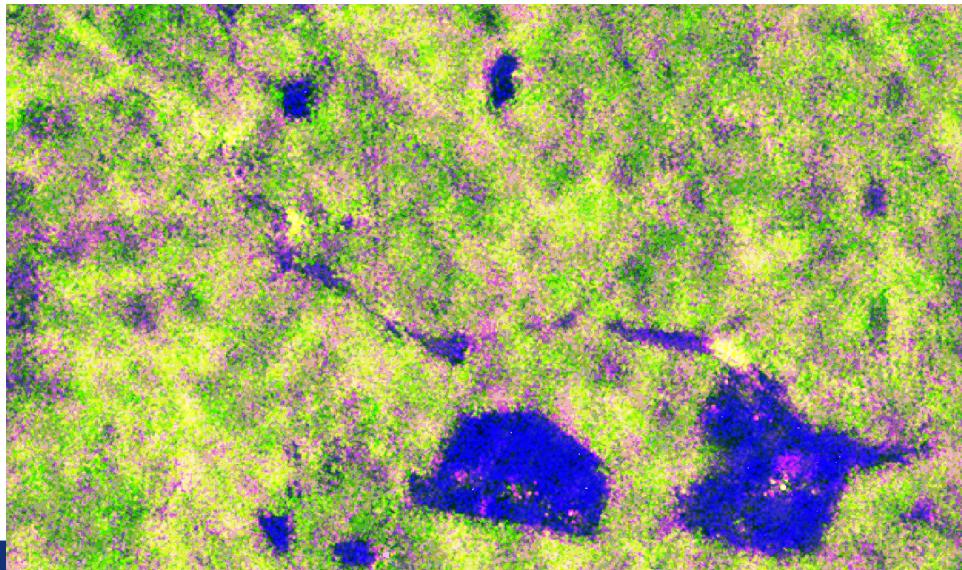
To enhance the visibility of differences between the channels, it is possible to create **false color composites** of SAR scenes, assigning RGB colors to different channels

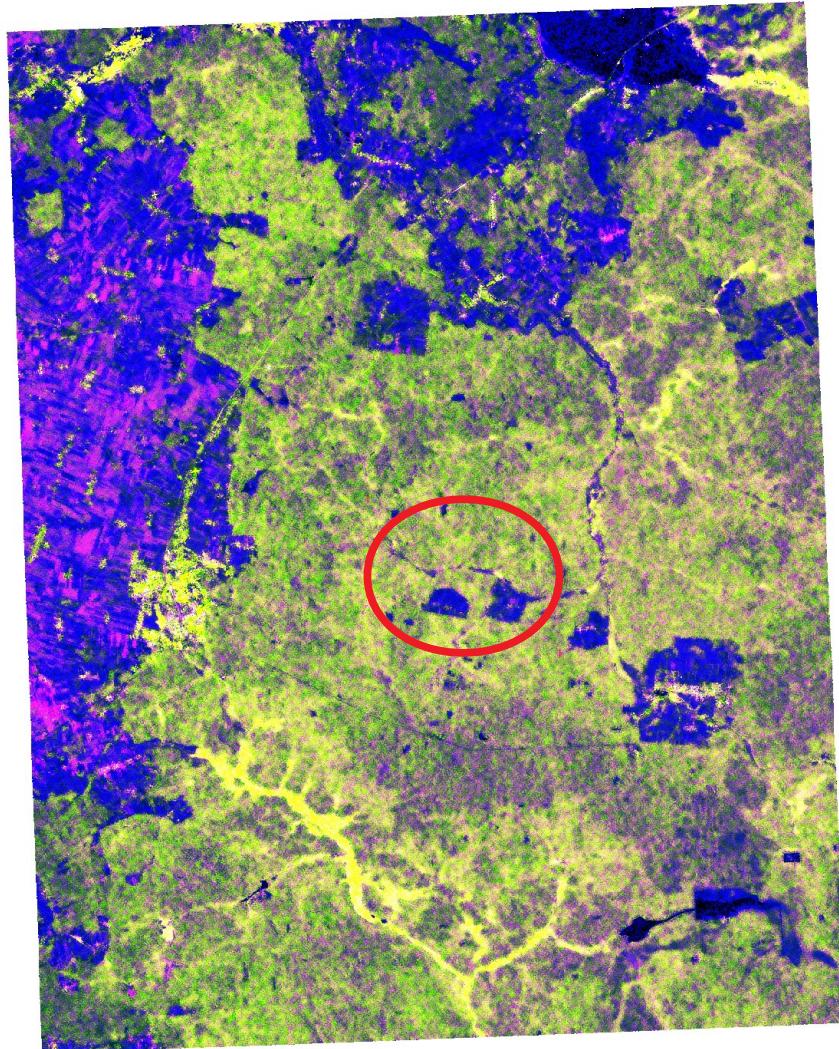
Sentinel-1 : VV and VH channels



An RGB color combination of C-band
SAR scenes from the Sentinel-1 sensor
over Białowieża on 14/11/2018

VV - red
VH - blue
VV-VH - green



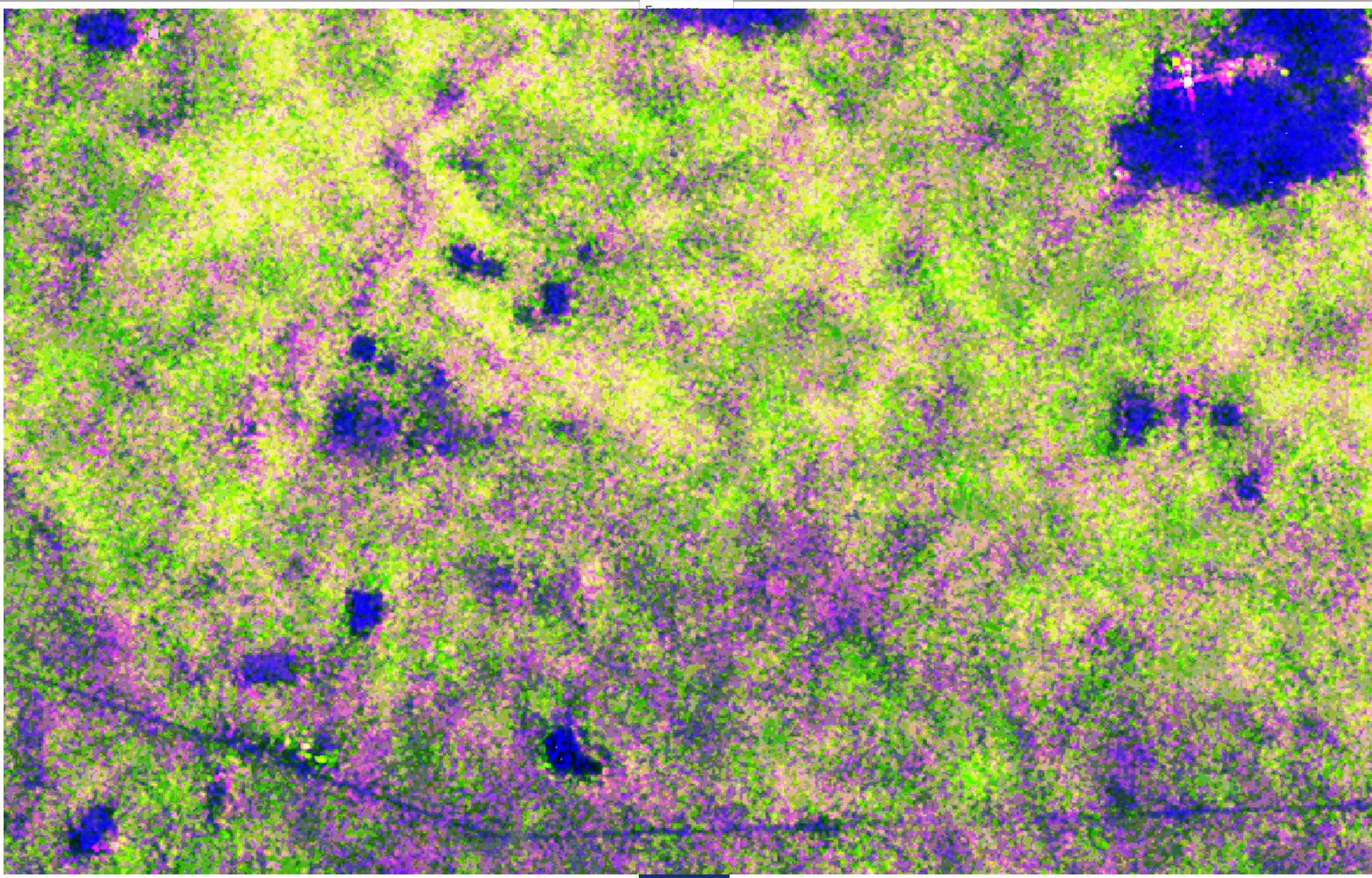


An RGB color combination of C-band
SAR scenes from the Sentinel-1 sensor
over Białowieża on 14/11/2018

VV - red
VH - blue
VV-VH - green









Using SAR data for mapping deforestation

Sensor parameters – wavelength / frequency, incidence angle, look directions, transmit and receive polarization – are typically **stable** from a satellite SAR, **backscatter variations over time can be attributed only to change in forest structure and moisture conditions.**

Complementary to optical data:

- (1) Not affected by **atmospheric** or sun **illumination** variations
- (2) Longer wavelengths and active penetration into forest canopies interact directly with **structure and moisture conditions**



Using SAR data for mapping deforestation

Response to deforestation

Change **from volume to surface scattering**. Cross-polarized backscatter decreases significantly.

Moisture conditions of soils become more visible. Can enhance signals at C-band and introduce ambiguities.

Getting the data

JEO-DPP

JEO-lab:

Filter over

- Study area
- Time

datemin: 2015-08-01

datemax: 2019-06-14

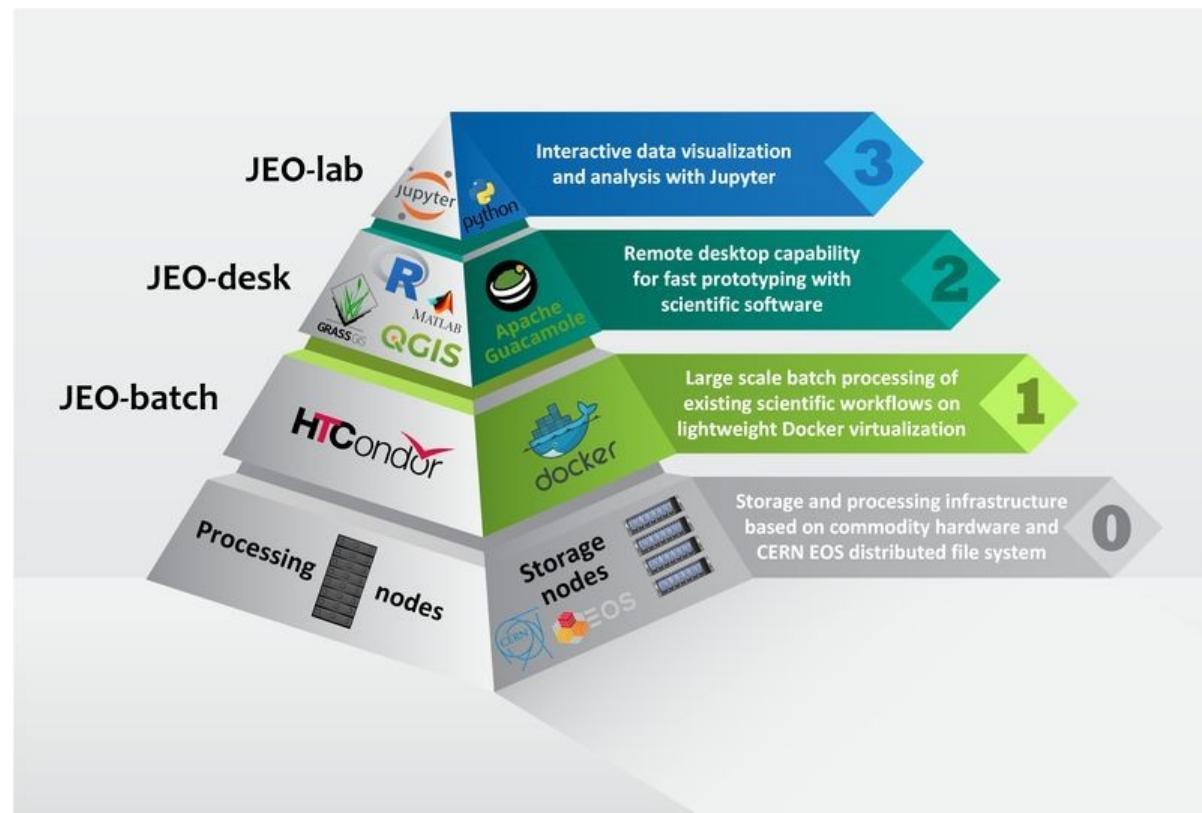
- Orbit

- Polarization

Preprocessing (geocoding)

JEO-desk:

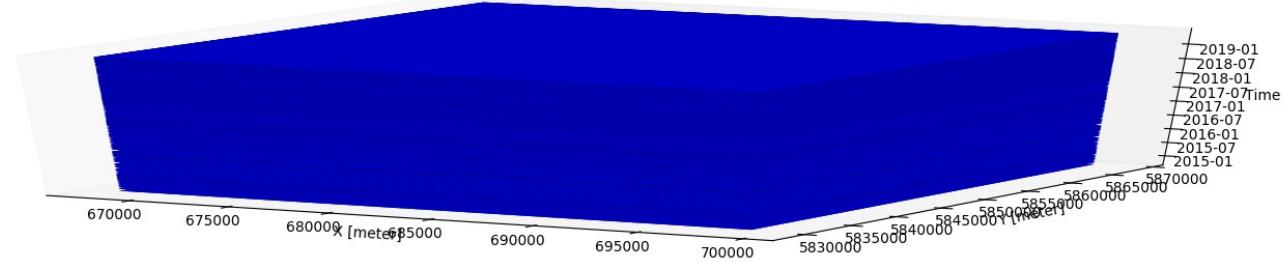
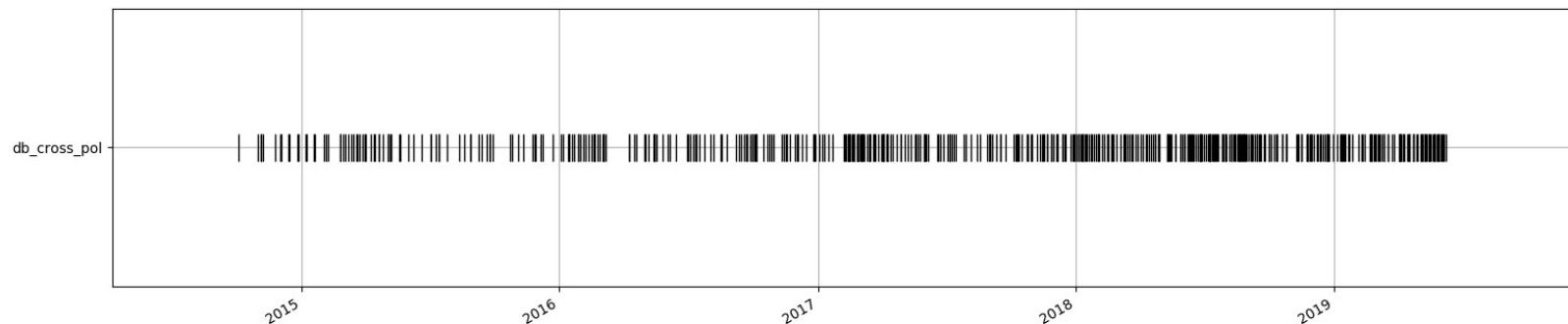
Crop the data to AOI





Creating time series data stack

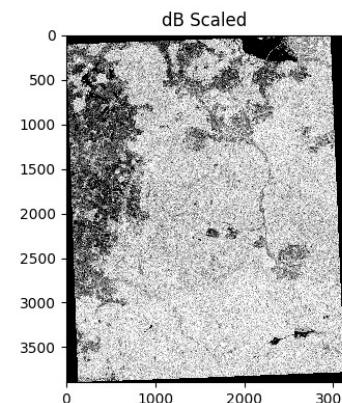
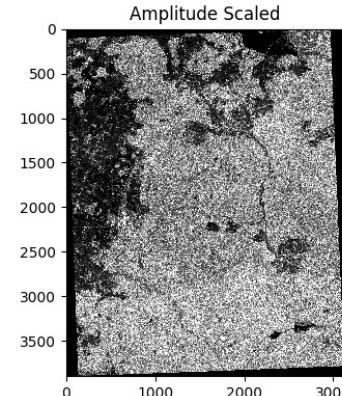
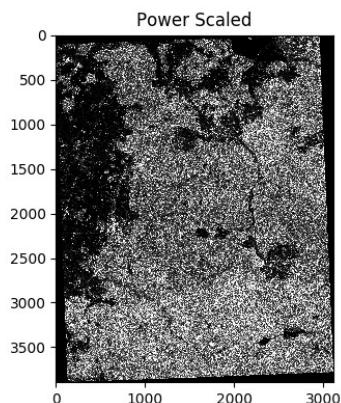
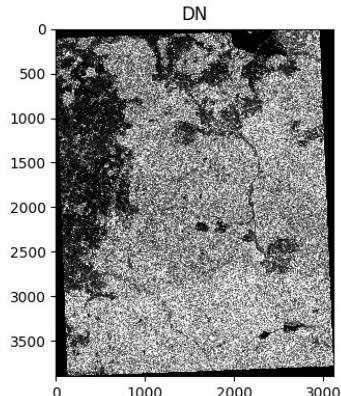
GRASS GIS: Space time raster dataset (strds)



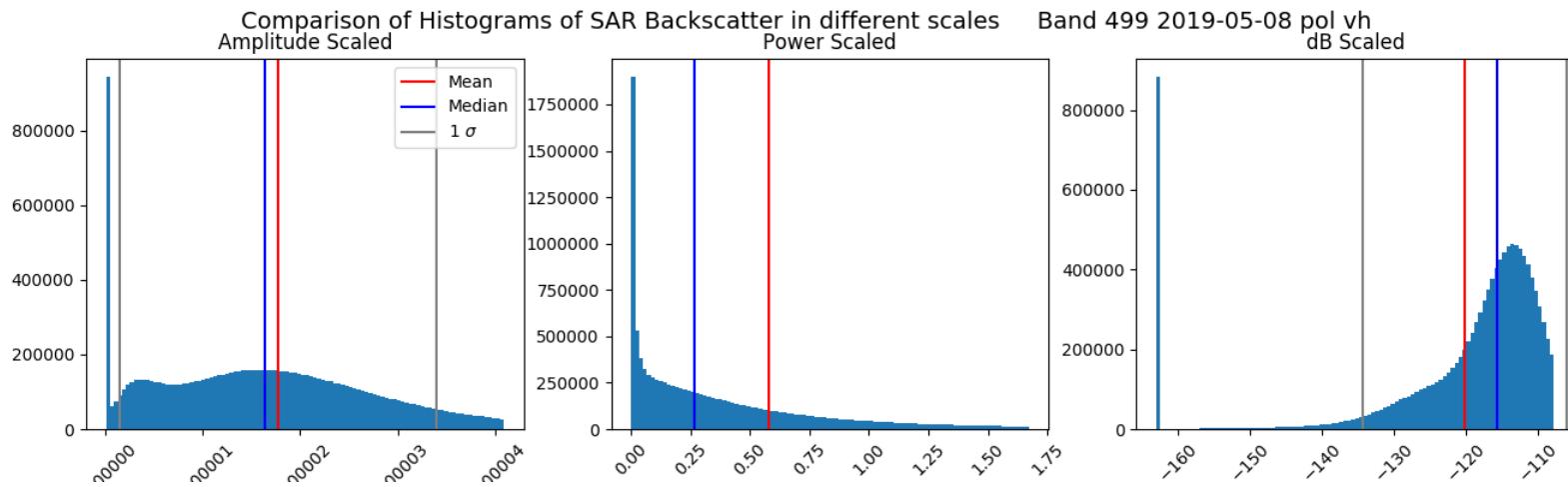


Scaling the data

Band 499 2019-05-08 pol vh



Scaling the data



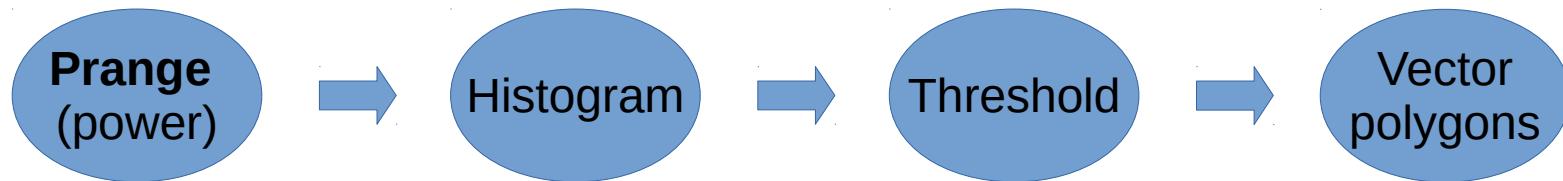


Computation of metrics for change detection

In the resulting map, each pixel value is a function of the values of the corresponding pixels in the time series

Considering the whole time series from 2015-08-01 to 2019-06-14

Mean, Median, Max, Min, Variance, Range (Max-Min), 5th Percentile, 95th Percentile, **Prange (95th - 5th Percentile)**, Coefficient of Variation (Variance/Mean)



Results



Base image WV3 10-01-2018

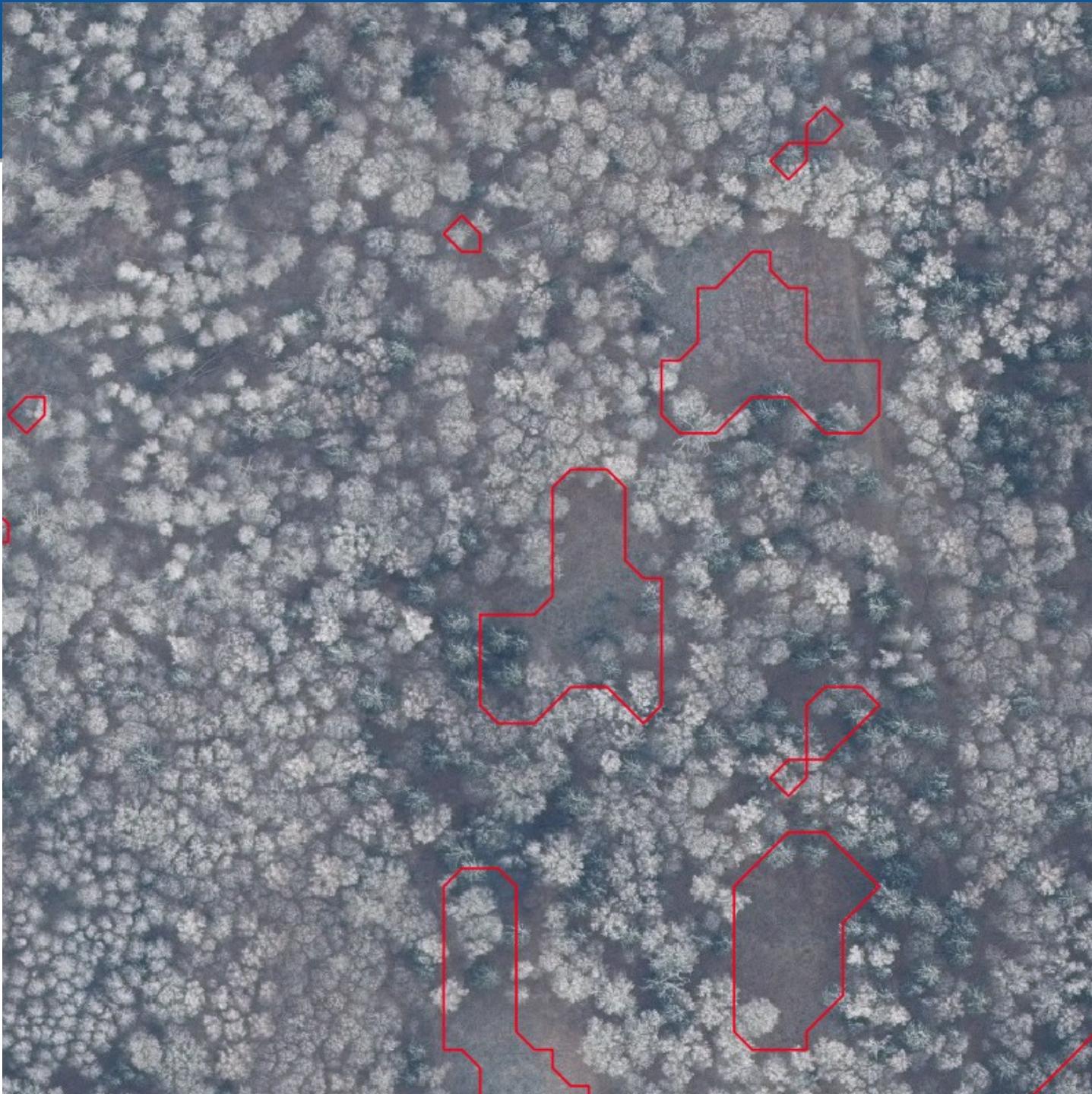


Results

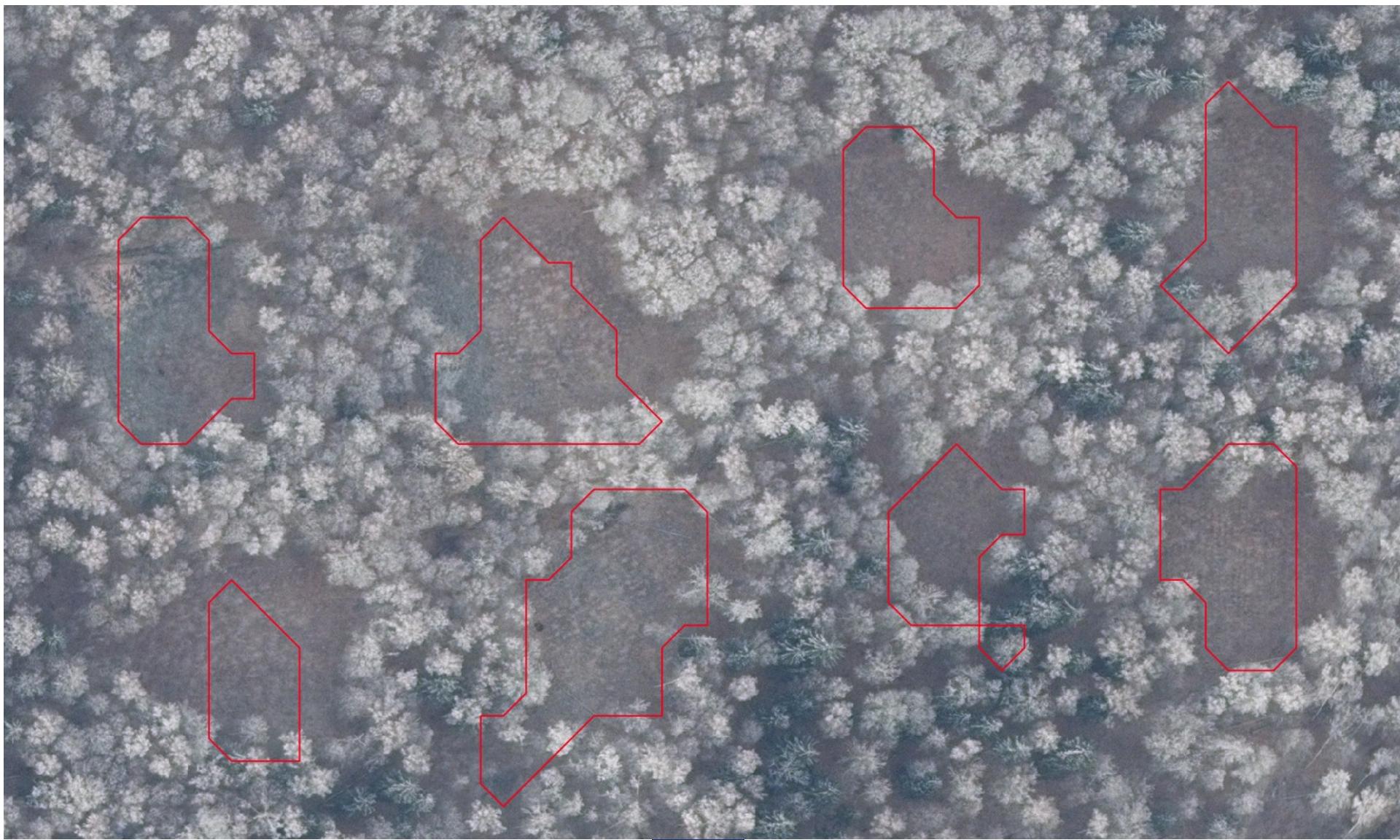


Base image WV3 31-08-2017





Results





18338
0.09 Ha

18373
0.7 Ha

Basemap WV3 2018-01-10



18338
0.09 Ha

18373
0.7 Ha

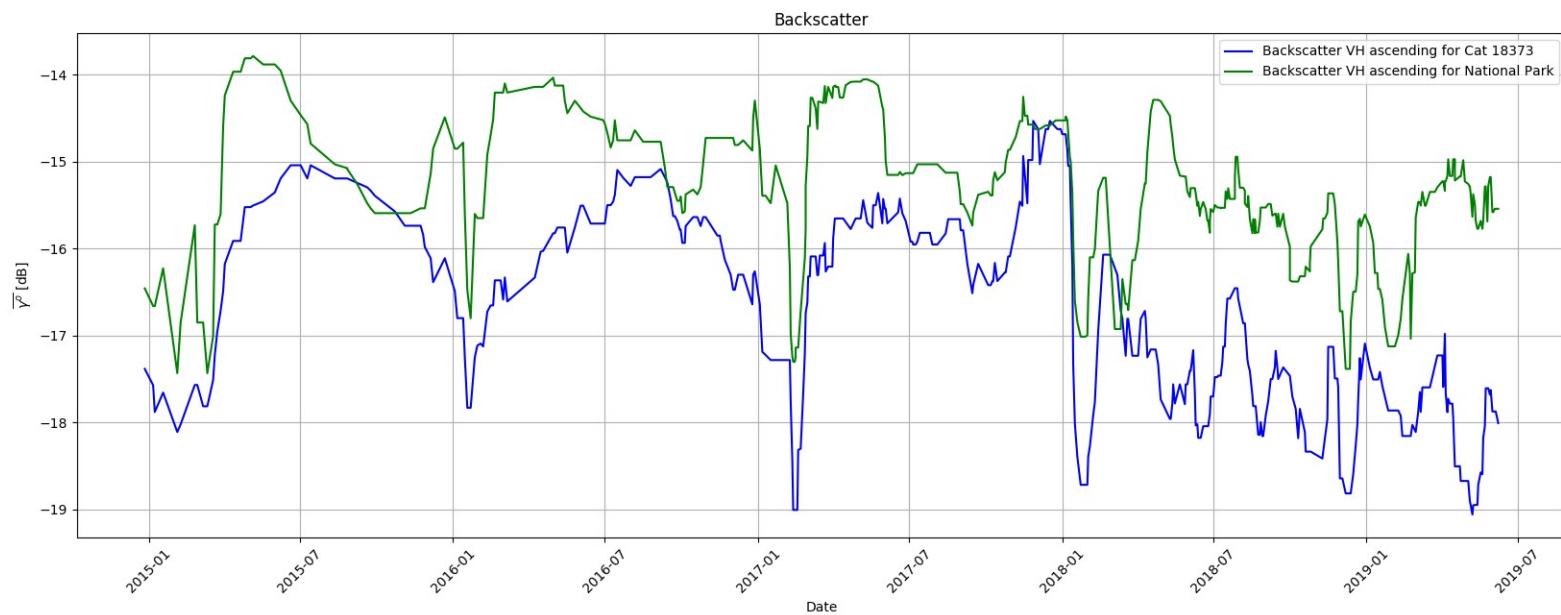
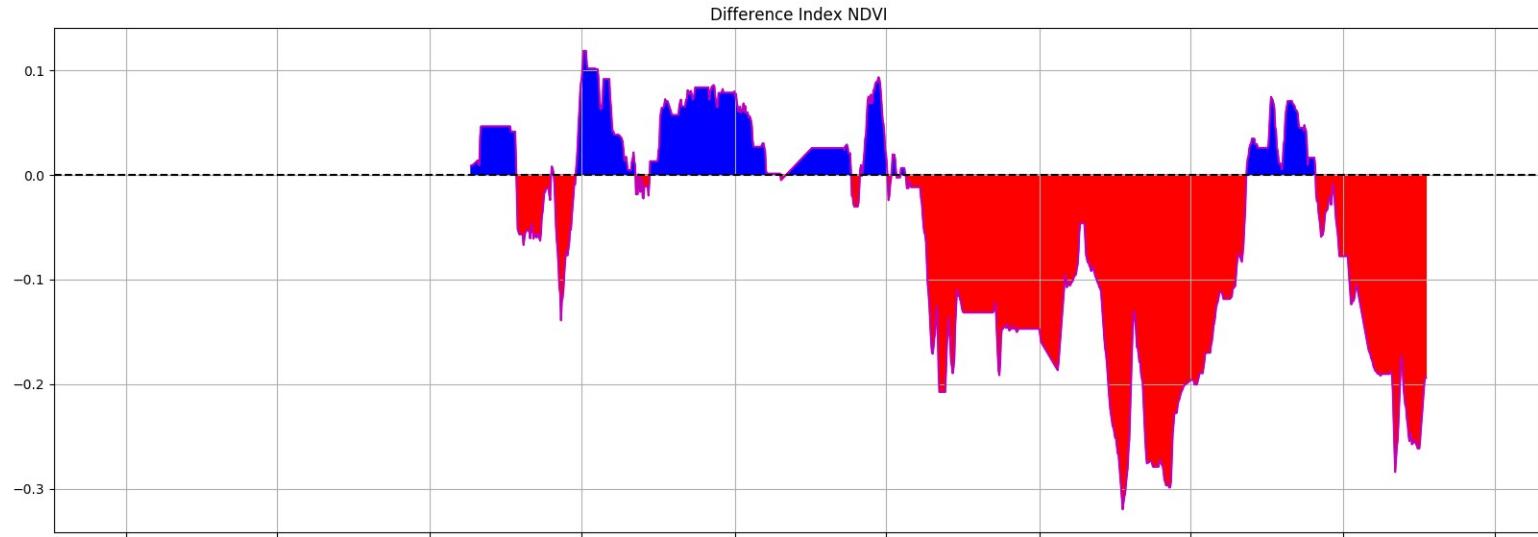
Basemap WV3 2017-08-31

18338
0.09 Ha

18373
0.7 Ha

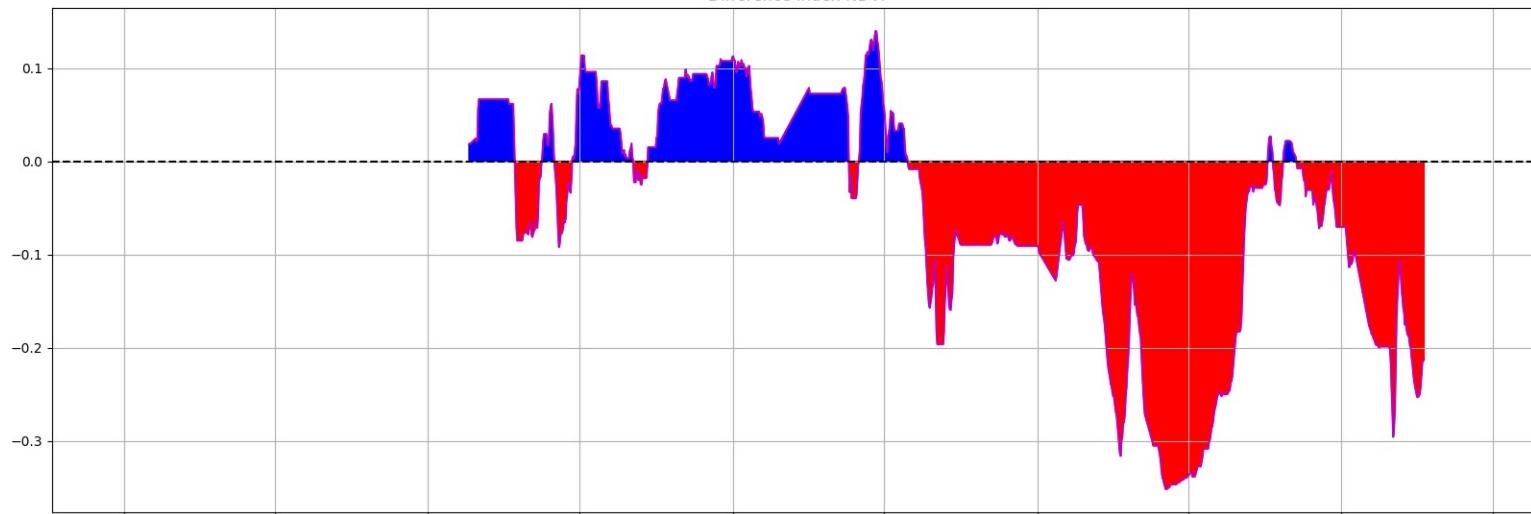
Basemap WV3 2017-06-27

Signal time series for Cat 18373

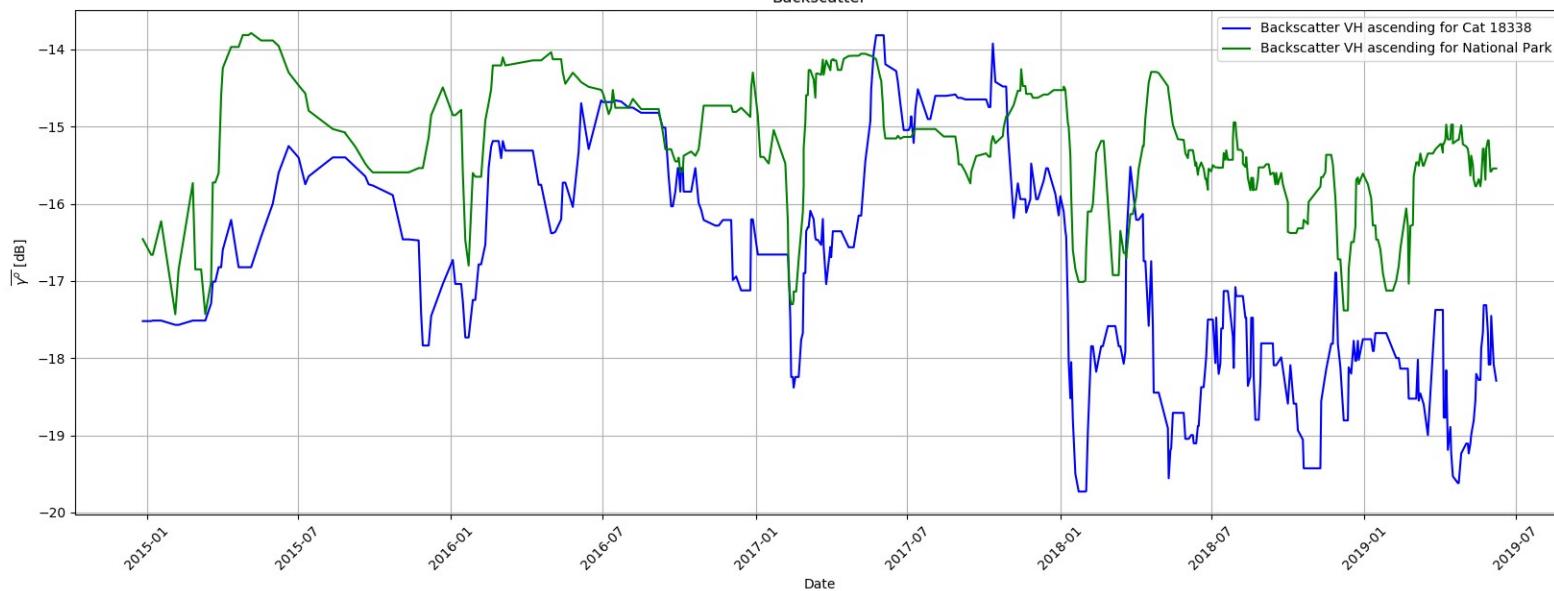


Signal time series for Cat 18338

Difference Index NDVI



Backscatter



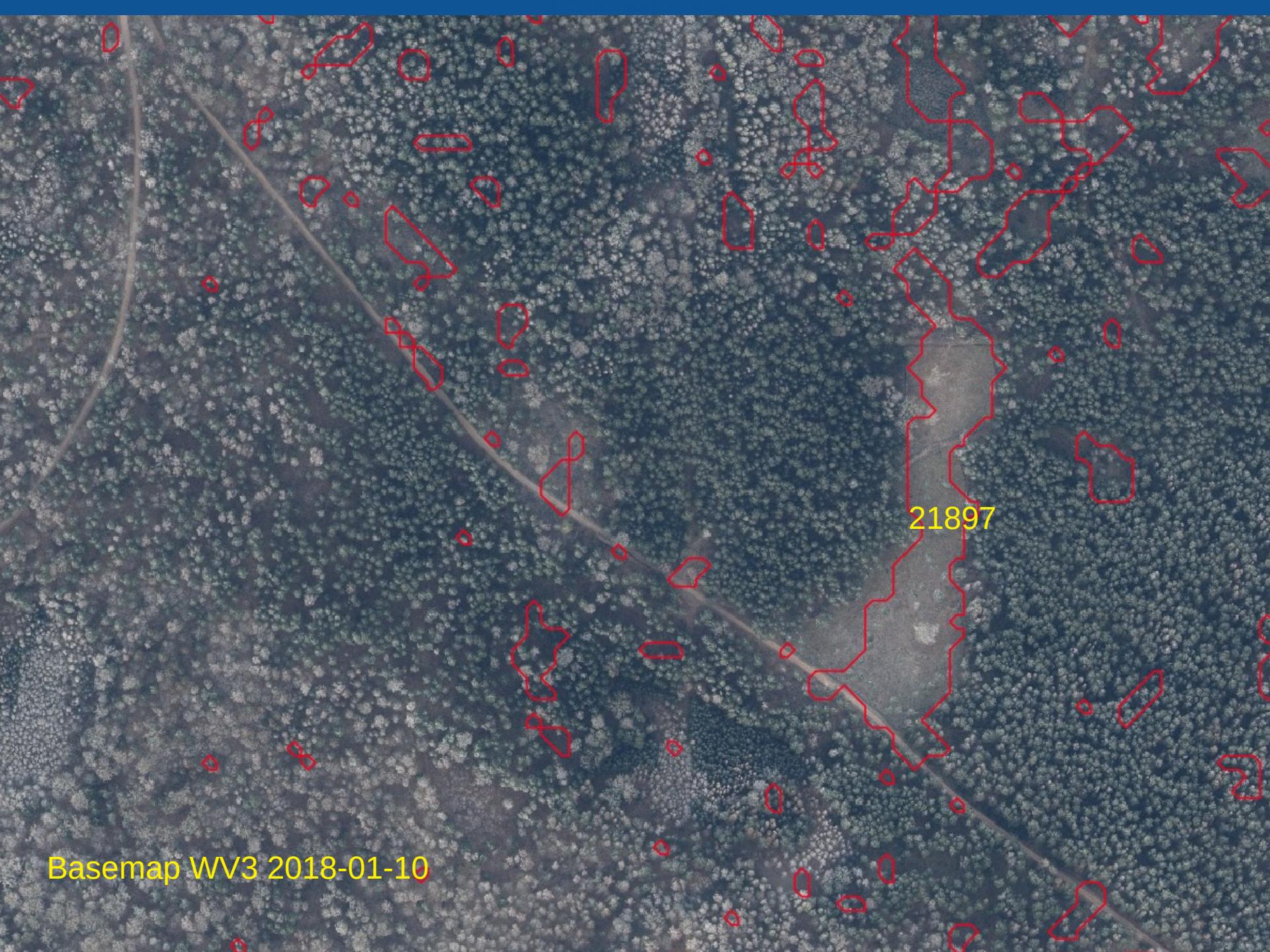
Conclusions



- False color composites of SAR scenes → snapshots in time of forest coverage
- Small scale changes in forest coverage → mapped using pixel-level statistics of time series S-1 imagery
- Zonal statistics of backscatter over polygons through time → insights of sudden changes in forest coverage such as clearcuts
- Zonal statistics of NDVI → add information regarding vegetation condition



Thank you!



21897

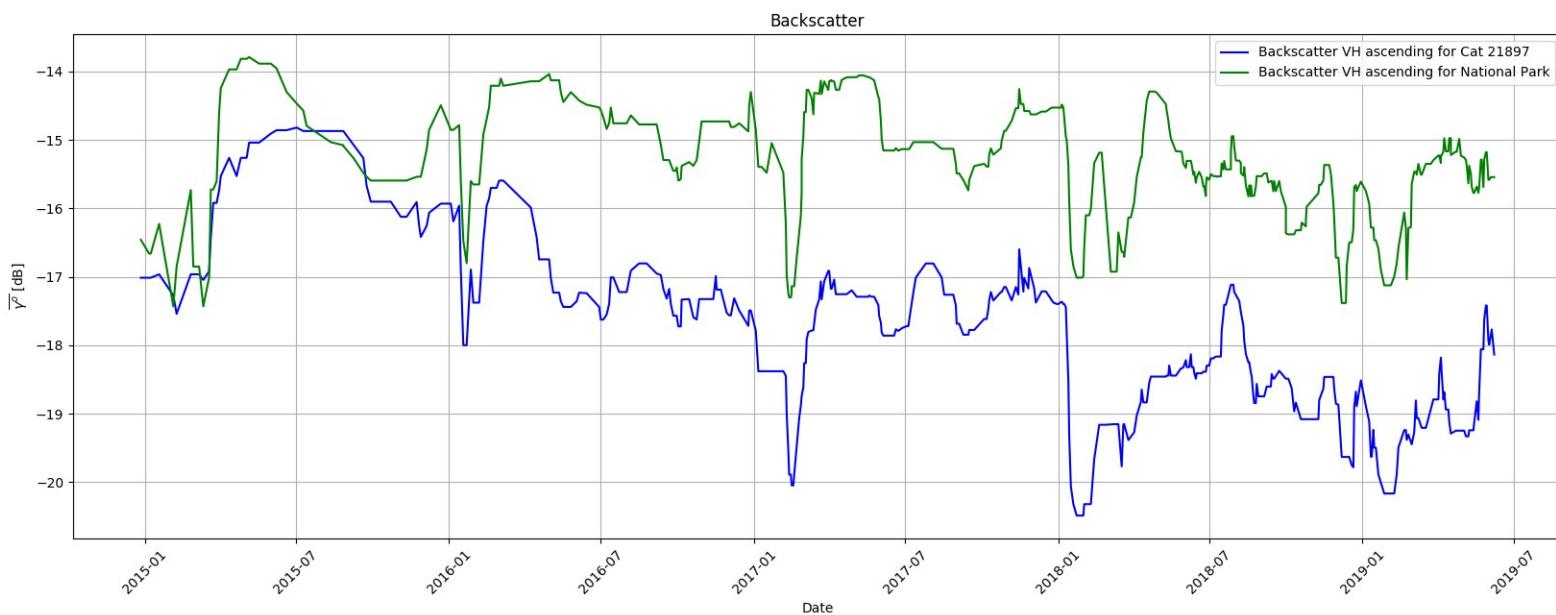
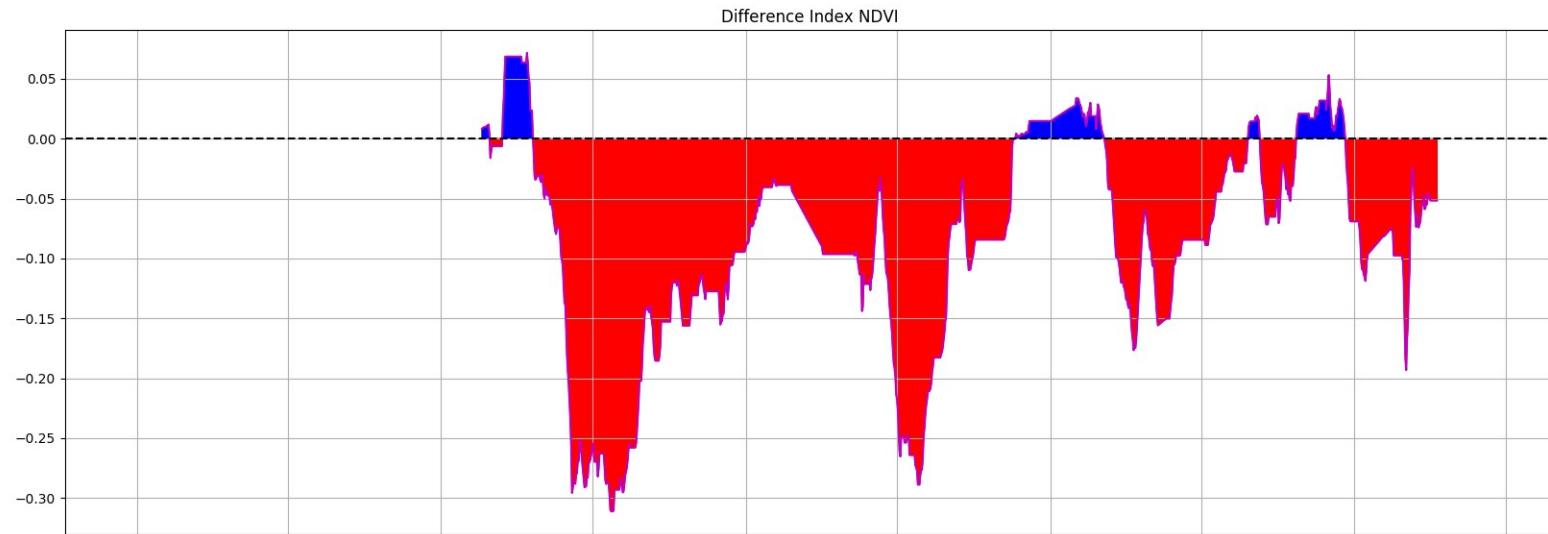
Basemap WV3 2018-01-10



21897

Basemap WV3 2017-06-27

Signal time series for Cat 21897





Upcoming space missions relevant for forest monitoring

NISAR (NASA-ISRO Synthetic Aperture Radar) : Launch 2021; L-band ($\lambda = 24.6\text{cm}$); Repeat cycle: 12 days; access: Free and Open

Biomass mission (ESA) : Launch 2021; P-band ($\lambda = 70.0\text{cm}$); Repeat cycle: 17 days; access: Free and Open

GEDI mission (NASA) (**LiDAR**) : Launch 2018



Forest height and vertical structure; habitat quality & biodiversity;
Forest carbon sinks & source areas; loss of carbon from extreme events
such as fires and hurricanes; parameterization of ecosystem models

Forest
Management &
Carbon Cycling

Canopy 3D structure that influences snowmelt, evapotranspiration,
canopy interception of precipitation. Glacier surface elevation change;
lake & river stage; snowpack elevation; coastal tides.

Water
Resources

Improved canopy aerodynamic profiles to parameterize weather
prediction models. Canopy and biomass products that initialize and
constrain climate models; impacts of land use change on climate

Weather
Prediction

Accurate bare earth and under canopy topographic elevations for
improved digital elevation models from radar. Calibration of satellite
based observations of surface deformation and earthquakes

Topography &
Surface
Deformation

GEDI'S CANOPY AND SURFACE 3D MEASUREMENTS ADDRESS KEY CHALLENGES IN A VARIETY OF SCIENTIFIC AREAS.

Source: <https://gedi.umd.edu/applications/application-overview/>