

# Exploring environmental variables based on ecotopes derived by remote sensing

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## Landscape delineation strategy impacts the performance of habitat suitability models

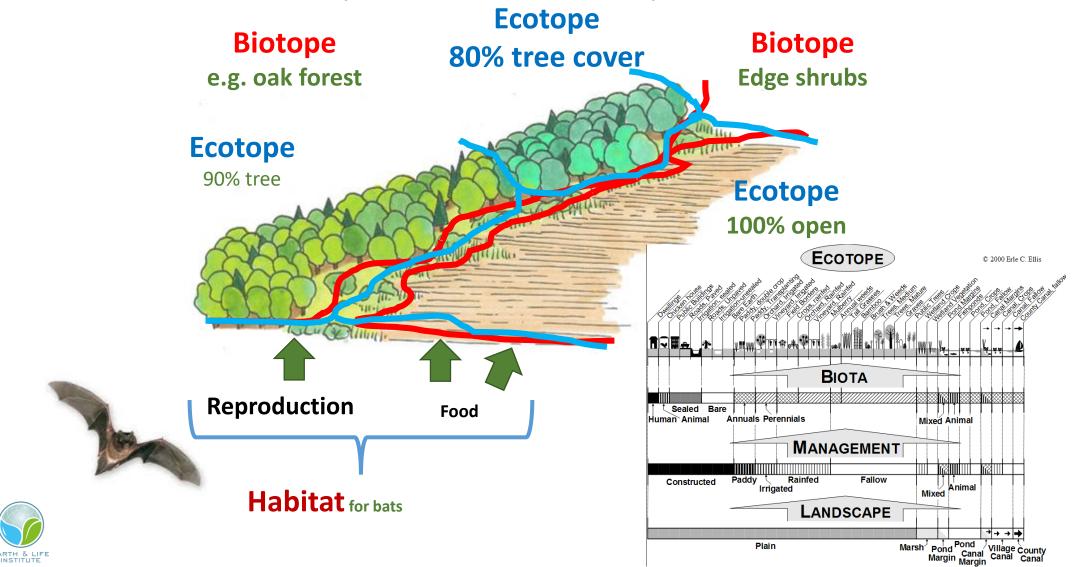
- Grid-based models have many advantages
  - Link with systematic inventory data
  - Possible hierarchical agregation
  - Easy to build
- ...but landscapes are not squares (well, not always)
  - Irregular polygons better match biotopes
  - GNSS-based observations are precisely located
  - Hierarchical agregation is less obvious







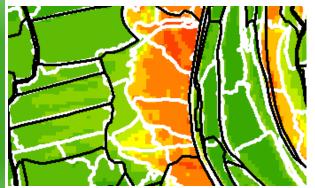
# Different partitions of landscape in ecology: Habitat, Biotope and Ecotope





### Automated delineation based on land cover and topography

Orthophotos + Hillshade + Height



Black lines: NO use of topography White lines : use of topography (steep slopes in orange and red)

Impact of delineation on AUC for ravine forest biotope

Weight of topography	0	0,5	1	2
RF	55	71	79	72
GAM	79	81	90	90



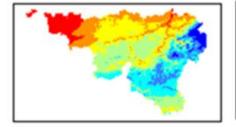




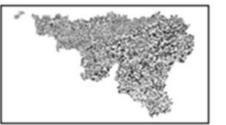


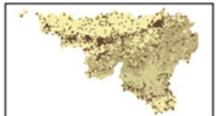
### More than a million polygons, enriched with GIS data

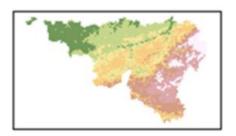
- Land cover ← Image analysis
- Climate ← Worldclim
- Topography ← LIDAR
- Soils ← soil map
- Light pollution ← DMSP
- Phenology ← SPOT, MODIS, PROBA
- Context (derived from land cover)
  - Distance to water, roads, forest, buldings...
  - Proportion within 250 et 500m radius



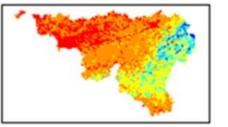


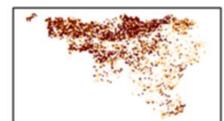


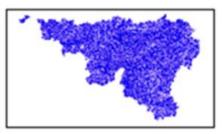


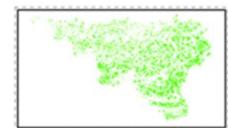












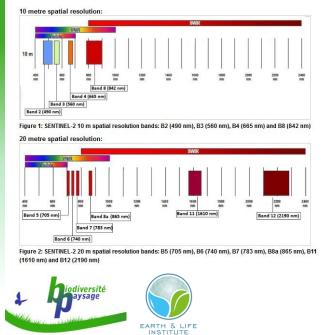




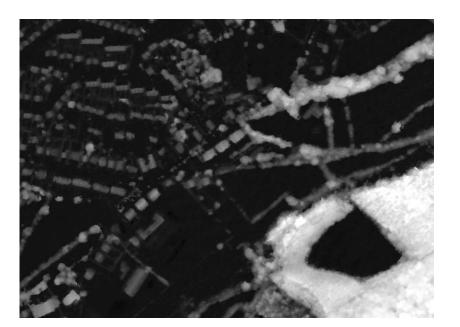
For accurate land cover,

combining RS data is the key

- Sentinel-1 and 2 time series: high temporal and spectral resolution
- Orthophotos: very high spatial resolution (25 cm)
- Height and elevation from LIDAR and photogrammetry



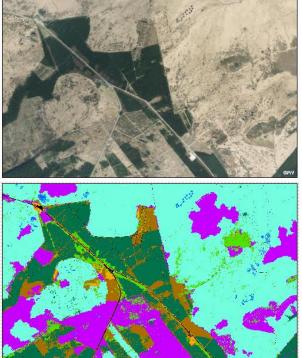


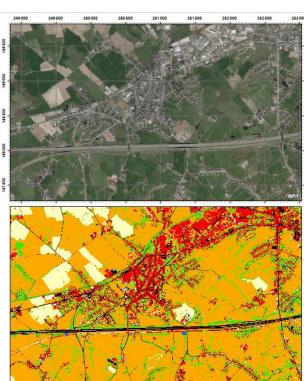


### 2m pixel-based classification as a source of land cover

- RS image analysis consolidated in low confidence areas
  - Ancillary data and photointerpretation still needed for specific cases
- Final product with 94,5 +/- 1,2 overall accuracy with 11 classes











### So how to represent a 90+ quantitative variables database?

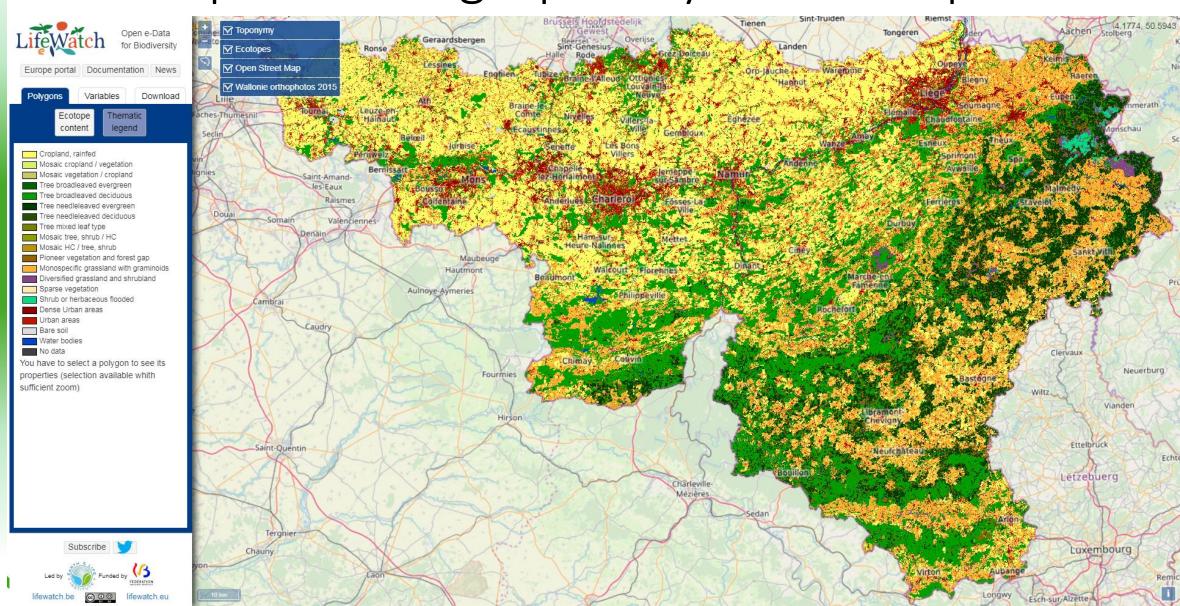
- Selecting a legend to create categories
  - Advantage : easy to conceptualize
  - Disadvantage: frozen definition
- Graduated colours for quantitative variables
  - Advantage: a lot of information
  - Disadvantage: only one variable at the time
- Advanced tools to get more...
  - Advantage: more flexibility to users
  - Disadvantage: more difficult to use
- For even more: download database or extract by points







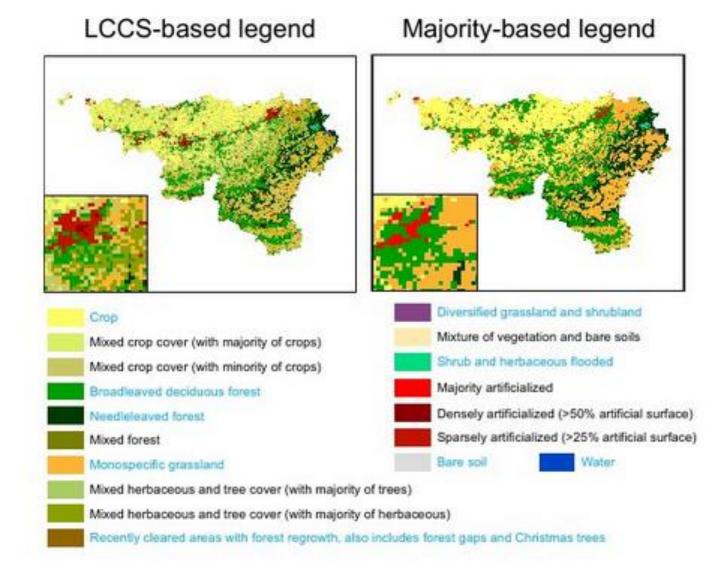
#### Web portal using OpenLayer and MapServer





#### Inside the ecotopes: categorical legend

- Standard classification systems
  - LCCS
  - Majority based
- Choice based on the ESA land cover CCI legend
  - Lower threshold on urban areas
  - Larger threshold for forests
  - New class for diversified grasslands

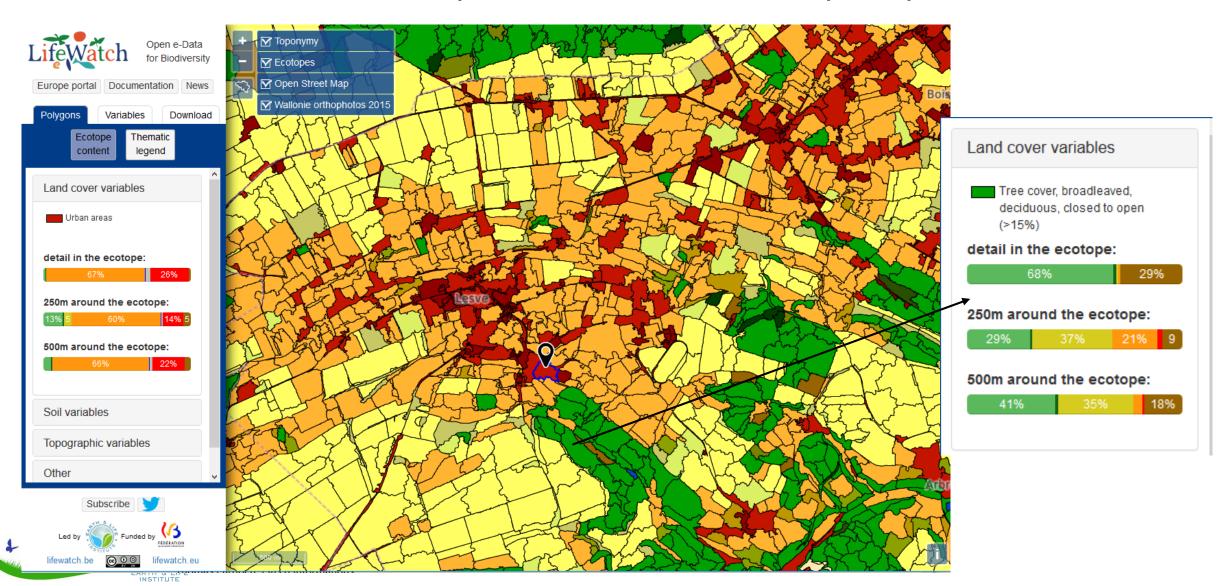








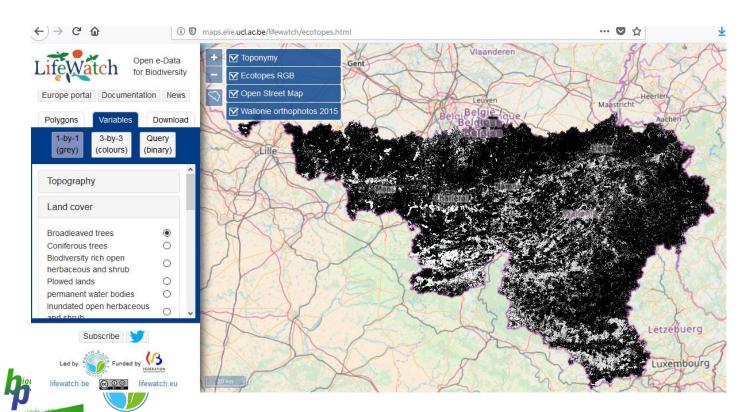
#### Inside the ecotopes: land cover proportions





#### Inside the ecotopes: single variables

- Intensity proportional to the value
- Values need to be rescaled between 0 and 255 (8bit, screen colours)



White = 100% broadleaved

Black = 0 % broadleaved

Grey = in between



#### Inside the ecotopes: color composite

• Human eyes has a spectral resolution of 3 colours: Red, Green and Blue



Red + Green = Yellow

Red + Blue = Magenta

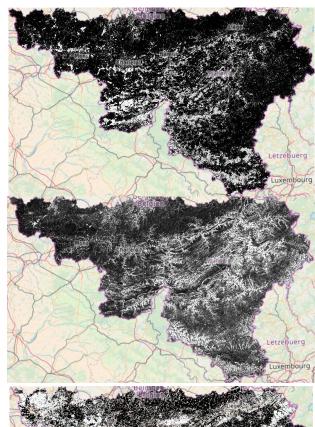
Blue + Green = Cyan

Red + Green + Blue = White

No light = Black

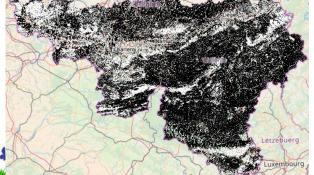


### Inside the ecotopes: color composite

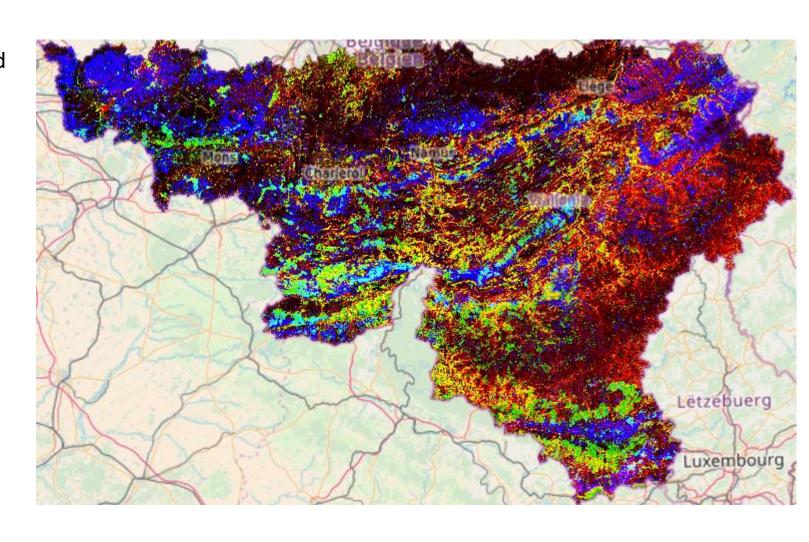


Broadleaved proportion (Green)

Slope percentage (Red)



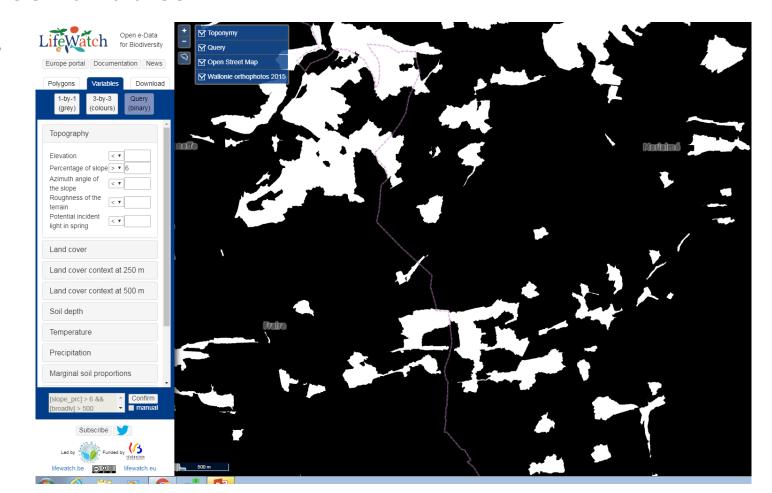
Imperfect drainage (Blue)





#### Inside the ecotopes: queries

- For more than three variables
- Binary thresholds

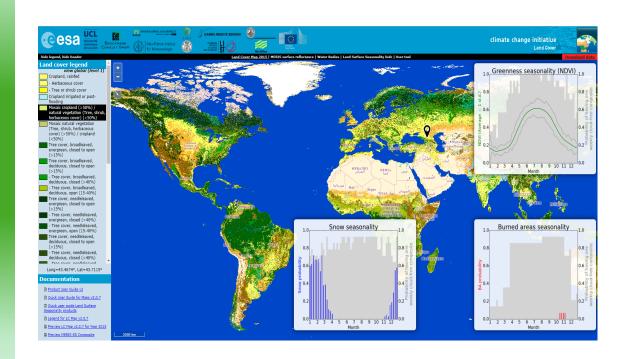


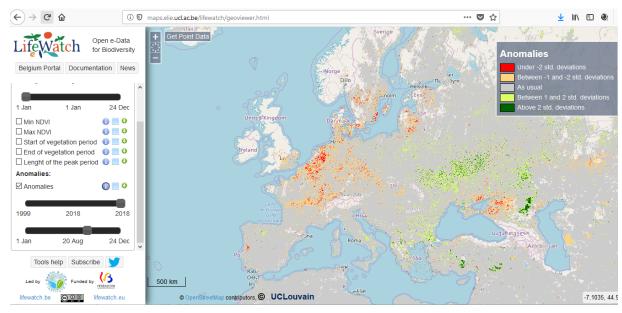






### maps.elie.ucl.ac.be/lifewatch





also extreme event for Europe...

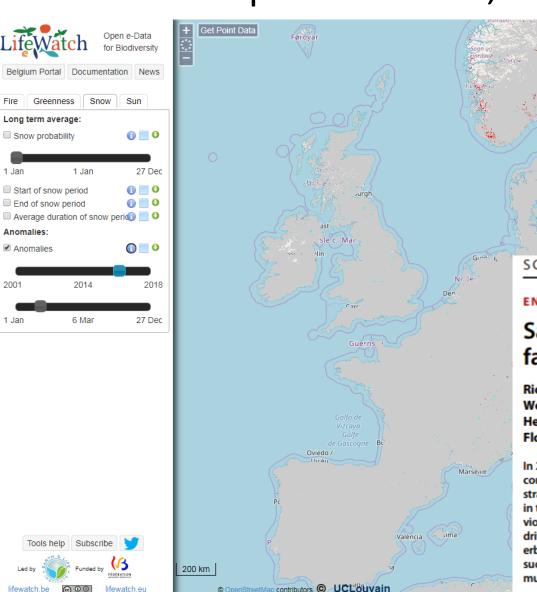
...and if you need global time series: maps.elie.ucl.ac.be/CCI







#### At European level, statistics on the phenology



Anomalies

Exceptional presence
Unusual presence
As usual
Unusual absence
Exceptional absence
Exceptional absence

#### **ENVIRONMENTAL STUDIES**

#### Saigas on the brink: Multidisciplinary analysis of the factors influencing mass mortality events

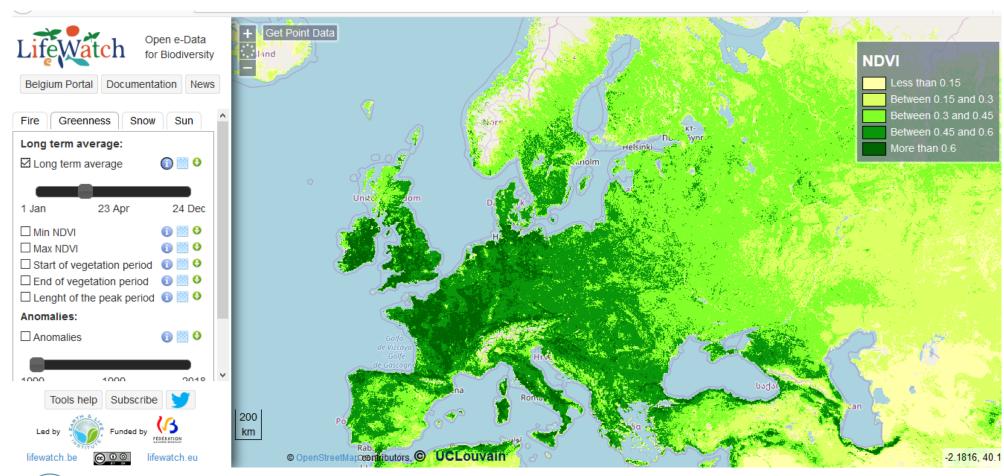
Richard A. Kock, <sup>1\*</sup> Mukhit Orynbayev, <sup>2</sup> Sarah Robinson, <sup>3</sup> Steffen Zuther, <sup>4,5</sup> Navinder J. Singh, <sup>6</sup> Wendy Beauvais, <sup>1†</sup> Eric R. Morgan, <sup>7</sup> Aslan Kerimbayev, <sup>2</sup> Sergei Khomenko, <sup>8</sup> Henny M. Martineau, <sup>1</sup> Rashida Rystaeva, <sup>2</sup> Zamira Omarova, <sup>2</sup> Sara Wolfs, <sup>1</sup> Florent Hawotte, <sup>9</sup> Julien Radoux, <sup>9</sup> Eleanor J. Milner-Gulland <sup>3</sup>

In 2015, more than 200,000 saiga antelopes died in 3 weeks in central Kazakhstan. The proximate cause of death is confirmed as hemorrhagic septicemia caused by the bacterium *Pasteurella multocida* type B, based on multiple strands of evidence. Statistical modeling suggests that there was unusually high relative humidity and temperature in the days leading up to the mortality event; temperature and humidity anomalies were also observed in two previous similar events in the same region. The modeled influence of environmental covariates is consistent with known drivers of hemorrhagic septicemia. Given the saiga population's vulnerability to mass mortality and the likely exacerbation of climate-related and environmental stressors in the future, management of risks to population viability such as poaching and viral livestock disease is urgently needed, as well as robust ongoing veterinary surveillance. A multidisciplinary approach is needed to research mass mortality events under rapid environmental change.

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#### Looking at time series: Long term average

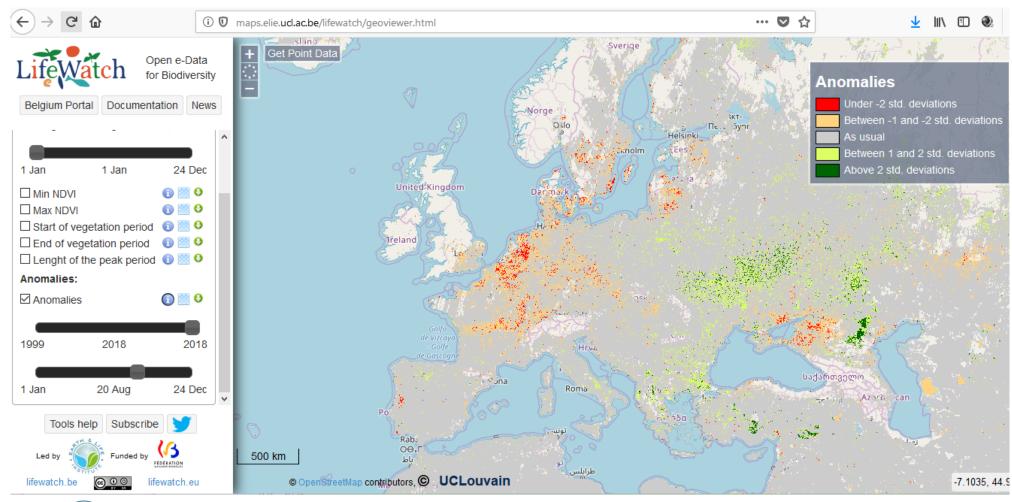








#### Looking at time series: Anomalies

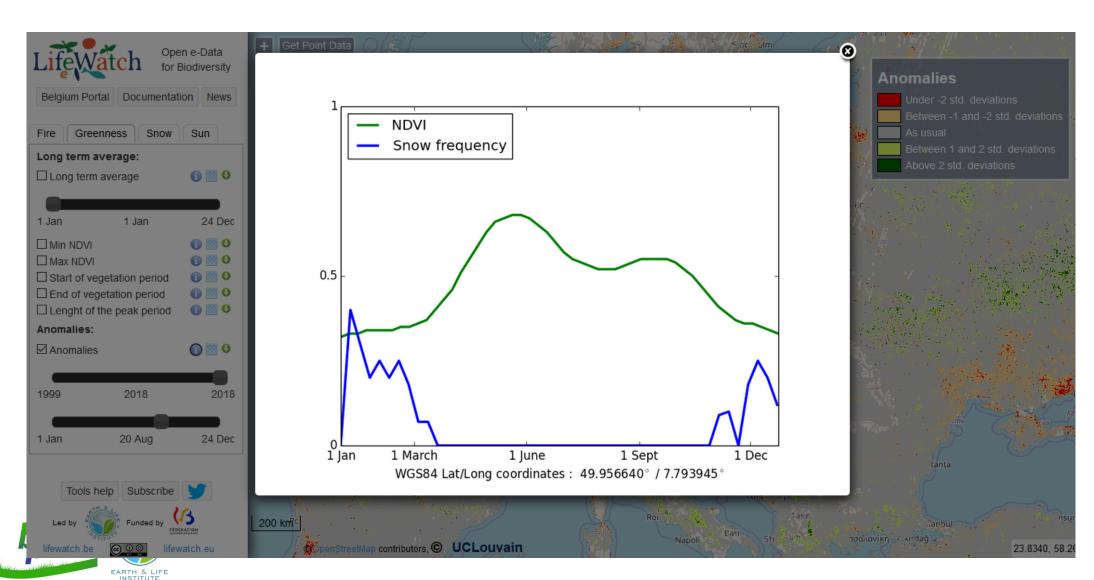








#### Looking at time series: temporal signature













Cropland, rainfed

- Herbaceous cover

- Tree or shrub cover

Cropland irrigated or postflooding

Mosaic cropland (>50%) / natural vegetation (Tree, shrub, herbaceous cover) (<50%)

Mosaic natural vegetation (Tree, shrub, herbaceous cover) (>50%) / cropland (<50%)

Tree cover, broadleaved, evergreen, closed to open (>15%)

Tree cover, broadleaved, deciduous, closed to open (>15%)

- Tree cover, broadleaved, deciduous, closed (>40%)

- Tree cover, broadleaved, deciduous, open (15-40%)

Tree cover, needleleaved, evergreen, closed to open (>15%)

- Tree cover, needleleaved, evergreen, closed (>40%)

- Tree cover, needleleaved, evergreen, open (15-40%)

Tree cover, needleleaved, deciduous, closed to open (>15%)

- Tree cover, needleleaved, deciduous, closed (>40%) Trop cover poodleleswed

Long=43.4674°, Lat=43.7115°

#### **Documentation**

Product User Guide v2

A Quick User Guide for Maps v2.0.7

Quick user guide Land Surface Seasonality products

A Legend for LC Map v2.0.7

Preview LC Map v2.0.7 for Year 2015

Preview MERIS SR Composite

