

Image Analysis and Machine Learning for Monitoring Climate Change

Sebastiano B. Serpico and Gabriele Moser



Synoptic and time-repetitive view at the global, regional, and local scales

A **precious** source of data to monitor and mitigate the effects of **climate change**

But also a **huge amount of highly complex data**

To extract useful and quantitative information, **computer-based algorithms are necessary**



(d) 10 m

(e) 2 m

(f) 5 cm (da drone)

climate change initiative

→ HIGH RESOLUTION LAND COVER



- Funded by the **European Space Agency**
- **Mapping land cover at subcontinental scale and its evolution along the past decades**, through satellite image analysis, to support climate change studies
- The project involves **research labs** in remote sensing and climatology as well as **private companies**



high resolution
land cover
cci



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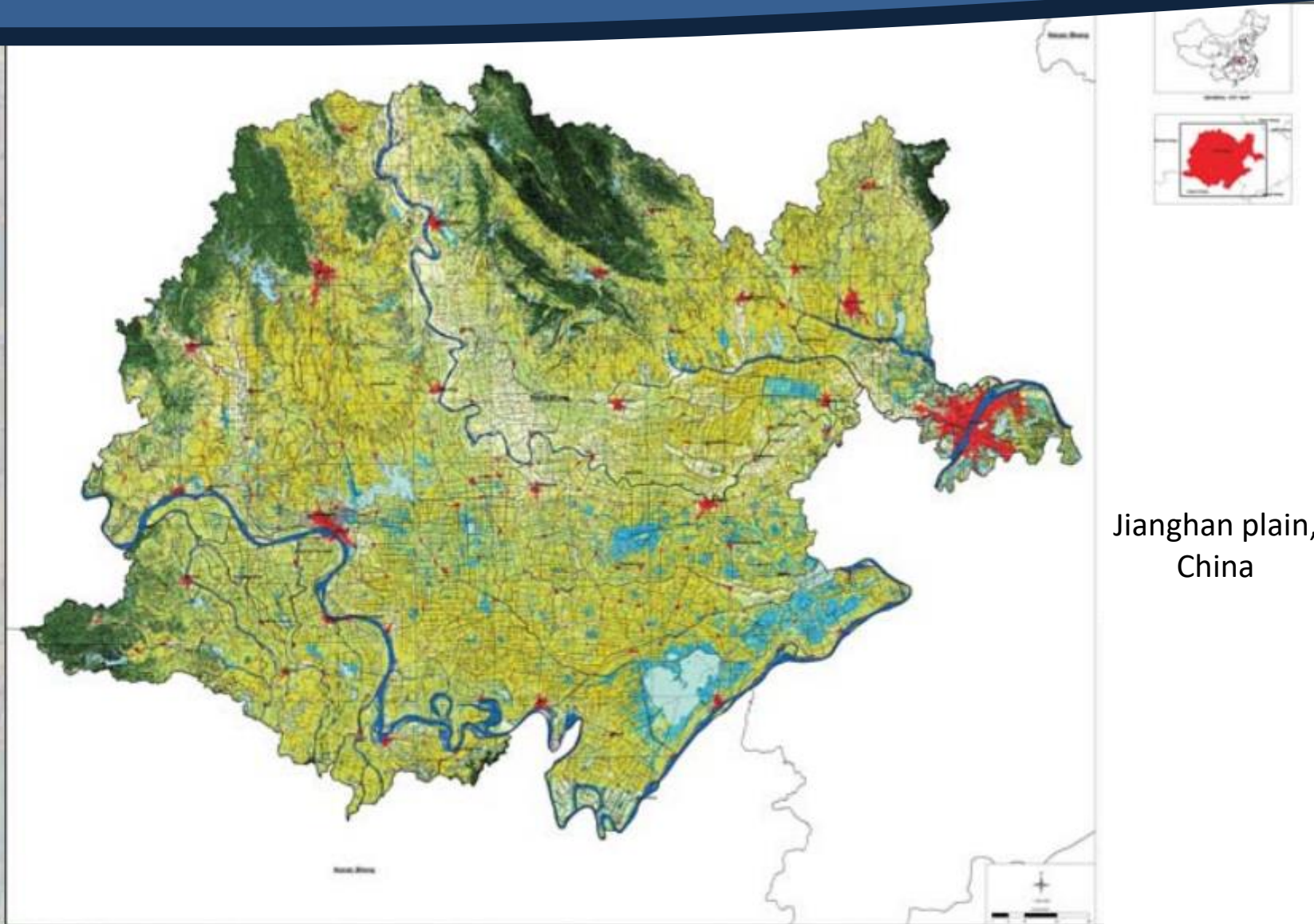


POLITECNICO
MILANO 1863





Land Cover from a Satellite



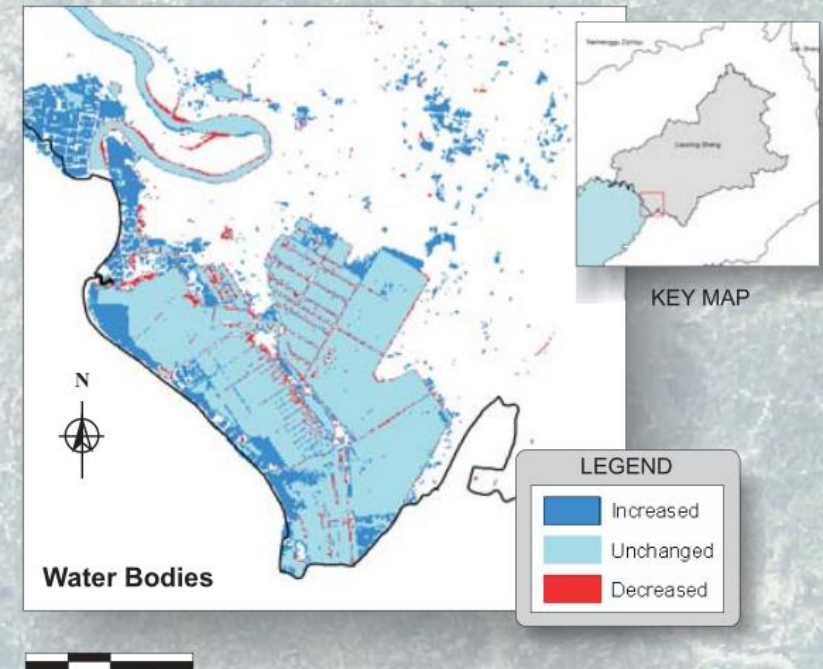
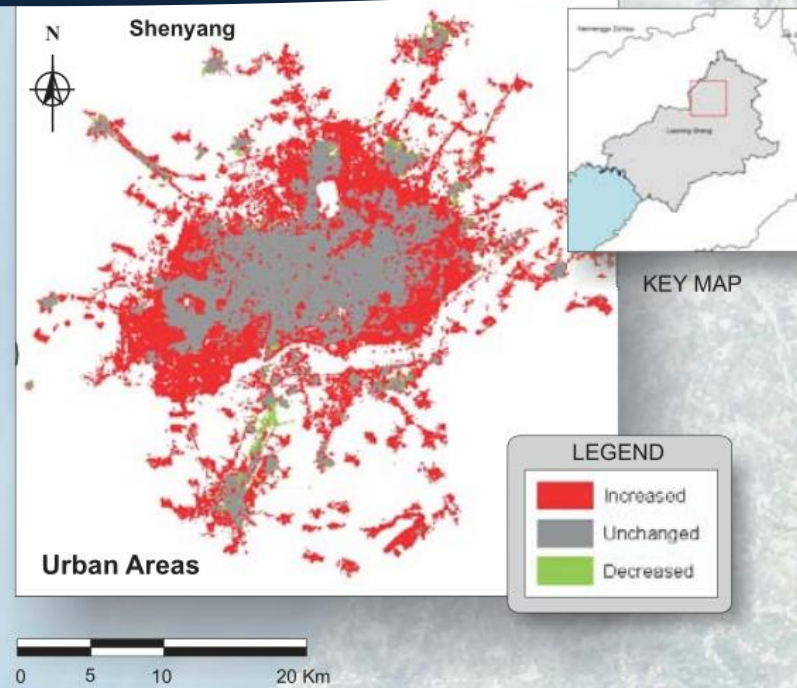
Jiangnan plain,
China

LEGEND

- 0 - Unclassified
- 1 - Urban/Industrial or Built-up Areas
- 2 - Cultivated Land/Agricultural
- 3 - Grassland/Pasture Land
- 4 - Forest Land/Woodland
- 5 - Water Bodies
- 6 - Wet Fields
- 7 - Bare Soil/Unexploited Land
- 8 - Clouds and Shadows

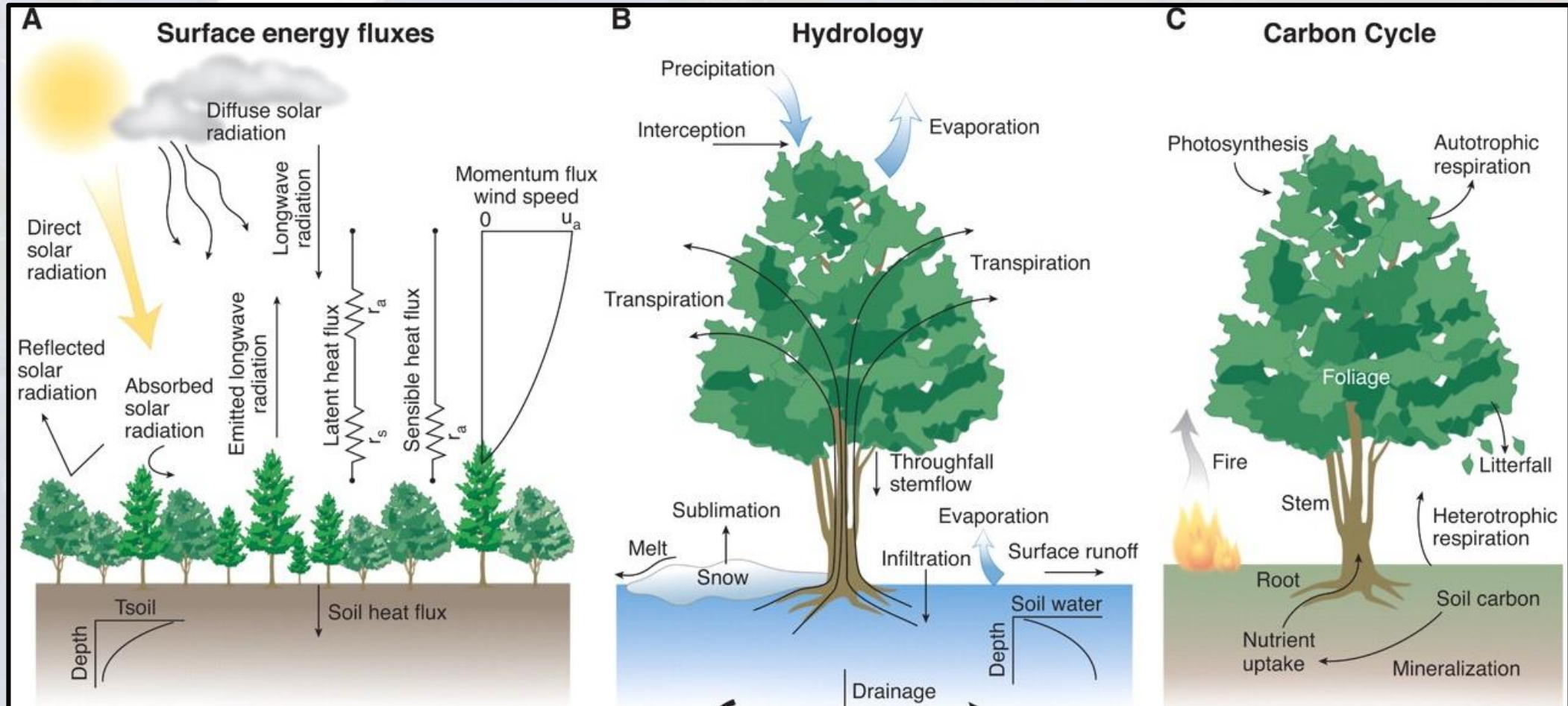
Land Cover Change from a Satellite

Temporal evolution of
urban areas and water
bodies



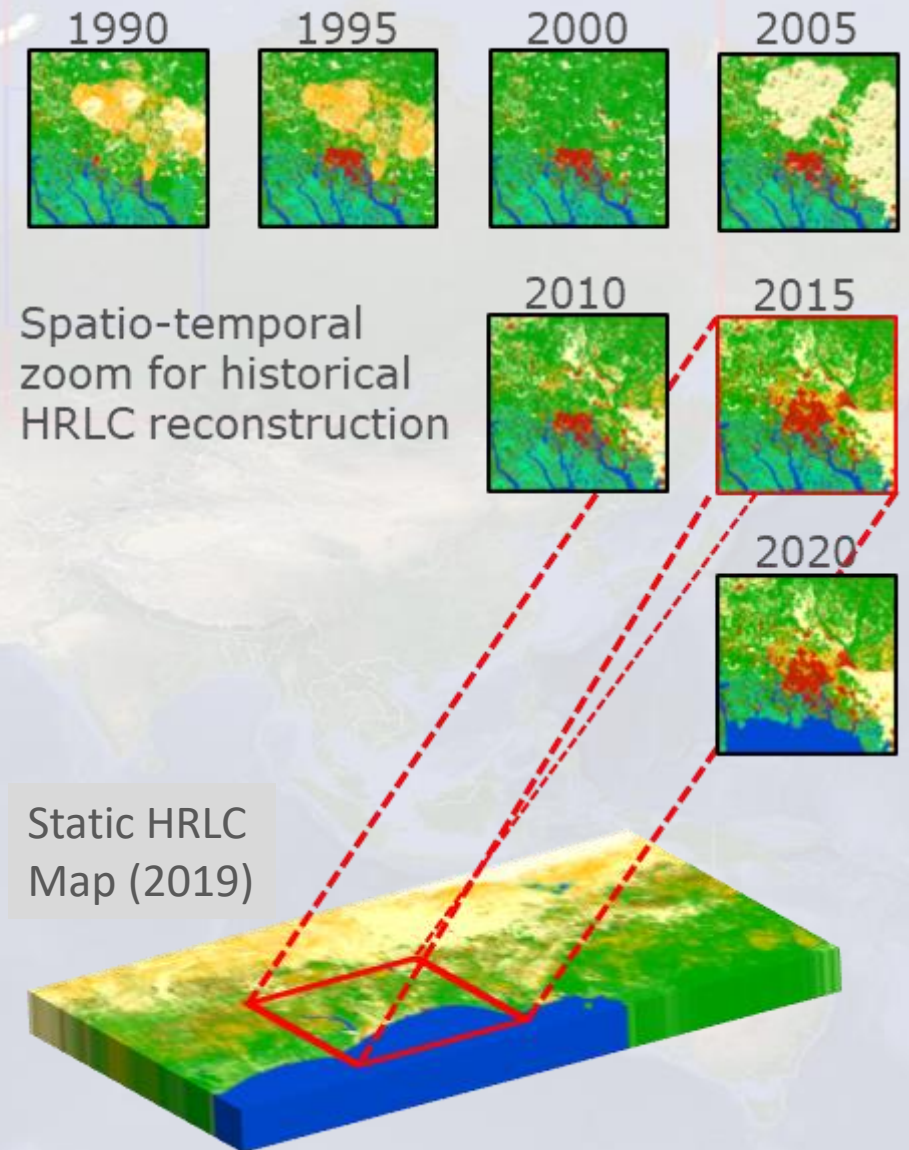
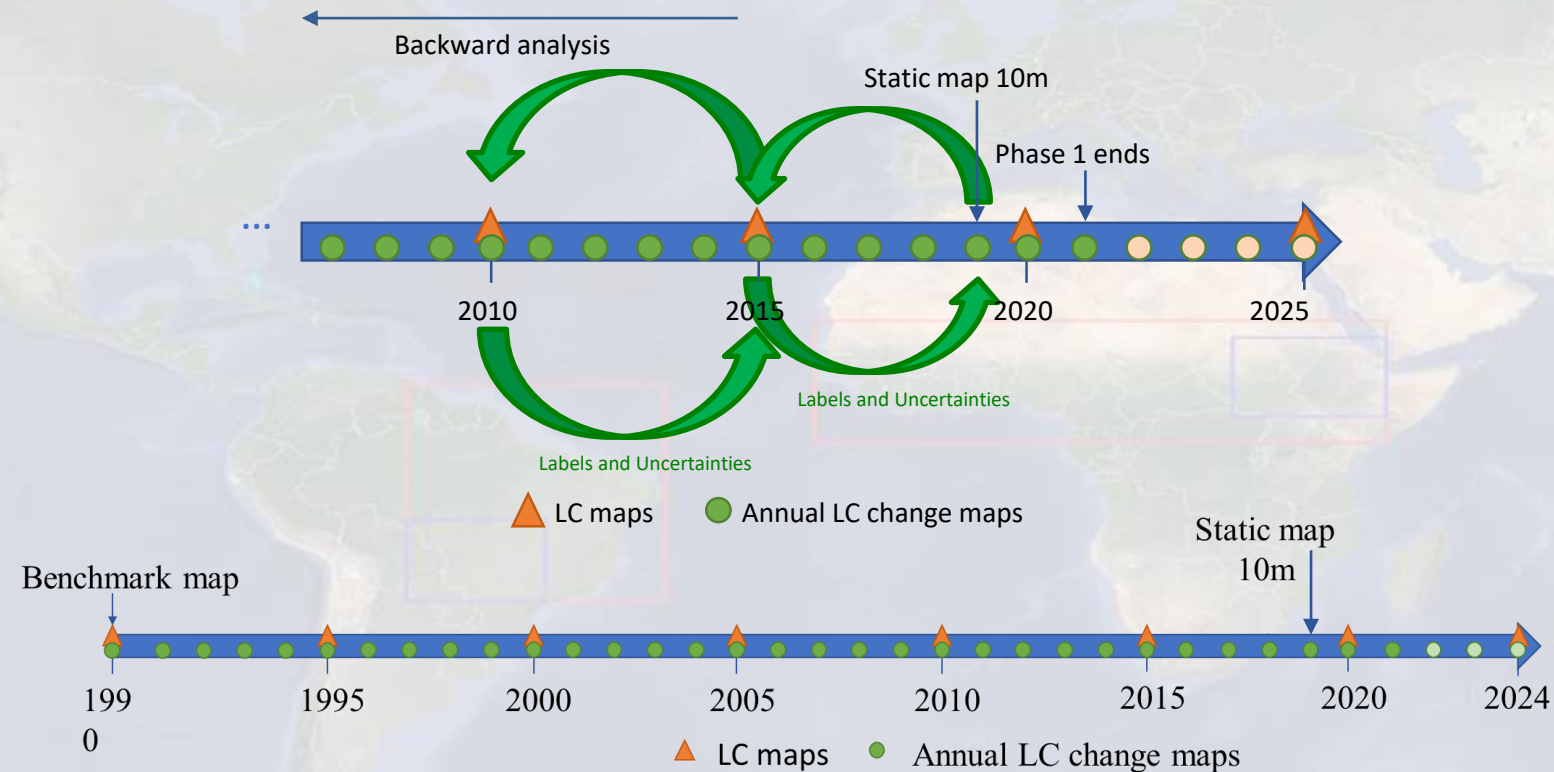
What's the Matter with Land Cover for Climate Change?

Land cover affects reflective (albedo) and aerodynamic properties, evapotranspiration, carbon storage, land surface temperature, and precipitation.



Project Products

- “Static” map of **land cover** at the **spatial resolution of 10 m** in **2019**
- “**Historical**” **evolution of land cover since 1990**, observed every five years at the spatial resolution of 30 m, also assessing annual changes





Study Areas

Regions characterized by **extreme climate conditions** or subject to **significant climate change in the past decades**

- Static map • 2019
- Historical evolution and change • 1990-2019

Impact of deforestation, forest fires, and agricultural development on water and carbon cycles

Impact of climate change on the shift of the forest-shrubs-grassland transition zone, and evolution of carbon storage in permafrost

Impact of climate change on drought, floods, and monsoon dynamic

About 35 million km²,
115 times the area of Italy

More than 400 terabytes
of images

(Geospatial) **Big Data**

A few Ideas on the Methods

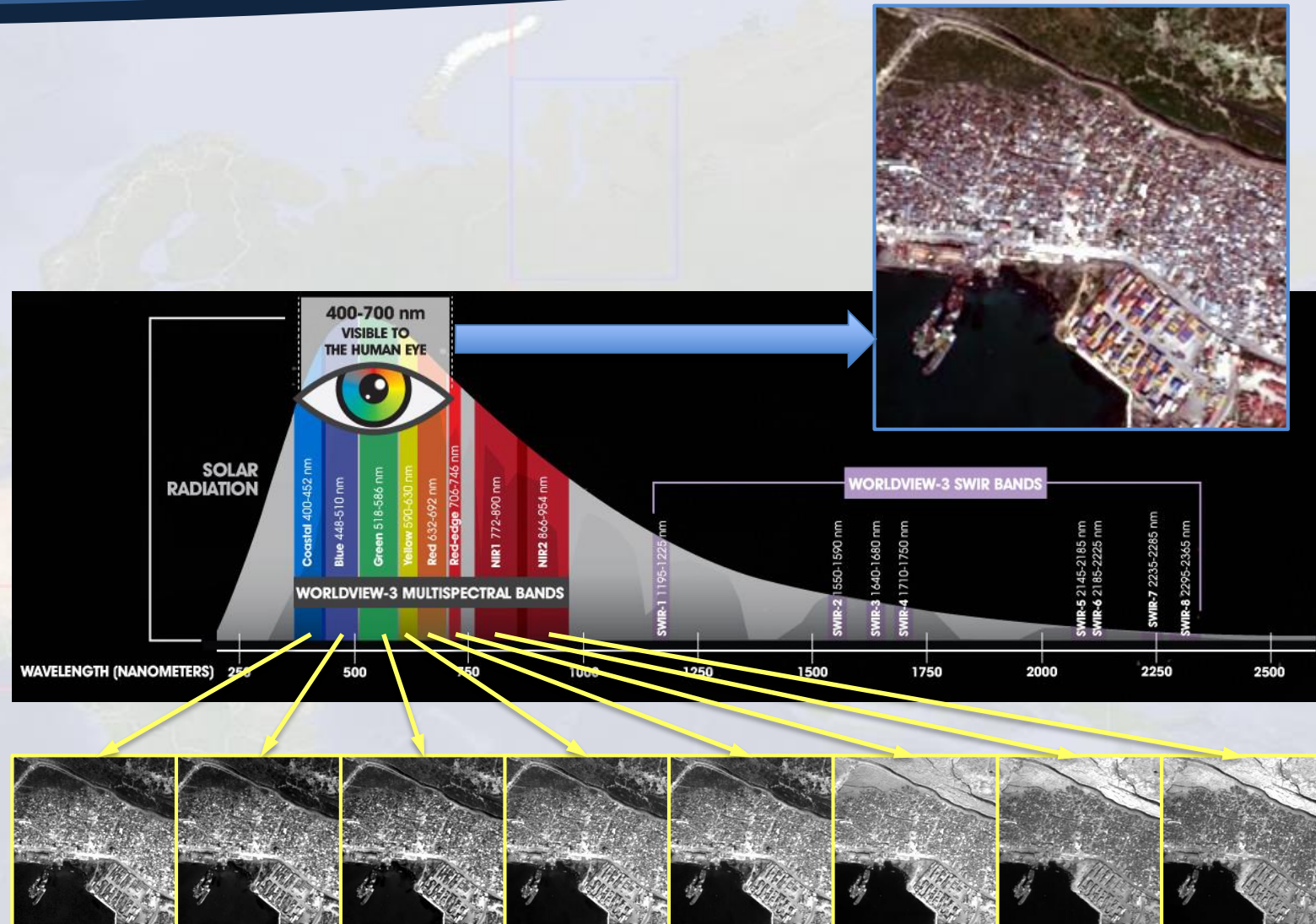
In general, satellite sensors collect **multiple measurements for each pixel**, for example in relation to different **wavelength ranges**.

The “measurements” taken by the human eye in the wavelength range of visible light determine what we usually call “**color**.”

A satellite sensor generally acquires **more measurements** than the human eye including at wavelengths outside the visible range (e.g., infrared, microwave).

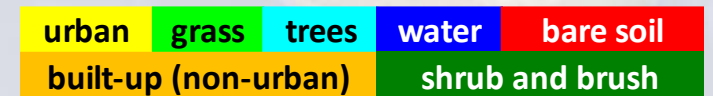
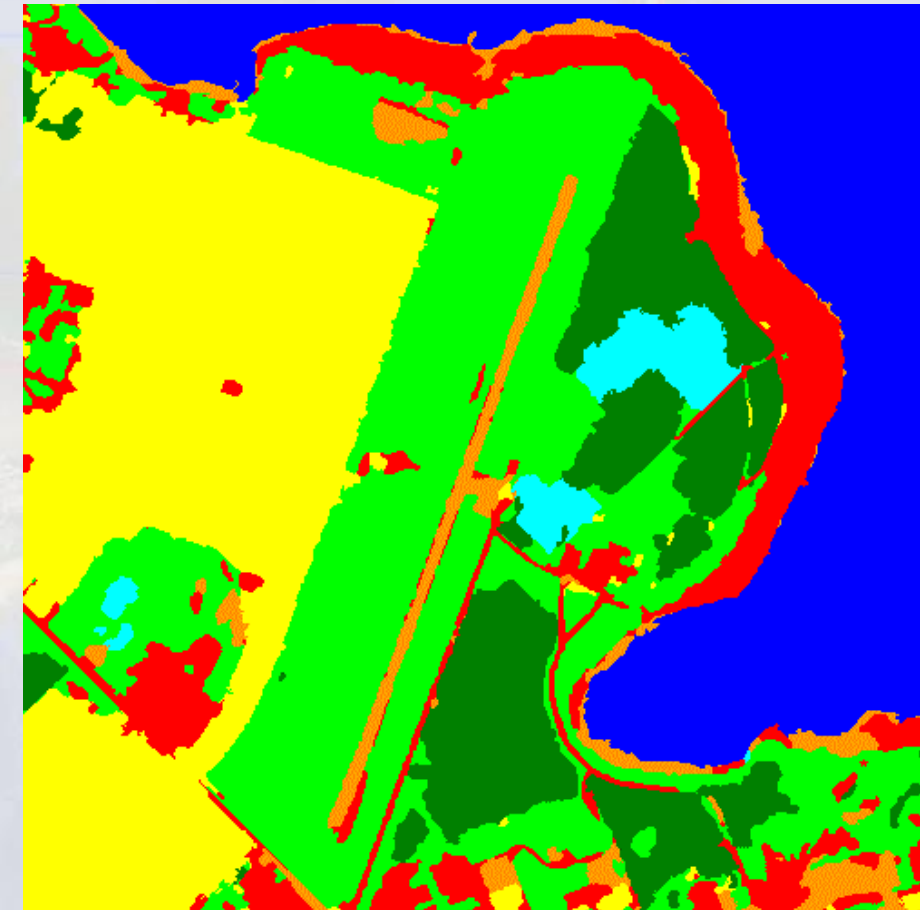
All these measurements (“**spectral information**”) are relevant to identify the land cover of the portion of Earth surface associated with each pixel.

But it's not enough...

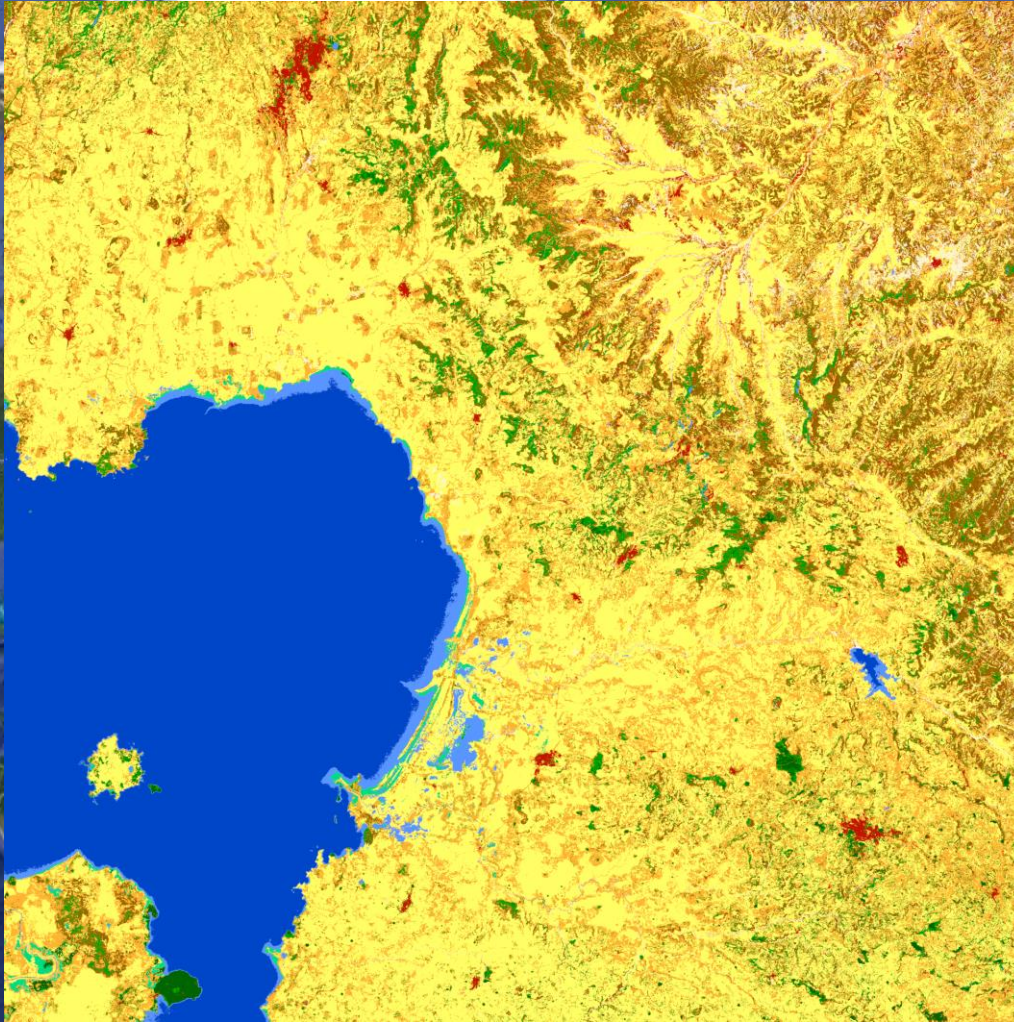


A few Ideas on the Methods

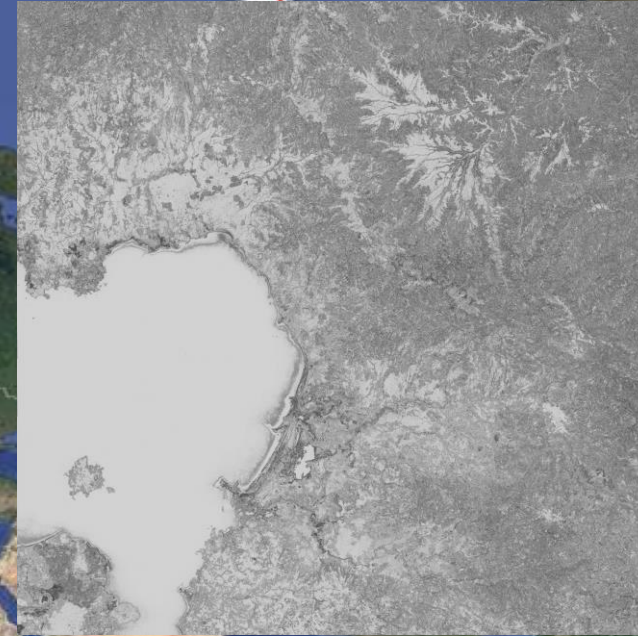
- Neighboring pixels in an image correspond to nearby portions of the Earth surface. Therefore, they are expected to be correlated (**“spatial context”**).
- If we do not take this into account and map land cover using only the spectral information, then the result can be **inaccurate and spatially irregular**.
- But if **we also use the spatial context...**
- Advanced **image processing** and **machine learning** methods model the spatial context and fuse it with the spectral information.
- This can be accomplished using advanced **probabilistic graphical models** (e.g., **Markov random fields**) and recently with **neural networks** (deep learning).



Examples of Results: Africa



Classification map



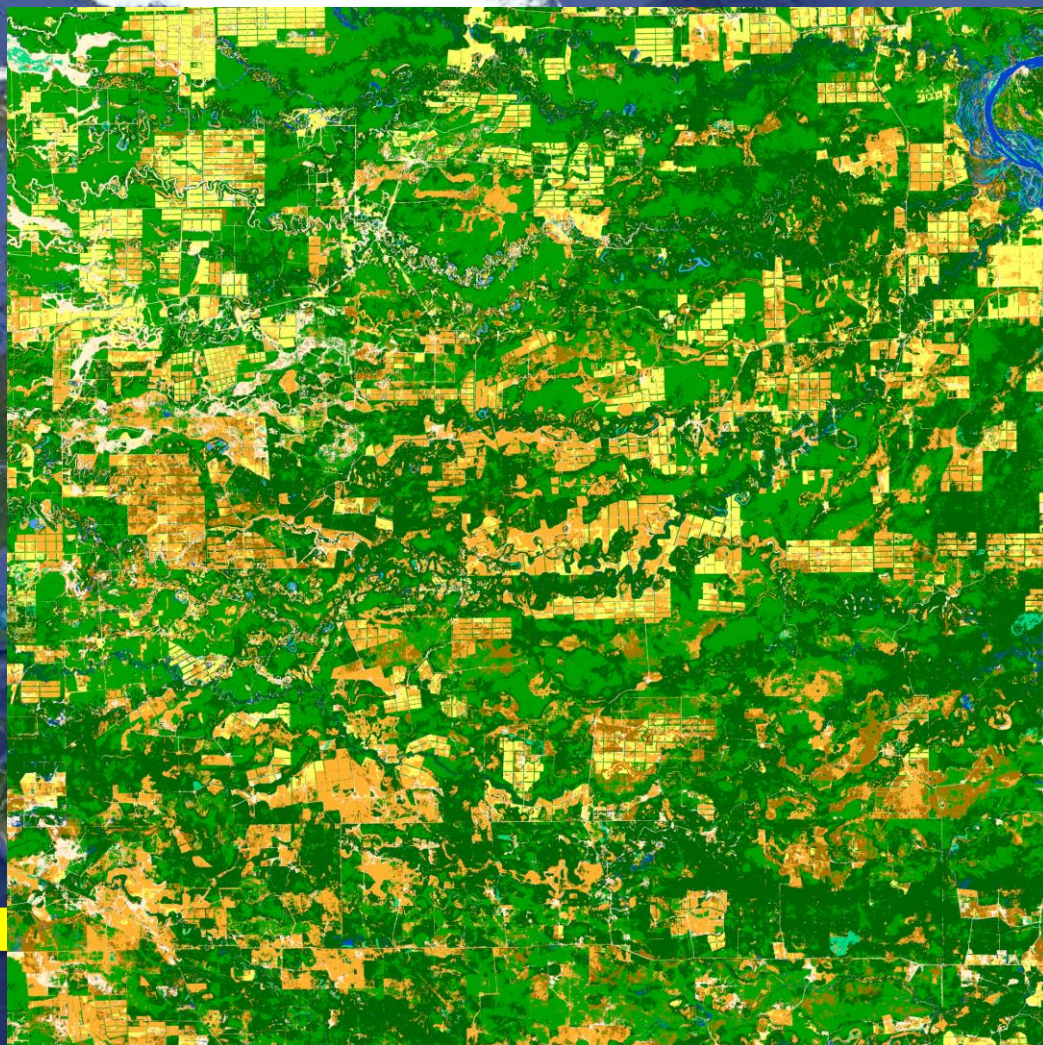
100%

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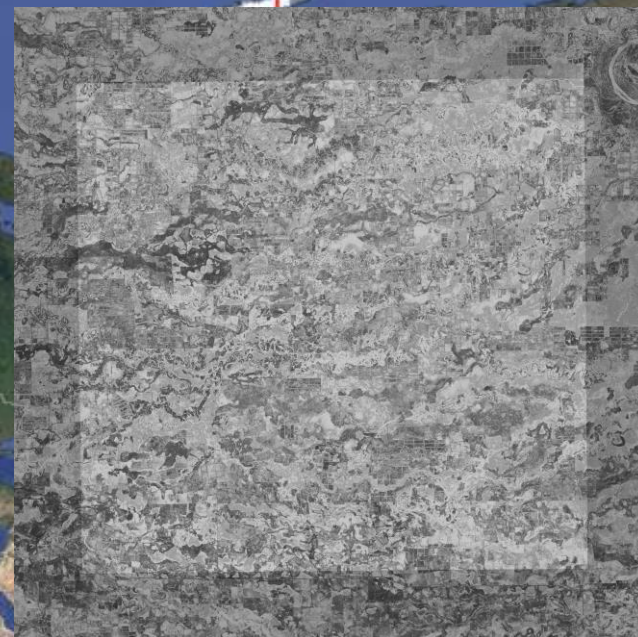
Confidence

CODE	DN	1 st LEVEL	CODE	DN	1 st LEVEL
10	1	Tree cover evergreen broadleaf	90	9	Woody vegetation aquatic or regularly flooded
20	2	Tree cover evergreen needleleaf	100	10	Grassland vegetation aquatic or regularly flooded
30	3	Tree cover deciduous broadleaf	110	11	Lichens and Mosses
40	4	Tree cover deciduous needleleaf	120	12	Bare areas
50	5	Shrub cover evergreen	130	13	Built-up
60	6	Shrub cover deciduous	140	14	Open Water seasonal
70	7	Grasslands	150	15	Open Water permanent
80	8	Croplands	160	16	Permanent snow and/or ice

Examples of Results: Amazonia (1 of 2)



Classification map



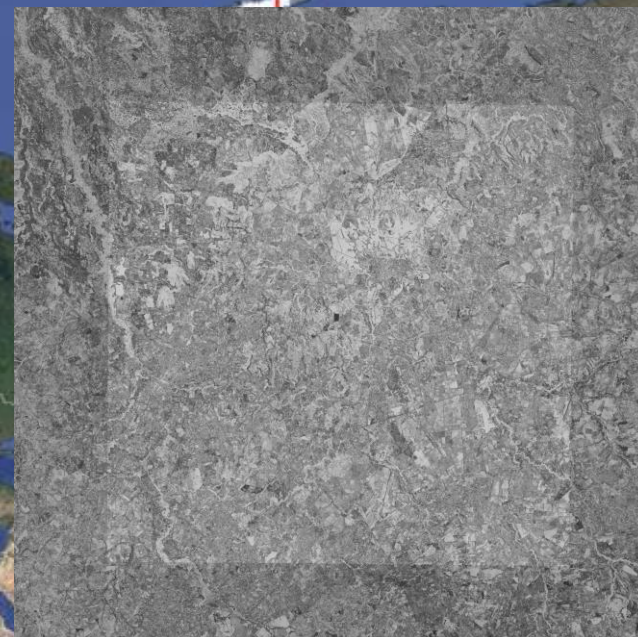
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Examples of Results: Amazonia (2 of 2)



Classification map



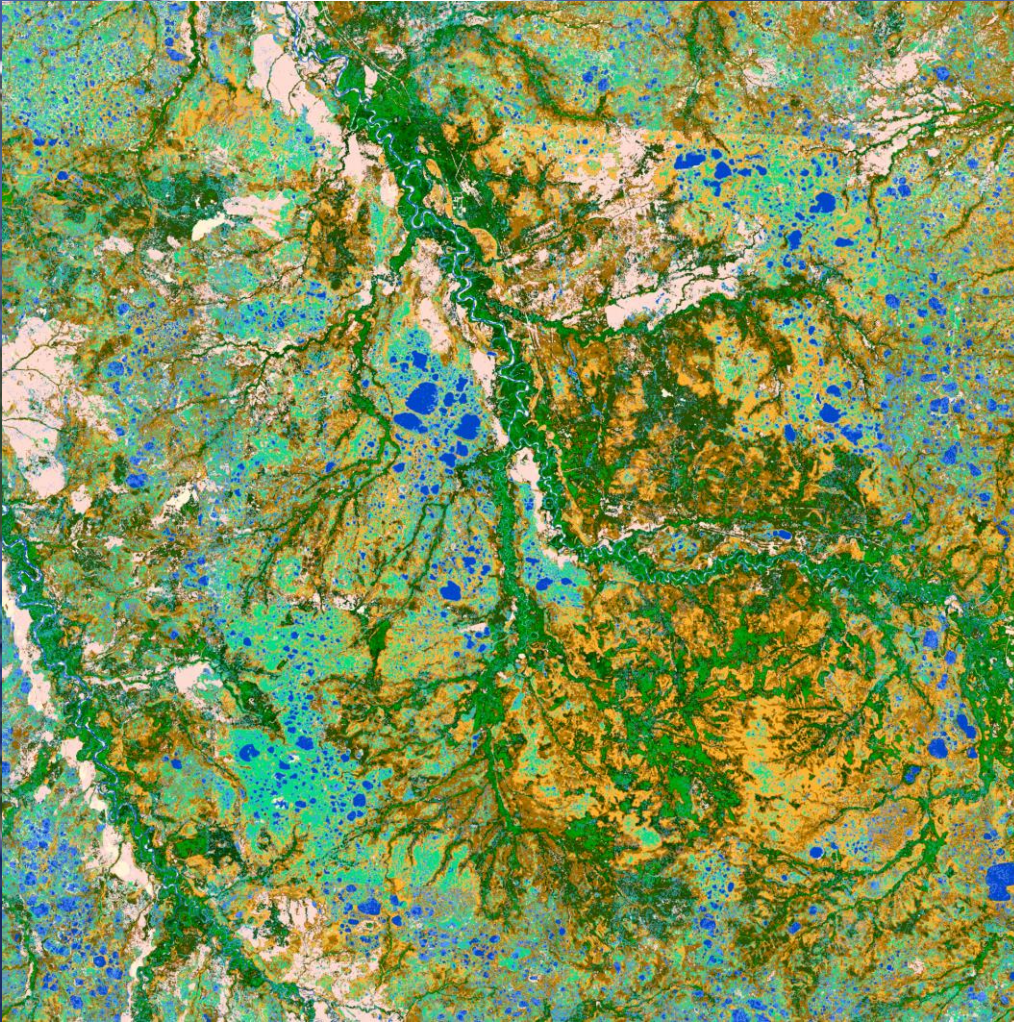
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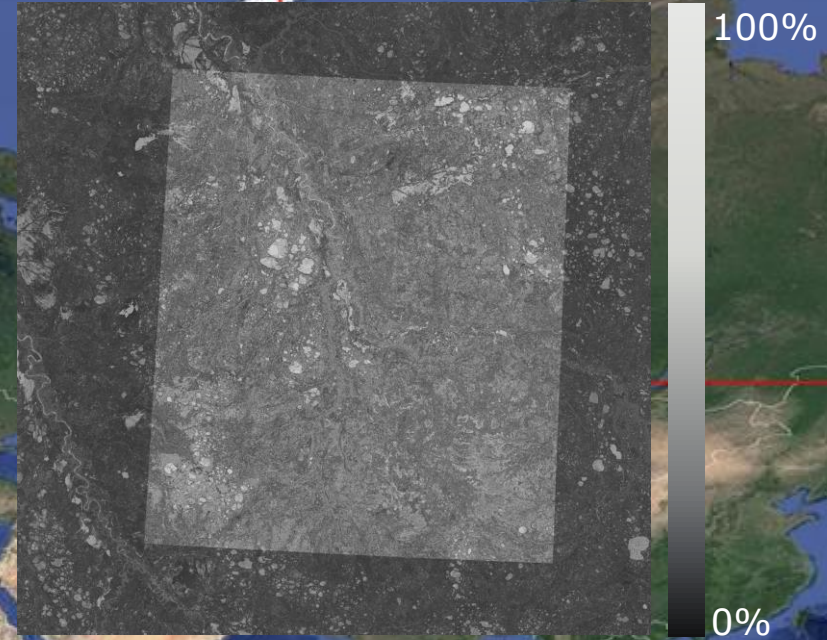
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Examples of Results: Siberia

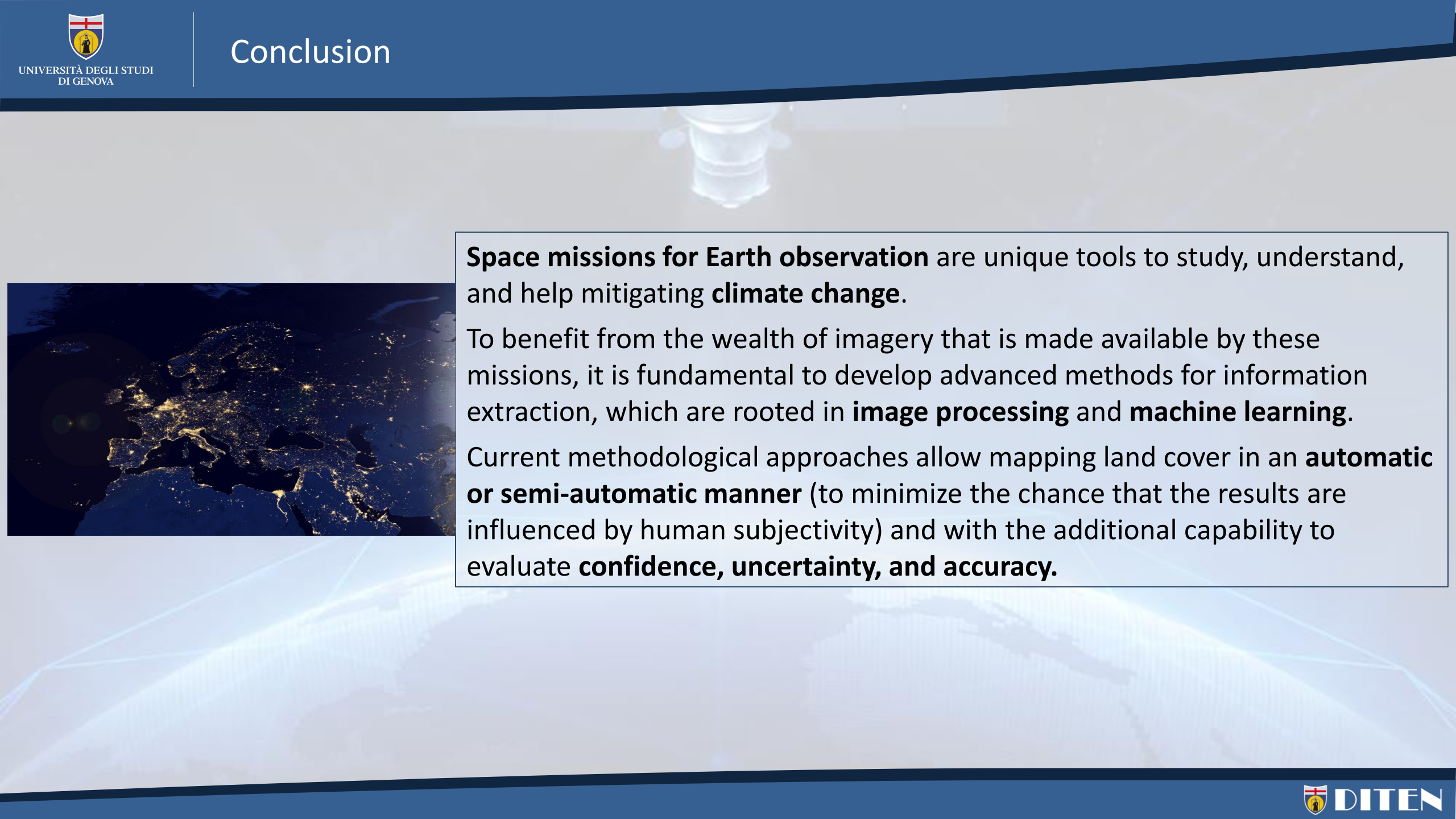


Classification map



Confidence

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Space missions for Earth observation are unique tools to study, understand, and help mitigating **climate change**.

To benefit from the wealth of imagery that is made available by these missions, it is fundamental to develop advanced methods for information extraction, which are rooted in **image processing** and **machine learning**.

Current methodological approaches allow mapping land cover in an **automatic or semi-automatic manner** (to minimize the chance that the results are influenced by human subjectivity) and with the additional capability to evaluate **confidence, uncertainty, and accuracy**.

