

Applications of AI in urban data spaces - the USAGE EU project -

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and contributions from the entire USAGE EU project

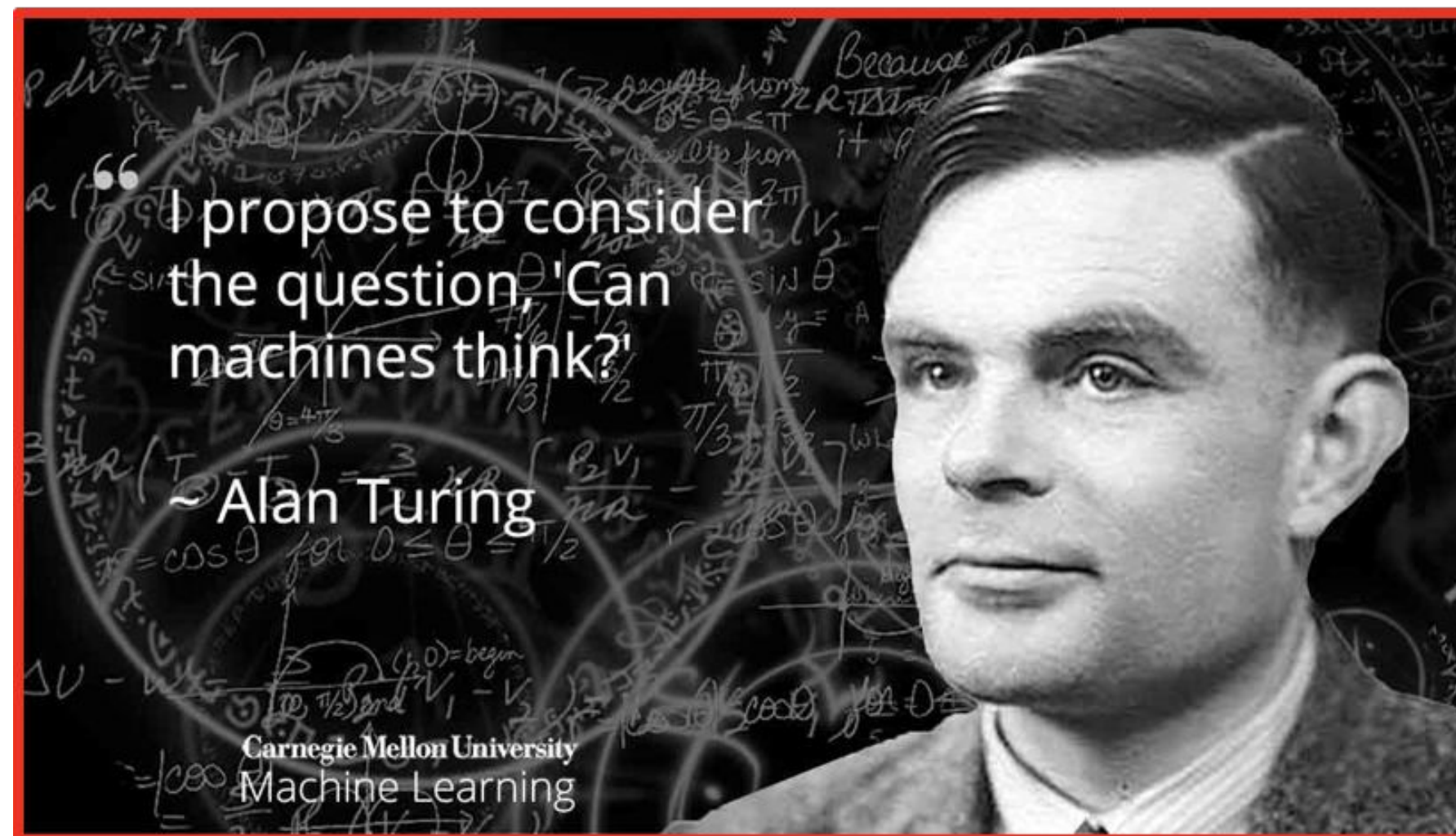
The EU USAGE project (2022-2025)

- Urban Data Space for Green Deal
- Call HORIZON-CL6-2021-GOVERNANCE-01-17 (IA)
- Support the implementation of the EU strategy for **data** and **European Green Deal** priority actions fostering the use of geospatial data and **AI solutions**
- 4 pilot cities: Ferrara (Italy), Graz (Austria), Leuven (Belgium), Zaragoza (Spain)
- <https://www.usage-project.eu/>



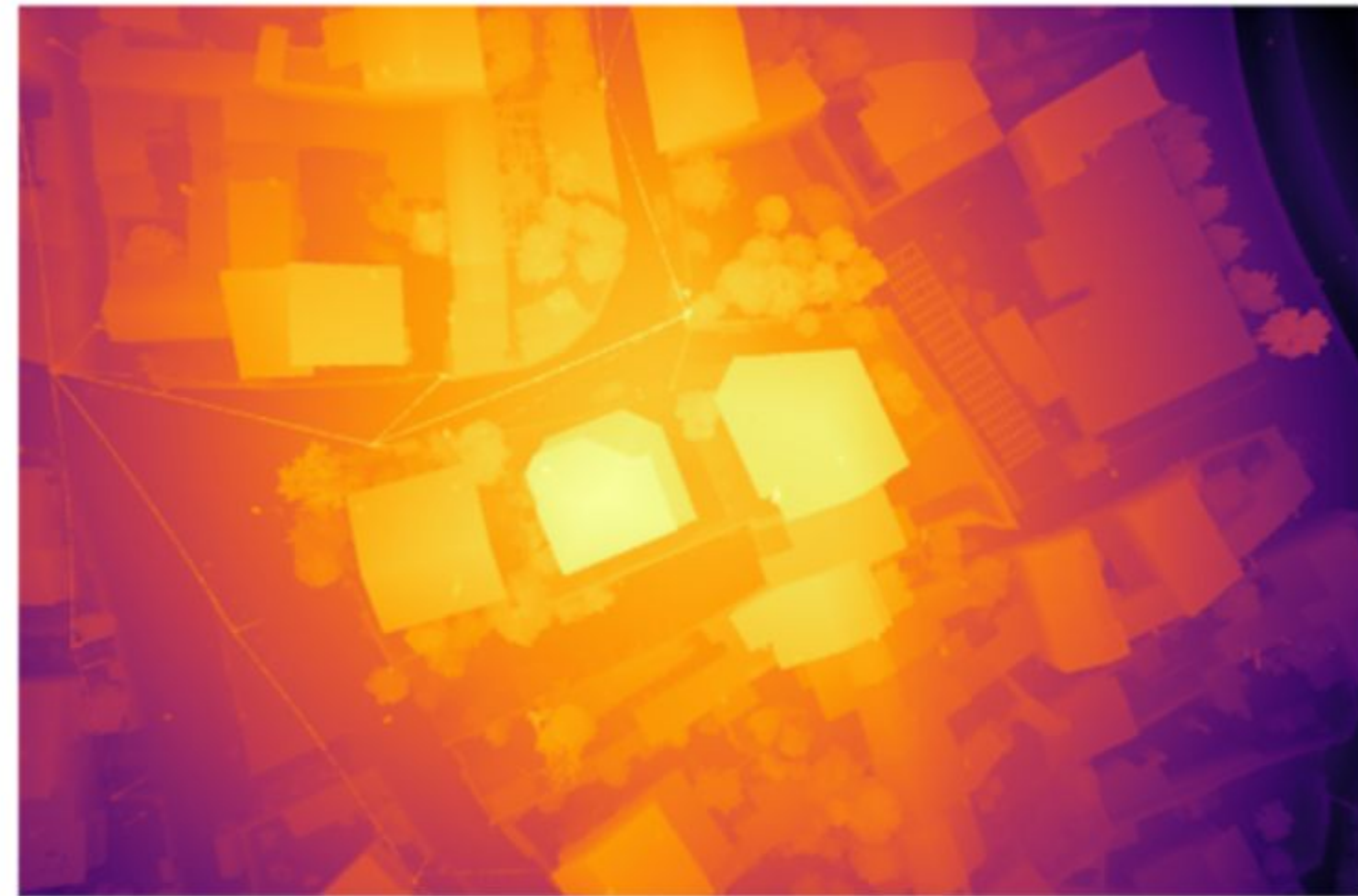
AI ...what?

- Artificial Intelligence or Absence of Intelligence?
- Sometimes used as a **Black Box**, sometimes you get lost into the **Black Box**
- Lack of explainability, replicability and generalization
- We need to move towards an Intelligent use of AI



AI for geospatial data

- **Dream**: extract meaningful information (2D or 3D) from geodata better and faster than humans
- Current: for *radiometric* data still not faster than human but for *geometric* data machines / algorithms are almost unbeatable



AI for geospatial data

- Image colorization
- Image segmentation / classification
- Tie points extraction / Image triangulation
- 3D reconstruction (monocular, stereo/MVS, NeRF)
- DSM inpainting / editing / cleaning / super-resolution
- Point cloud segmentation / classification
- Orthophoto correction
- Photovoltaic panel identification / counting
- Urban heat island prediction
- Detection of building footprints
- Tree identification / counting
- etc.



AI for image colorization

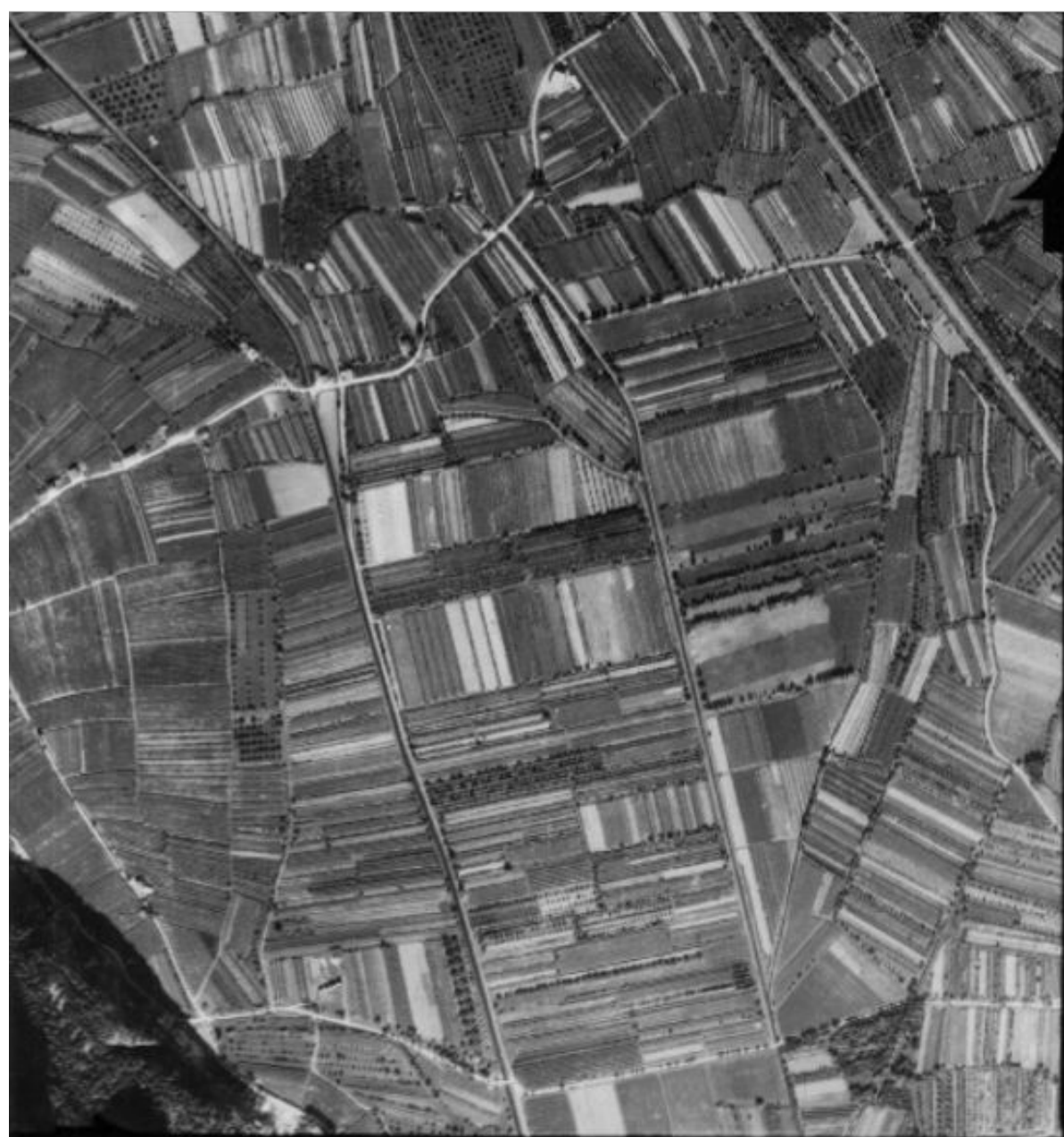
- Historical aerial images: AI to **restore** colors and **improve** interpretation and understanding
- Hyper-U-NET (https://github.com/3DOM-FBK/Hyper_U_Net)



[Farella, E.M., Malek, S., Remondino, F., 2022: *Colorizing the past: Deep learning for the automatic colorization of historical aerial images*. Imaging, 8, 269]

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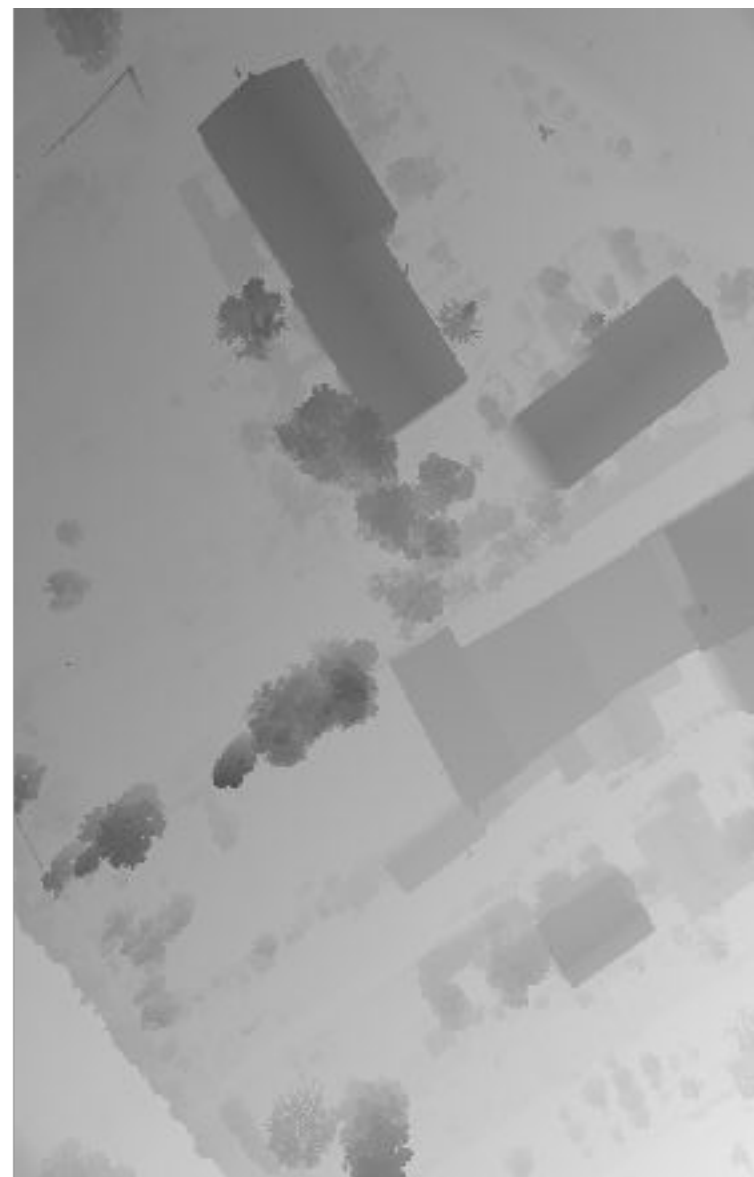
AI for Multi-View Stereo (MVS) 3D reconstruction

- Learning-based MVS methods complementary (?) to conventional approaches

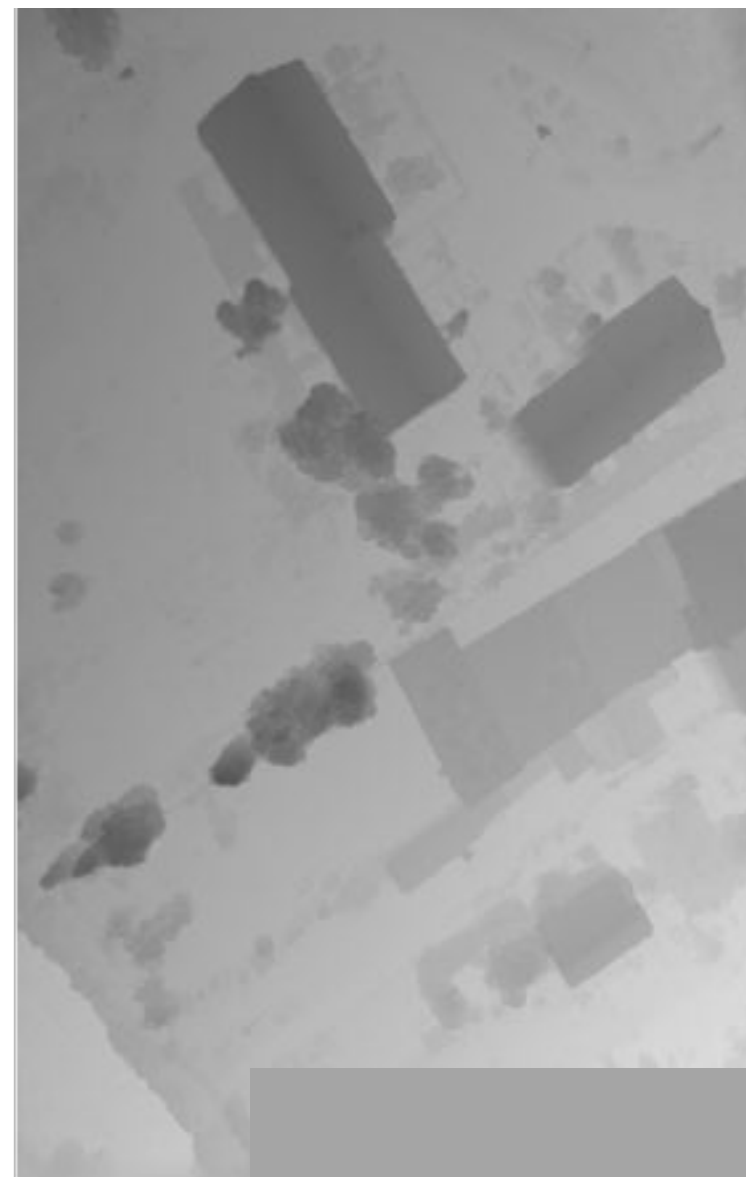
UAV



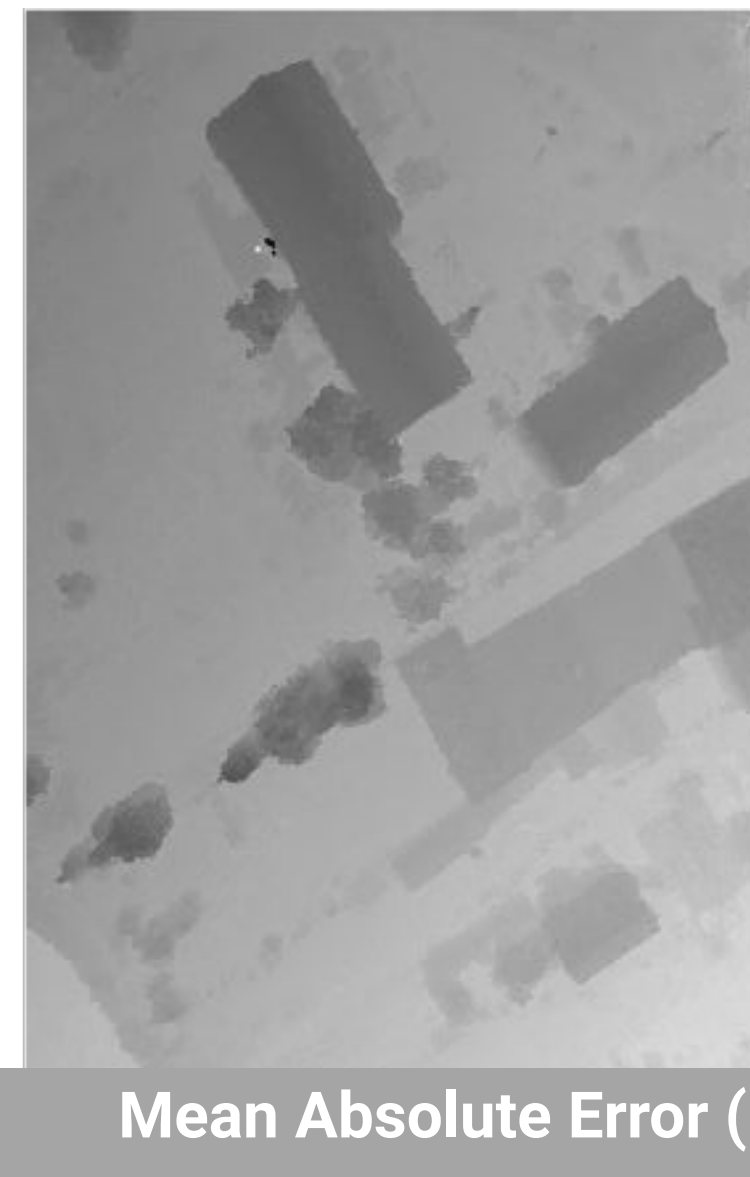
LiDAR GT



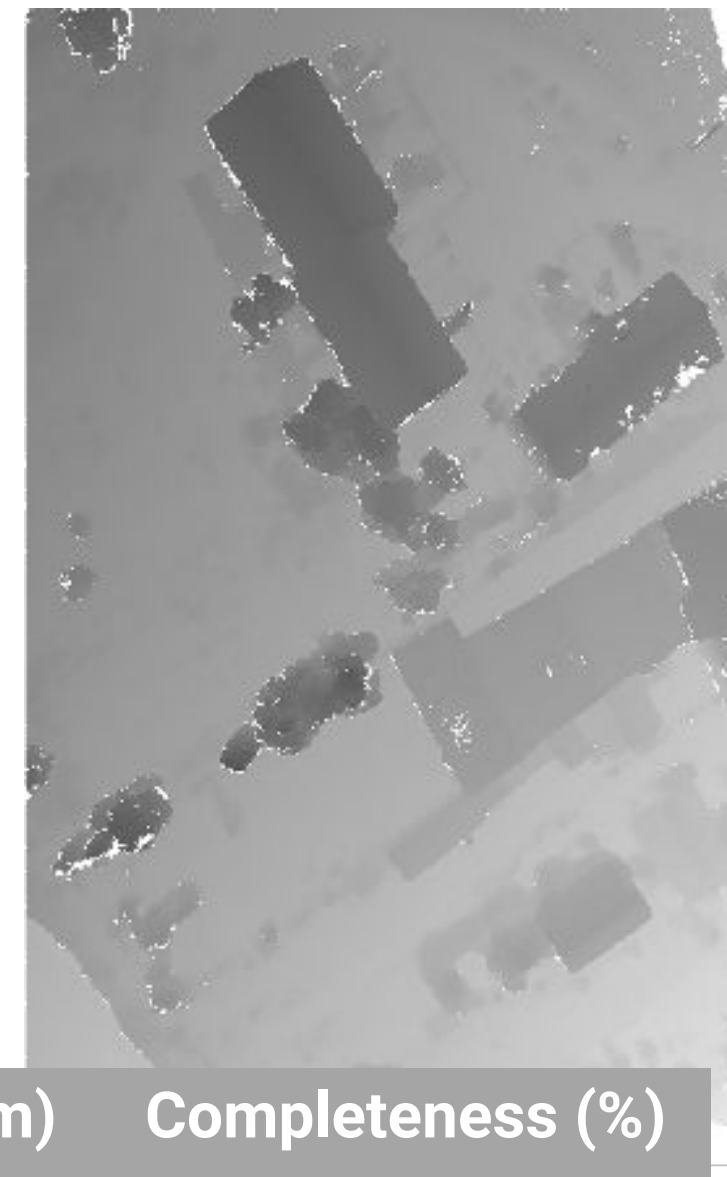
MVSFormer



Unimvs



Conventional (SGM)



Mean Absolute Error (m) Completeness (%)

MVSFormer	0.48	98
Unimvs	0.485	97
Conventional	0.35	94

[Nex, F., Zhang, N., Remondino, F., Farella, E.M., Qin, R., Zhang, C., 2023. **Benchmarking the extraction of 3D geometry from UAV images with deep learning methods**. ISPRS Int. Arch. Photogramm. Remote Sens. Spatial Inf. Sci., XLVIII-1/W3-2023.]

AI for point cloud segmentation / classification (1/2)

- Lack of generalization
- No standards in classes definition

Name	Reference	Year	Classes	Points (mil)	Spatial Size (m ²)	RGB	Sensor
ISPRS	Niemeyer et al. (2014)	2012	9	1.2	1.6 x 10 ⁵	No	ALS / LiDAR
DublinCity	Zolanvari et al. (2019)	2019	13	260	2 x 10 ⁶	No	ALS / LiDAR
DALES	Varney et al. (2020)	2020	8 (9)	505	10 x 10 ⁶	No	ALS / LiDAR
LASDU	Ye et al. (2020)	2020	5	3.12	1.02 x 10 ⁶	No	ALS / LiDAR
Campus3D	Li et al. (2020)	2020	24	937	1.58 x 10 ⁶	Yes	UAV Photo
SensatUrban	Hu et al. (2022)	2020	13 (31)	2847	7.64 x 10 ⁶	Yes	UAV Photo
Swiss3DCities	Can et al. (2021)	2021	5	226	2.7 x 10 ⁶	Yes	UAV Photo
Hessigheim 3D	Kölle et al. (2021)	2021	11	73	8 x 10 ⁴	Yes	UAV LiDAR
STPLS3D	Chen et al. (2022)	2022	6 (18)	-	6 x 10 ⁶	Yes	Synthetic + UAV Photo
HRHD-HK	Li et al. (2023)	2023	7	273	9 x 10 ⁶	Yes	UAV Photo

AI for point cloud segmentation / classification (2/2)

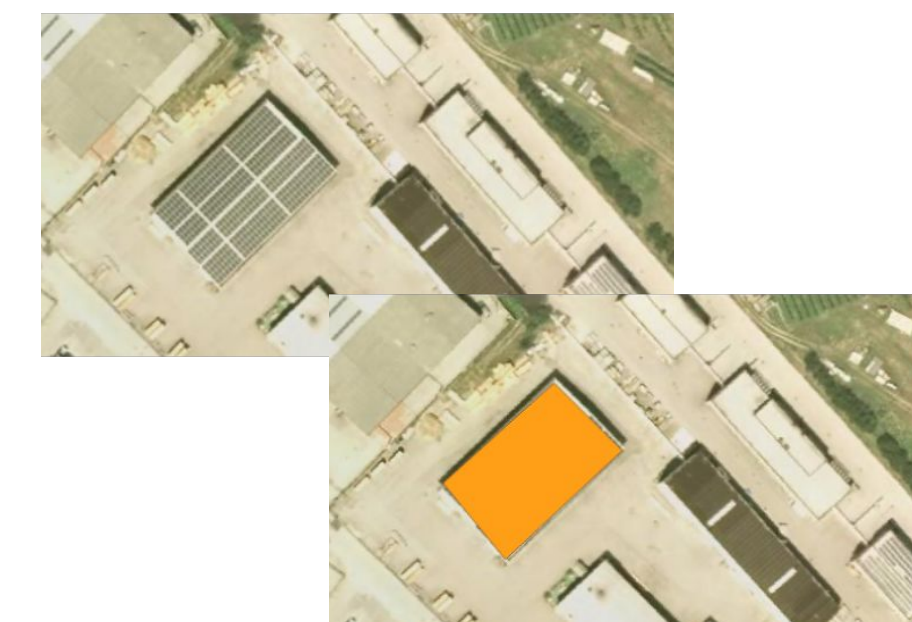
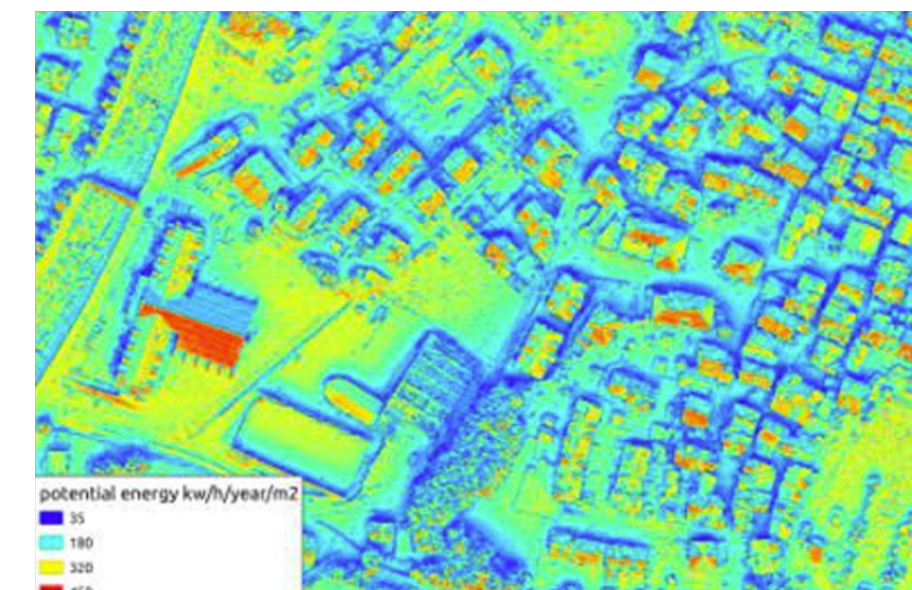
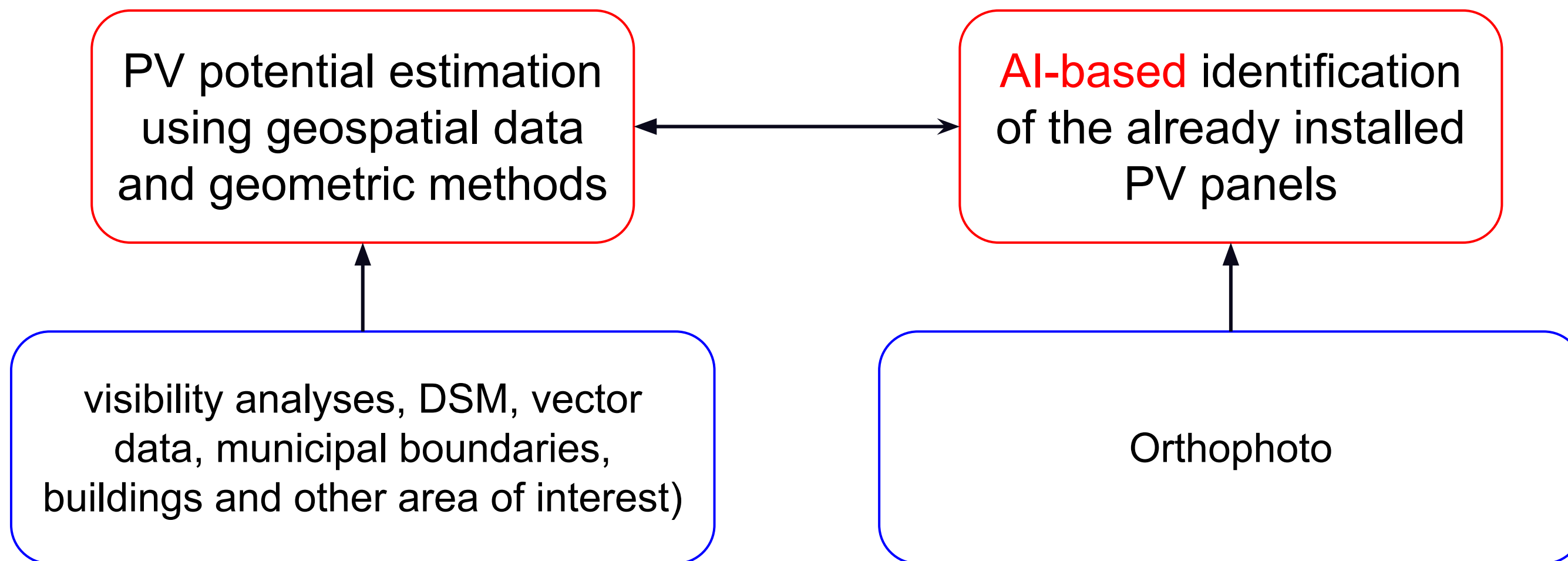
- Lack of generalization
- No standards in classes definition (task force for NMCAs?)

		ground	roof	facade	water	high vegetation	medium vegetation	low vegetation	noise	Bridge structure	Bridge deck	viaducts	pylons and cables	powerline	car	fence
Netherlands	AHN4	blue	red	red	yellow	dark brown	medium brown	light brown	deleted	grey	grey	grey	grey	grey	grey	grey
	AHN3	blue	red	red	yellow	dark brown	medium brown	light brown	deleted	grey	grey	grey	grey	grey	grey	grey
Poland	ISOK	blue	red	red	yellow	dark green	medium green	light green	yellow	grey	red	red	red	red	red	red
	Other	blue	red	red	yellow	dark green	medium green	light green	yellow	grey	red	red	red	red	red	red
Switzerland	Surface3D	blue	red	red	yellow	light green	medium green	light green	deleted	grey	grey	grey	grey	grey	grey	grey
France	-	blue	red	red	yellow	light green	medium green	light green	deleted	grey	grey	grey	grey	grey	grey	grey
Austria	Vienna	blue	red	red	yellow	light green	medium green	light green	yellow	grey	grey	grey	grey	grey	grey	grey

[Walicka, A., Pfeifer, 2023: *Classification of point clouds for transnational data*. EuroSDR Workshop on Point Cloud Processing, Stuttgart, Jan 2023]

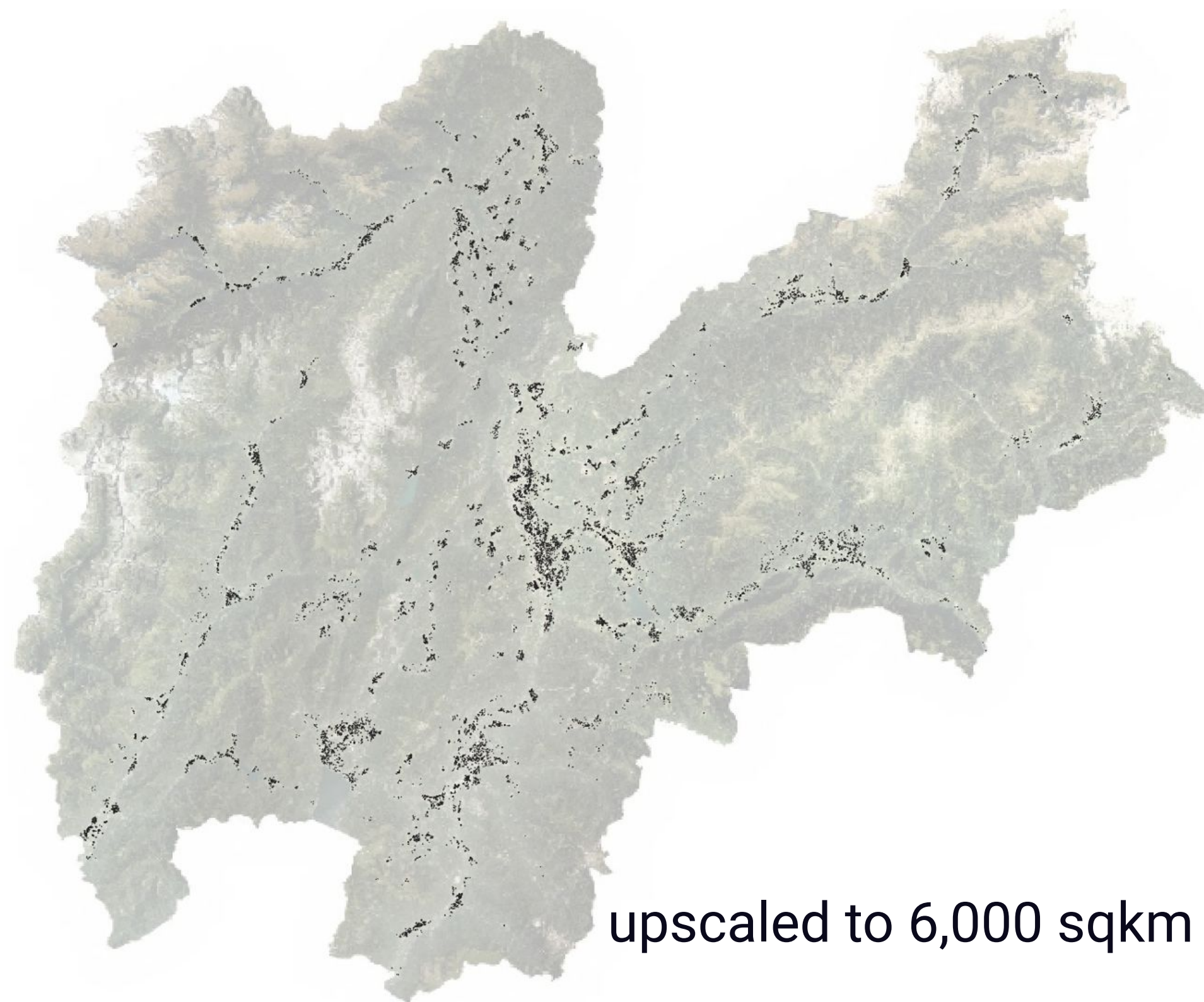
AI for solar panel identification (1/2)

- Identification of existing solar panels to better estimate PV potential of buildings / ground



AI for solar panel identification (2/2)

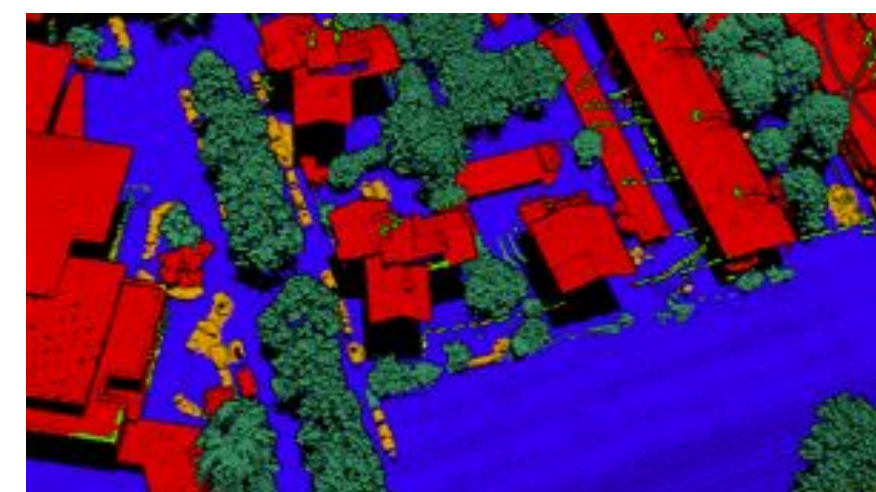
- U-net-like network for panel identification



upscaled to 6,000 sqkm

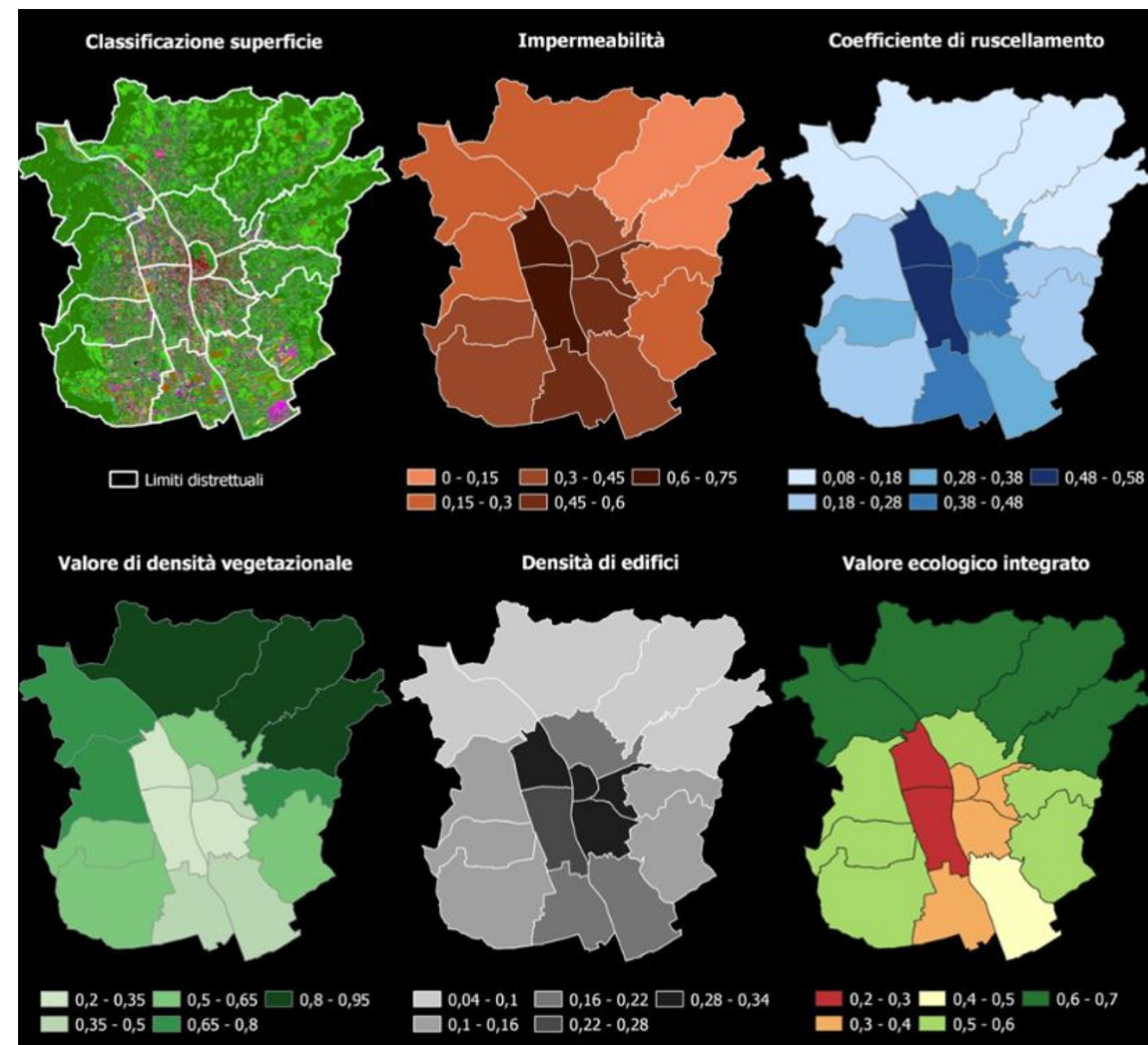
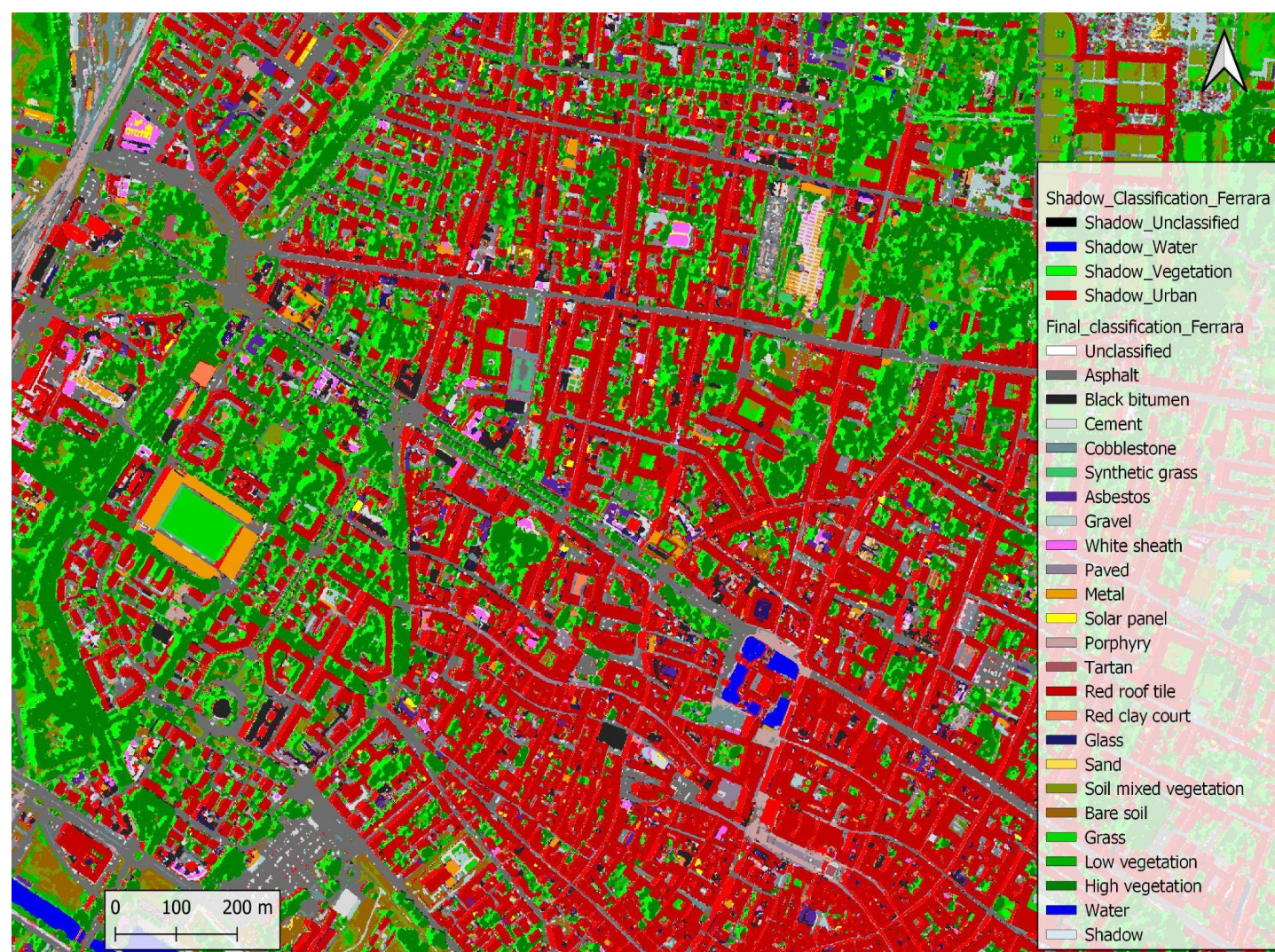
AI for geospatial data within USAGE pilot activities

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AI within USAGE pilot activities: image classification

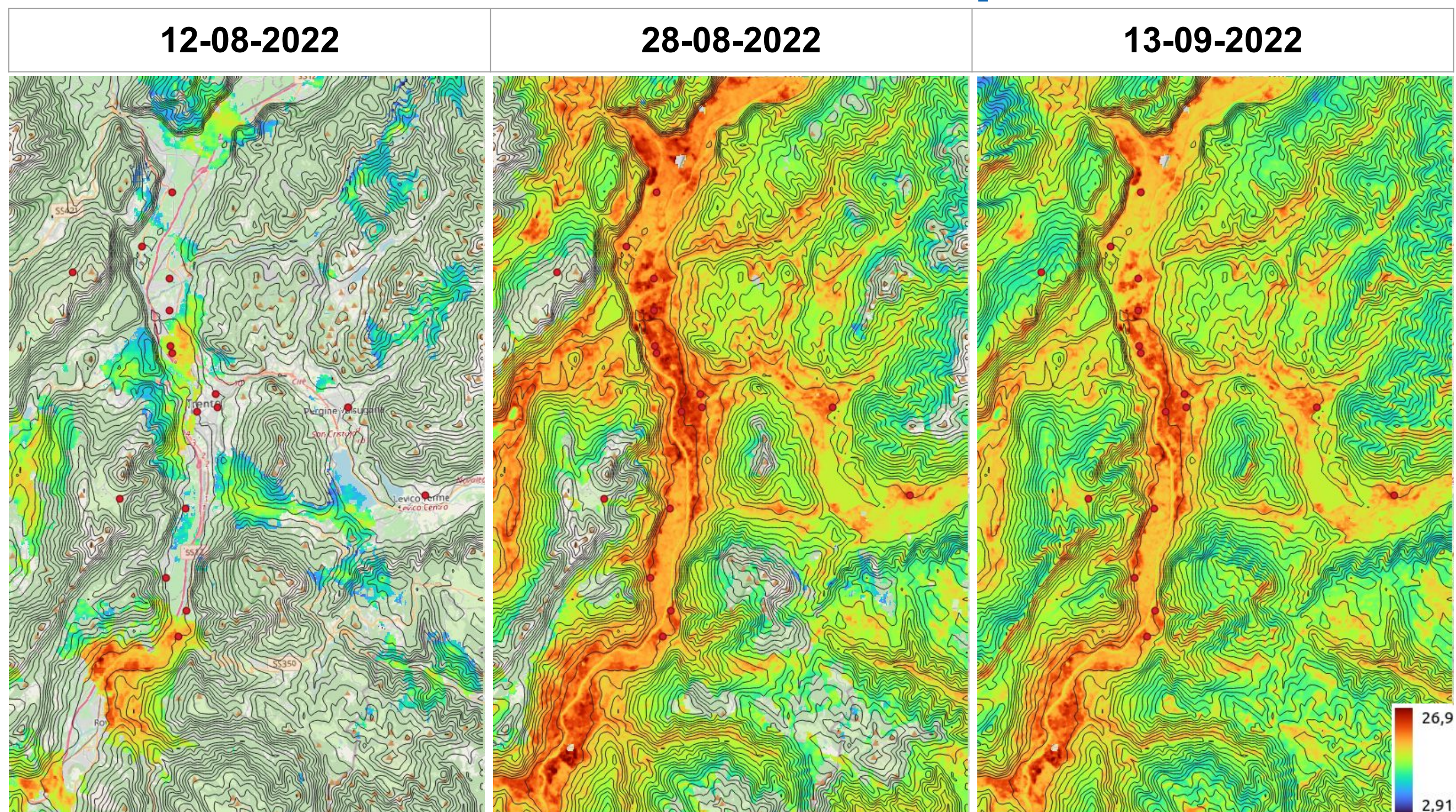
- Hyperspectral images (AVT): 1m, 364 bands, VNIR and SWIR (0,4 - 2,5 μm) [AisaFENIX384 by Specim]
- AI-based supervised classification
- 26 classes



[Beber et al., 2023. *Multi-modal geospatial and thematic data to foster green deal applications*. ISPRS Int. Arch. Photogramm. Remote Sens. Spatial Inf. Sci., XLVIII-1/W3-2023.]

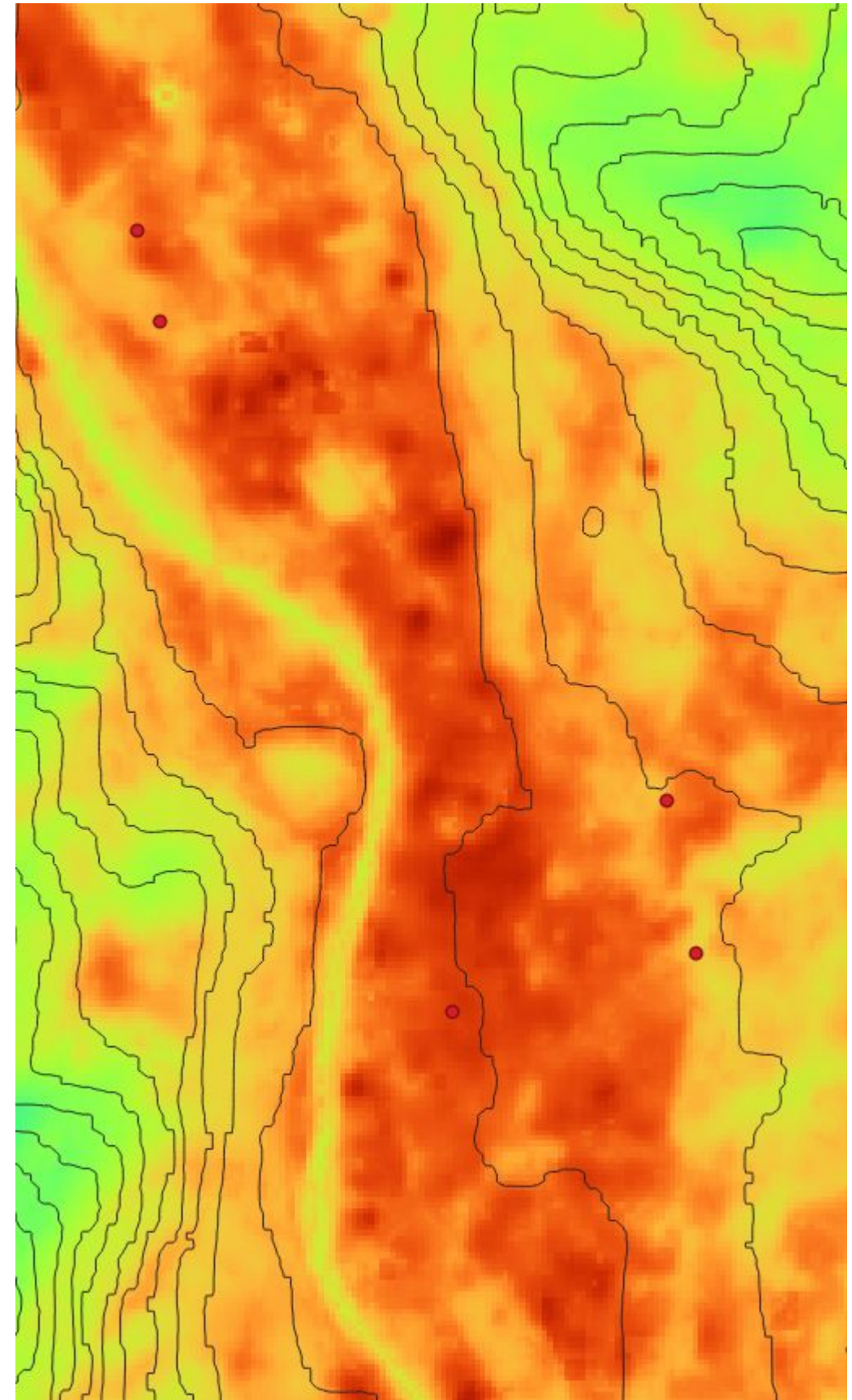
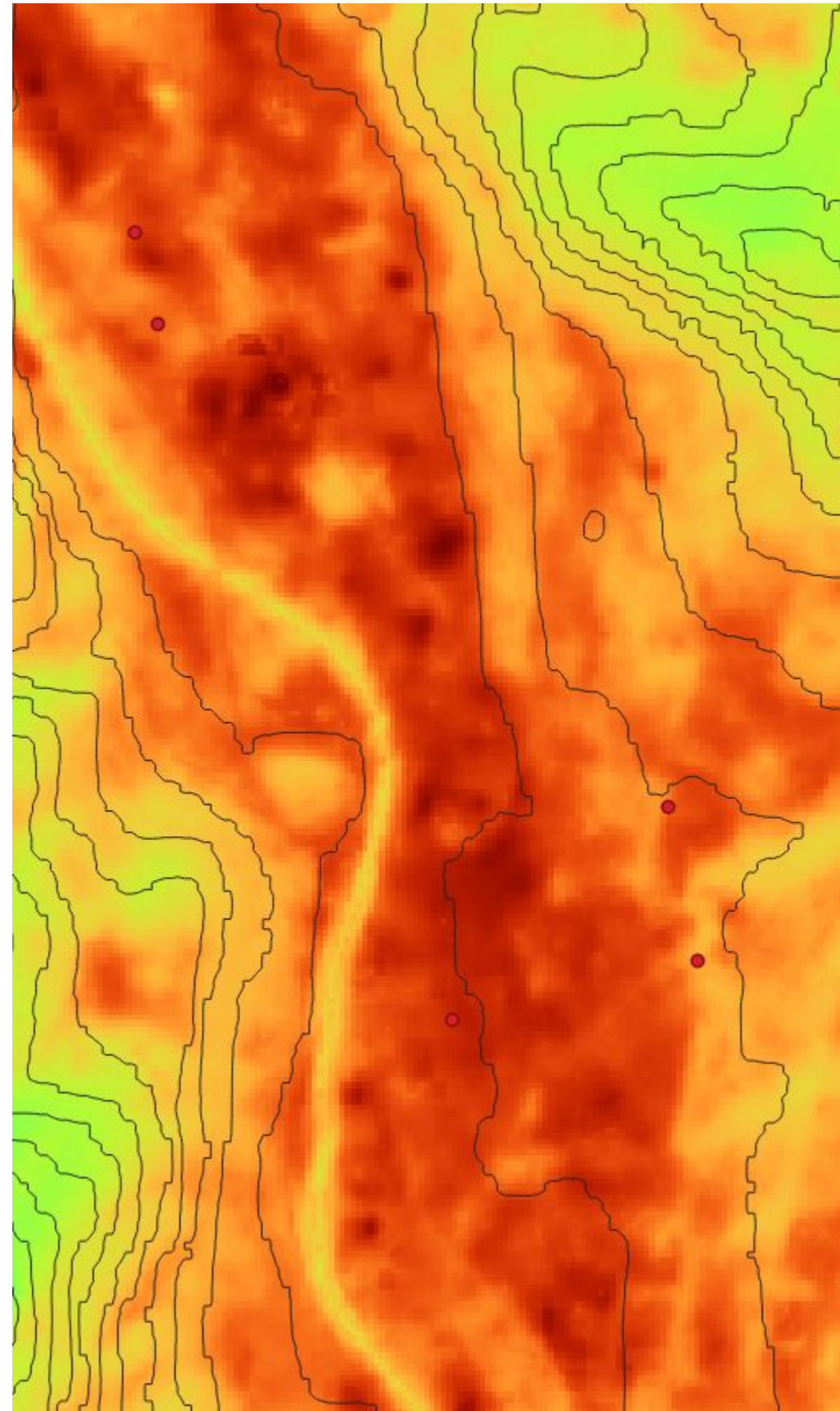
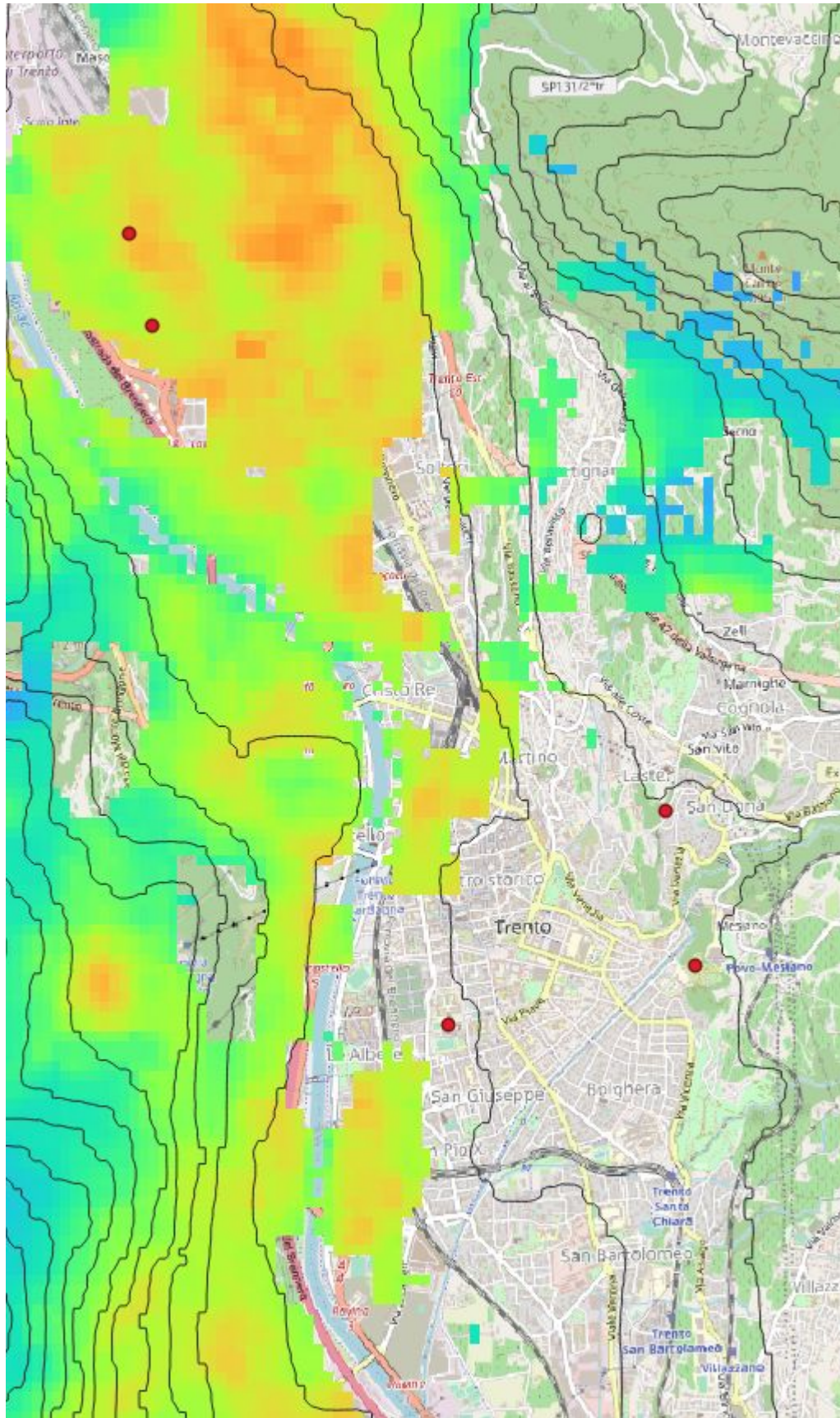
AI within USAGE pilot activities: urban heat island prediction

- **What:** Air temperature play an essential role in UHI prediction but is measured in few location (i.e. red dots)
- **With:** land cover properties, buildings morphology, detail vegetation volumes and satellite optical & thermal images
- **To:** City & neighborhood level *Comfort index & heat wave intensity* prediction 48h ahead.



Fusion of weather stations *time-dense* & **spatially sparse** data with *resolute (30-70m)* LST satellite data that are **time sparse**

AI within USAGE pilot activities: urban heat island prediction



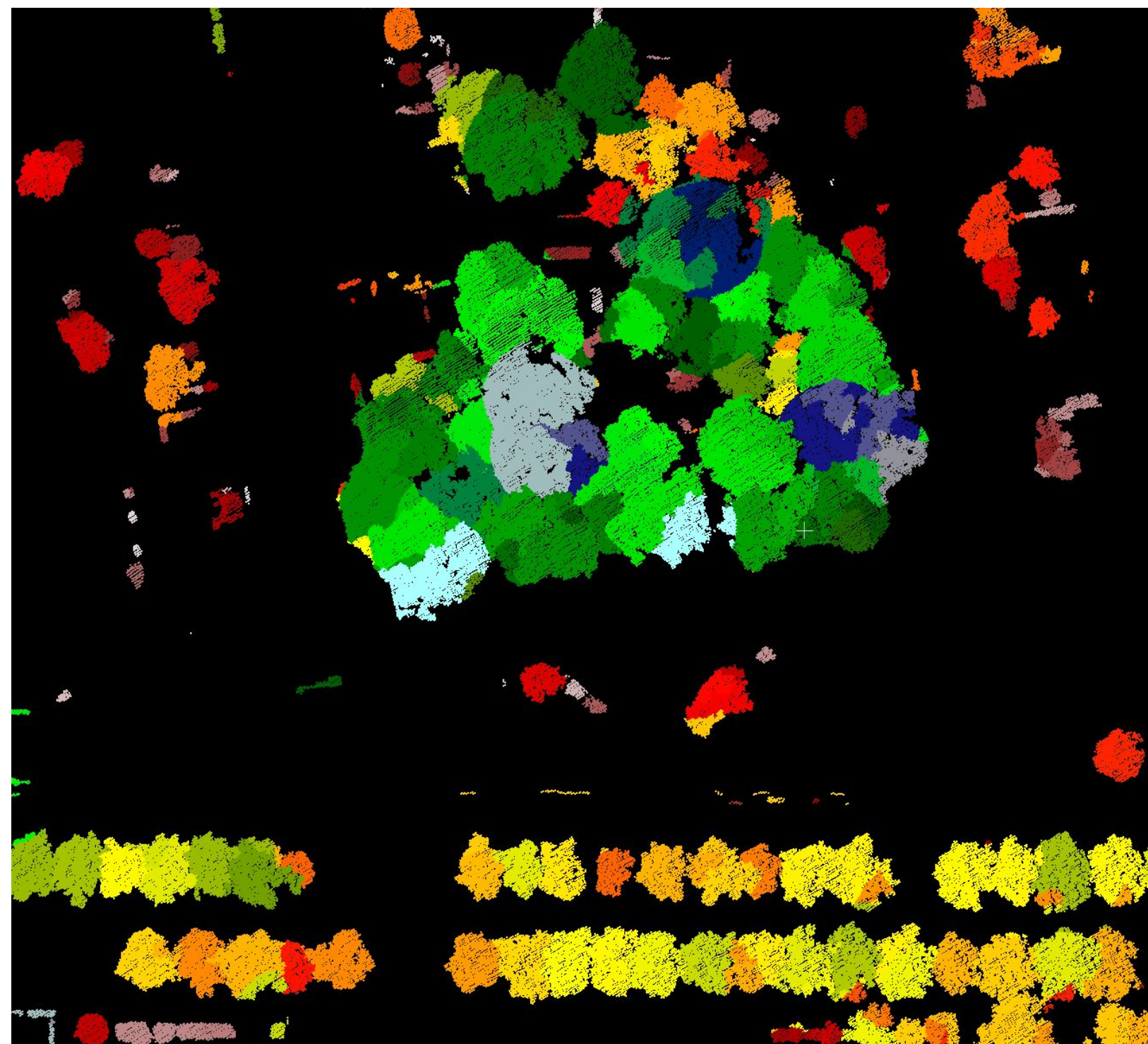
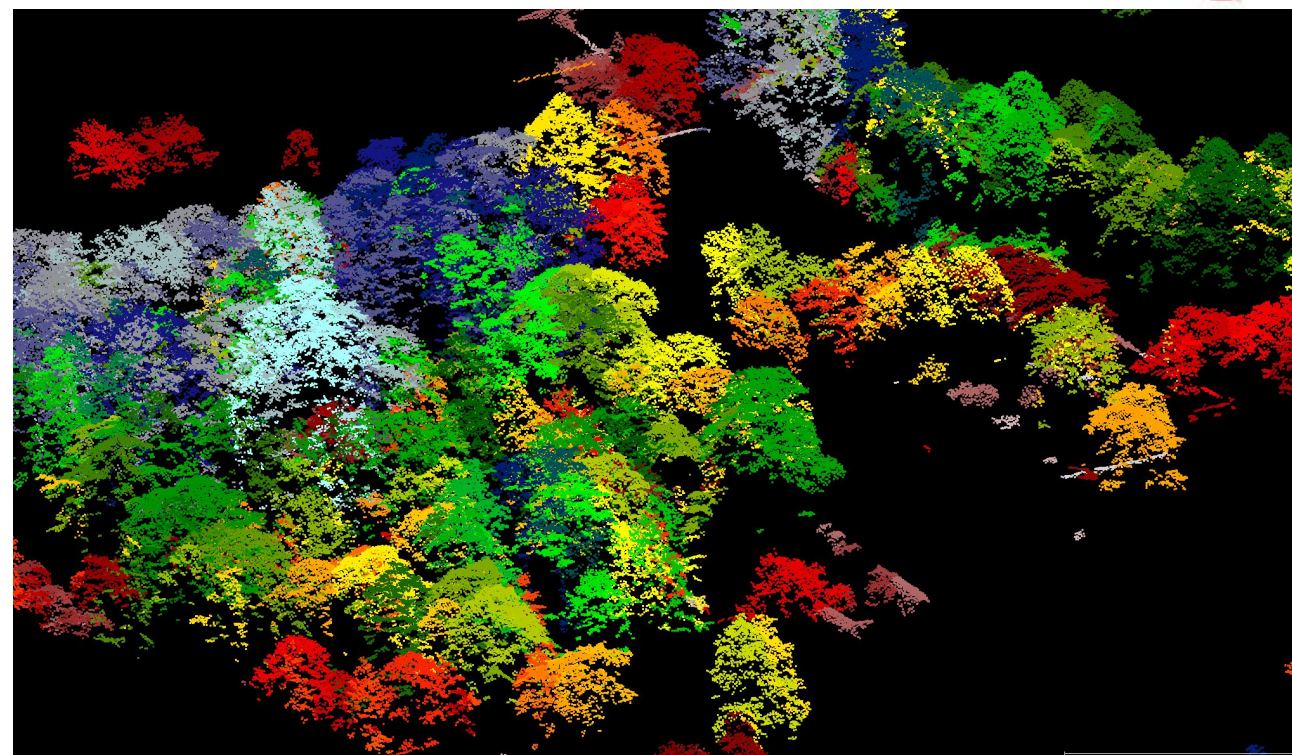
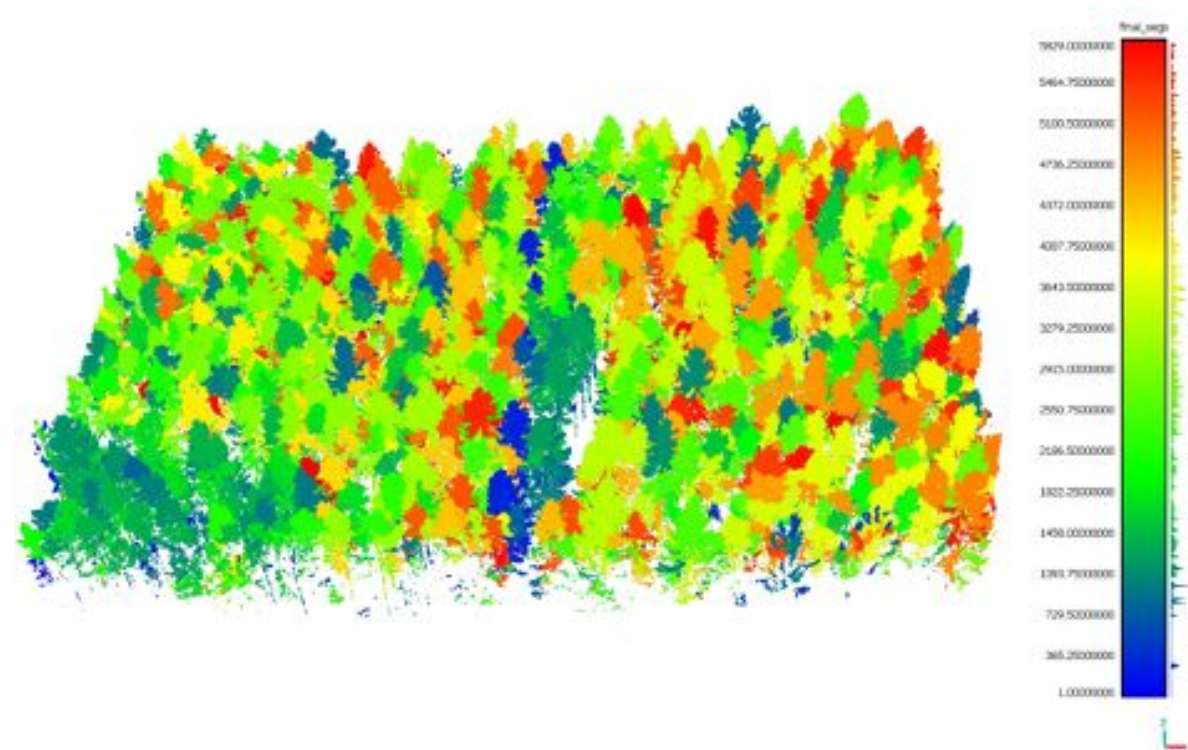
AI within USAGE pilot activities: building footprints detection

- Unet-like method
- Ferrara results



AI within USAGE pilot activities: individual tree segmentation

- color-coded ITS results from airborne LiDAR on Ferrara



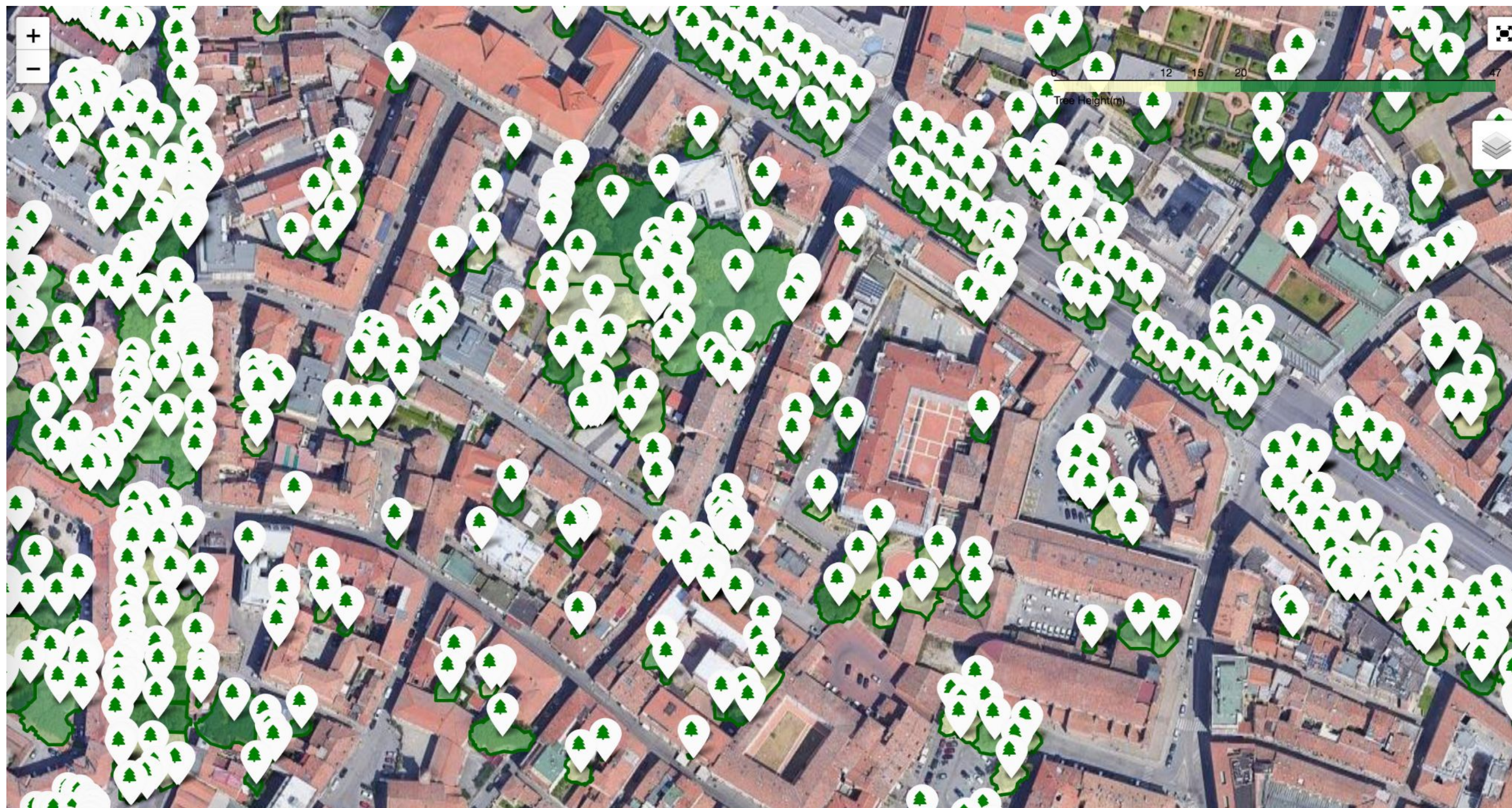
AI within USAGE pilot activities: individual tree segmentation

- GIS-based viz of ITS results from airborne **LiDAR** on Ferrara



AI within USAGE pilot activities: individual tree segmentation

- GIS-based viz of ITS results from airborne **LiDAR + hyperspectral** images on Ferrara



AI within USAGE pilot activities: individual tree segmentation

- GIS-based viz of ITS results from airborne **LiDAR + hyperspectral** images on Ferrara

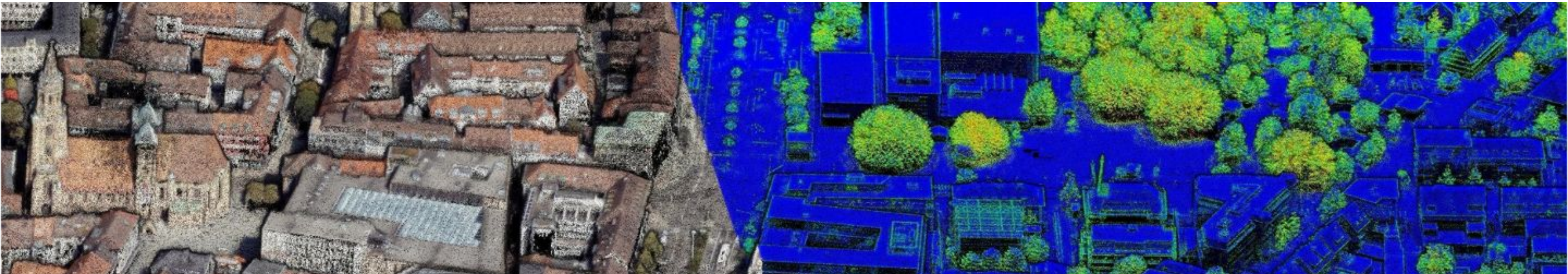


Take away messages

- AI methods are surely **boosting many tasks** in geospatial data processing pipelines
- AI still **suffers** when dealing with large image resolutions and point clouds
- AI should **not** replace traditional (geometric) methods but should **support** them in order to speed up processes and improve results (**complementarity**)
- In the mapping field, the **great variety of datasets** from different sensor types (terrestrial, multi-head aerial cameras, multi/hyperspectral images, LiDAR, SAR, etc.) and scales (terrestrial, UAV, aerial, satellite), leads to **poor availability of training / labelled samples**
- AI methods are rapidly changing but **lack of generalization**, hence it is important for us to have our feet grounded and use common sense, while selecting these technologies

Take away messages

- AI will offer **more valuable solutions for the geospatial field in the near future**, inspiring and impacting research in our field through collaboration with colleagues in neighboring disciplines
- NMCAs are **carefully monitoring** new AI-based solutions and **gradually adopting** them as their potential is huge
- For sure, fully automated general image understanding and geometric processing remain an **elusive problem for many years to come**





thank you!



<https://3dom.fbk.eu/>
<https://www.usage-project.eu/>