

# AI and GIS for Urban Biodiversity Monitoring

## Tree Census from Location Estimation to Ecosystem Services Assessment

Thomas Martinoli<sup>1,2</sup>, Piero Fraternali<sup>1</sup>

<sup>1</sup> Politecnico di Milano, Dipartimento Di Elettronica, Informazione e Bioingegneria (DEIB), Milano, Italy

<sup>2</sup> National Biodiversity Future Center (NBFC), Piazza Marina 61, 90133 Palermo, Italy



CONTACT

### Research Context

**Biodiversity loss** is particularly noticeable in **urban** and **peri-urban areas**, where **urbanization processes** exert significant strain on urban biodiversity and urban ecosystem services (ES). As a result, cities are **less sustainable** and **resilient**, with negative consequences for **public health** and **urban environment quality** [1].

**Urban trees** provide several ES that positively impact urban ecosystems [2]. Therefore, **accurate urban tree inventories** are effective instruments for **properly estimating the ES**.

### Open Challenges

Creating urban tree inventories poses challenges:

- ❖ **Temporal** and **spatial resolution** of data acquisition
- ❖ **Expensive** and **time-consuming** field data collection campaigns
- ❖ A **large amount** of **labeled data** is required to exploit **data-driven approaches** (Deep Learning) to support tree census automation
- ❖ The **ongoing** evolution of cities makes public inventories **incomplete** or **outdated**
- ❖ **Private green areas** are less accurately mapped

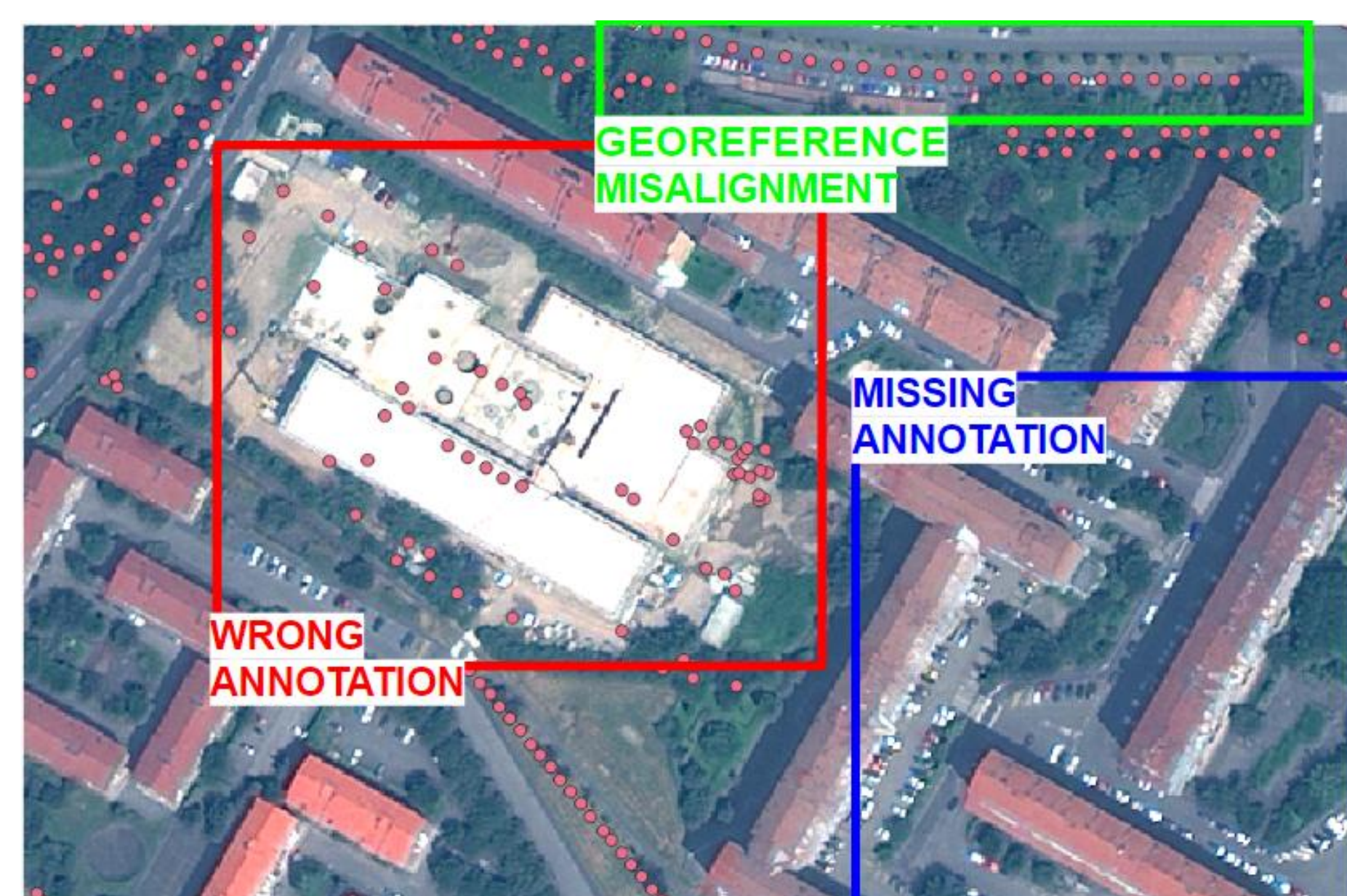


Figure1. Satellite images with tree locations (red points) extrapolated from the tree urban inventory of Milan (2023).

### Data

Very **High Spatial** and **Temporal Resolution (VHR) Satellite products** and **GIS data** can be exploited to describe **fragmented** and **dynamic environments** at scale over time.

- ❖ Pleiades Neo (**PNEO**) product was considered:
  - Six **multispectral** bands with **1.2 m** as spatial resolution;
  - **Panchromatic** band with **30 cm** as spatial resolution;
- ❖ **Tree locations** were collected from the Geoportal of the Lombardy region.

### Preliminary Results

The original model of [2] was evaluated in a study area of the Lombardy region. Annotated images were analysed and **Ground Truth (GT)** was manually corrected. The best results are obtained by **Fine-Tuning (FT)** the model in the study area.

Experiment	Training Configuration		Test set Evaluation		
	# Images	# tree	Precision	Recall	F-Score
Original Model [2]	1651	95972	0.587	0.451	0.510
FT Model GT Uncorrected	426	24477	0.799	0.420	0.550
FT Model GT corrected	218	17927	<b>0.670</b>	<b>0.660</b>	<b>0.665</b>

Table1. Training configuration and the results of the experiments performed, evaluated on the same test set.

### Conclusions

**DL** combined with **VHR Satellite data** can support **tree census** in **urban** environments. However, the **geographical context**, the **tree crown size** and the presence of **dense groups of trees** influence the performance of the models.

### Methods

**Deep Learning (DL)**, combined with **Remote Sensing (RS)** and **GIS data**, can be leveraged to **automate operations** for **monitoring urban environments at scale**.

In the early stages of the study, DL models are exploited to **localize** and **count** urban trees within the study area.

Given an image, trees are identified by employing a **binary segmentation** task [2].



Figure2. Confidence Map (model output)



Figure3. Prediction extrapolated from Confidence Map

### Next Steps

**New data** covering the whole Milan metropolitan area have been acquired to improve and test the performances of DL models at scale. Other **DL architectures** to **segment** and **detect trees** will be explored. **DL models** for **tree species** classification will be investigated.

[1] Yanhua Chen et al; Urban Remote Sensing: Monitoring, Synthesis, and Modeling in the Urban Environment, Chapter 20; (2021)

[2] Ventura et al, Individual tree detection in large-scale urban environments using high-resolution multispectral imagery, (2024)

