

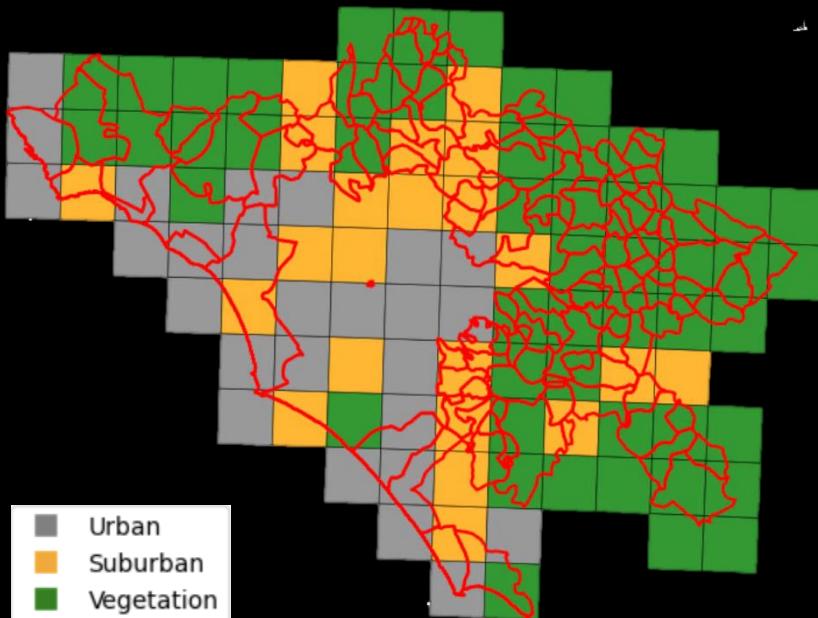
# GENHACK CHALLENGE: THE URBAN HEAT ISLAND EFFECT

Focus: visualize and explain the Urban Heat Island effect in Rome

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# TASK 1: MULTI-SCALE LAND COVER ZONATION

**OBJECTIVE:** Classification of the Rome Province into distinct climatological zones (Urban, Suburban and Vegetation) to spatially aggregate high-resolution surface data into the ERA5-Land grid (9km×9km).



## Dataset

### GADM 4.1

Administrative boundaries (Rome Province).

### Sentinel-2 (L2A):

High-resolution multispectral imagery (resampled to 80mx80m) for Normalized Difference Vegetation Index (NDVI) computation.

### ERA5-Land

Provides the target geometry ( $0.1^\circ \times 0.1^\circ$  grid) for the final upscaling process

# Resolving Spectral Ambiguity via Data Fusion

Deep water and dense urban surfaces both exhibit low NDVI values, causing spectral confusion and potential river misclassification.

## Spectral Ambiguity



## Sentinel-2 Acquisition

We derive NDVI from summer Sentinel-2 imagery (80m x 80m) to serve as the primary metric for land cover segmentation.



## Temporal Integration

We integrate JRC Global Surface Water data, leveraging its 37-year history to validate water presence independent of spectral noise.



## Enhanced Zonation

This data fusion accurately separates hydrography from urban areas, creating a robust baseline for the subsequent grid aggregation.

A priority mask overrides NDVI: pixels with >10% historical water occurrence are deterministically classified as "Water".

## Priority Masking

# Classification Logic: From Pixel (80m80m) to Macro-Grid (9km9km)

## Micro-scale Classification (80m x 80m)

Pixels are categorized using an handmade priority-based decision tree:

- **Water:**  $J_{occurrence} > 10\%$  or  $NDVI \leq -0.8$
- **Urban:**  $-0.8 \leq NDVI \leq 0.33$
- **Park/Grassland:**  $0.33 < NDVI \leq 0.6$
- **Forest:**  $NDVI > 0.60$

## Upscaling to Macro-Grid (9km x 9km)

Grid cells are assigned a dominant class based on internal pixel composition ( $P$ ):  
Let  $P_{nature} = P_{grassland} + P_{water}$

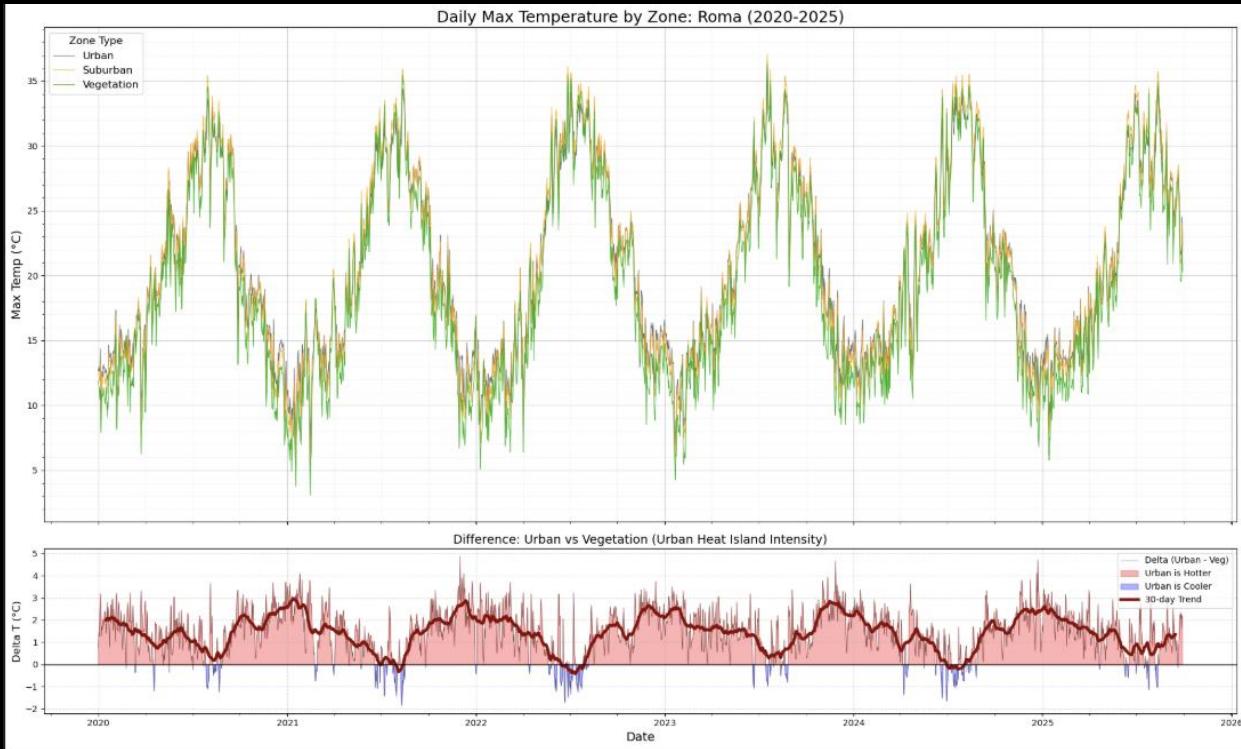
IF ( $P_{nature} > 70\%$ ) then **VEGETATION**

ELIF ( $P_{forest} > P_{urban}$ ) and  $P_{forest} > 30\%$   
then **VEGETATION**

ELIF  $P_{urban} > P_{nature}$  or  $urban > 40\%$   
then **URBAN**

ELSE **SUBURBAN**

# Task 3: Temporal Dynamics

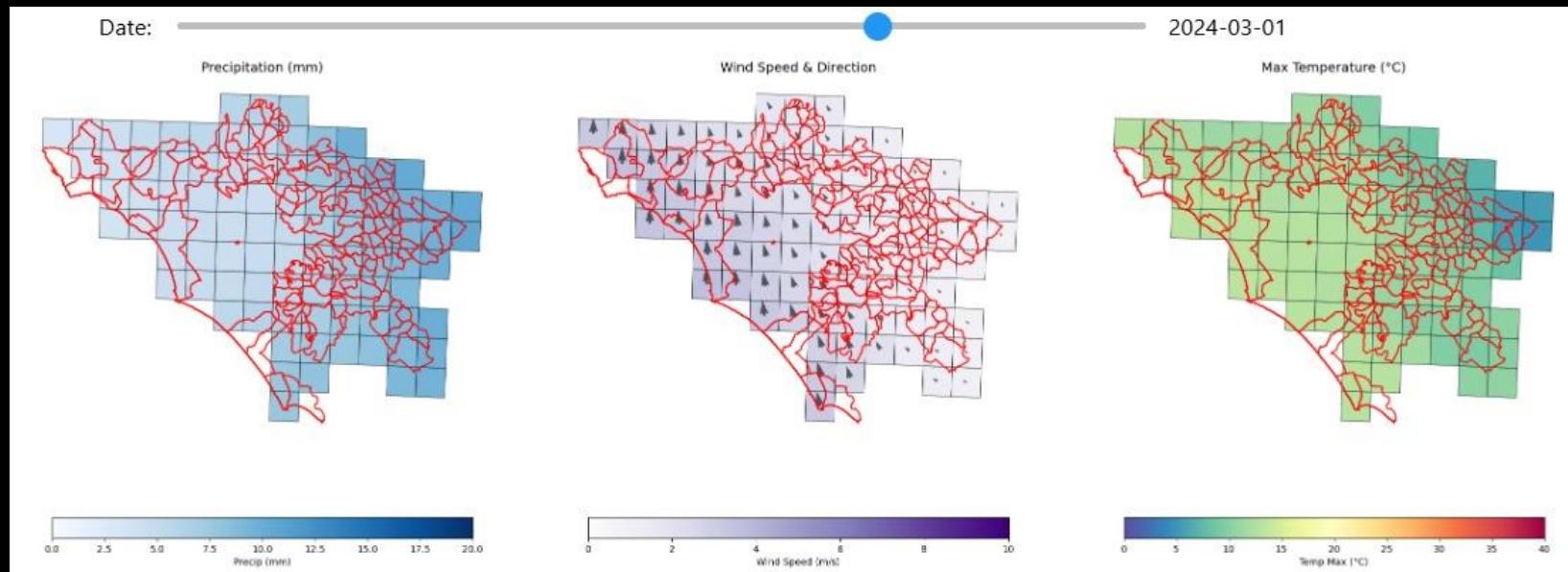


**Spatial Aggregation:**  
Daily maximum temperatures ( $T_{max}$ ) from ERA5-Land were averaged spatially for each classified zone

**Differential Metric** To isolate the anthropogenic effect, we computed the instantaneous delta:  
$$\Delta_{UHI} = T_{Urban} - T_{Vegetation}$$

# Task 2: Interactive Spatio-temporal Climate Inspector

An interactive dashboard developed to explore the temporal dynamics of the ERA5-Land dataset. The tool visualizes daily spatial distributions of **Total Precipitation**, **Wind Vectors**, and **Maximum Temperature** across the 9km x 9km grid, allowing for the granular inspection of specific meteorological events and the validation of data consistency prior to statistical aggregation.



# Modeling spatial temperature Anomalies

$$\Delta T_{Anomaly} = -65.95 + 0.67(\%Urban) + 0.64(\%Forest) + 0.66(\%Water) + 0.11(Wind) - 0.03(Precip)$$



## Objective

Predict local  
temperature deviation  
(Anomaly) for every 9  
 $Km^2$

$$\Delta T_{Anomaly} = T_{cell} - T_{CityAverage}$$



## Interpretation

Under identical weather  
conditions, a zone  
classified as Urban is,  
on average,  $1.23^\circ C$   
hotter than a zone  
classified as Vegetation.



## Accuracy

$R^2 = 0.400$  :  
explains 2x more  
variance than simple  
zoning labels.