

**XII giornata sulla modellistica in aria(net)**  
**Milano, 26 marzo 2025**

# **CoKer [Convolutor of dispersion **K**ernels]**

A microscale PMSS-based tool for **Digital Twins**

Application on the city of **Taranto** in the context of **CALLIOPE** project

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# The Digital Twin of Taranto: CALLIOPE project

CTE (*Casa delle Tecnologie Emergenti*) **CALLIOPE** is a research project financed by the **Ministry of Enterprises**

## The scope of the CALLIOPE project

A sandbox platform to study and understand the relationships between environmental, chemical, physical agents and the **human health** in an **urban** context (city of Taranto).

## The role of ARIANET

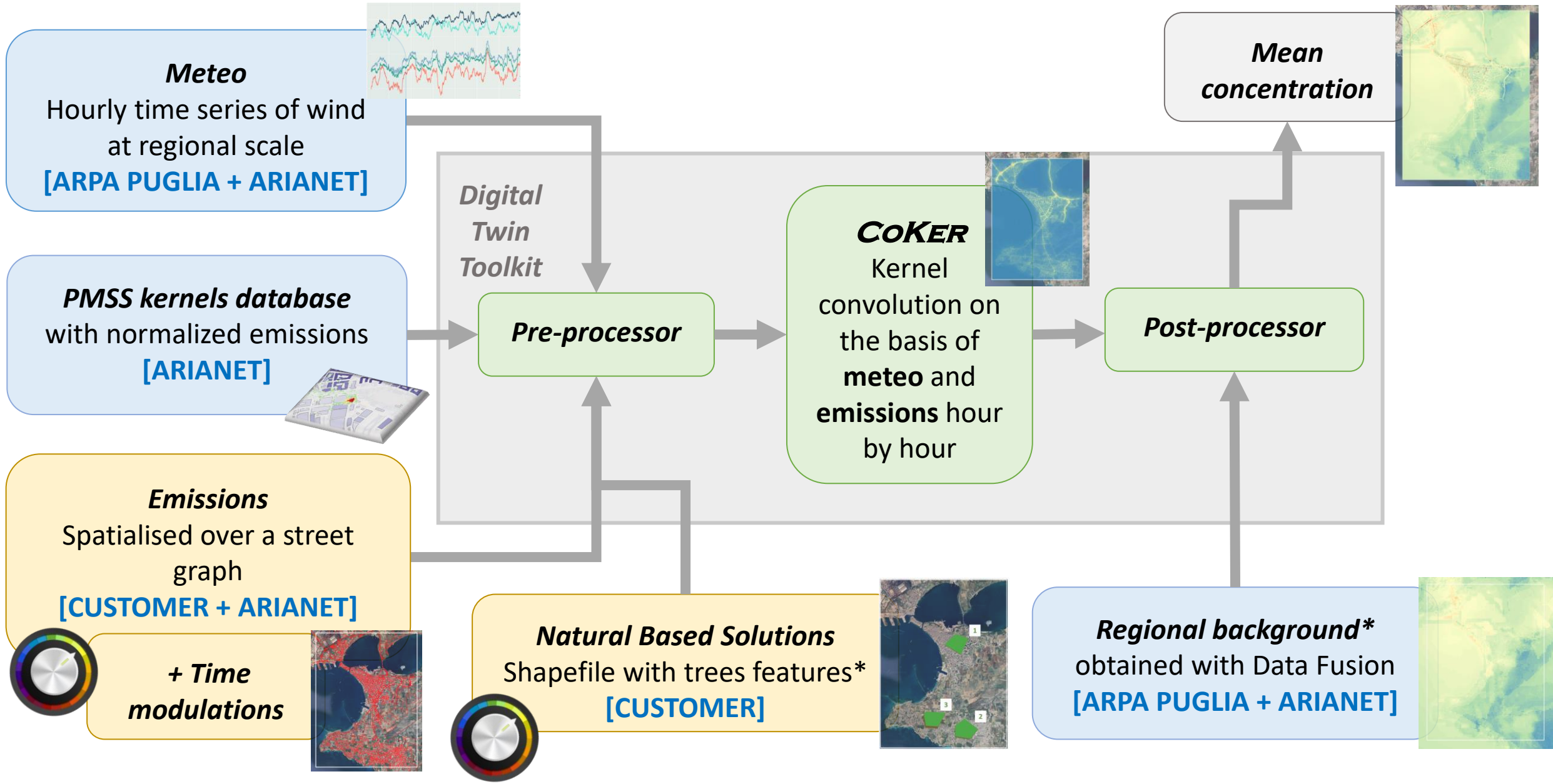
- Develop a **digital twin toolkit** of the city of Taranto to assess multiple exposition factors and eventually preserve the population from **sanitary risks**
- The toolkit enables the realization of **“what if”** traffic emission scenarios, allowing end users to apply mitigation factors, such as **variations in the road traffic** or the adoption of **natural based solutions (NBS)**
- Contextually, provide regional concentration background fields at 100m with data fusion techniques, for epidemiologic studies

### Microscale domain

Lx = 9 km - Ly = 12 km  
Grid size = 5 m



# The Digital Twin: a tool for traffic emission scenarios



# Normalized dispersion kernels construction

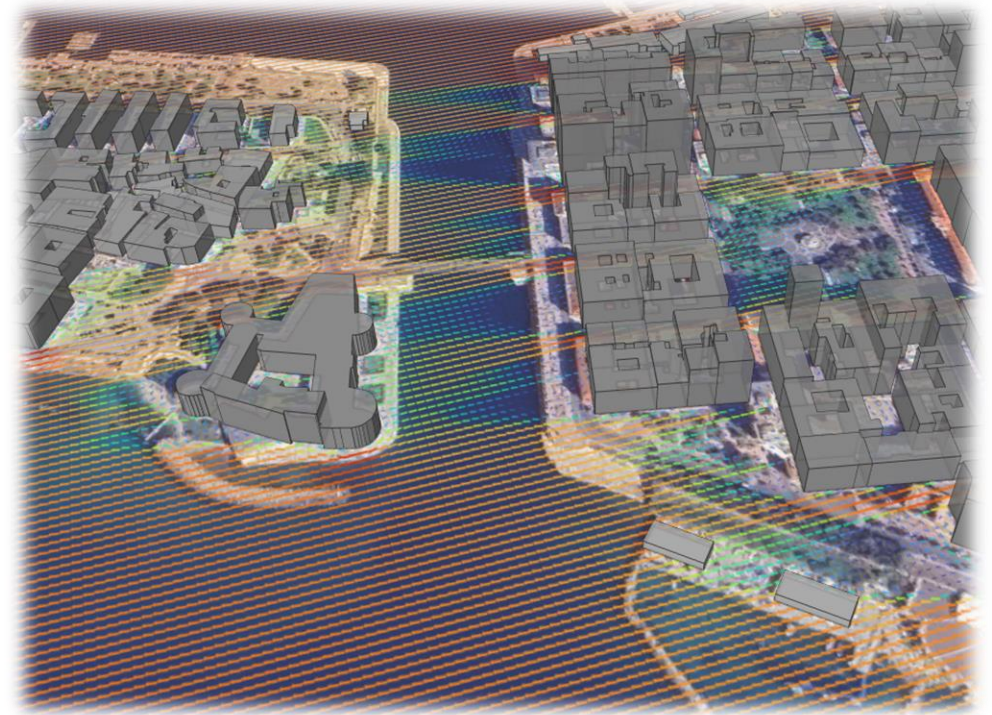
## Inlet wind profiles

6 wind speeds at 10m  
8 wind directions at 10m  
5 stability classes  
**240 combinations**

## 3D Meteo downscaling

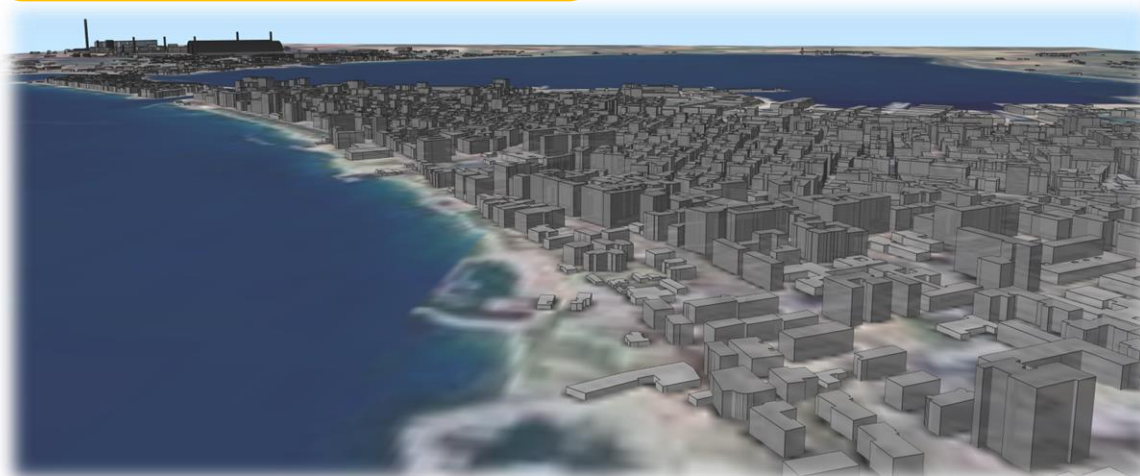
### *PSWIFT*

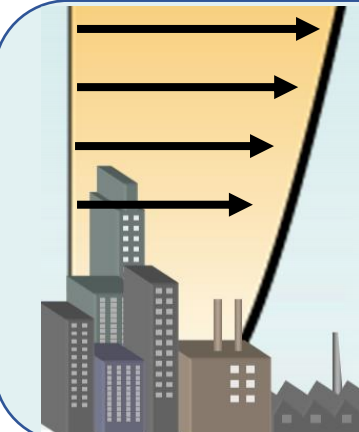
Adapt the wind profiles  
to buildings  
**240 simulations**



## Buildings

Vectorial Shapefile




$$U = U_{10} \left( \frac{z}{10} \right)^\alpha$$

Wind profiles are built raising the wind at 10 m using **power laws** which depend on **stability** and imposing **urban surface roughness**

# Normalized dispersion kernels construction

## Inlet wind profiles\*

6 wind speeds  
8 wind directions  
5 stability classes  
**240 combinations**

## 3D Meteo downscaling *PSWIFT*

Adapt the wind profiles to buildings  
**240 simulations**

## 3D Dispersion simulations *PSPRAY*

Stationary dispersion from traffic emission at microscale  
**240 simulations**

*PMSS  
DRIVEN  
KERNELS*

## Buildings

Vectorial Shapefile

Particles are saved in a buffer  
2km around each

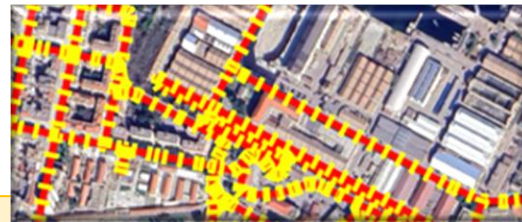
## Dispersion

A ground cover for each road edge for each meteorological combination

**6 millions of files (150 GB)**



**Street Graph**  
4000 polylines



**Discretized graph**  
26000 road segments with normalized emissions



# Dispersion kernels at microscale – meteo combinations

Dispersion kernels are realized performing 3D **PMSS** simulations for each **combinations** of inlet meteorological variables\*

Stationary regime  
Normalized emissions

Variable	Number of classes	Classes
<b>Wind direction</b> [degrees from North]	<b>8</b>	0° - 45° - 90° - 135° - 180° - 225° - 270° - 315 °
<b>Wind speed at 10 m</b> [m/s]	<b>6</b>	1 – 2 – 3 – 5 – 7 – 9
<b>Atmospheric stability class</b> [Pasquill]	<b>5</b>	A – B – C – D – E/F

For each meteorological combination, we extract the contribution of each **source** to the **ground concentration field** in a **square buffer of side 2km** around the segment baricenter

Total meteo combination:  $8 \times 6 \times 5 = 240$   
Sources number: around **26000**



about **6 millions** kernels  
(concentration field at ground for street segment and meteo combination)

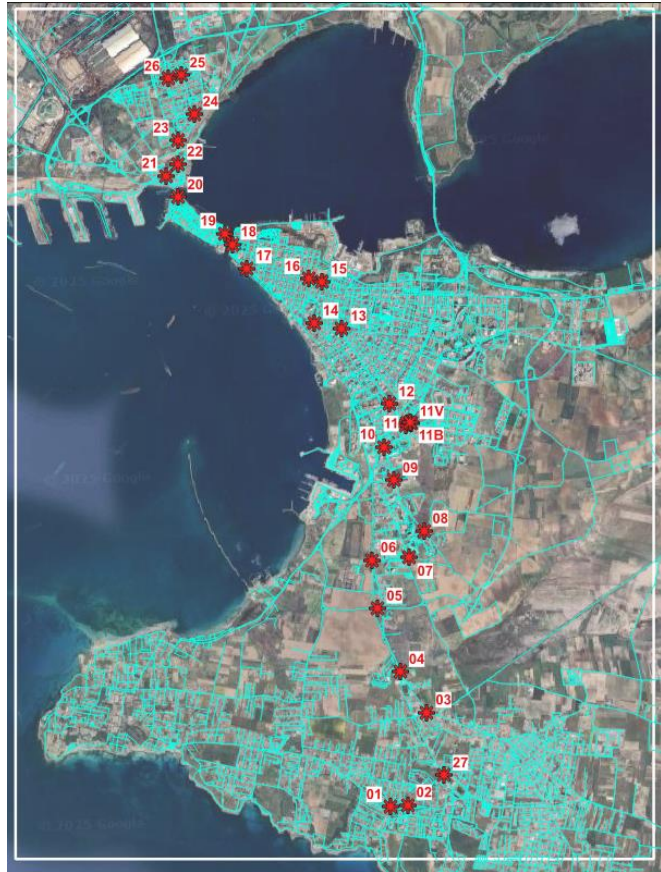
The kernel database built in such way weights about **180 GB**.

# Dispersion kernels at microscale – examples

Kernel examples, with fixed source e by varying of meteorological conditions\*



\* Fields cropped in a domain 600x400 m around a source are displayed to highlight the plume at the source, but kernels extend by 2000x2000 m  
Buildings are displayed in purple scale, with intensity proportional to height

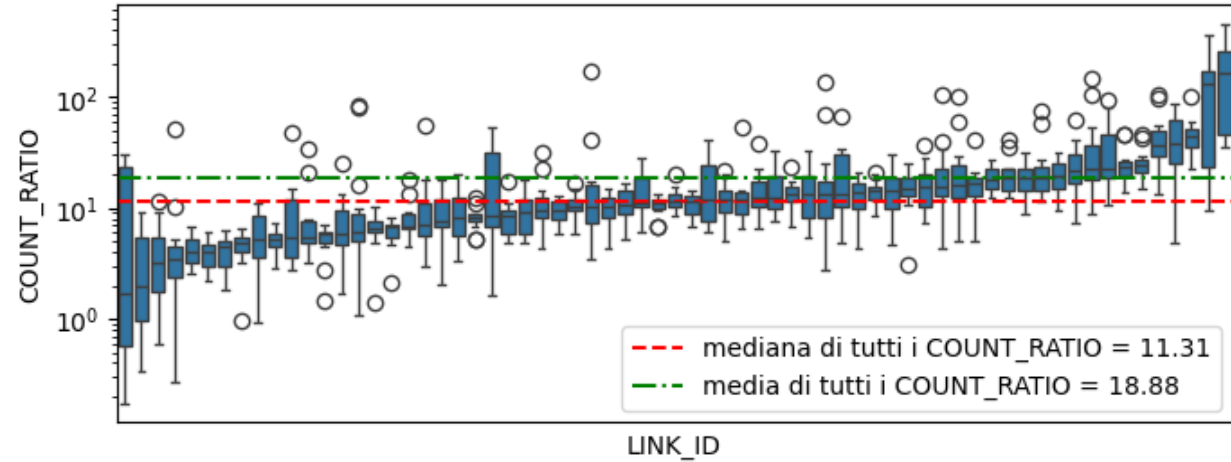


## HERE Historical data

- **Vehicle counts**
- **Vehicle speeds**
- Frequency 5 mins
- Years 2022/2023
- Spatialized over the street graph

## Traffic light observations

- **Vehicle counts**
- Hourly frequency
- One week (jan 2024)
- Localized over few streets



- Average **HERE** observations and **traffic light observations** have been compared in the same reference week (but different years)
- The ratio between traffic light and HERE observations shows a **homogeneous bias** with median equal to **11**.
- Such bias is used to rescale HERE counts over the whole street network

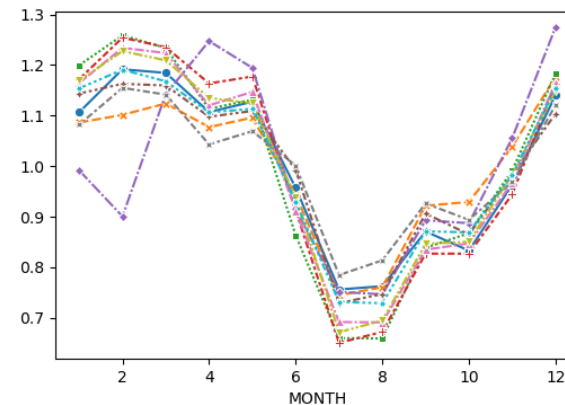
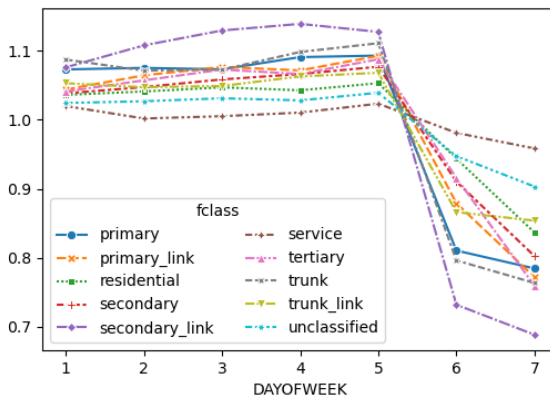
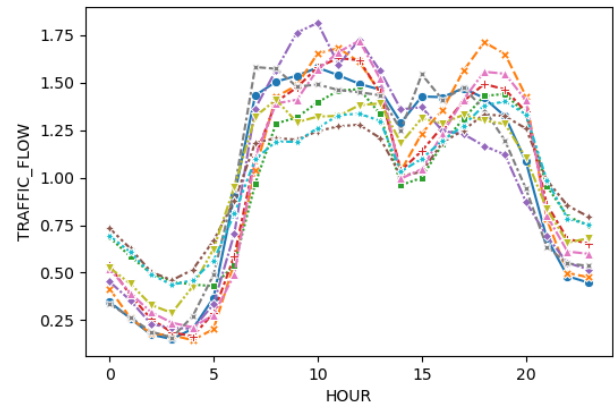
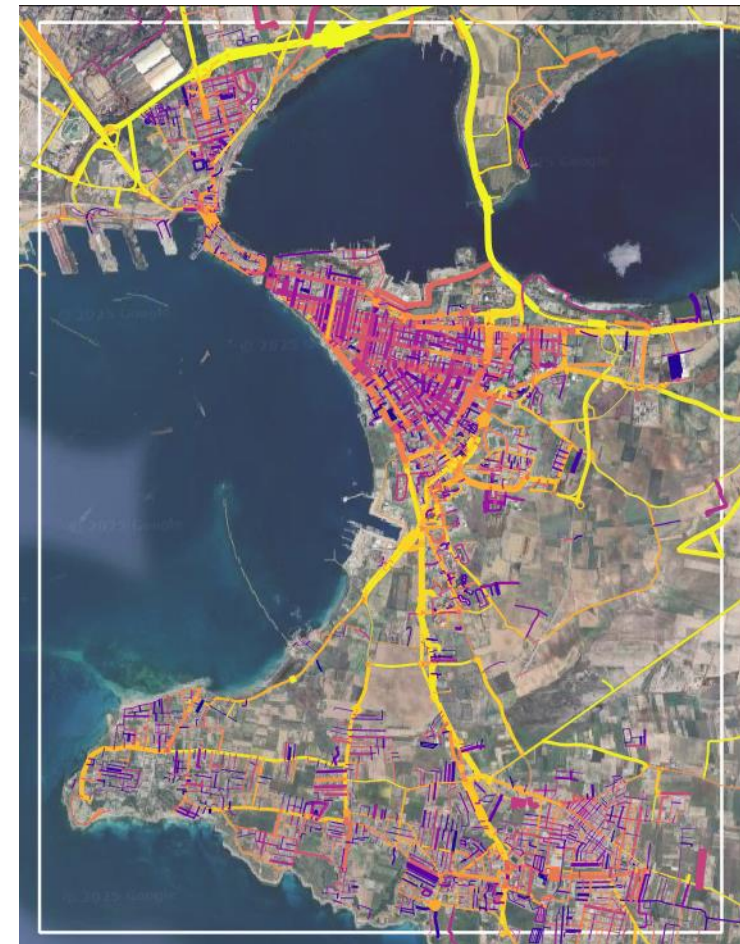
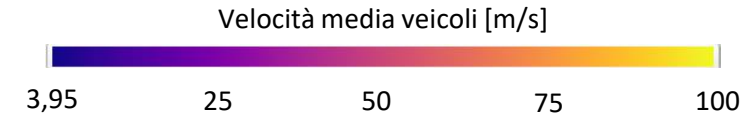
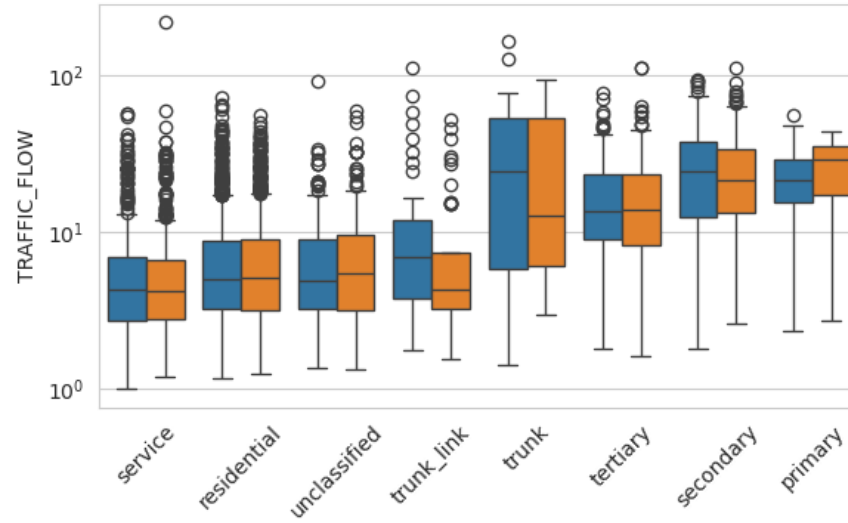
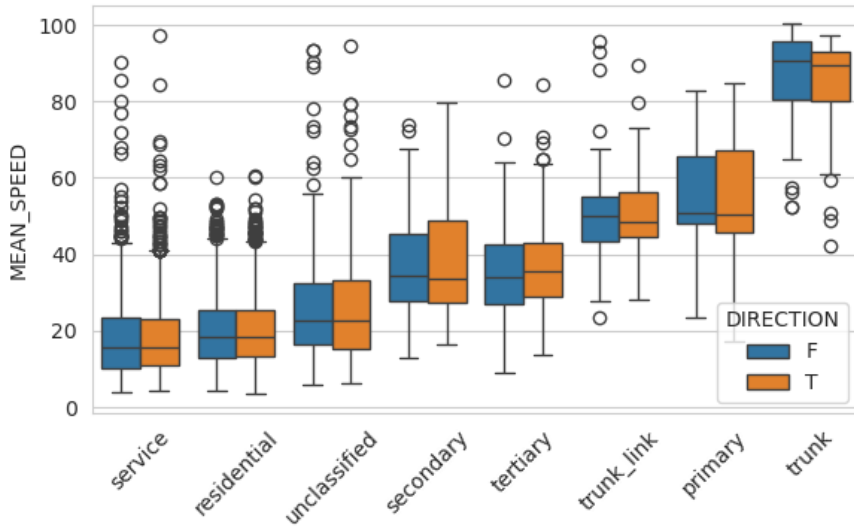
- Streets with HERE measurements
- Traffic lights



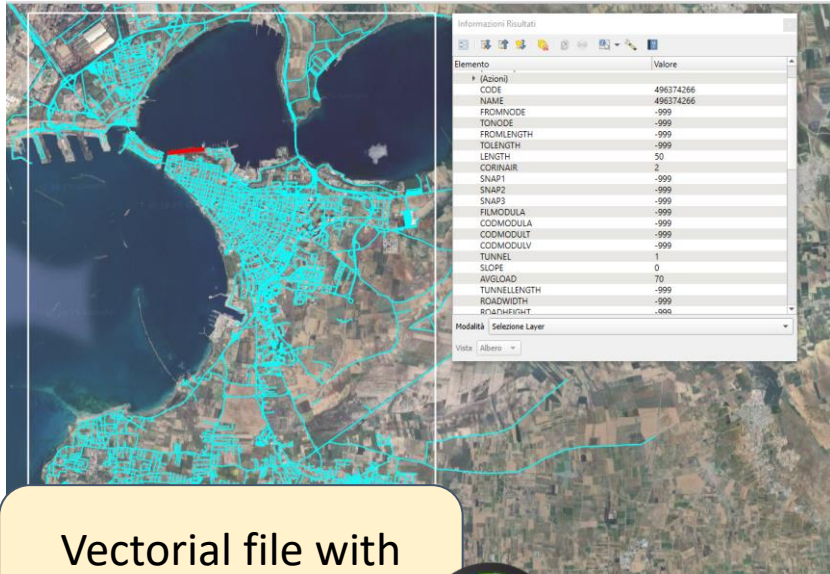
# Traffic emissions: Distribution of traffic flows and speeds

HERE data, properly re-calibrated are used to build the input of the emission model (TREFIC)

Flows and speeds are *averaged across time* and spatialized over the OSM network.  
For each street type, **time modulations** are computed.



# Traffic emission: calculation with TREFIC

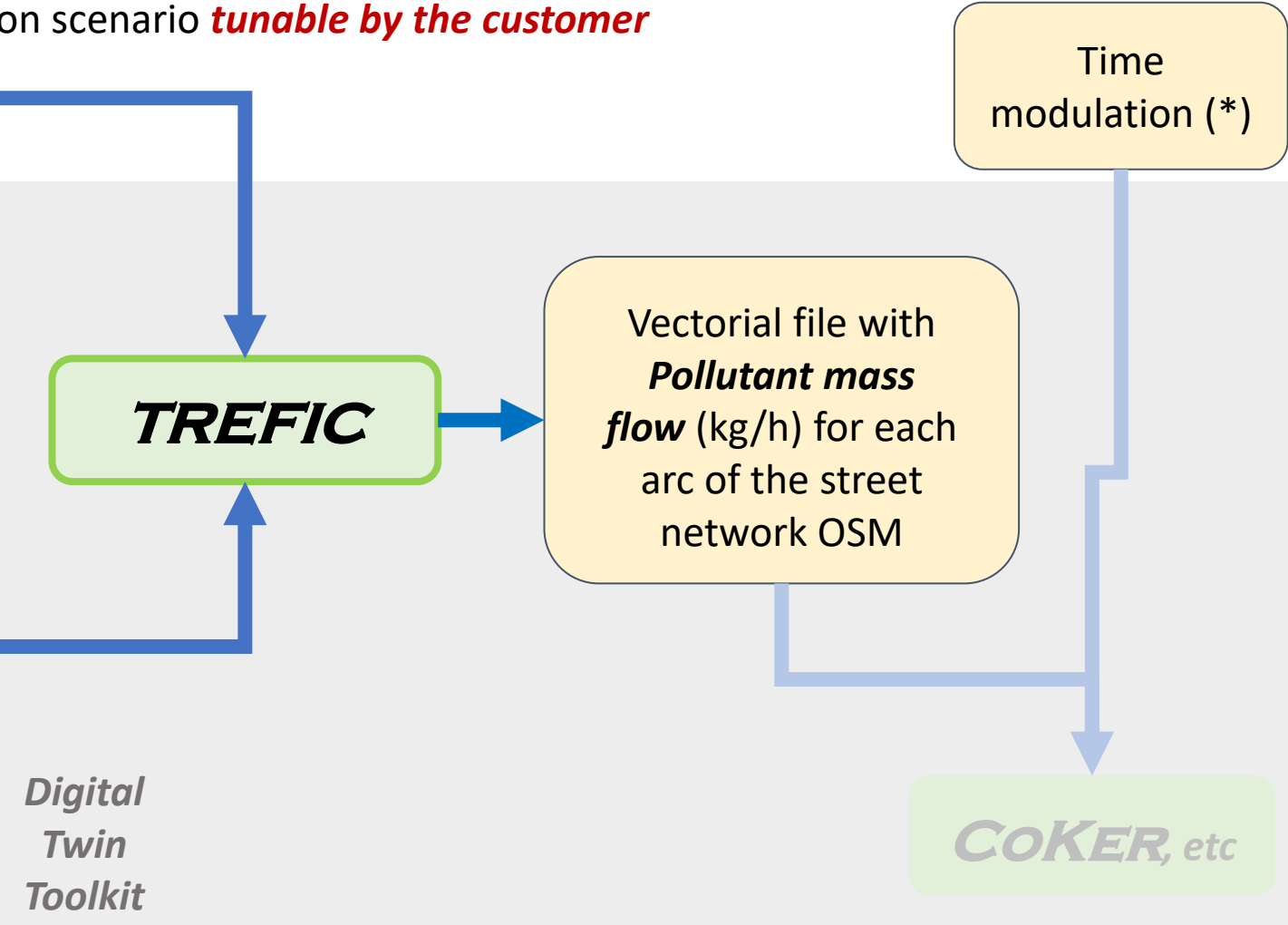


Vectorial file with *flows and speeds*

Traffic flows and speeds are formatted in a **vectorial format (.mif)**. Together with the **vehicular fleet**, they are an input for the emission scenario *tunable by the customer*

		urb	%
* cars			
1	Petrol Mini Euro 4	0	0.00
2	Petrol Mini Euro 5	0	0.00
3	Petrol Mini Euro 6 up to 2016	0	0.00
4	Petrol Mini Euro 6 2017-2019	0	0.00
5	Petrol Mini Euro 6 2020+	0	0.00
...	.....	.....	.....
36	Petrol Large-SUV-Executive PRE ECE	0	0
37	Petrol Large-SUV-Executive ECE 15/00-01	0	0
38	Petrol Large-SUV-Executive ECE 15/02	0	0
39	Petrol Large-SUV-Executive ECE 15/03	0	0
40	Petrol Large-SUV-Executive ECE 15/04	0	0
41	Petrol Large-SUV-Executive Euro 1	0	0
42	Petrol Large-SUV-Executive Euro 2	0	0
	Large-SUV-Executive Euro 3	21062269	0.1
	Large-SUV-Executive Euro 4	53122291	0.25
	Large-SUV-Executive Euro 5	14244499	0.07
	Large-SUV-Executive Euro 6 up to 2016	13115574	0.06

Table file with *Vehicular fleet*

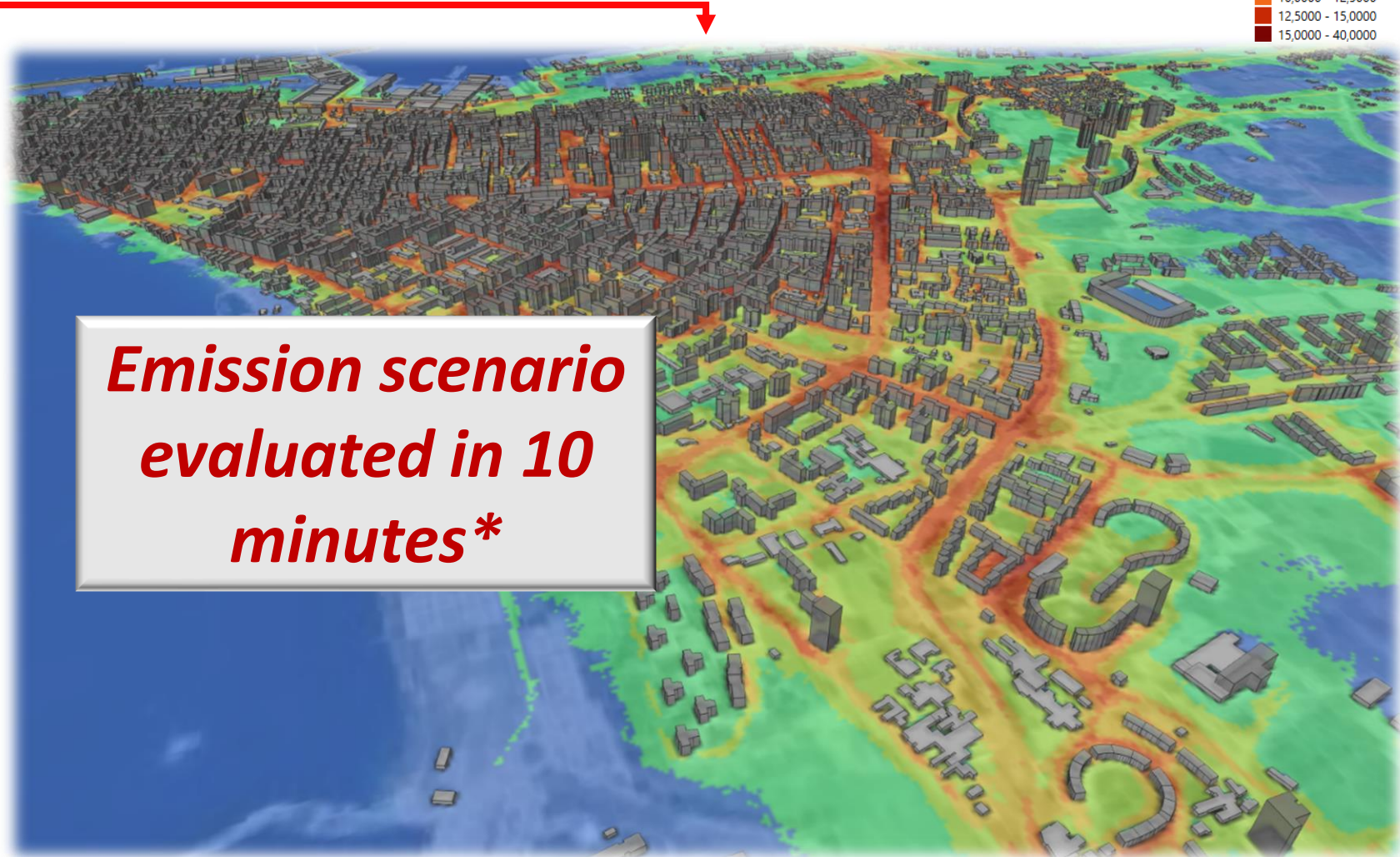
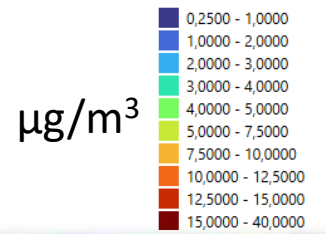


(\*) Time modulation for different road categories are another tunable input

# Output of CoKer - Mean NO2 annual average at microscale (2023)

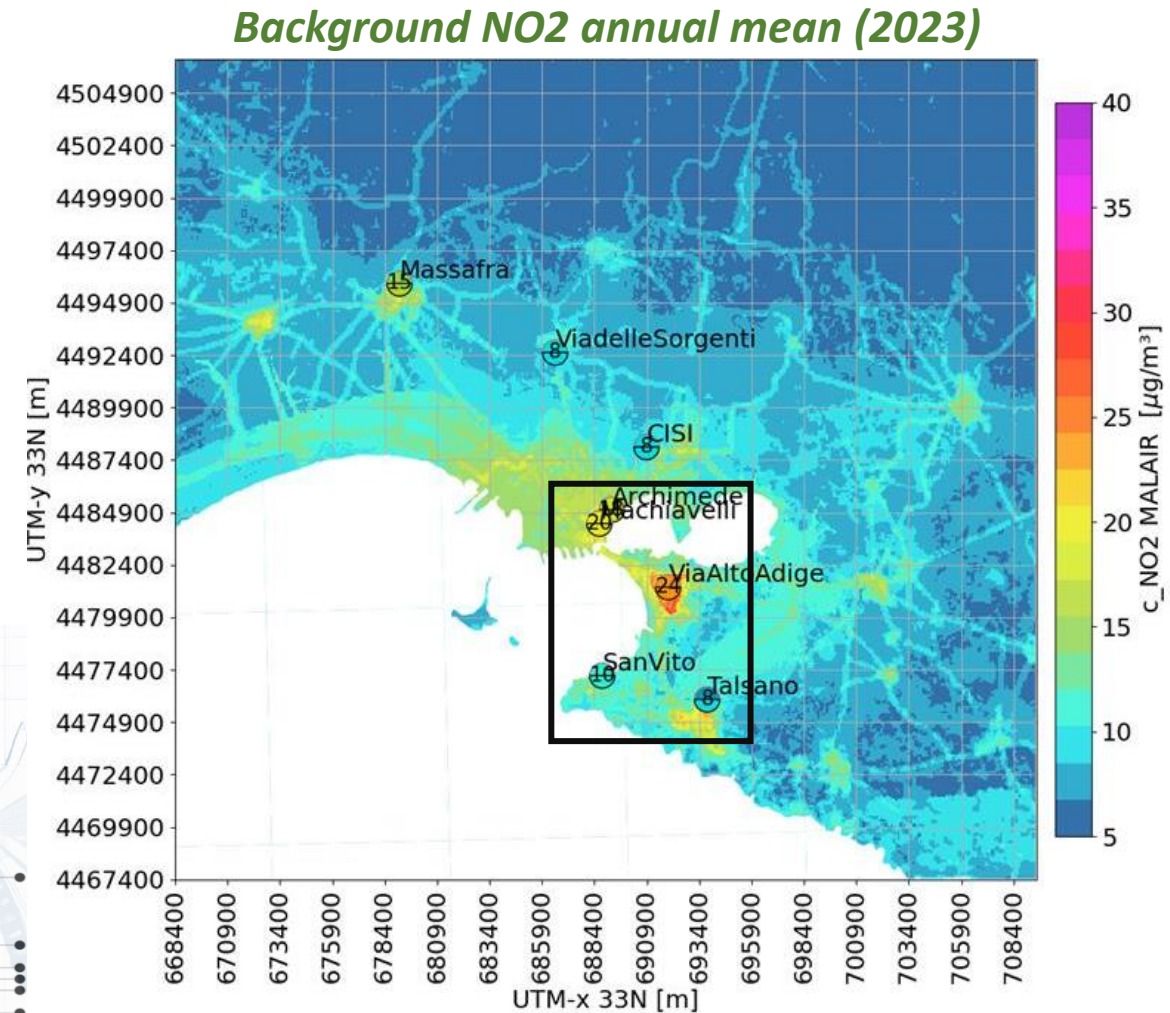
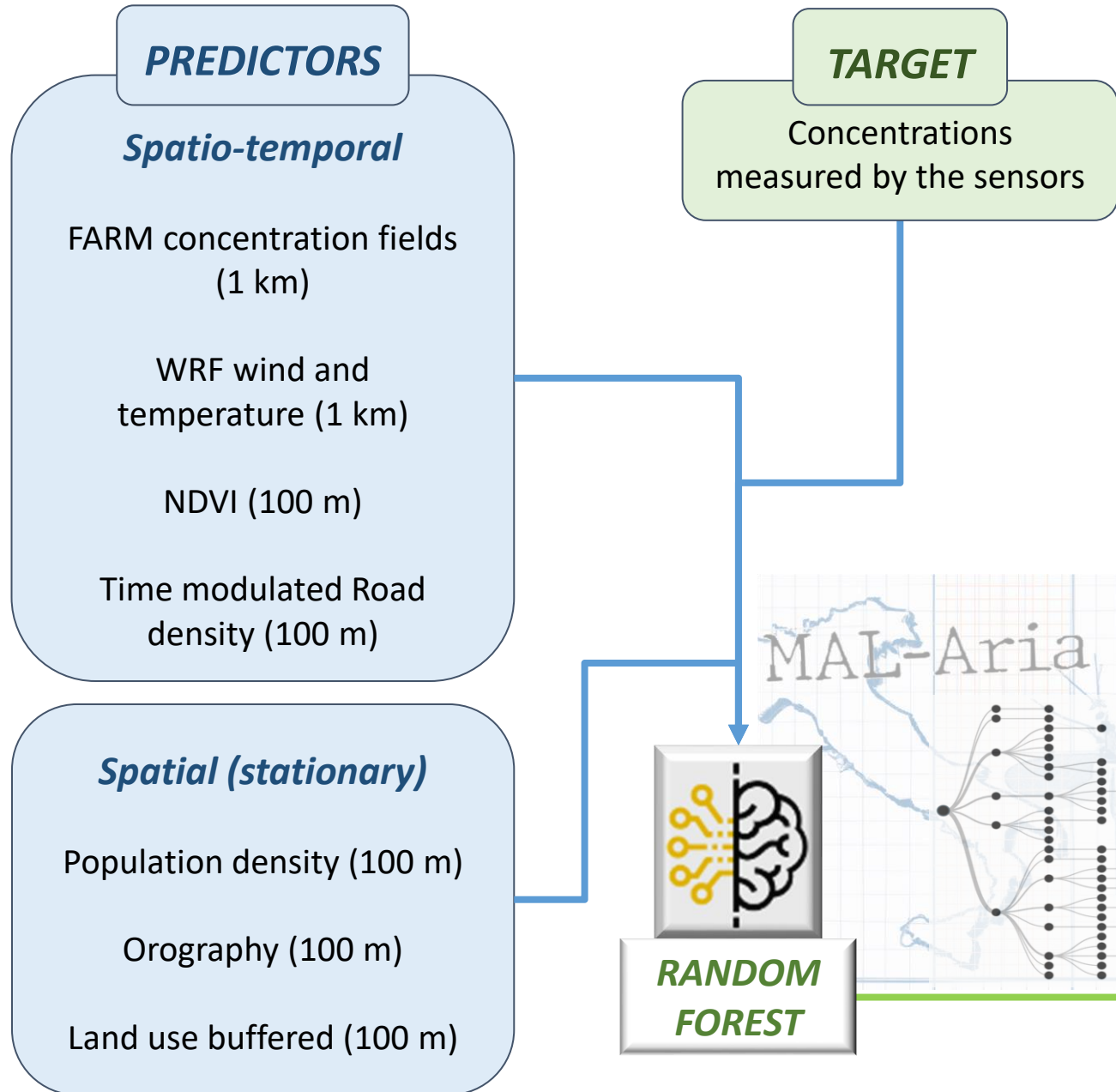
Hour by hour, kernels are **selected from the database** for each source on the basis of the driving *meteorology*, and rescaled according to *emissions*.

The convolution and the following time average give rise to the final concentration field.



[\*] OpenMP parallelization, with 192 Threads and 200+ GB RAM. For long periods, loading all kernels in memory improves a lot the performance. If not enough RAM is available, 1-year simulation may be 5 times slower.

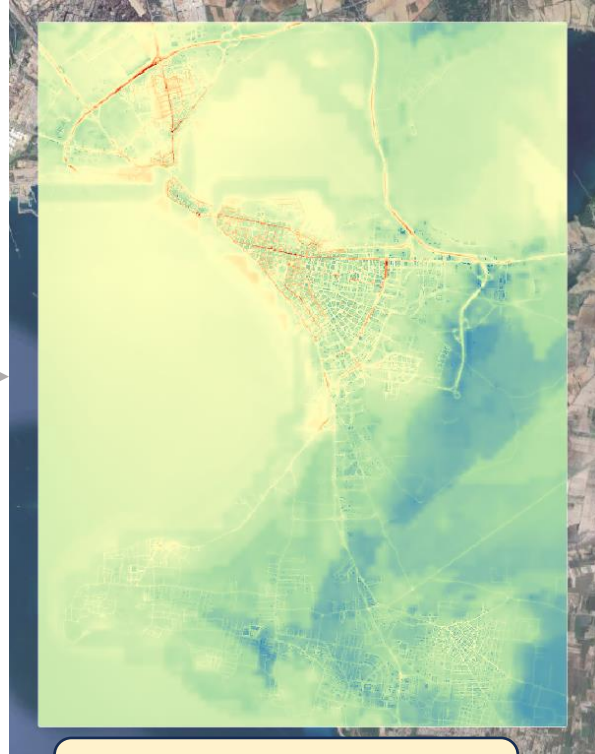
# Data fusion with Random Forest for the regional background



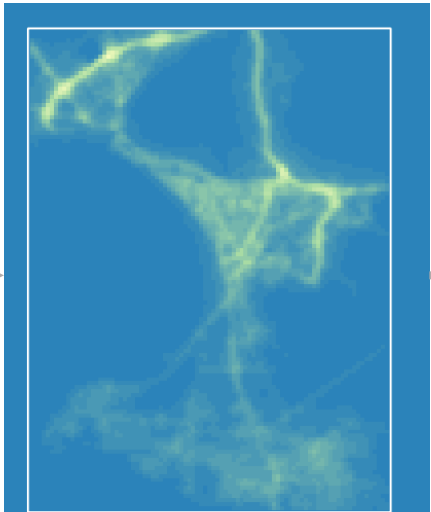
# Merge between microscale and regional background



**A**  
Microscale convolution [5m]



**F**  
Sum A + E [5m + 100m]

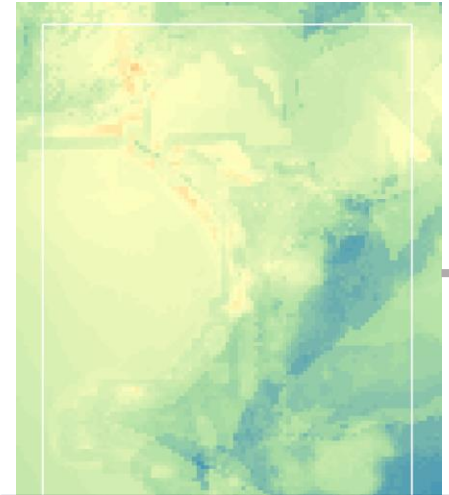


**C**  
Microscale regrided [100m]

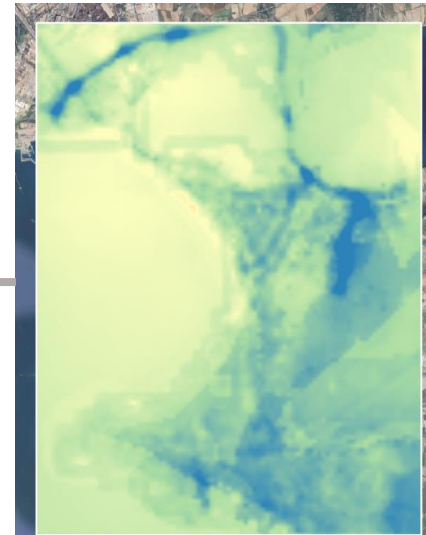
The tool combines yearly means of microscale concentration from traffic and regional background

The regional background contains all the sources, fusing **FARM** output with station observations via Machine **Learning**

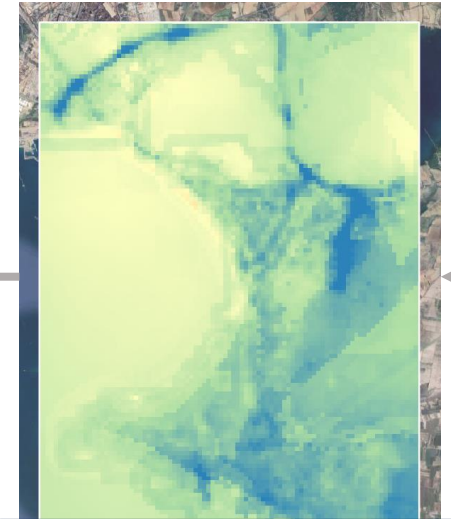
A combination procedure which avoids the **double counting** is implemented.



**B**  
Regional background [100m]



**E**  
Difference B-C regrided [5m]

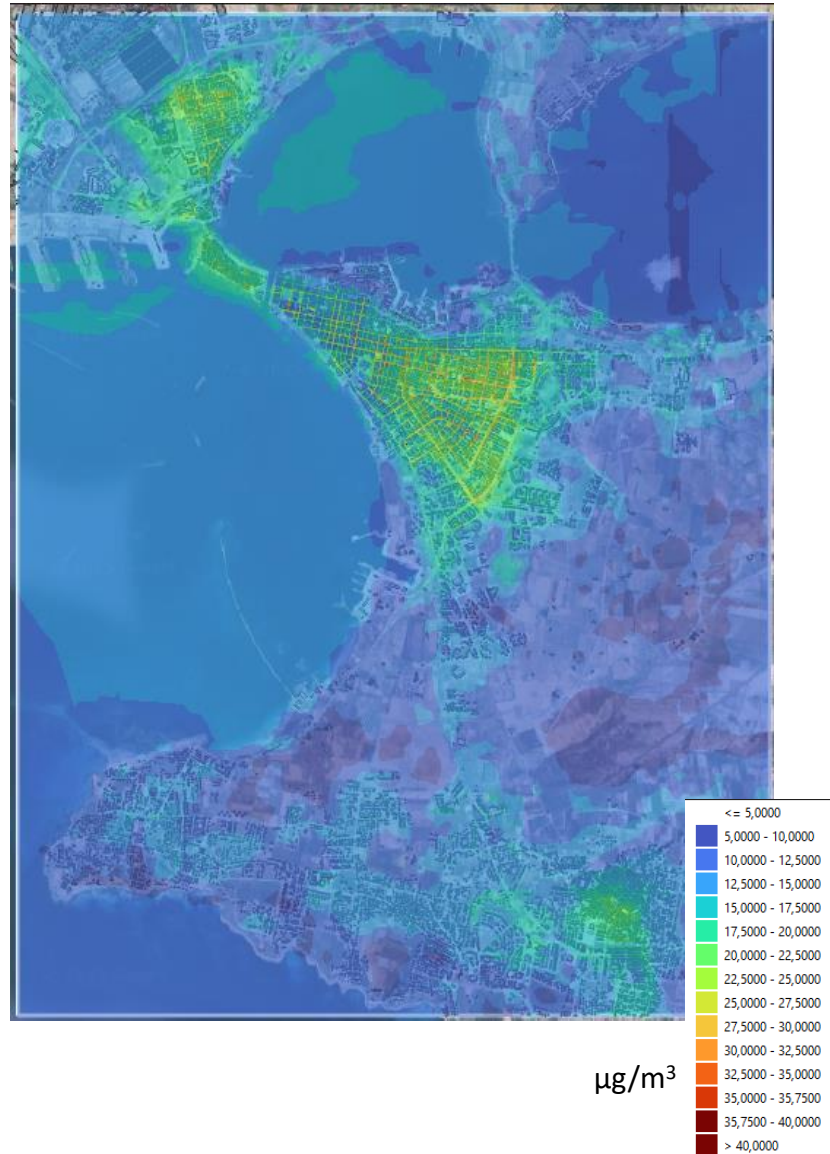


**D**  
Difference B - C [100m]

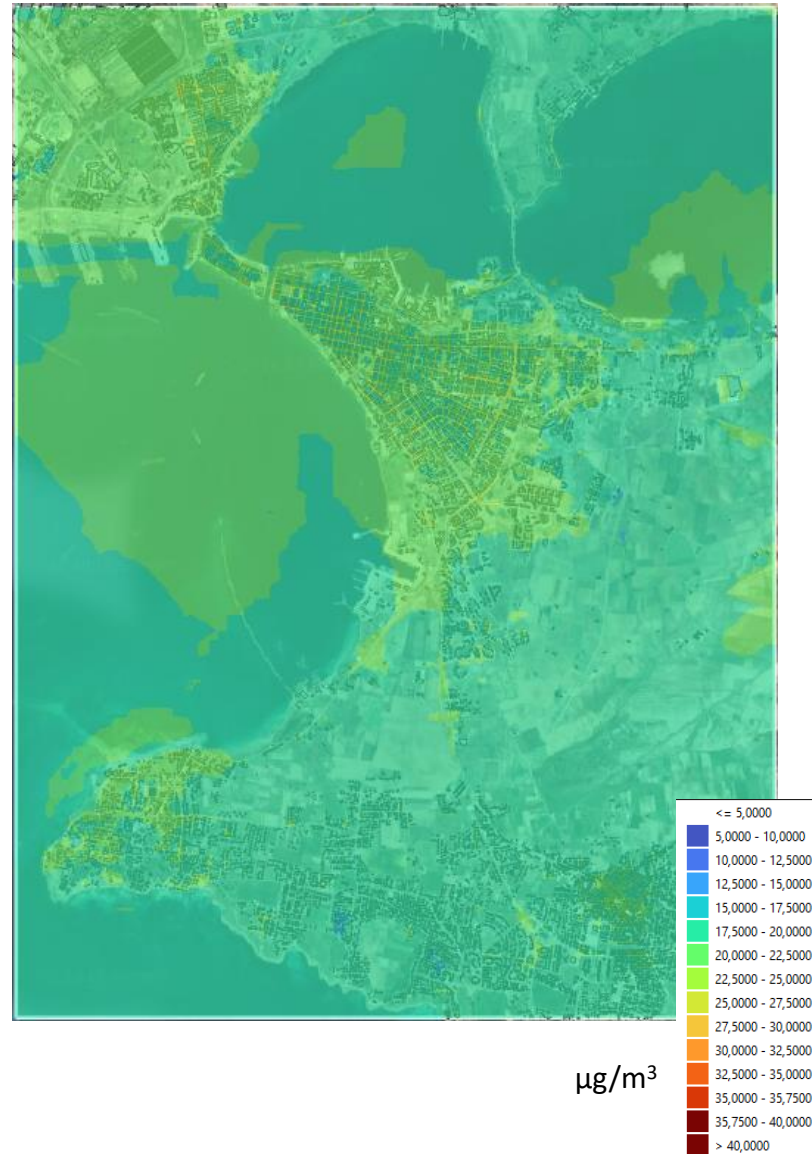
\*Data refer to preliminar PM10 simulations on 2019. Traffic emission data yet to be modelled.

# Merge between microscale and regional background

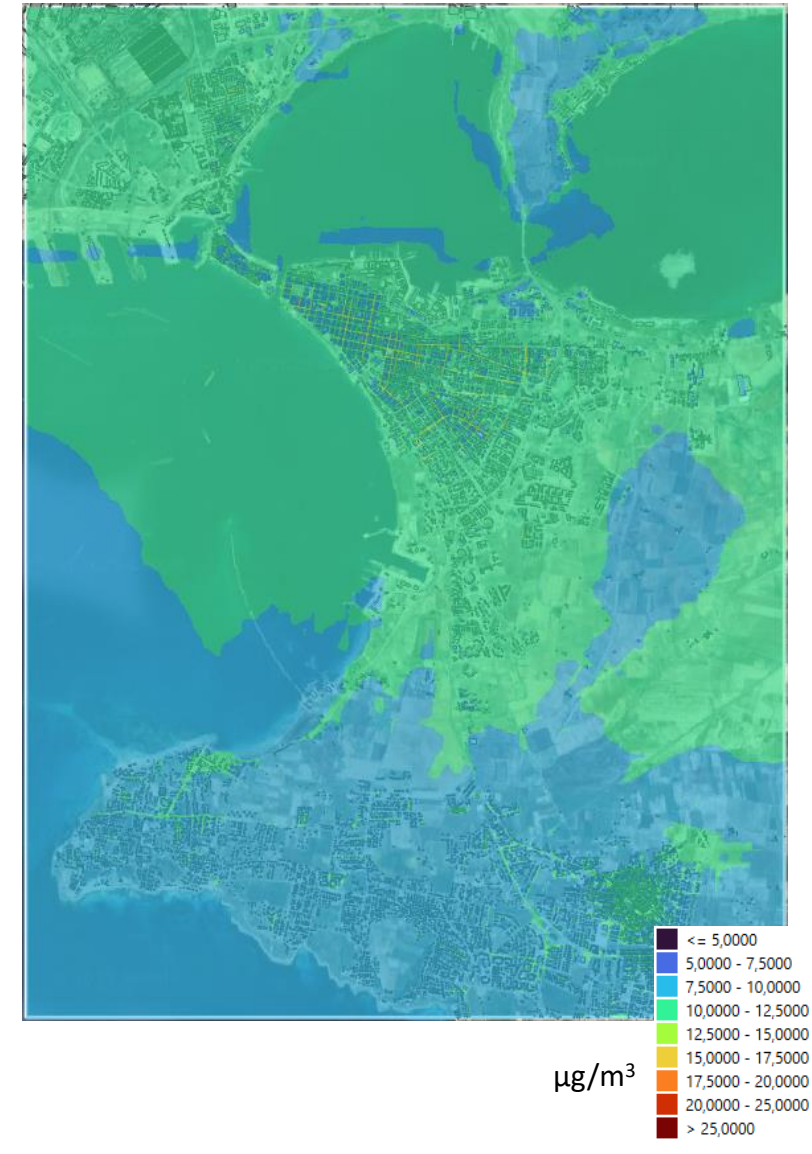
*NO2 annual mean (2023)*



*PM10 annual mean (2023)*



*PM25 annual mean (2023)*



# THANK YOU

