



UNIVERSITÀ  
DEGLI STUDI  
FIRENZE

**DINFO**  
DIPARTIMENTO DI  
INGEGNERIA  
DELL'INFORMAZIONE

**DISIT**  
DISTRIBUTED SYSTEMS  
AND INTERNET  
TECHNOLOGIES LAB



<https://www.Snap4City.org>

# *Digital Twin with AI/XAI for Management and What-if Analysis*

**Paolo Nesi**

DISIT Lab, <https://www.disit.org>, <https://www.sanp4city.org>, University of  
Florence, Florence, Italy

## **Convegno Nazionale**

**Sul Presente e il Futuro delle Smart Cities e Smart Communities:  
Un Dialogo Interdisciplinare**









# SMART SOLUTIONS AND DECISION SUPPORT SYSTEMS

CONTROL ROOMS - DECISION SUPPORT SYSTEMS - WHAT-IF ANALYSIS - BUSINESS INTELLIGENCE - SIMULATIONS - SMART APPLICATIONS



DASHBOARDS - VISUAL ANALYTICS - SYNOPTICS - DIGITAL TWIN - GRAPHICAL WIDGETS - ANALYTICS - GUI CUSTOM STYLES - VISUAL PROGRAMMING



DASHBOARDS, WIDGETS  
TEMPLATES

PREDICTION - ANOMALY DETECTION - CLUSTERING - ROUTING - SENTIMENT NLP - TRAFFIC FLOW  
PEOPLE FLOWS - SDG - 15 MIN CITY INDEX - KPI - HEATMAPS - ORIGIN DESTINATION - ETC...

API - MICROSERVICES - GIS - BPM  
VIDEO - REPORTS - MAPS - 3D ...

ANY: DATA, BROKER, NETWORK AND VERTICAL

EXPERT SYSTEM, KNOWLEDGE BASE  
SEMANTIC REASONING  
SMART DATA MODEL  
IOT DEVICE MODELS, STORAGE

BIG DATA ANALYTICS, ARTIFICIAL INTELLIGENCE  
EXPLAINABLE AI, MACHINE LEARNING  
OPERATIVE RESEARCH, STATISTICS

VISUAL PROGRAMMING, ADAPTERS  
DATA FLOWS, WORKFLOWS  
PARALLEL DISTRIBUTED PROCESSING  
DATA DRIVEN

**Native and External Applications**

- Smart Parking
- Smart Light
- Smart Waste
- Smart Energy
- Social Media Analysis

METHODOLOGIES  
LIVING LABS  
COURSES AND COMMUNITY  
DEVELOPMENT TOOLS



Powered by  
**FIWARE**

**FREE  
TRIAL**

**PEN Test  
Passed**

**EU GDPR  
COMPLIANT**

**SNAP4**  
Appliances and Dockers  
**Installations**

**EUROPEAN OPEN  
SCIENCE CLOUD**

**Node-RED**

**JS Foundation**

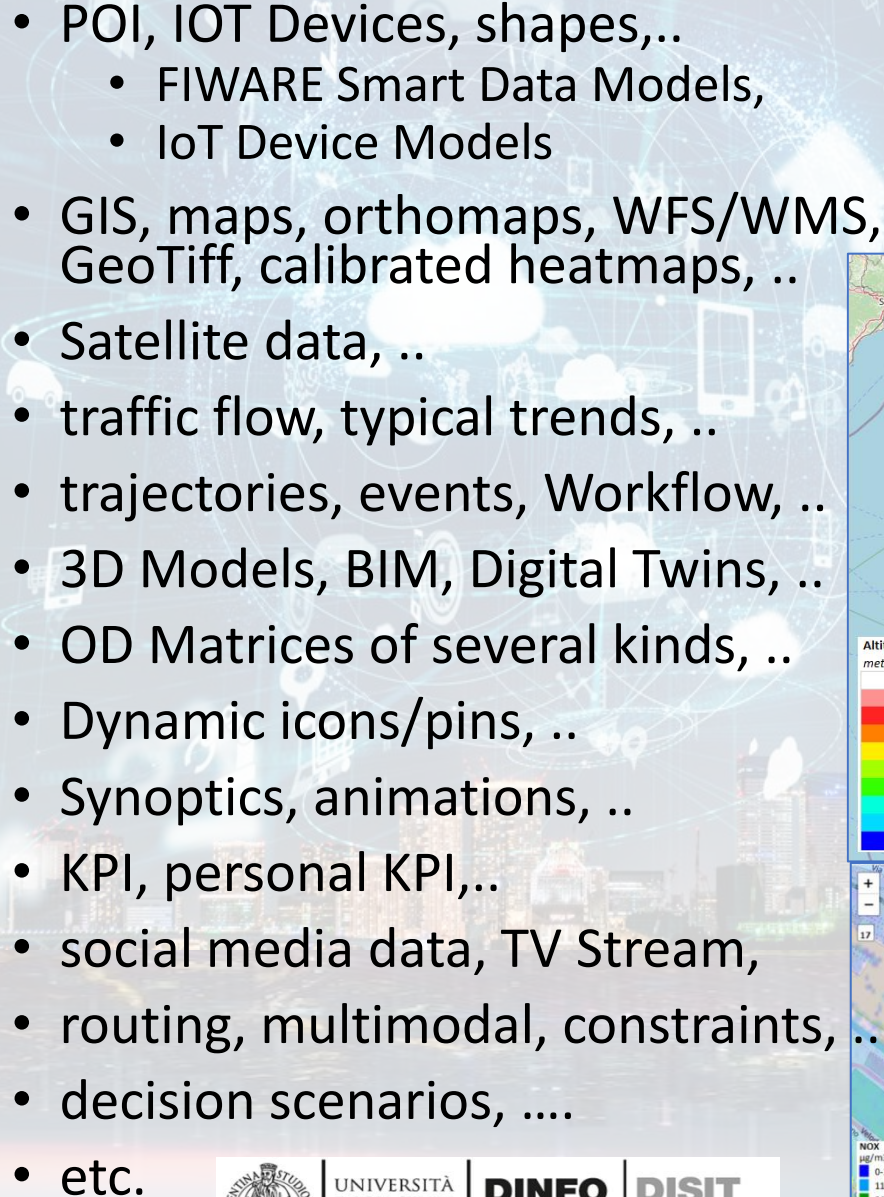

**E015**  
digital ecosystem

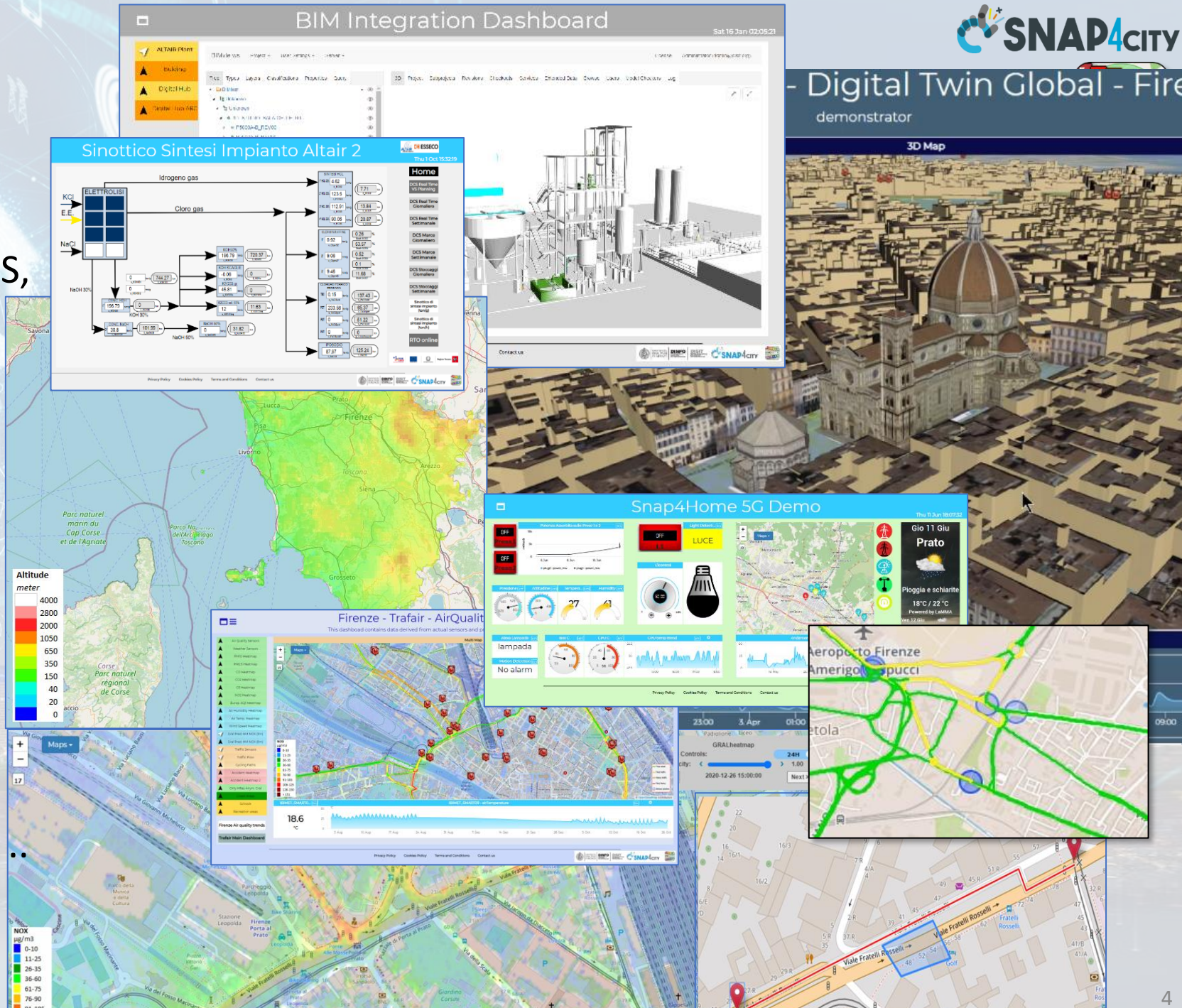
**NVIDIA**




# High Level Types

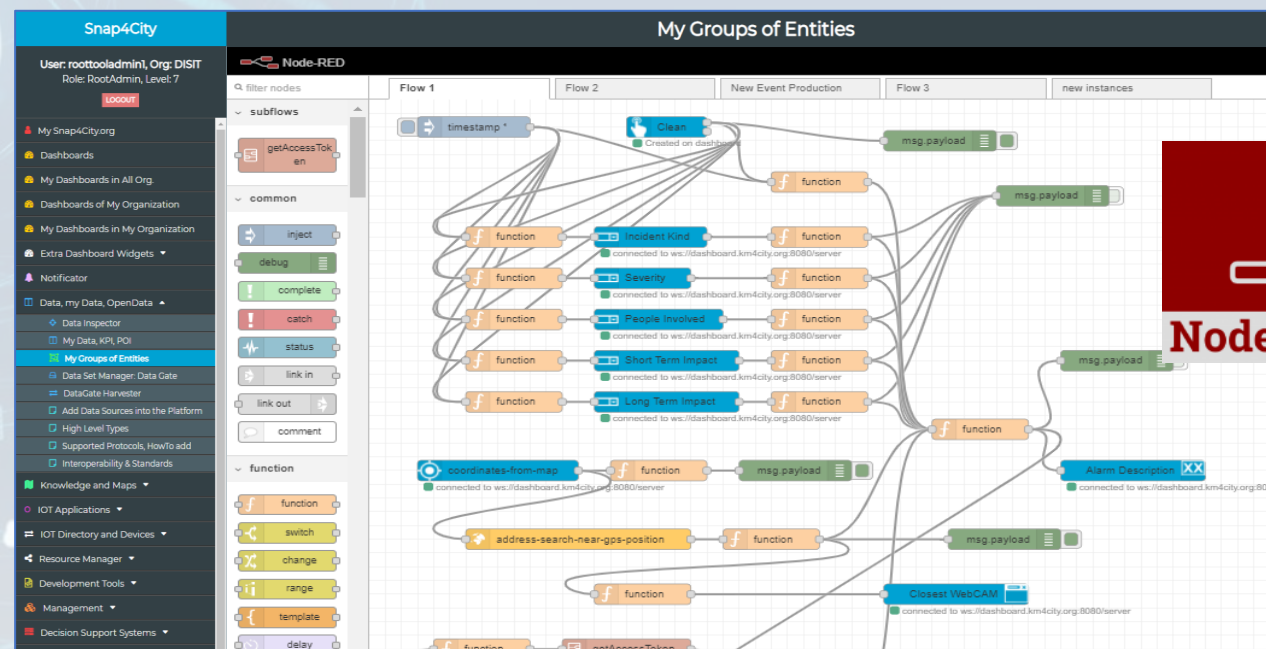
Snap4City (C), Cagliari 2023, July 2023

- 
- POI, IOT Devices, shapes,..
    - FIWARE Smart Data Models,
    - IoT Device Models
  - GIS, maps, orthomaps, WFS/WMS, GeoTiff, calibrated heatmaps, ..
  - Satellite data, ..
  - traffic flow, typical trends, ..
  - trajectories, events, Workflow, ..
  - 3D Models, BIM, Digital Twins, ..
  - OD Matrices of several kinds, ..
  - Dynamic icons/pins, ..
  - Synoptics, animations, ..
  - KPI, personal KPI,..
  - social media data, TV Stream,
  - routing, multimodal, constraints, ..
  - decision scenarios, ....
  - etc.
- 
- UNIVERSITÀ
- DINEO
- DISIT





- 
- Node-RED**



-



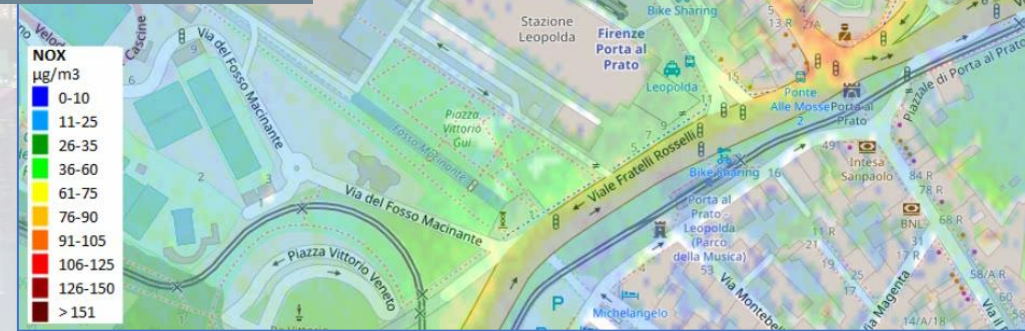
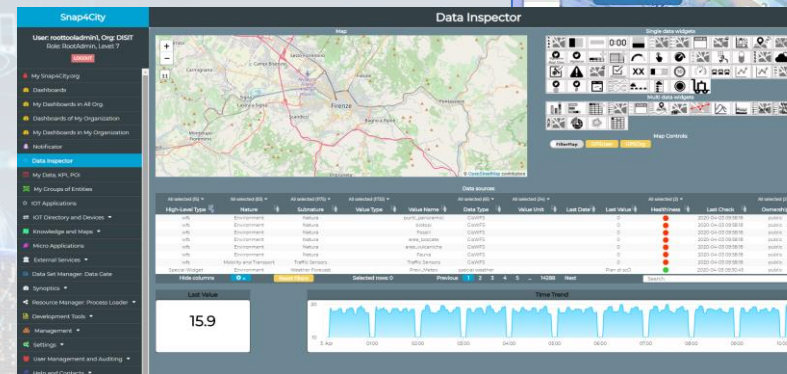
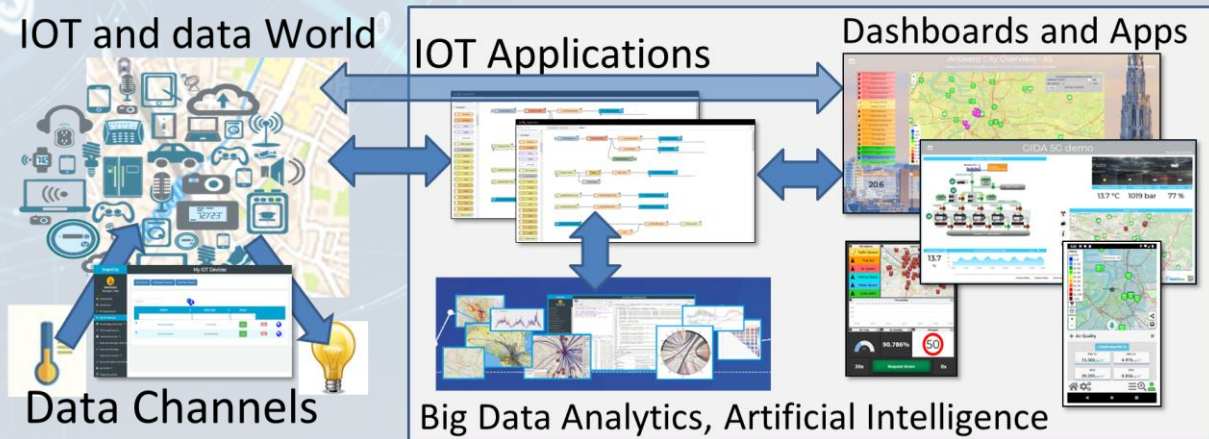
# Solutions: reliable, secure and fast to realize

- **Via Snap4City tools**

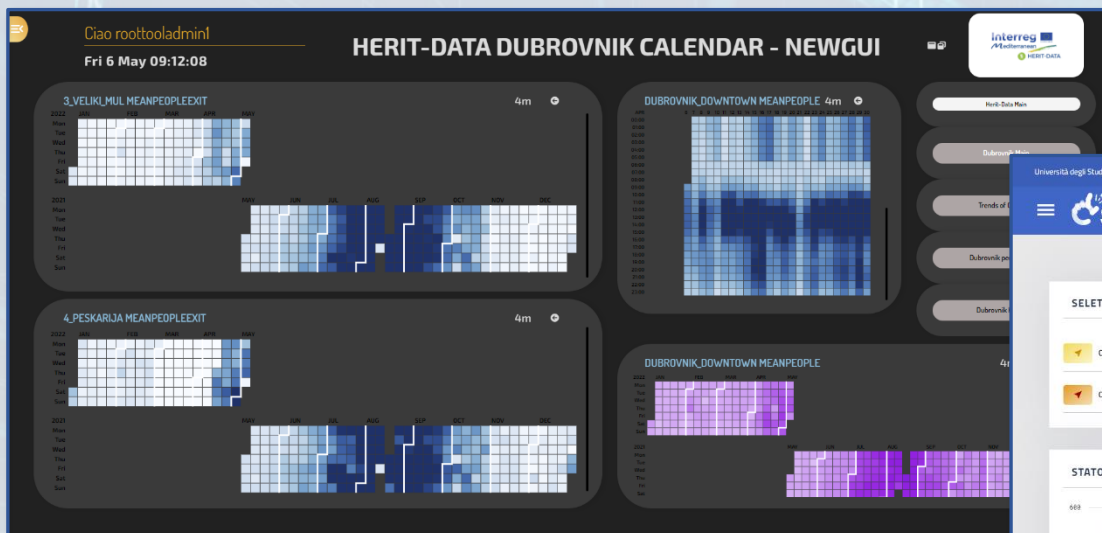
- **Dashboard Wizard**
- **Dashboard Builder**
- **Data/Visual Analytic**

- **Smart Solutions results to be**

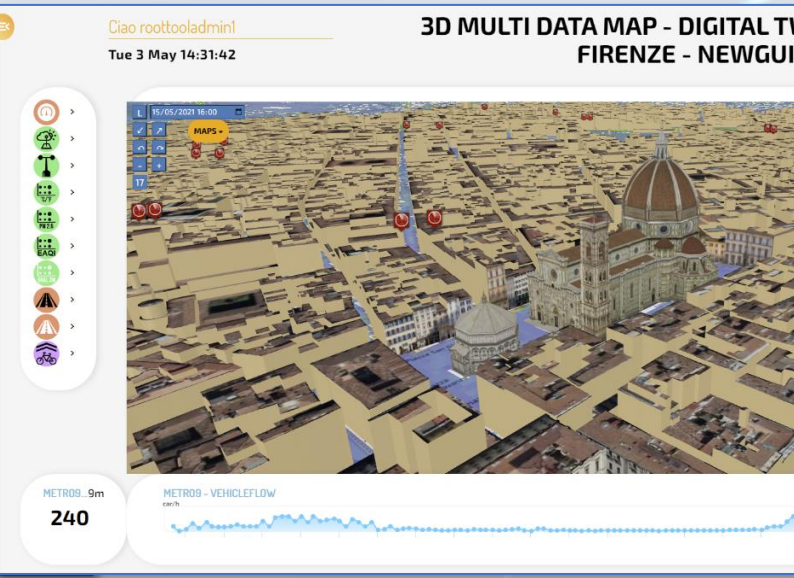
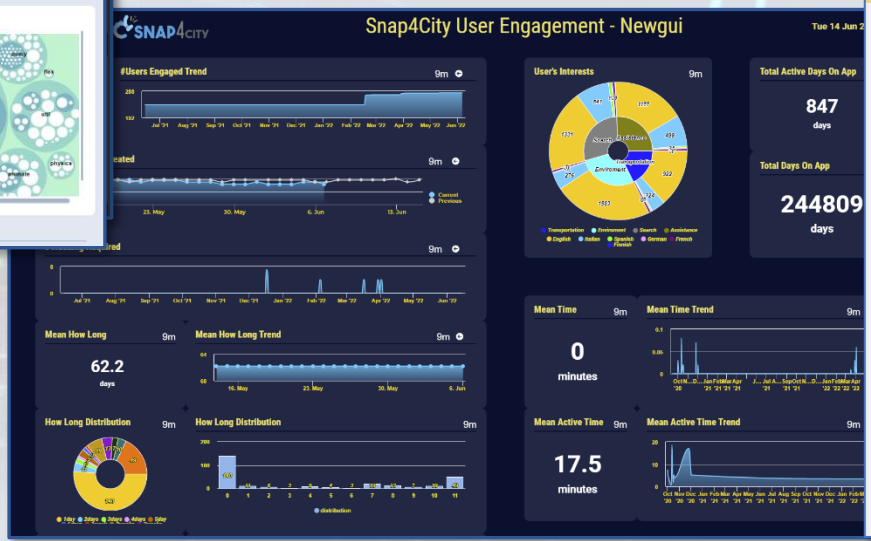
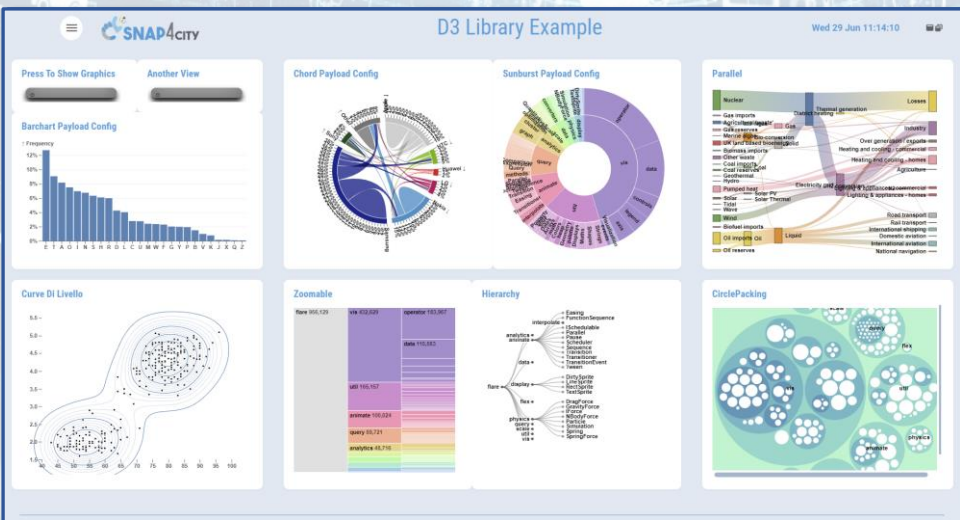
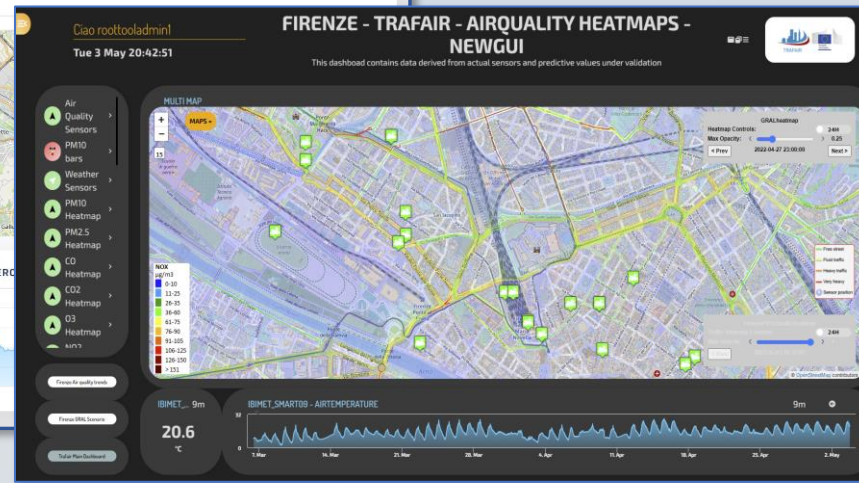
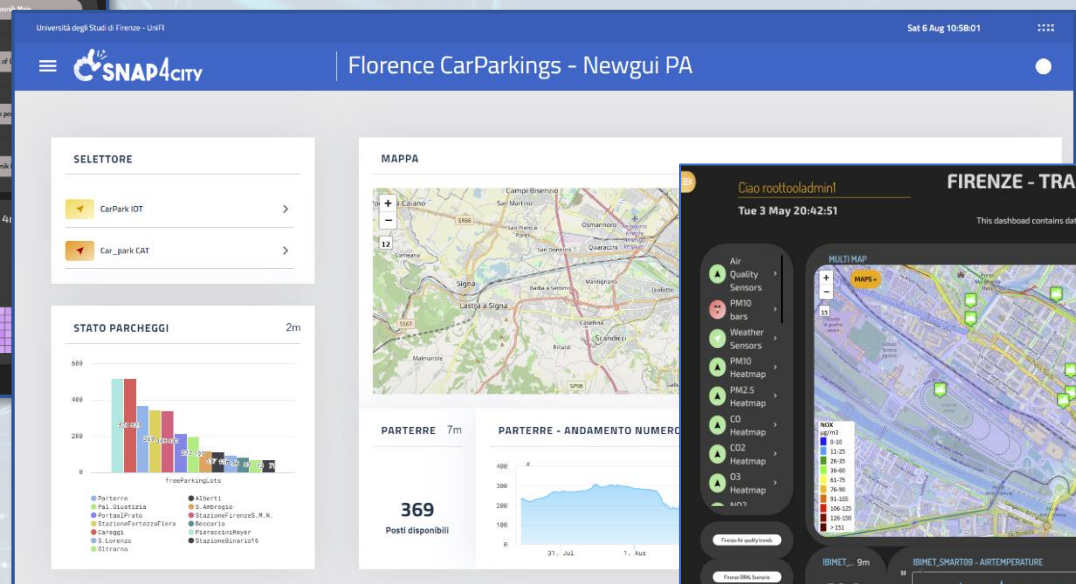
- **Real time data drive**
- **Secure end-to-end**
- **GDPR compliant**
- **Reliable, interoperable**
- **Auditable, marketable**







# Different Themes



New styles/themes can be developed by specializing a few files from open source

<https://www.snap4city.org/793>





## 3D Map



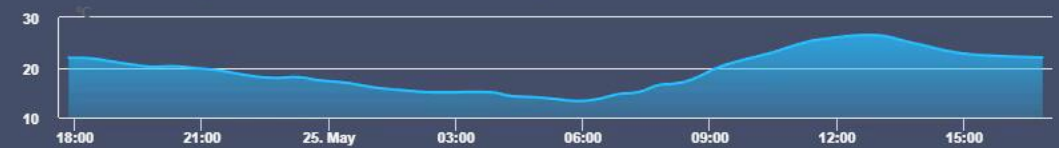
## Air Quality FI-BASSI - NO2

6m



## Weather\_sensor\_Open Weather 3176959 - Air Temperature

6m



<https://youtu.be/JLzT9k3Xbc0>







Ciao

Sun 2 Jul 12:58:18

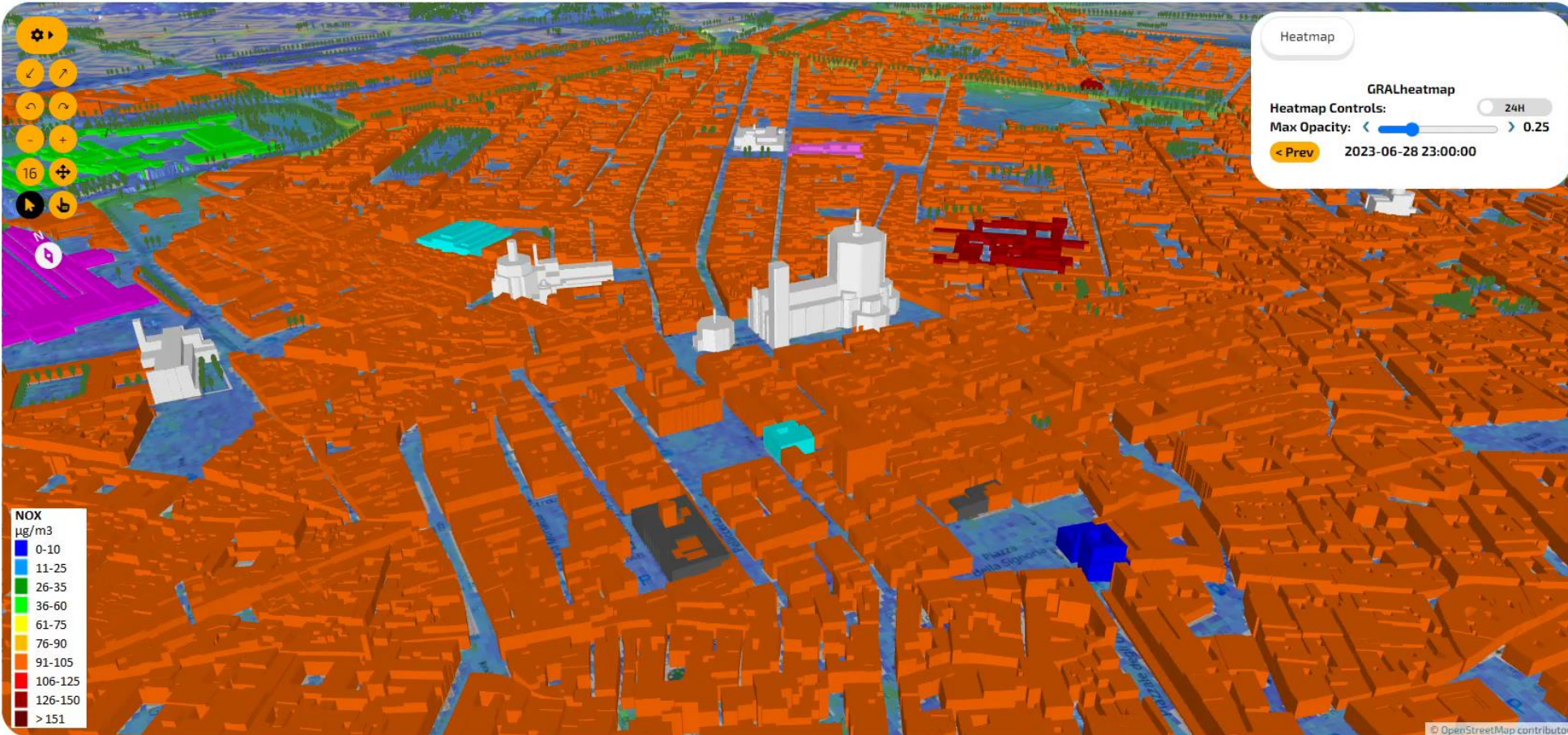
# FLORENCE SCDT



SELECT...

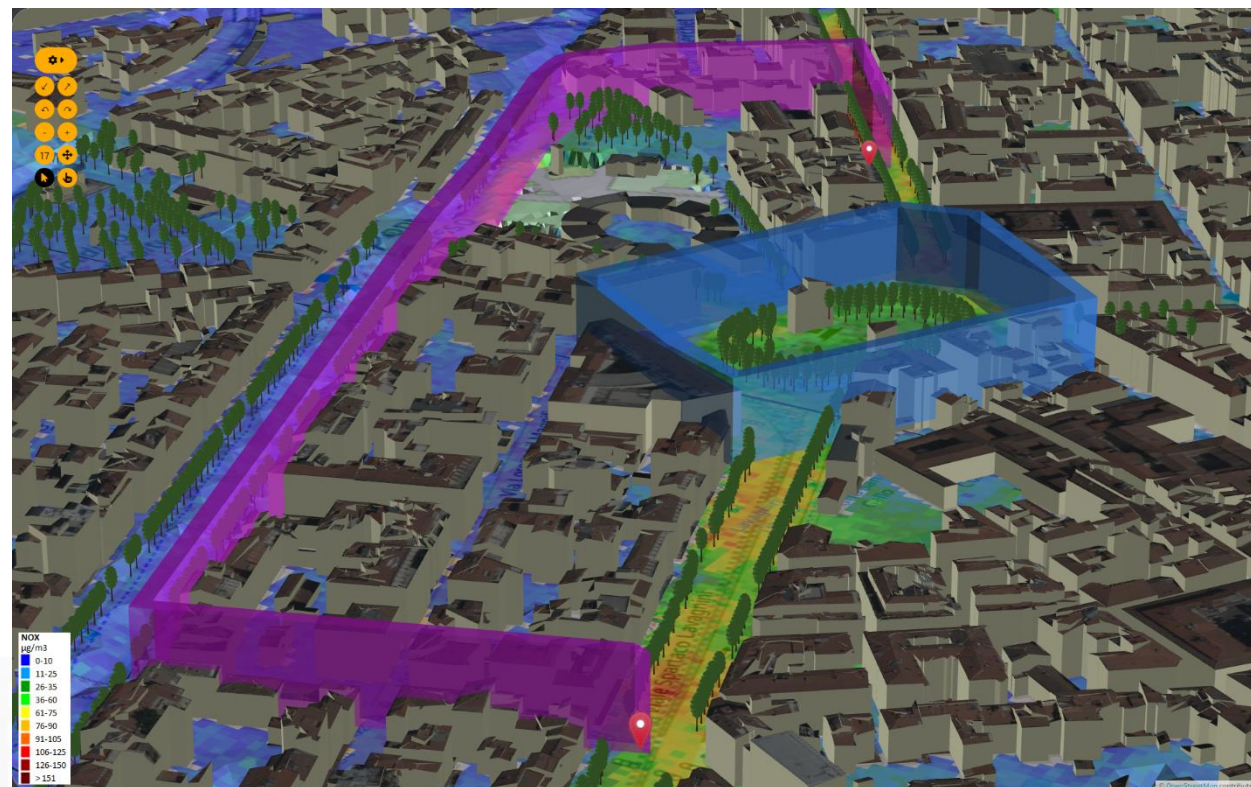
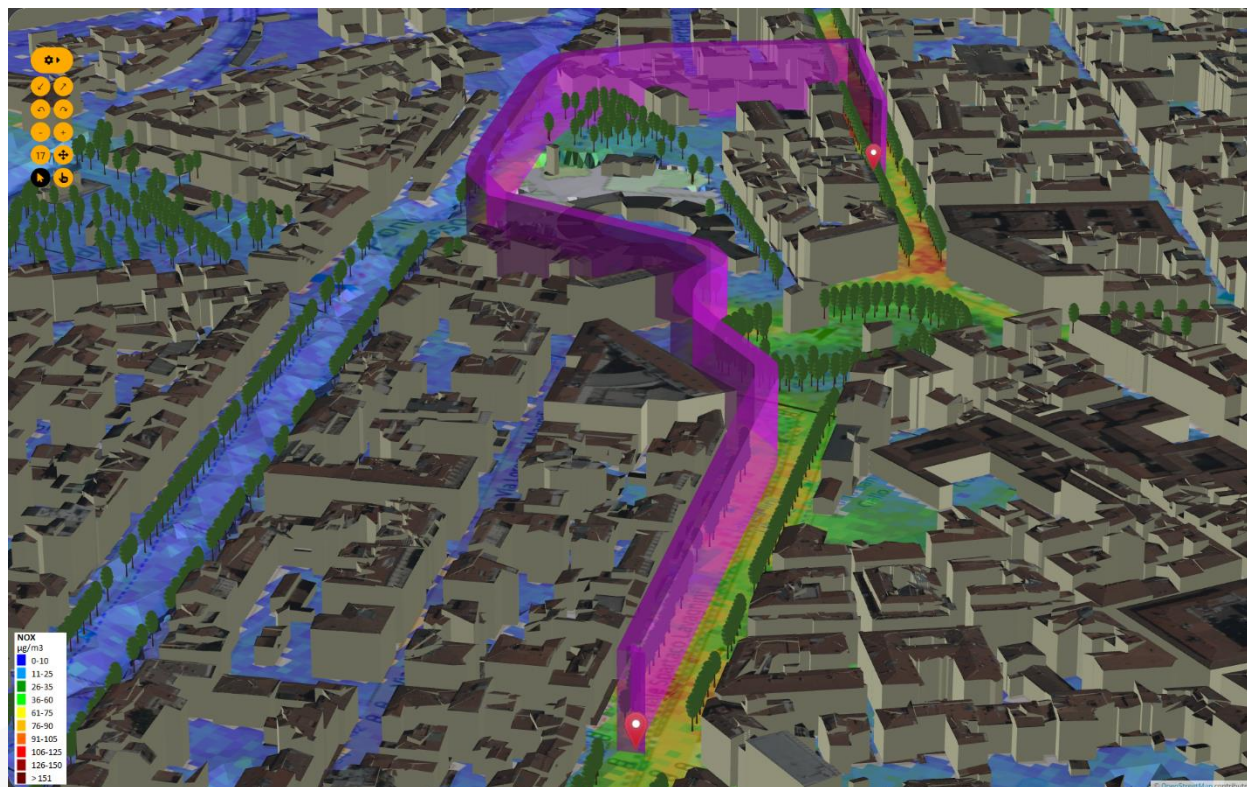


DOUBLE MAP





# Dyamic Routing in 3D space





# Requirements

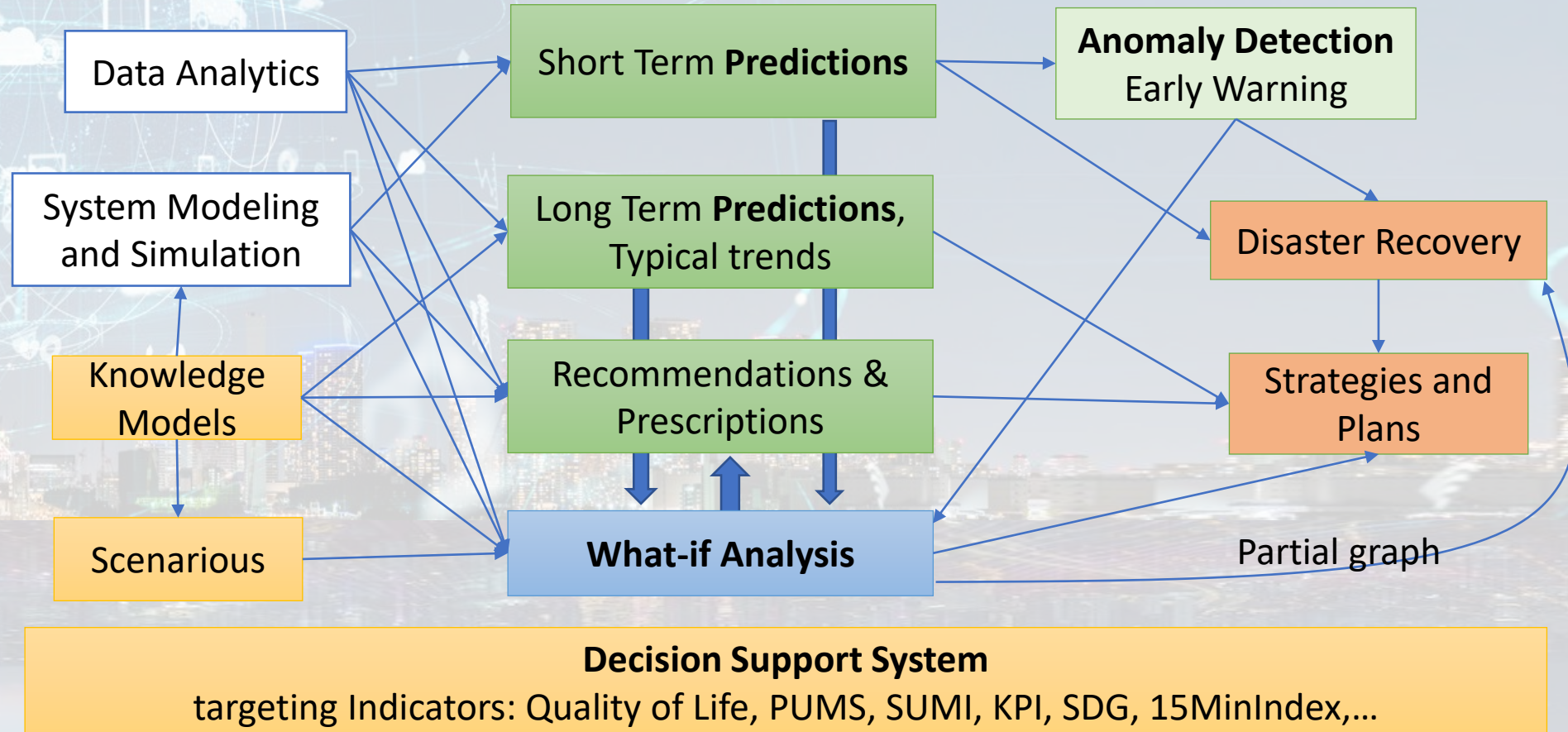
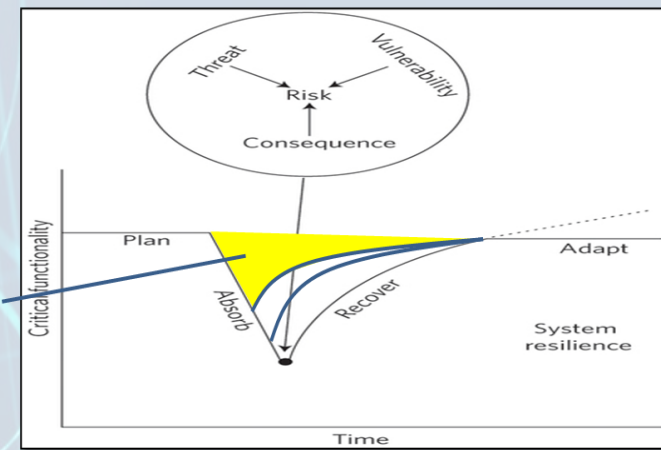
	CityGML [3]	Helsinki [19]	Rotterdam [20]	Berlin [21]	Stockholm [22]	Wellington [23]	DUET [24]	Snap4 City (our)
RD1.i	Yes	No	No	No	No	No	Yes	Yes
RD1.ii	Yes (LoD3)	Yes	Yes (LoD2)	Yes (LoD2)	Yes (LoD3)	Yes (LoD3)	Yes (LoD2/LoD3)	Yes
RD1.iii	No	No	No	No	No	Probably	No	Yes
RD2	No	Yes	Yes (C)	Yes (C)	Yes	Yes	Yes	Yes
RD3	No	No	No	Yes	No	Probably	Yes	Yes
RD4	No	Yes (C)	Yes (C)	No (x)	Yes	Yes	No	Yes
RD5	No	No	No	No	Yes	Yes	No	Yes
RD6	Yes	Yes	No	No	Yes	Yes	Yes	Yes
RD7.i	Yes	Yes	No	No	Yes	Yes	Yes	Yes
RD7.ii	Yes	No	No	No	No	Yes	No	No
RI1	No (*)	Yes	Yes	Yes	Yes	Yes	Yes	Yes
RI2	No (*)	No	No	No	No	Yes	No	Yes
RI3	No (*)	No	Yes	No	Yes	Yes	No	Yes
RI4.i	Not clear (maybe)	Yes (s)	Yes	Yes	No	Probably	No	Yes
RI4.ii	No	No	No	No	No	Probably	No	No
RI4.iii	No	No	No	No	No	Yes	No	No
RI5	No	No	No	Yes	No	Yes	No	Yes
RI6	No (*)	No	No	No	Yes	No	Yes	Yes
RI7	No (*)	Yes (**)	Yes (**)	No (x)	No	No	No	No
RI8	No	No	No	No	Yes	Yes	No	Yes
RO1	No	No	No	No	No	No	No	Yes
RO2	No	No	No	No	No	No	No	Yes
RO3	No	No	No	No	No	No	No	Yes
RO4	No	No	No (not specified)	No	No (not specified)	No (not specified)	No	Yes
RO5	No	No	No	No	No	No	No	No
RO6.i	No	Yes	Yes	Yes	Yes	Yes	Yes	Yes
RO6.ii	n/a	Non-free	Free	Free	Non-free	Non-free	Non-free	Free
RO7	No	Yes	Possible (x)	Possible (x)	Possible	Yes	Possible	Yes
RO8.i	No	Yes	Yes	No	Yes	Yes	Not clear	Yes
RO8.ii	n/a	Non free	Non free	n/a	Non free	Non free	Not clear	Yes



# Snap4City Analytics

- Decision support systems
- Improvement of life quality
- Sustainable Solutions
- Reduction of costs
- Risk Assessment
- Resilience

**P**repare  
**A**bsorb  
**R**ecover  
**A**dapt





# indicators

- **United Nations Sustainable Development Goals, SDGs** (for which cities can do more to achieve some of the 17 SDGs, <https://sdgs.un.org/goals>);
- **15 minutes cities** (where primary services must be accessible within 15 minutes on foot);
- **objectives of the European Commission** in terms of pollutant emissions for: NO<sub>2</sub>, PM<sub>10</sub>, PM<sub>2.5</sub> ([https://environment.ec.europa.eu/topic/s/air\\_en](https://environment.ec.europa.eu/topic/s/air_en));
- **PUMS: mobility and transport vs wnv**
- **SUMI: mobility and transport vs env**
- **ISO indicators: city smartness, digitization. Tech level**
- ....

Global  
Vs  
Local



## SUSTAINABLE DEVELOPMENT GOALS



				WHO guidelines	
Pollutant	Averaging period	Objective and legal nature and concentration	Comments	Concentration	Comments
PM <sub>2.5</sub>	One day			25 µg/m <sup>3</sup> (*)	99 <sup>th</sup> percentile (3 days/year)
PM <sub>2.5</sub>	Calendar year	Target value, 25 µg/m <sup>3</sup>	The target value has become a limit value since 1 January 2015	10 µg/m <sup>3</sup>	
PM <sub>10</sub>	One day	Limit value, 50 µg/m <sup>3</sup>	Not to be exceeded on more than 35 days per year.	50 µg/m <sup>3</sup> (*)	99 <sup>th</sup> percentile (3 days/year)
PM <sub>10</sub>	Calendar year	Limit value, 40 µg/m <sup>3</sup> (*)		20 µg/m <sup>3</sup>	
O <sub>3</sub>	Maximum daily 8-hour mean	Target value, 120 µg/m <sup>3</sup>	Not to be exceeded on more than 25 days per year, averaged over three years	100 µg/m <sup>3</sup>	
NO <sub>2</sub>	One hour	Limit value, 200 µg/m <sup>3</sup> (*)	Not to be exceeded more than 18 times a calendar year	200 µg/m <sup>3</sup> (*)	
NO <sub>2</sub>	Calendar year	Limit value, 40 µg/m <sup>3</sup>		40 µg/m <sup>3</sup>	

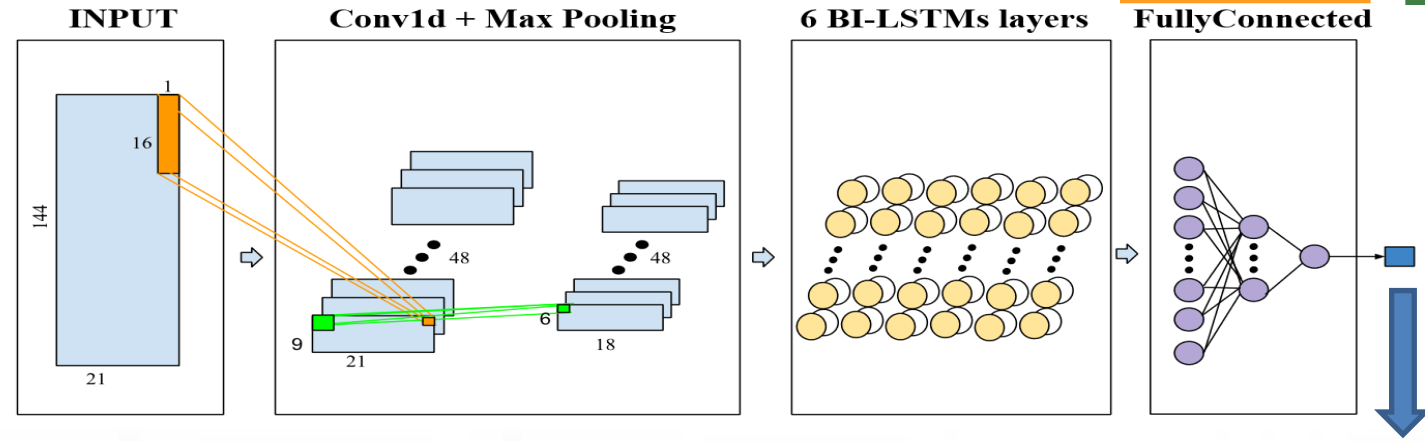


# Mobility and Transport

- **Public Transportation:** Ingestion and modelling of GTFS, Transmodel, etc. (DP)
  - Analysis of the **demand mobility vs offer transport** of according to public transportation and multiple data sources (Simulation)
  - Assessing **quality of public transportation** (analysis)
- **Accidents** heatmaps, anomaly detection (analysis, ML)
- **Predictions** for: traffic flow, smart parking, smart bike sharing, people flows, etc. (ML, DL)
- **What if analysis:** routing, traffic flow, demand vs offer, pollutant, etc. (Simulation + ML)
- **Traffic flow reconstruction** from sensors and other sources (simulation + ML)
- **Tracking fleets**, people, via devices: OBU, OBD2, mobile apps, etc. (DP)
- **Routing** and multimodal routing (multistop travel planning), constrained routing, dynamic routing (DA)
- Computing **Origin Destination Matrices** from different kind of data (analysis, DP, DP)
- Computing **typical trajectories** on the basis of tracks (analysis, ML)
- Computing Messages for Connected drive (DP)
- Slow and Fast Mobility **15 Minute City Indexes** (analysis, DP, ...ML)
- Computing and comparing traffic flow on devices and at the city border (analysis)
- **Typical time trends** for traffic flow and IoT Time series. (analysis, ML)
- **Impact of COVID-19** on mobility and transport
- Computing **SUMI, PUMS**, etc. (mainly DP)
- Etc.



# Short-Term Prediction of City Traffic Flow via Convolutional Deep Learning



Urban data:

- Date-time
- Traffic
- Temporal
- Seasonality
- Pollution
- Weather

RF

XGBOOST

DNN

LSTM

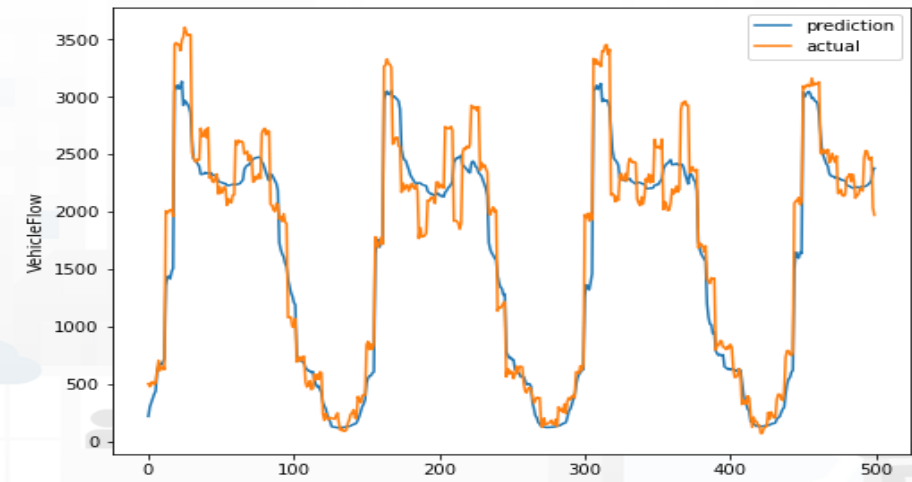
BI-LSTM

Autoencoder BI-LSTM

Attention CONV-LSTM

CONV-BI-LSTM

CONV-BI-LSTM

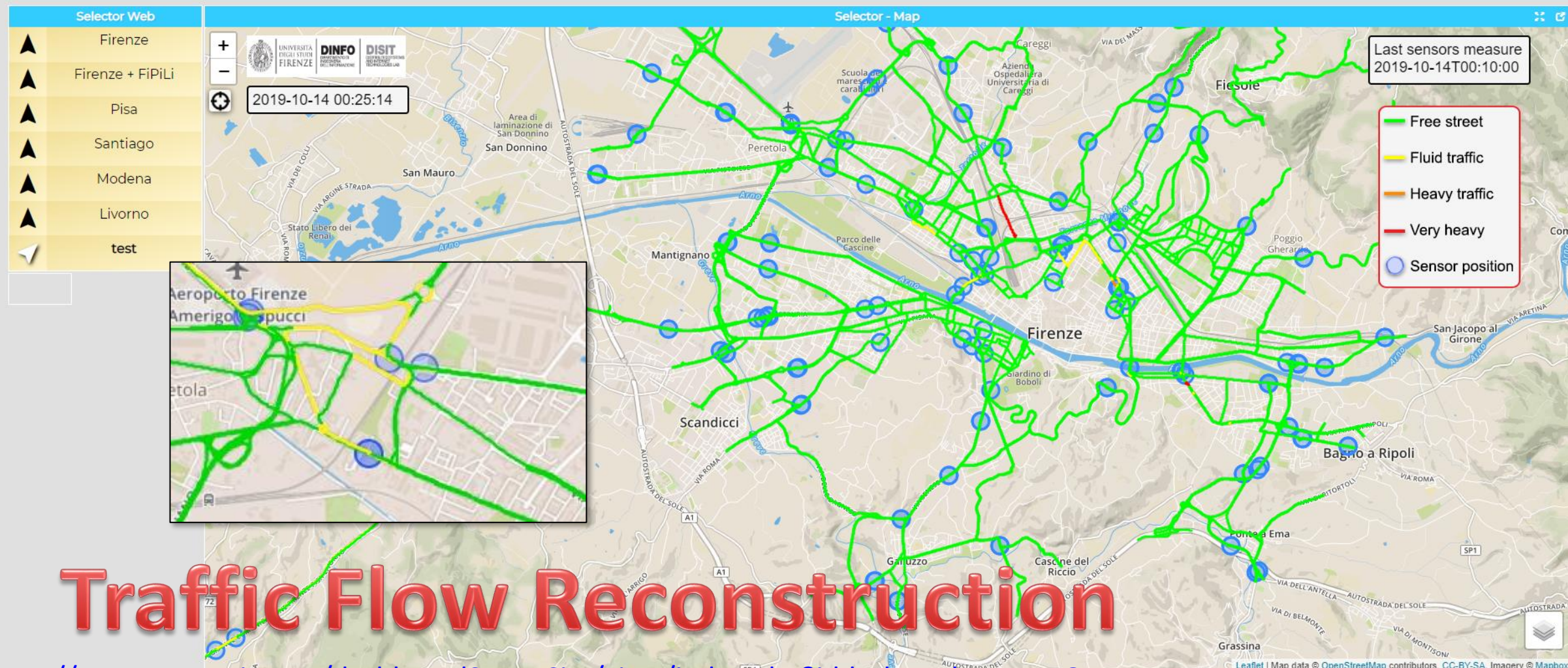






# Traffic Flow Reconstruction for the cities

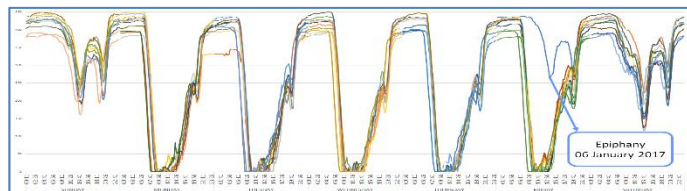
Mon 14 Oct 00:25:15



<https://www.snap4city.org/dashboardSmartCity/view/index.php?iddasboard=MTc5NQ==>



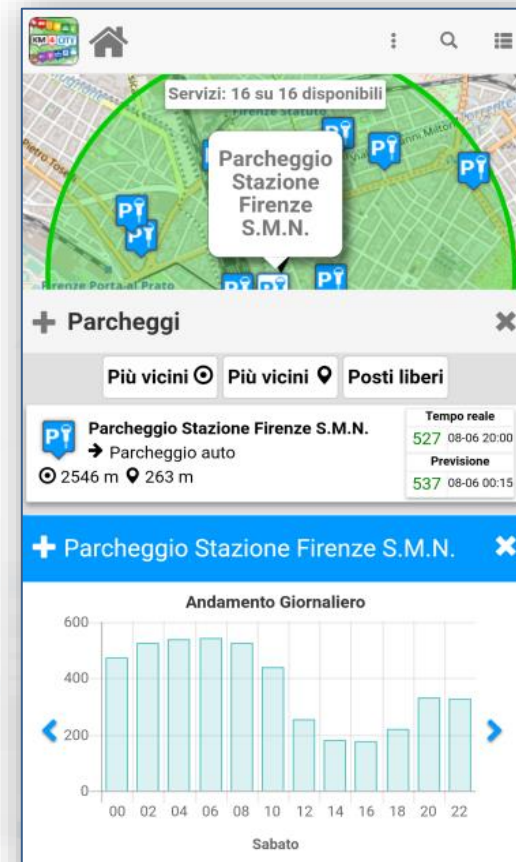
## I would arrive to surely Park in 45 Minutes??



Category	Features	Description of features variable
Baseline features of free slot data	Free parking slots	Real number of available slots recorded every 15 minutes
	Time	Hours and minutes
	Month	Month of the year (1-12)
	Day	Day of the month (1-31)
	Day week	Day of the week (0-6)
	Weekend	0 for working days, 1 else
	Previous observation's difference (POD)	Difference between the number of free spaces at time $i$ and number of free spaces at time $(i - 15 \text{ minutes})$ recorded in the previous week
Weather features	Subsequent observation's difference (SOD)	Difference between the number of free spaces at time $i$ , and the number of free spaces at time $(i + 15 \text{ minutes})$ recorded in the previous week
	Temperature	City temperature measured one hour earlier than Time ( $^{\circ}\text{C}$ )
	Humidity	City humidity measured one hour earlier than Time (%)
Traffic Sensors features	Rainfall	City rainfall measured one hour earlier than Time (mm)
	Average Vehicle Speed	Average speed of vehicles on the road being closest to the parking, over one-hour period (km/h)
	Vehicle Flow	Number of vehicles passing by closest to the parking, over one-hour period
	Average Vehicle Time	Average of distance between vehicles, over one-hour period
	Vehicle Concentration	Number of vehicles per kilometer, over one-hour period

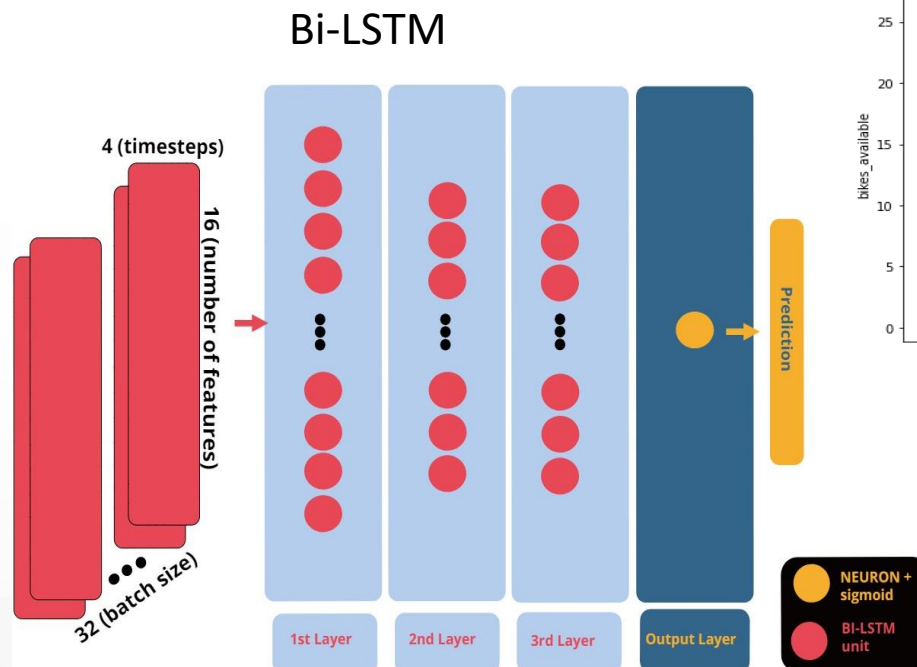
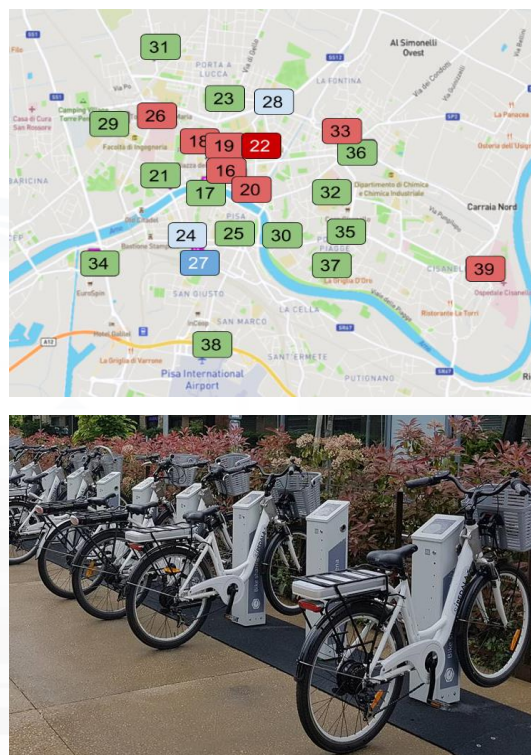
Artificial Intelligence  
Predictions

97% of precision





# Deep Learning for Short-Term Prediction of Available Bikes on Bike-Sharing Stations



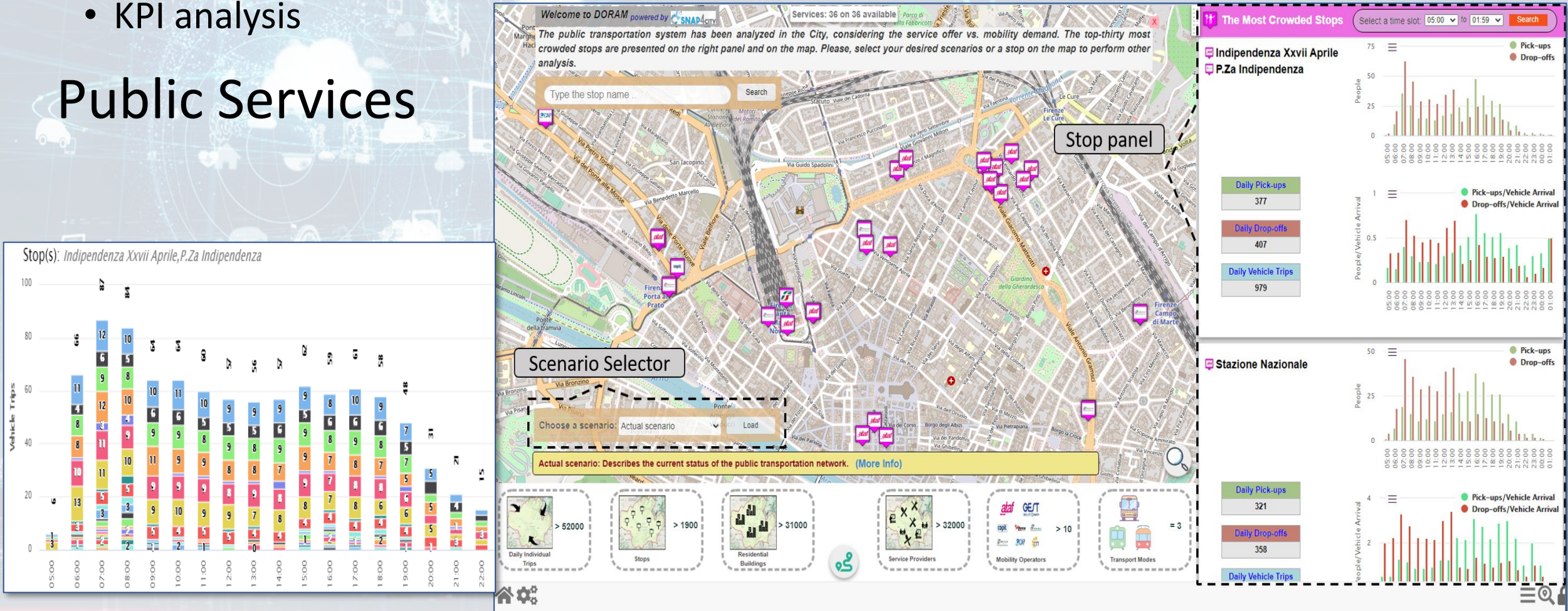


# What-if Analysis on Pub Transport



- Definition of scenarios impact on
  - Traffic, Pollutant, parking, public transport, private flows, etc.
- KPI analysis

## Public Services





# City Users Behavior and Social Analysis

- **People detection and classification:** persona, strollers, bikes, etc. (ML, DL)
- **people counting** and tracking, head counting (via thermal cameras, ML, DL)
- **People flows prediction** and reconstruction, (ML, DL)
  - Wi-Fi data, mobile apps data, Mobile Data, etc.
- **User's behaviour analysis, People flow analysis** from PAX Counters and heterogenous data sources (ML, AI)
  - origin destination matrices, hot places, time schedule,
  - Recency and frequency, permanence, typical trajectory, etc.
- **Computing User engagement and suggestions** for sustainable mobility (Rule Based, ML)
- **Social media analysis** on specific channel, specific keywords: see Twitter Vigilance,
  - Reputation, service assessment: MultiLingual NLP and Sentiment Analysis, SA
  - Tweet proneness, retweet-ability of tweets, impact guessing
  - Audience predictions on TV channels and physical events, locations
  - Prediction of attendance of events and on attractions
- **Virtual Assistant construction, LLM, NLP, Sentiment Analysis (DL, NLP)**
- **15 Minute City Index** , etc. (modeling and computability)
- Computing SDG, etc., (DP)
- Etc.



# User Behaviour Analysis

Where

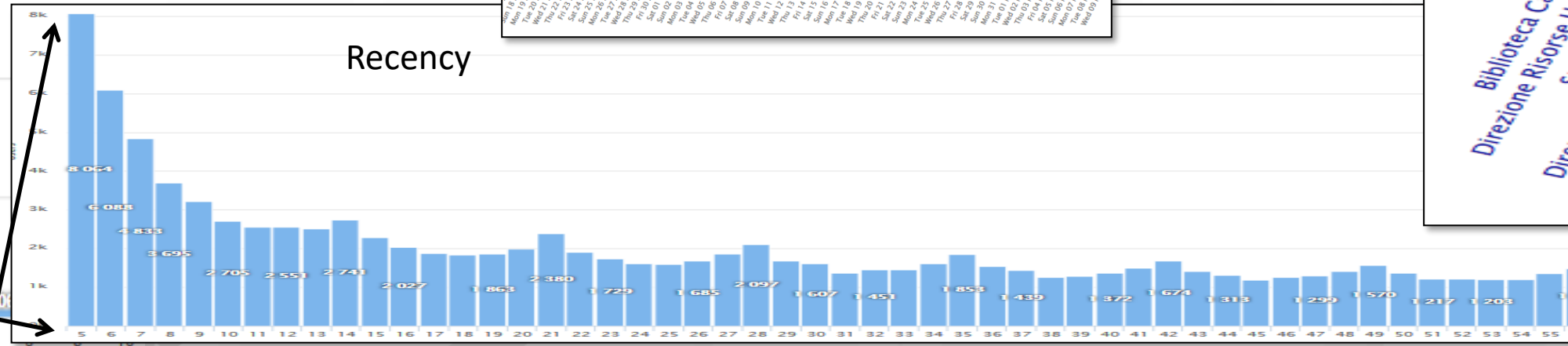
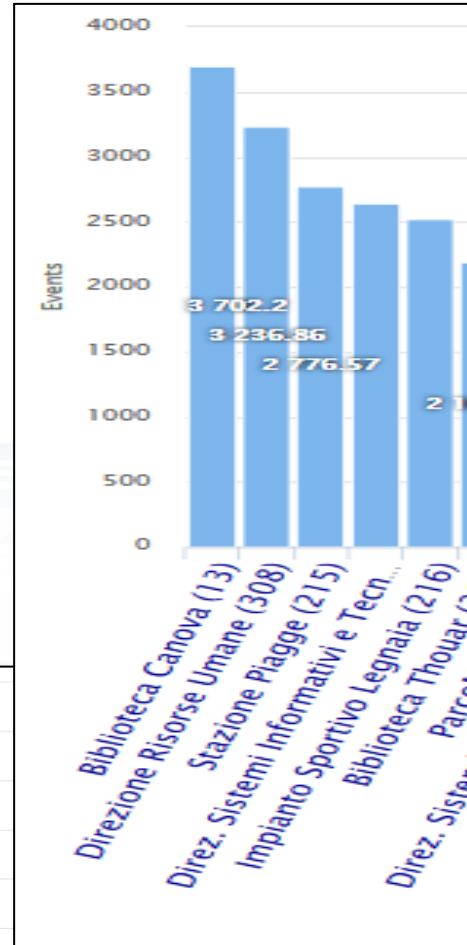
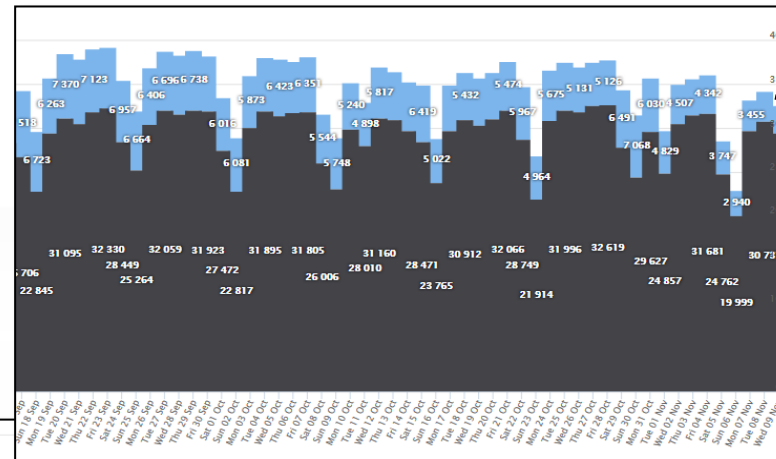
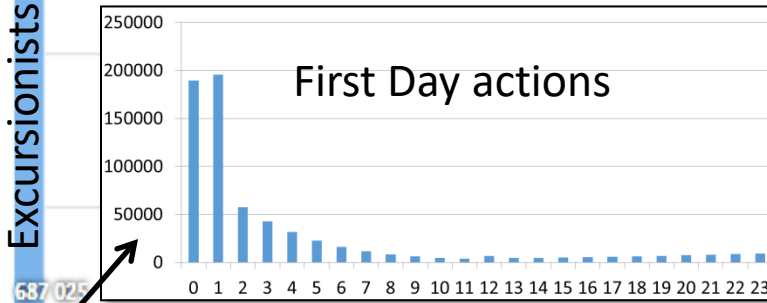
Distinct APs: 343

Distinct APs (last 24 hours): 311

Distinct Users (last 180 days): 1102098

Distinct Excursionists (last 180 days, < 24 h): 687025

Excursionists





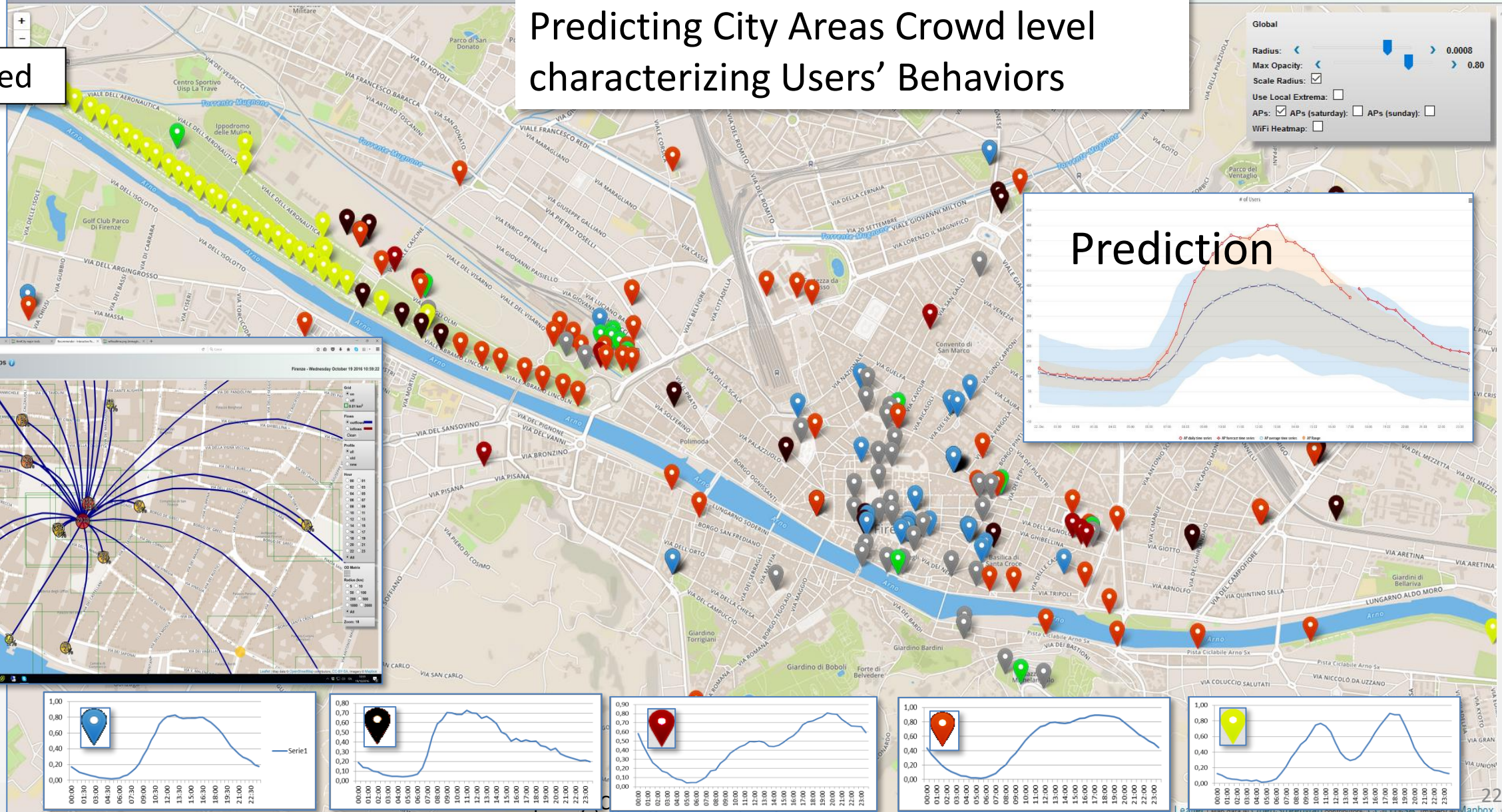
# Characterizing City Areas

DisIT Firenze Wi-Fi: Access Points Clusters Coverage Map  
DisIT - Distributed Systems and Internet Technologies Lab

Firenze - Saturday November 12 2016 19:16:33

Wi-Fi based

Predicting City Areas Crowd level  
characterizing Users' Behaviors





# A view and data from the Thermal Camera



## Detection BOX Snap4Thermal PV Firenze Tue 15 Mar 13:30:41



**11** SUSTAINABLE CITIES  
AND COMMUNITIES





# Environment and Weather

- **Pollutant Predictions: short, long and very long term** European Commission KPIs
  - NOX, PM10 pollution on the basis of traffic flow, 48 hours (ML, AI, DL)
  - Cumulated NO2 average value over the year, ..... (ML, AI, DL)
- **Computation of CO2** on the basis of traffic flows (DP), computing emission factor (DA)
  - each road for each time slot of the day
- **Prediction of MicroClimate** conditions for diffusion (ML, AI)
  - NO2, PM10, PM2.5, etc.
- **Prediction of landslides**, 24 hours in advance (AI, DL)
- **Heatmaps production**, dense data interpolation (DP) for
  - Weather conditions: temperature, humidity, wind, DEW
  - Pollutants and Aerosol: NO, NO2, CO2, PM10, PM2.5, etc.
- **Impact of COVID-19** on Environmental aspects (DP)
- Optimisation of **waste collection** schedule and paths (DP, ML)
- Computing **SDG, SUMI, PUMS**, .. (mainly DP)
- Etc.



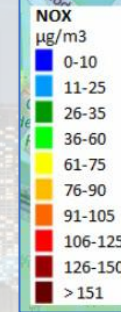
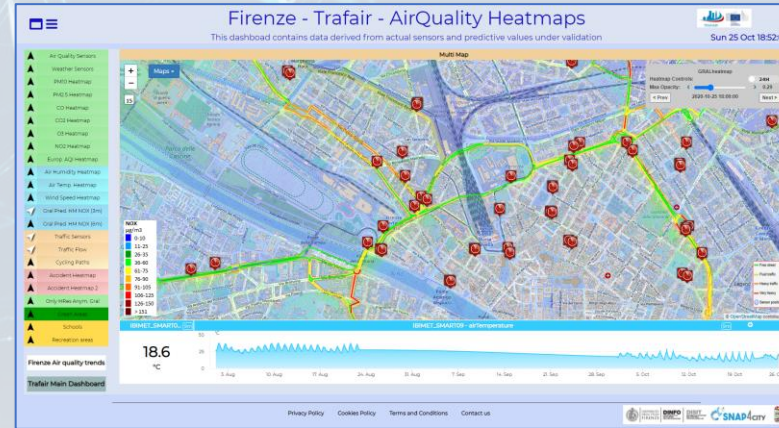
# Environment and Quality of Life

## Air Quality Predictions

Cities of:  
Firenze, Pisa, Livorno



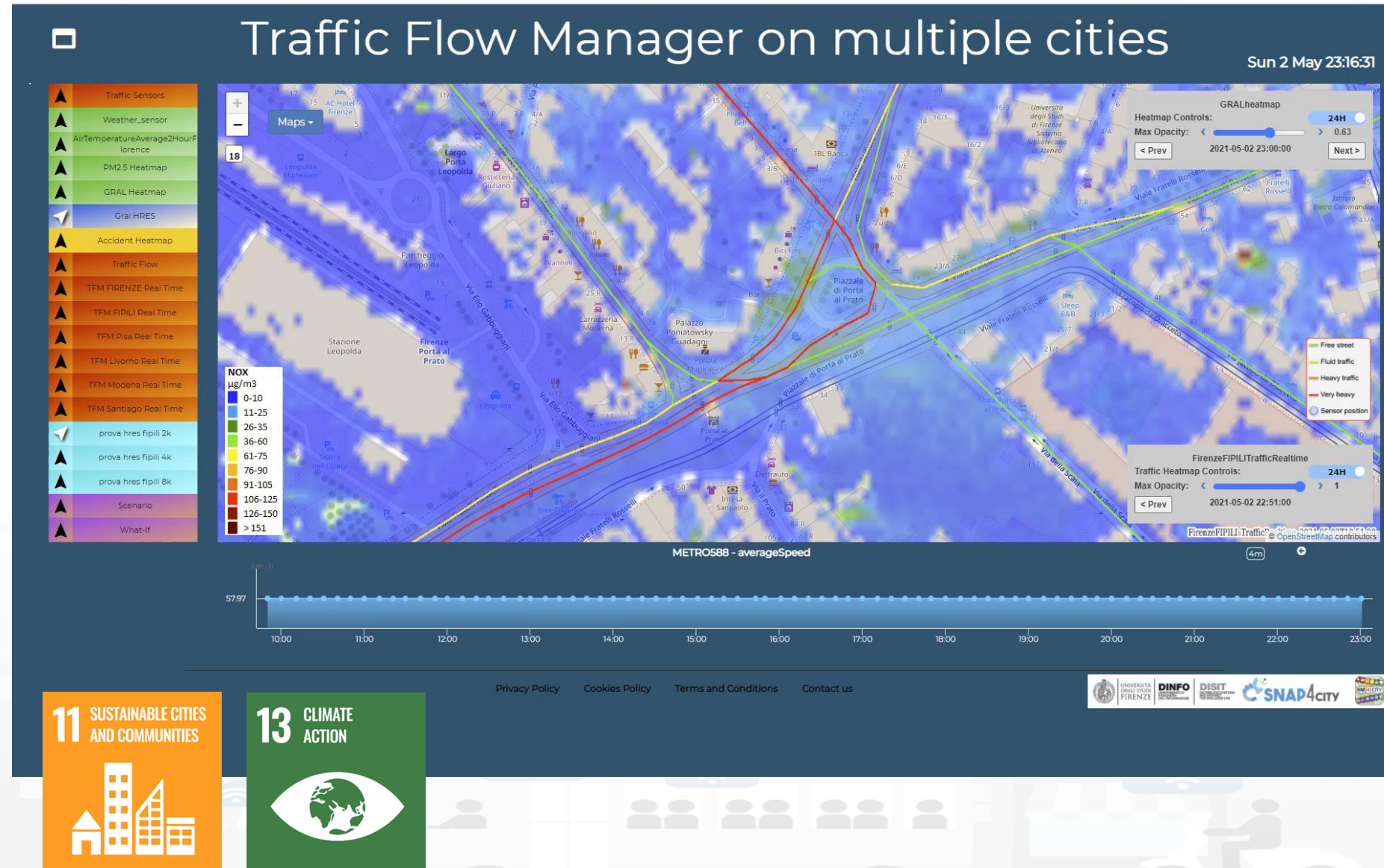
- **Multiple Domain Data**
  - Traffic Flow data, Pollutant: NOX, CO2, PM10, PM2.5, O3, ....
  - 3D City structure, weather, ...
- **Multiple Decision Makers**
  - Pollutant Predictions: NOX, NO2, ..
  - City officers, energy industries
  - Dashboards, What-IF analysis
  - Traffic Flow Reconstruction
- **Historical and Real Time data**
  - Billions of Data
- **Services Exploited on:**
  - Dashboards, Mobile App
- **Since 2020**



Air Quality Directive				WHO guidelines	
Pollutant	Averaging period	Objective and legal nature and concentration	Comments	Concentration	Comments
PM <sub>2.5</sub>	One day			25 µg/m³ (*)	99 <sup>th</sup> percentile (3 days/year)
PM <sub>2.5</sub>	Calendar year	Target value, 25 µg/m³	The target value should be achieved by 2015	10 µg/m³	
PM <sub>10</sub>	One day	Limit value, 50 µg/m³	It should be exceeded on more than 35 days per year.	50 µg/m³ (*)	99 <sup>th</sup> percentile (3 days/year)
PM <sub>10</sub>	Calendar year	Limit value, 40 µg/m³ (*)		20 µg/m³	
O <sub>3</sub>	Maximum daily 8-hour mean	Target value, 120 µg/m³	Not to be exceeded on more than 25 days per year, averaged over three years	100 µg/m³	
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NO <sub>2</sub>	Calendar year	Limit value, 40 µg/m³		40 µg/m³	

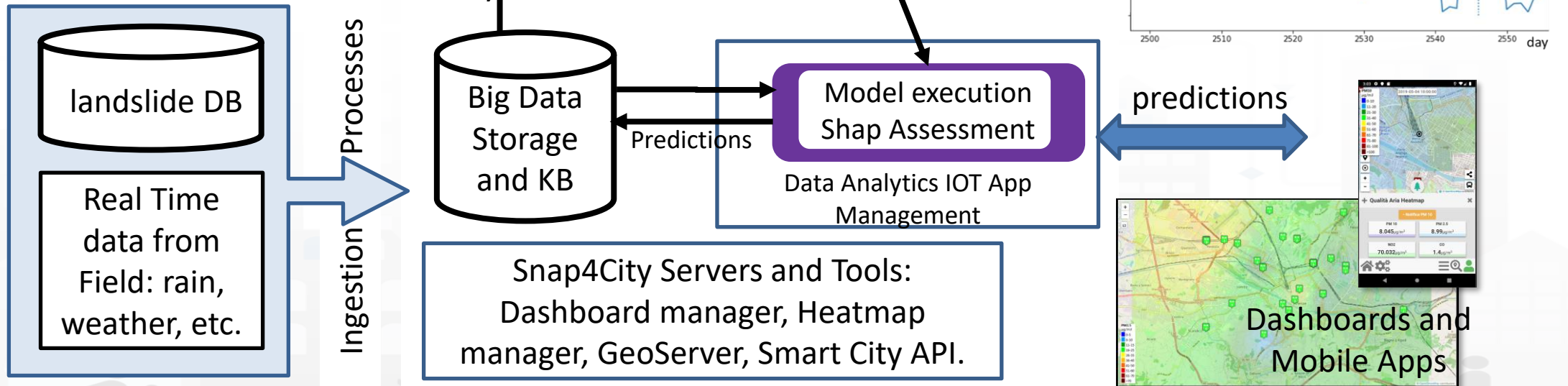
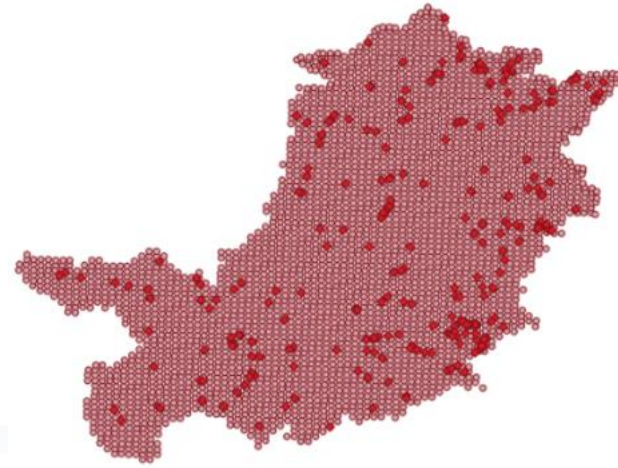


- **Prediction**
  - **NOX Pollutant** diffusion on the basis of Traffic Flow (prediction), weather and 3D structure
  - **NO2 progressive average** (Long term)
- **Project:**
  - **Trafair CEF EC**
  - Mixed solutions of Fluidinamics modeling and AI





# Predicting Land slides

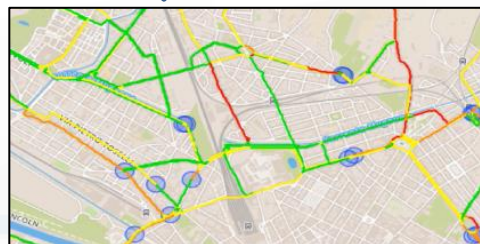




# Estimating City Local CO2 from Traffic Flow Data



Computing Traffic Flow  
into CO2 sensor area



Traffic Flow data

- Traffic Flow is one the main source of CO2
  - K1: Fluid Flow
  - K2: Stop and Go
- **Dense estimation of CO2 into the city** is very useful to know to target EC's KPIs

Computing CO2 on the basis of  
traffic flow data



CO2 estimation



S. Bilotta, P. Nesi, "Estimating CO2 Emissions from IoT Traffic Flow Sensors and Reconstruction", Sensors, MDPI, 2022. <https://www.mdpi.com/1424-8220/22/9/3382/>





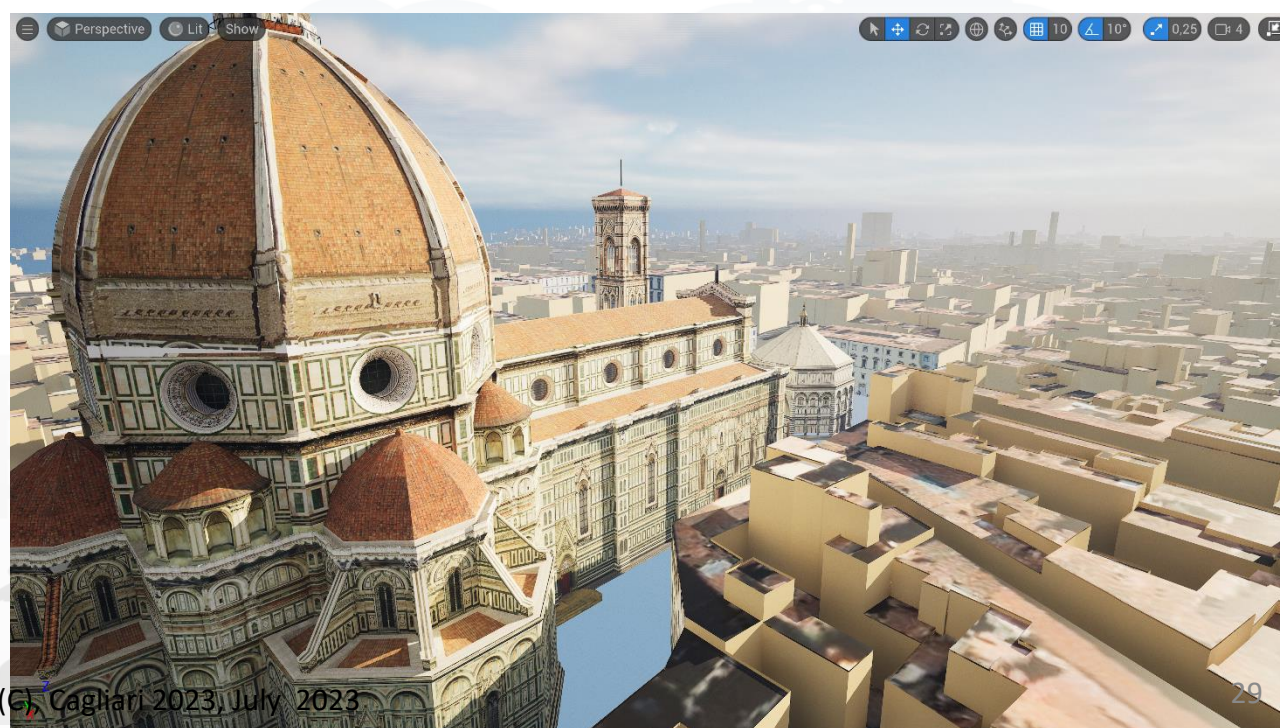
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# OCULUS









# 2022 booklets

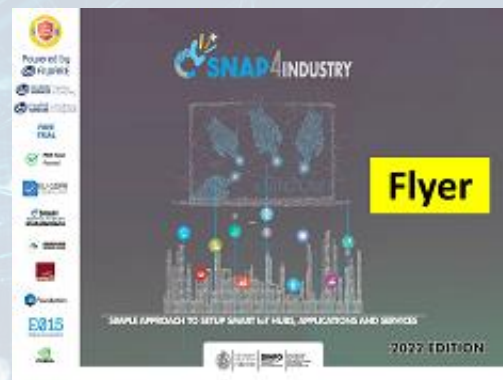


- Snap4City



[https://www.snap4city.org/download/video/DPL\\_SNAP4CITY\\_2022-v02.pdf](https://www.snap4city.org/download/video/DPL_SNAP4CITY_2022-v02.pdf)

- Snap4Industry



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- Solutions
- Data Analytics



[https://www.snap4city.org/download/video/DPL\\_SNAP4SOLU.pdf](https://www.snap4city.org/download/video/DPL_SNAP4SOLU.pdf)



- The **development of the SCDT of Florence** has been presented, highlighting its construction phases
- The development activities were carried out to fulfil a **series of requirements on data, on interactivity functionalities, and on the integration and distribution**
- Our SCDT is not limited to a **3D representation** of the city, but it includes **information coming from different sources**
- This capability to handle huge amount of data is powered by **Snap4City**, the IoT platform on top of which the SCDT is developed



TOP



*Be smart in a SNAP!*



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7-9 November 2023, Barcelona, Spain

Visit Snap4City in Hall 1

## CONTACT

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