











Powered by FIWARE





Paolo Nesi, 20/09/2022
<a href="mailto:Paolo.nesi@unifi.it">Paolo.nesi@unifi.it</a>
<a href="https://www.snap4city.org">https://www.snap4city.org</a>

A Framework for rapid implementation of - Sustainable Smart Solutions - Decision Support Systems as a no-coding, low-coding

SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES









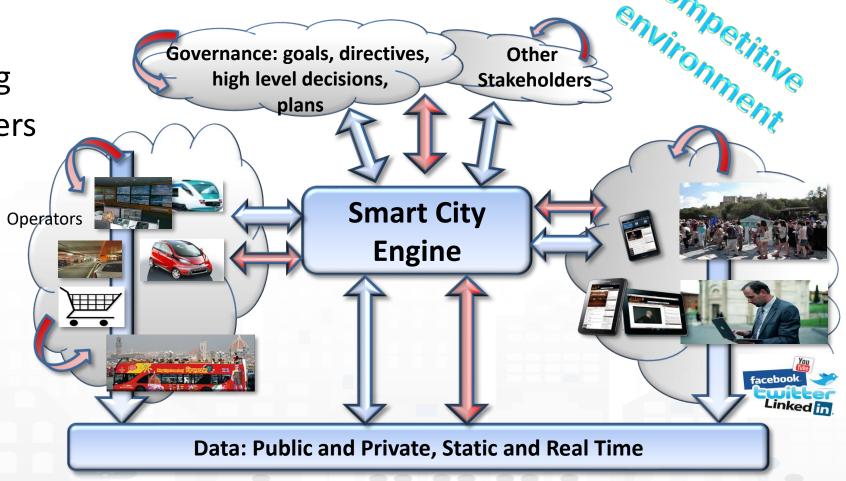






# From Strategies to (re-)Actions

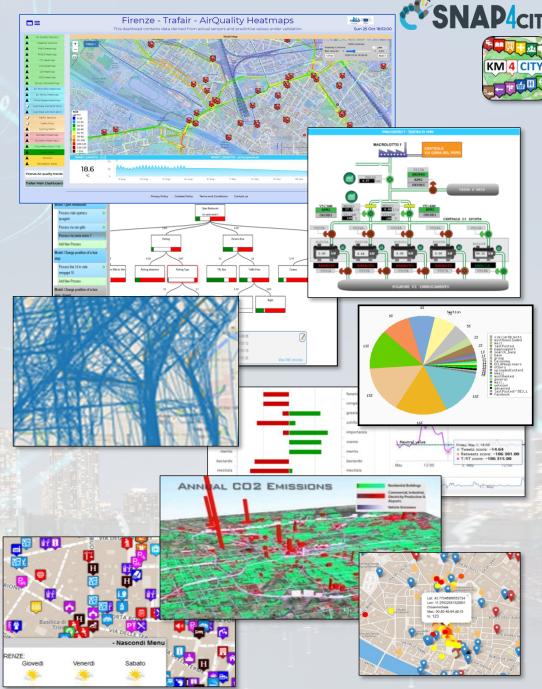
- Analyze
- Alerting, Early Warning
- Support Decision makers
- Plans
- Prescriptions
- Inform
- Suggest
- Engage
- Research



# Data Driven Decision Support

- Decision Support system
- Assessment / Strategies
- Data Rendering, visual analytics
- Data Processing
- Data aggregation, Storage, indexing
- Data Ingestion





### **Domains**

- Smart City, control room
- Green Deal, smart light, ..
- Environment, pollutant, ...
- Mobility and transport
- Tourism and People
- Energy , Industry
- Social Media
- Big Data
- Artificial Intelligence
- Public and private data

















































#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**





# https://www.Snap4City.org











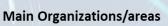


#### 8 running installations in Europe

- Toscana, Pisa, Sweden, ISPRA, Snap4.eu,
- Altair, Italmatic, Denmark, ....
- 13 projects, 12 pilots on 10 Countries
  - >40 cities/area

#### Wide MULTI-tenant deploy, e.g.,

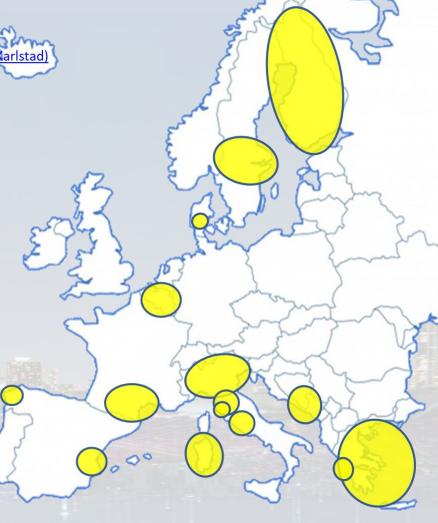
- 18 Organizations / tenant
- > 7400 users on
- > 1400 Dashboards
- > 16 mobile Apps
- > 2 Million of structured data per day
- > 520 IoT Applications/node-RED
- > 700 web pages with training
- > 60 videos, training videos



- Antwerp area (Be)
- Bologna (I)
- Capelon (Sweden: Västerås, Eskilstuna, Karlstad)
- DISIT demo (multiple)
- Dubrovnik, Croatia
- Firenze area (I)
- Garda Lake area (I)
- Greece (Gr)
- Helsinki area (Fin)
- Livorno area (I)
- Lonato del Garda (I)
- Modena (I)
- Mostar, Bosnia-Herzegovina
- Oslo & Padova (Impetus)
- Pisa area (I)
- Pistoia (I)
- Pont du Gard, Occitanie (Fr)
- Prato (I)
- Roma (I)
- Santiago de Compostela (S)
- Sardegna Region (I)
- Siena (I)
- SmartBed (multiple)
- Toscana Region (I), SM
- Valencia (S)

EUROPEAN OPFI

- Venezia area (I)
- WestGreece area (Gr)



9 Node-RED

Trials in Israel, Colombia, Brasile, Australia, India, Romania, etc.

### How to adopt Snap4City



Powered by





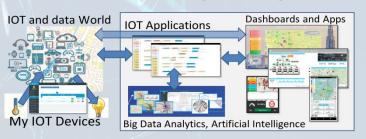


#### **Smart City as a Service**

- Supporting Org
- 100% Open Source Platform: Github
- Further developments
- **Publishing Appliances and Dockers**
- Training courses, docs
- Consulting
- **Forums**
- Etc.



#### On your premise

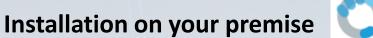






- Different configurations
  - From small to scalable
  - Exploiting your legacy tools
  - Interoperable with any tool

- - Migrate on premise on the fly
- Start on Cloud into a sand box
  - Pass to install on premise what you need





**Download** 

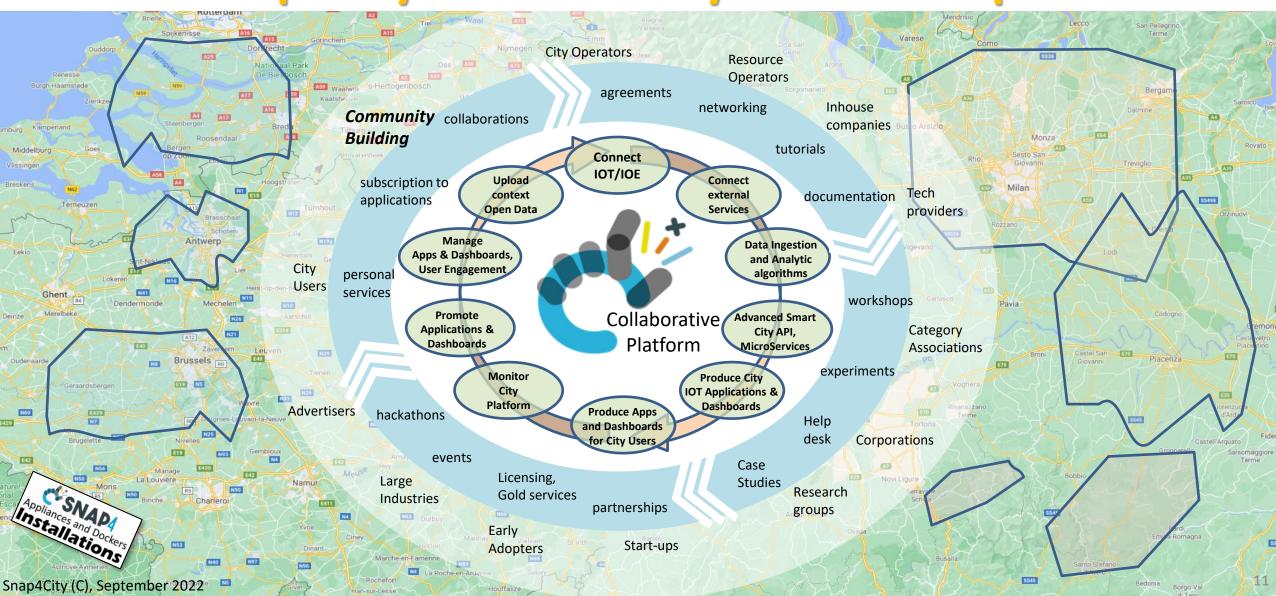
Snap4City (C), September 2022





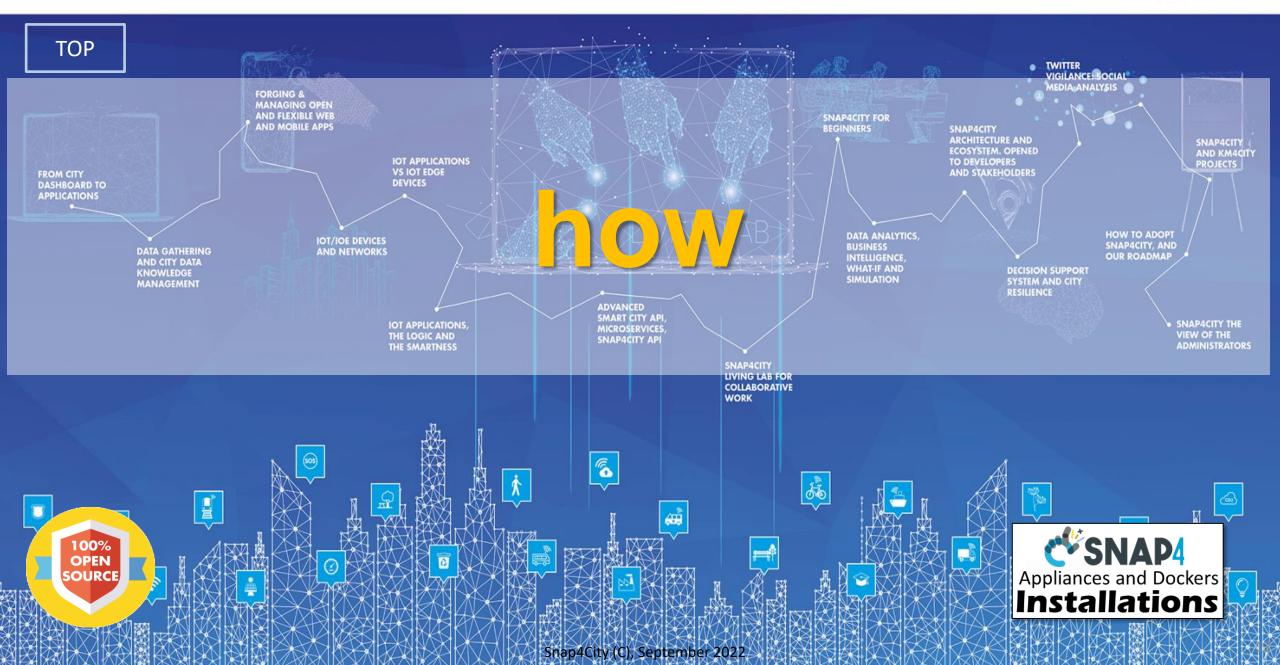


# DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB ACITY Platform may serve Multiple Cities



#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**







Powered by **S**FIWARE

> FREE TRIAL



















#### **SMART SOLUTIONS AND DECISION SUPPORT SYSTEMS**



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



**DASHBOARDS, WIDGETS TEMPLATES** 

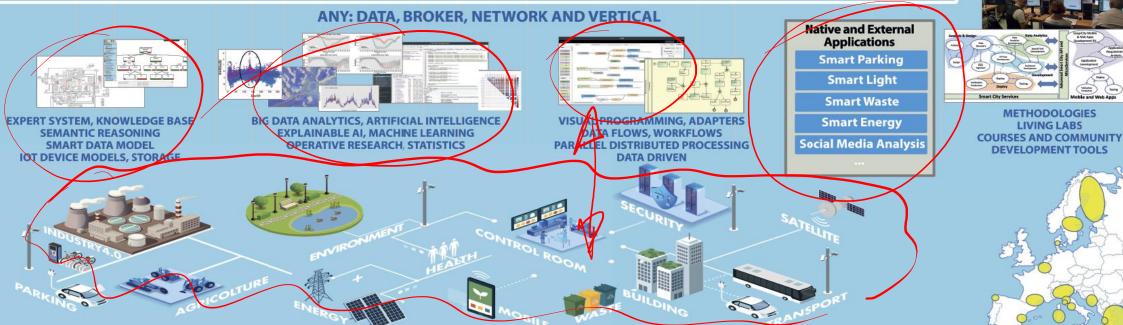
PREDICTION - ANOMALY DETECTION - CLUSTERING - ROUTING - SENTIMENT NLP - TRAFFIC FLOW PEOPLE FLOWS - SDG - 15 MIN CITY INDEX - KPI - HEATMAPS - ORIGIN DESTINATION - ETC...

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

**METHODOLOGIES** 

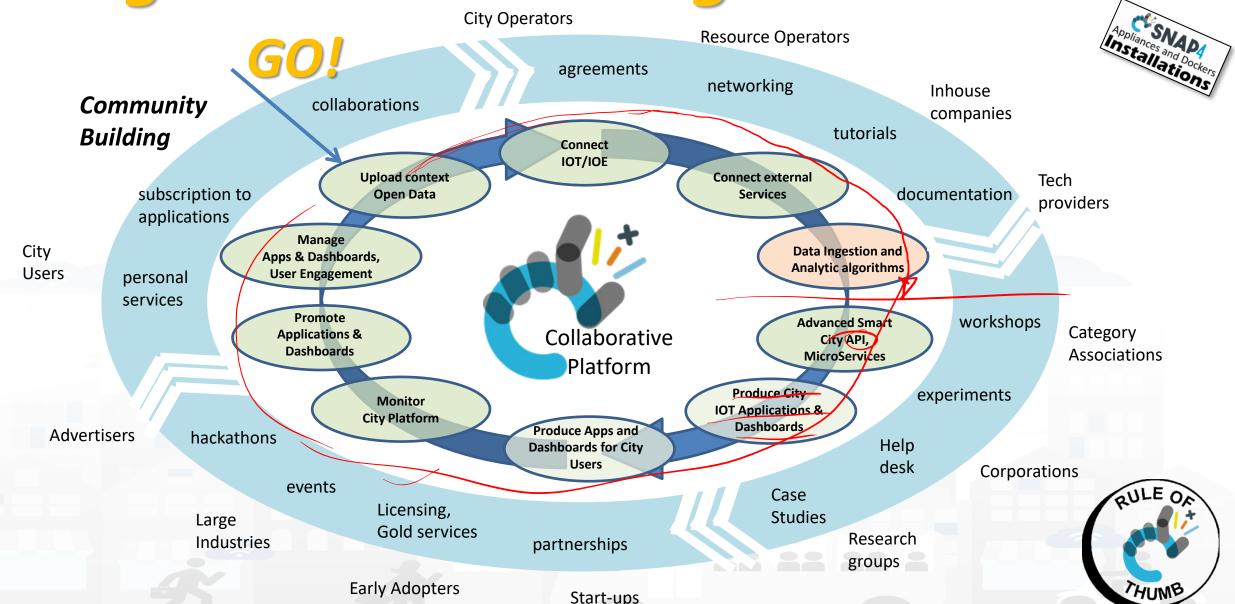
LIVING LABS

**DEVELOPMENT TOOLS** 



# Living Lab Accelerating





Ingestion, agg. -> exploitation

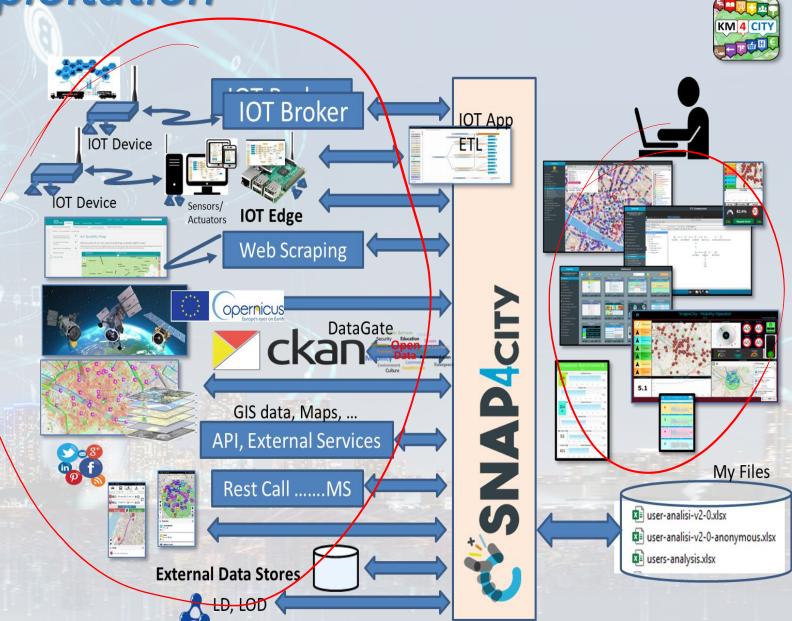








- Snap4City efficient tools for
  - Bidirectional data channels
  - Any format, any channel, any data, any broker, any protocol, ...
  - Km4City Knowledge base Ontology reasoning on geo, space, time, relationships



# Data Type Coverage

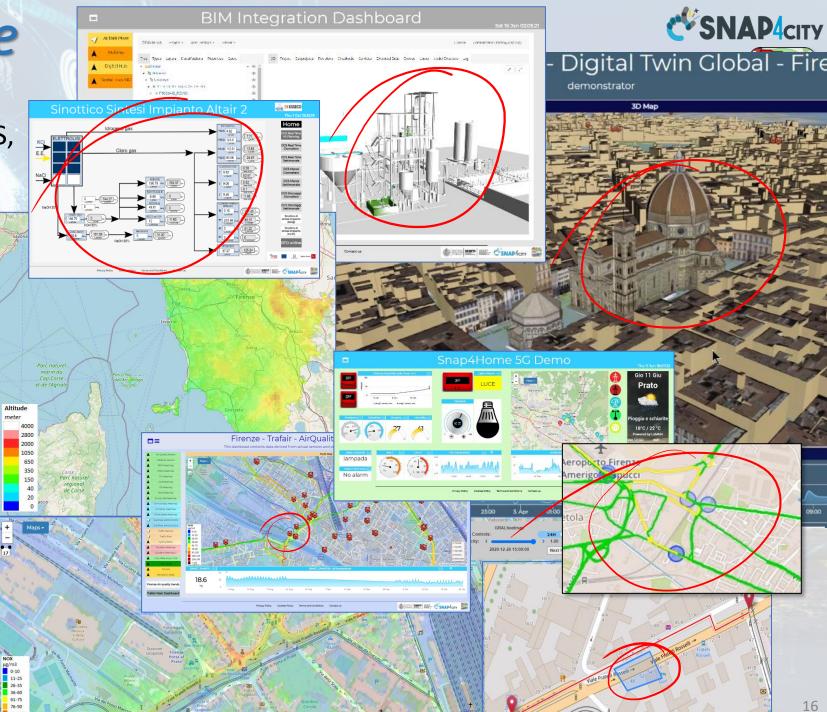
- POI, IOT Devices, shapes,...
- GIS, maps, orthomaps, WFS/WMS, GeoTiff, calibrated heatmaps, ..
- Satellite data, ...
- traffic flow, typical trends, ...
- trajectories, events, Workflow, ...
- 3D, BIM, ..
- Dynamic icons/pins, ...
- OD Matrices of several kinds, ...
- Synoptics, animations, ...
- KPI/personal KPI...
- social media data, TV Stream,
- routing, multimodal, constraints,
- decision scenarios, ....
- prediction models, ....
- etc.









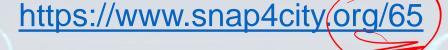


# Standards and Interoperability (9/2022)

### SNAP4city

#### **Compliant with:**

- IoT: NGSI V2/LD, LoRa, LoRaWan, MQTT, AMQP, COAP, OneM2M, TheThingsNetwork, SigFOX, Libelium, IBIMET/IBE, Enocean, Zigbee, DALI, ISEMC, Alexa, Sonoff, HUE Philips, Tplink, BACnet, TALQ, Protocol Buffer, KNX, OBD2, Proximus, ..
- **IoT model:** FIWARE Smart Data Model, Snap4City IoT Device Models
- **General**: HTTP, HTTPS, TLS, Rest Call, SMTP, TCP, UDP, SOAP, WSDL, FTP, FTPS, WebSocket, WebSocket Secure, GML, WFS, WMS, RTSP, ONVIF, AXIS TVCam, CISCO Meraki, OSM, Copernicus, The Weather Channel, Open Weather, OLAP, ....
- Formats: JSON, GeoJSON, XML, CSV, GeoTIFF, OWL, WKT, KML, SHP, db, XLS, XLSX, TXT, HTML, CSS, SVG, IFC, XPDL, OSM, Enfuser FMI, Lidar, glTF, GLB, DTM, GDAL, Satellite, D3 JSON, ...
- Database: Open Search, MySQL, Mongo, HBASE, SOLR, SPARQL, ODBC, JDBC, Elastic Search, Phoenix, PostGres, MS Azure, ...
- Industry: OPC/OPC-UA, OLAP, ModBUS, RS485, RS232,...
- Mobility: DATEX, GTFS, Transmodel, ETSI, ...
- Social:Twitter, FaceBook, Telegram, ...
- Events: SMS, EMAIL, CAP, RSS Feed, ...
- OS: Linux, Windows, Android, Raspberry Pi, Local File System, AXIS, ESP32, etc.

































# Expert System semantic queries

UNIVERSITÀ DEGLI STUDI FIRENZE DIPARTIMENTO DI DIPARTIMENTO DI

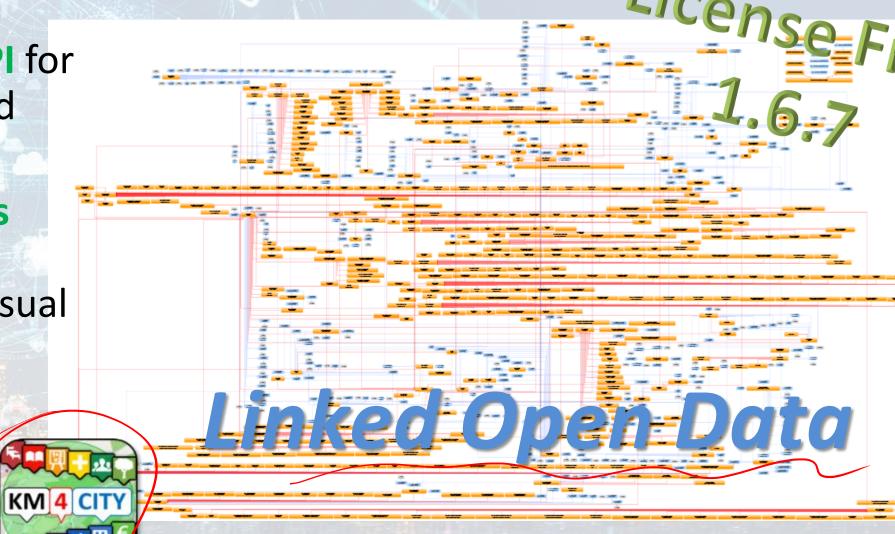




• via:

 Smart City API for Apps and third party

MicroServices
 data driven
 develop via visual
 language
 Node-RED



https://www.snap4city.org/19

# Ingestion, aggreg. > exploitation

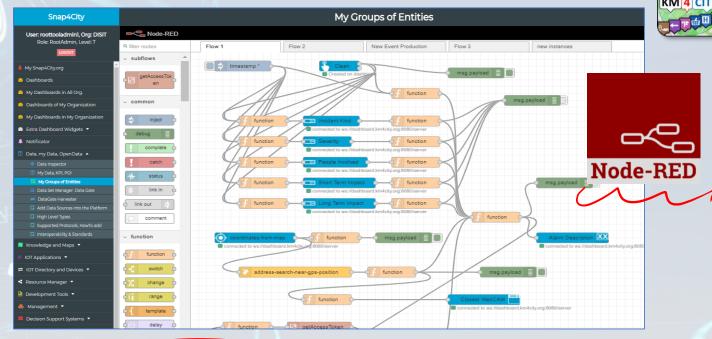


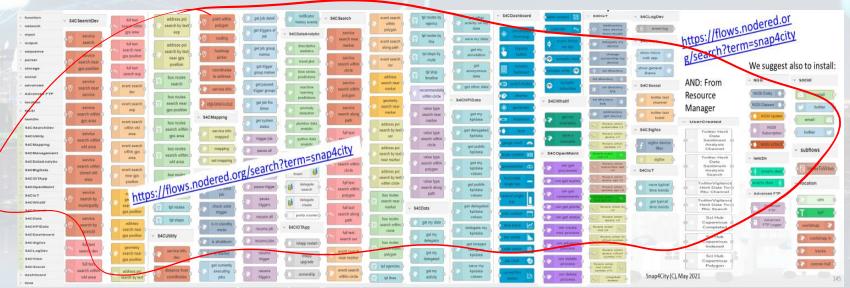






- IoT App Visual Programming, no coding
  - Data transformation
  - Integration, Interoperab.
  - Scripting Data Analytics
  - Data ingestion
  - Business logic
- Edge and Cloud
- MicroServices data driven develop via visual language Node-RED

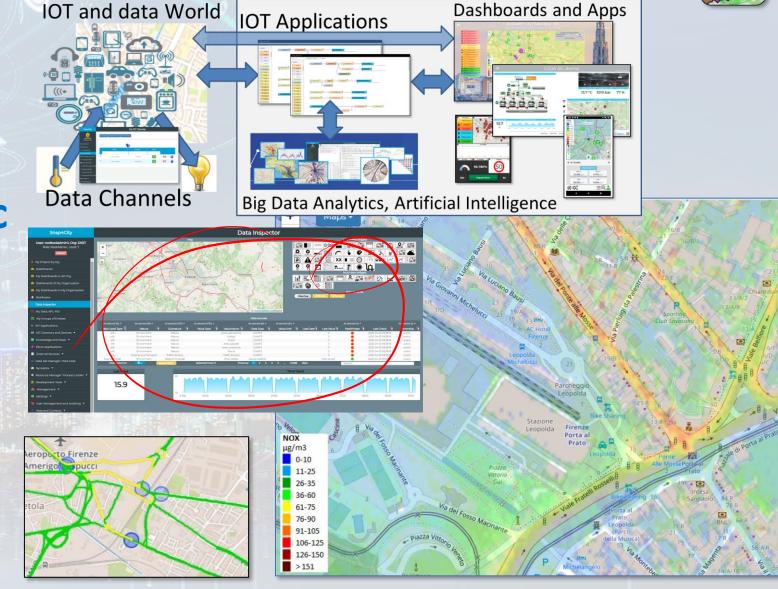




# Solutions: reliable, secure and fast to realize

KM 4 CITY

- 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



# Big Data Analytics + Artificial Intelligence

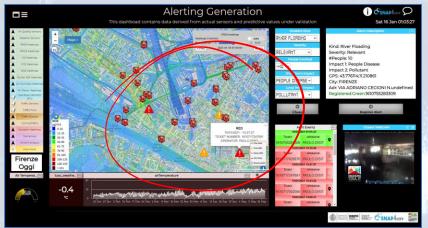
- Short and Long terms predictive models on:
  - traffic, parking, people flow, maintenance, land sliding NO2
- 3D Flow prediction: Pollutant (NOX, NO2, ...)
- Early warning, City Indexes, etc.
- AI & XAI:
  - RF, XGBoost, BRNN, RNN, SVR, DNN, LSTM, CNN-LSTM, BI-LSTM, Autoencoders, ...
  - Clustering: K-means, K-Medoid, ...
  - XAI: Shap, variations, ...
- Modelling, simulation, routing
  - Traffic Flow reconstruction
  - Constrained Routing
- What-IF analysis (simulation + AI + data)
- Based on several computational models:
  - trajectories, OD matrices, Typical Time Trends, etc.

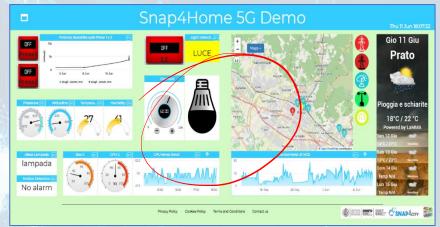


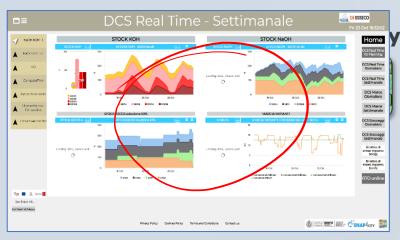
#### to cope with

- any data, format
- any channel, protocol
- any AI/ML
- any place
- online development
- multi-tenant
- Secure, PENTest
- GDPR, privacy
- → low costs
- → easy to evolve

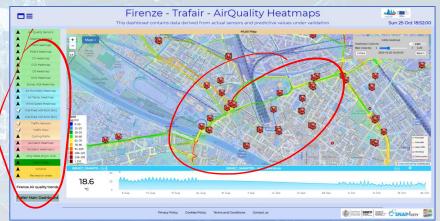






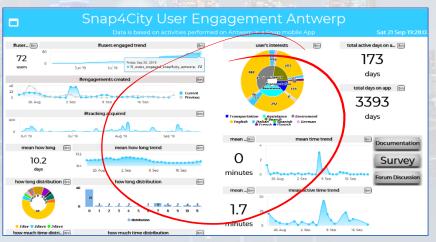


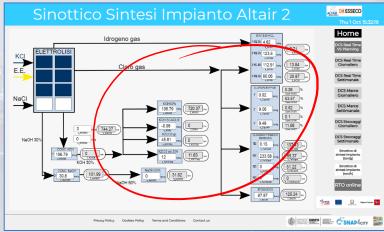








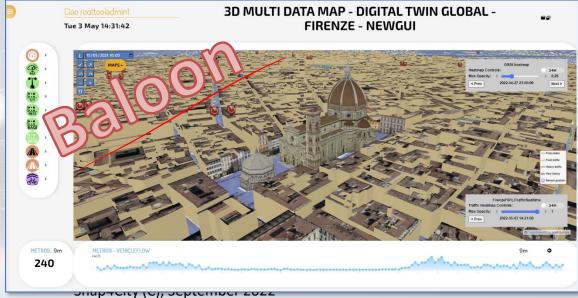




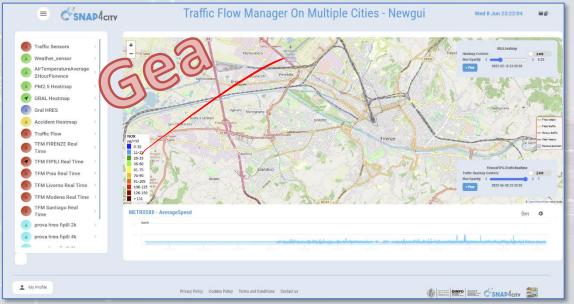
# Different Themes

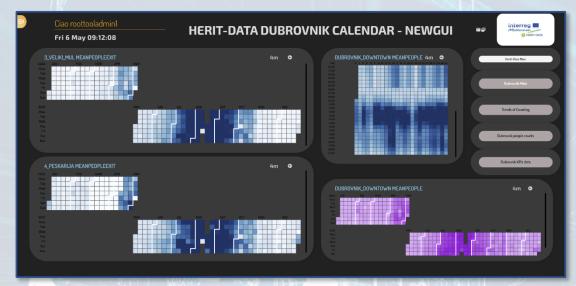


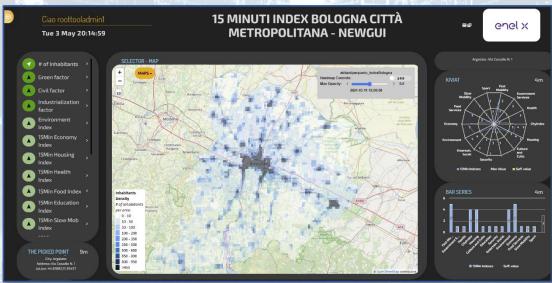






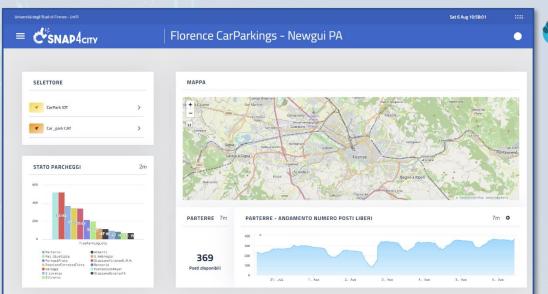




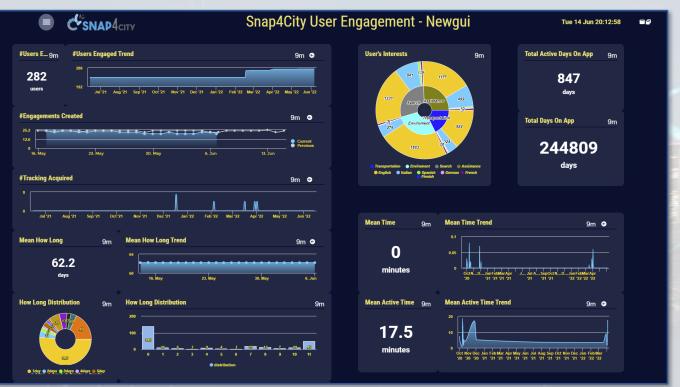


New styles/themes can be developed by specializing a few files from open source

https://www.snap4city.org/793







Snap4City (C), September 2022





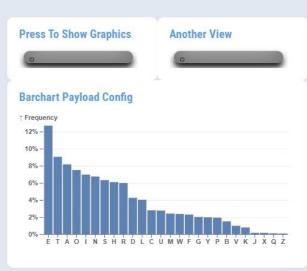


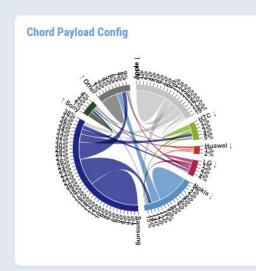


#### D3 Library Example

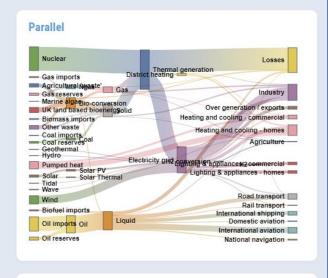


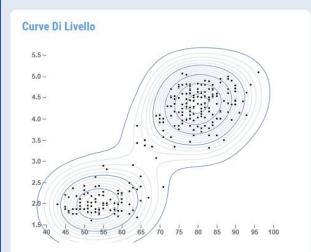


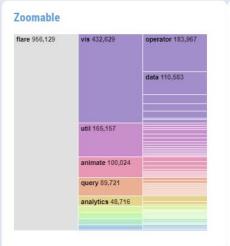


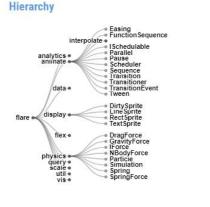


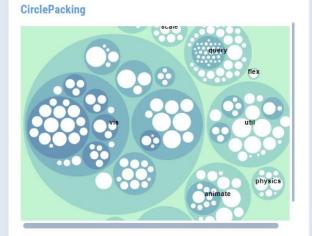












https://www.snap4city.org/dashboardSmartCity/view/Gea.php?iddasboard=MzQ4OQ==

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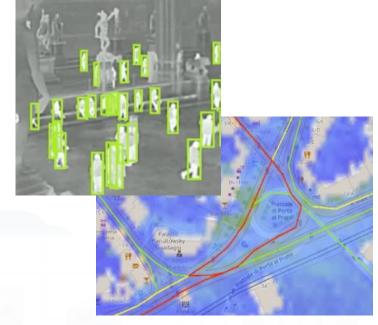


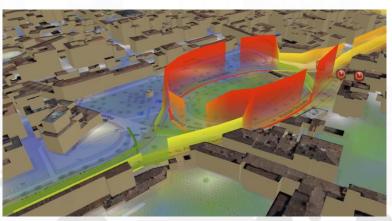


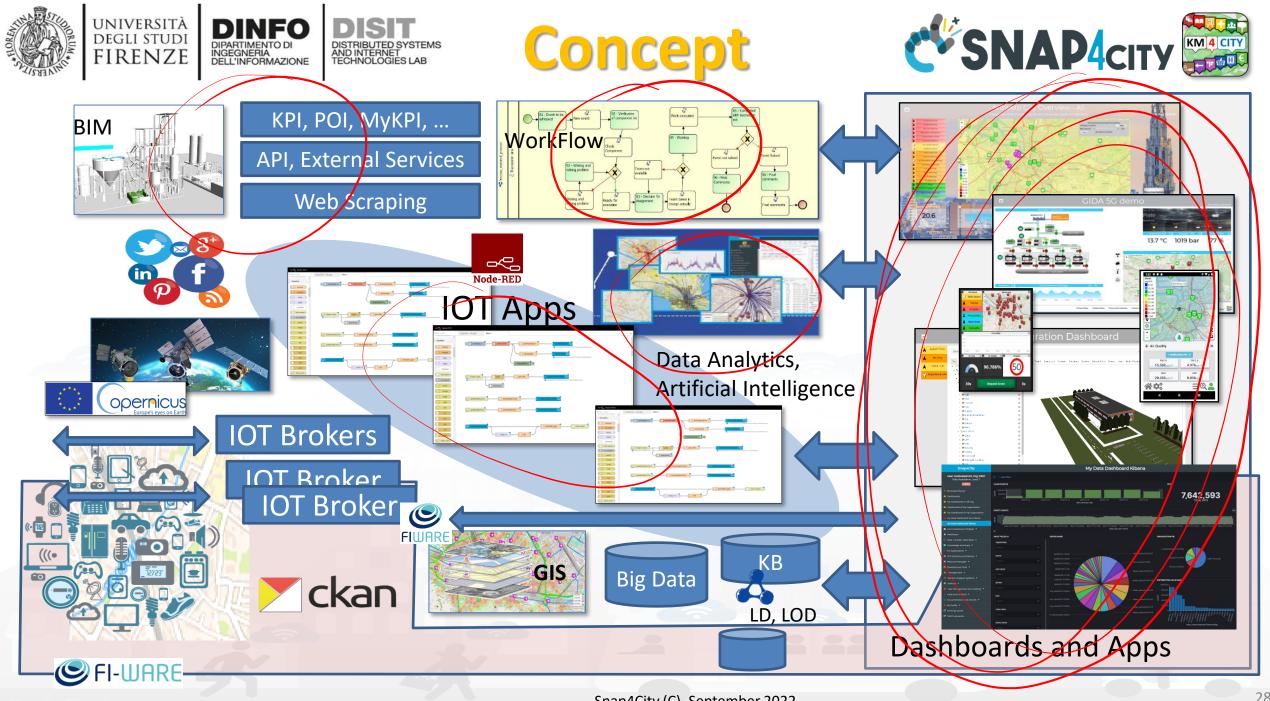


# L'SNAP4CI

- Awareness to manage and improve ....
- Infrastructures of the cultural cities:
  - Security and Safety: roads, buildings, squares
  - Mobility and Transport: traffic flow, parking, etc.
  - **Environment**: microclimate, predictions, assessment for acting
- Services / events: assessment and plan:
  - Most of the cities provide diffuse cultural heritage as a wall
    - Security, clean, public transport, environment, delivery, etc.
  - Global and Local: events vs actions
  - Local Structures: museums, events, shopping, attractions, ...
- People and Transport Means (city users: citizens, tourists, etc.) :
  - Understand:
    - flows, density, behaviour, classifications of user/means
    - reputation, appreciation Trip Advisor, Twitter, etc.
  - Suggest, Recommend, Engage, Guide...
    - Context based

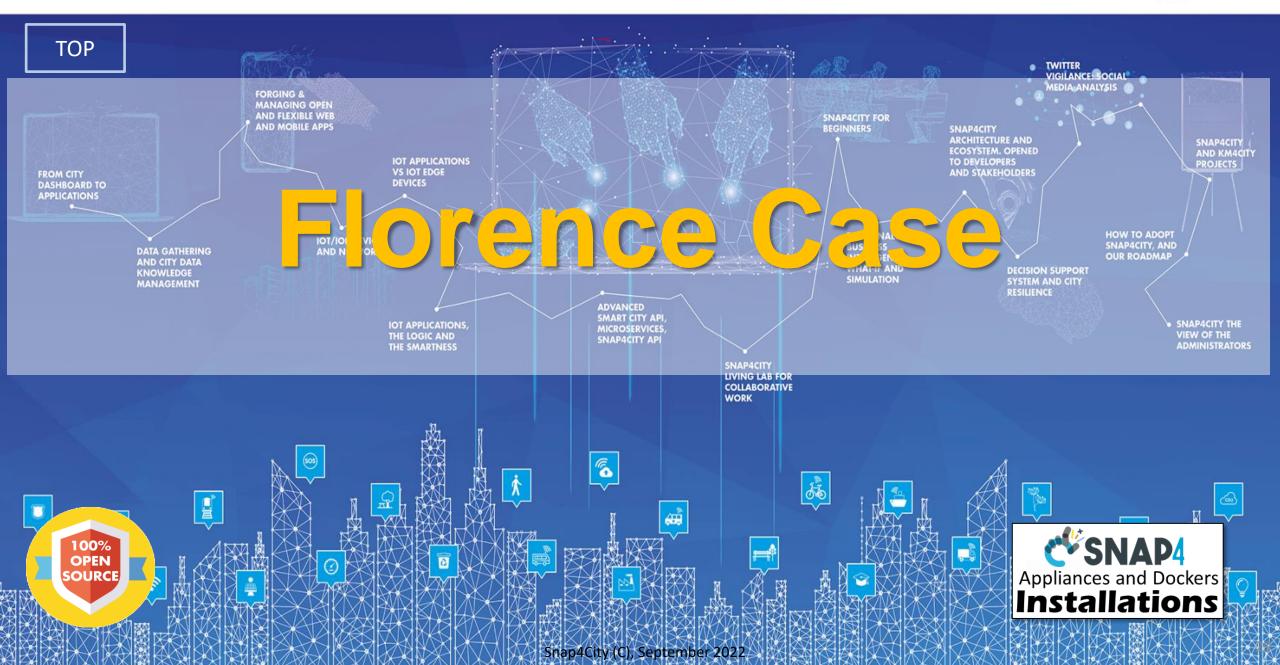






#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**





# **Smart City Control Room** Florence Metropolitan City







#### Multiple Domain Data

- Thousands of Open/Private data, POI, IOT, etc.
- mobility and transport: accidents, public transport, parking, traffic flow, Traffic Reconstruction, KPI, ...
- AND: environment, civil protection, gov KPI, covid-19, social & social media, people flow, tourism, energy, culture, ...

#### Multiple dash/tool Levels & Decision Makers

• Real Time monitoring, Alerting, quality assess.

Predictions, KPI, DSS, what-if analysis

#### Historical and Real Time data

Billions of Data

#### Services Exploited on:

Multiple Levels, Mobile Apps, API

• Since 2017















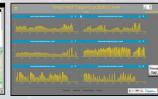




















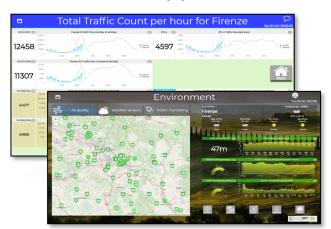




### Florence Case



- **Smart City Control Room**
- **Dashboards and Services**
- **Mobile App:** Firenze Where What





#### Mobility:

- quality of public transportation service (mean delay on bus-stops)
- public transport operators schedule and paths, routing, multimodal routing
- traffic flow reconstruction
- Smart parking: predictions
- Accidents and events, Log, heatmaps

#### **Environment:**

- smart irrigators
- smart waste
- Sensors: PM10. PM2.5,.....
- Heatmaps: PM10, PM2.5, ....
- **NOX** predictions

#### Energy:

- recharging stations (fast and reg.)
- consumption meters (smart info)
- smart light, street lights



#### Social:

- smart benches
- Twitter monitoring, Sentiment analysis, NLP text
- TV camera streams

#### **People Flows:**

- Wi-Fi, people flow
- Origin destination matrices

#### **Governmental and Communications:**

- KPI of the City
- **Digital Signage**
- Civil protection, Resilience (Resolute)

#### **Tourism and Culture:**

POI, etc.

#### **Analysis:**

- what-if routing, scenarios,
- traffic flow, environmental predictions



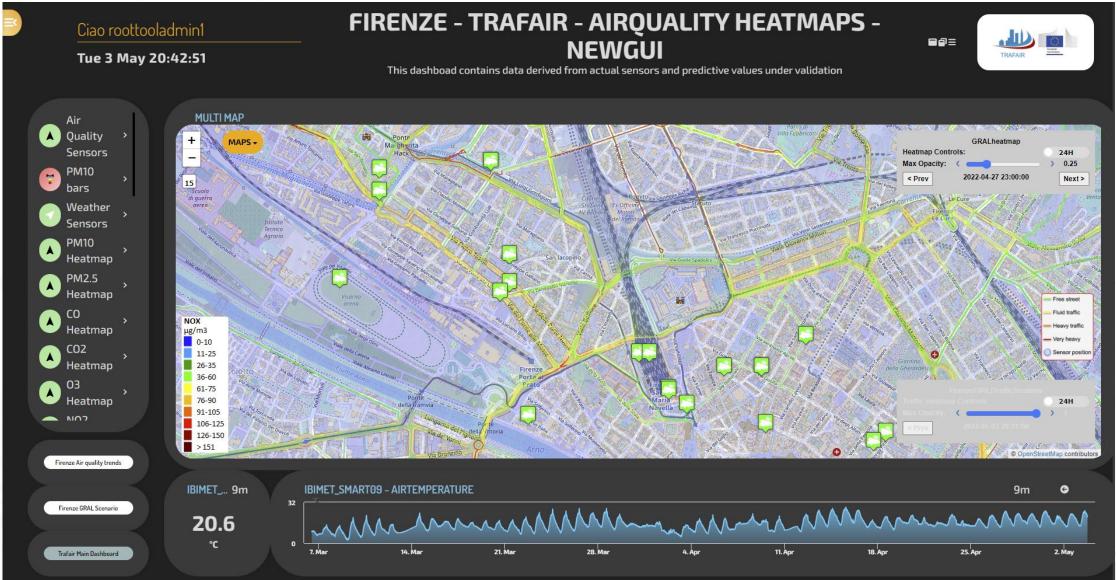
































#### Ciao roottooladmin1

**3D MAP DECK TEST-NEWGUI** demonstrator







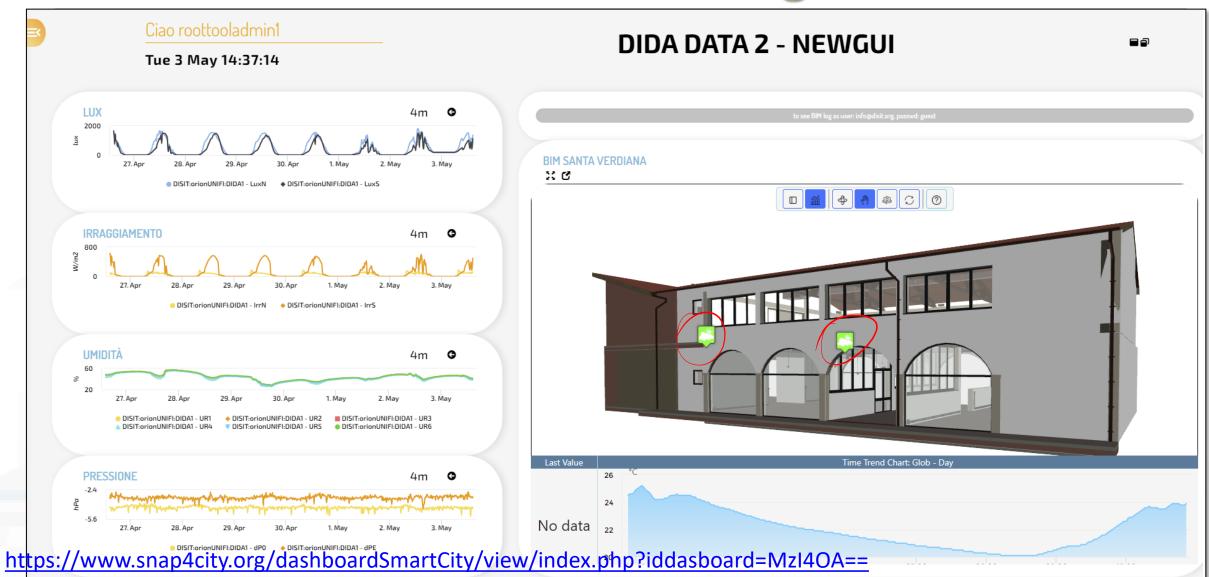






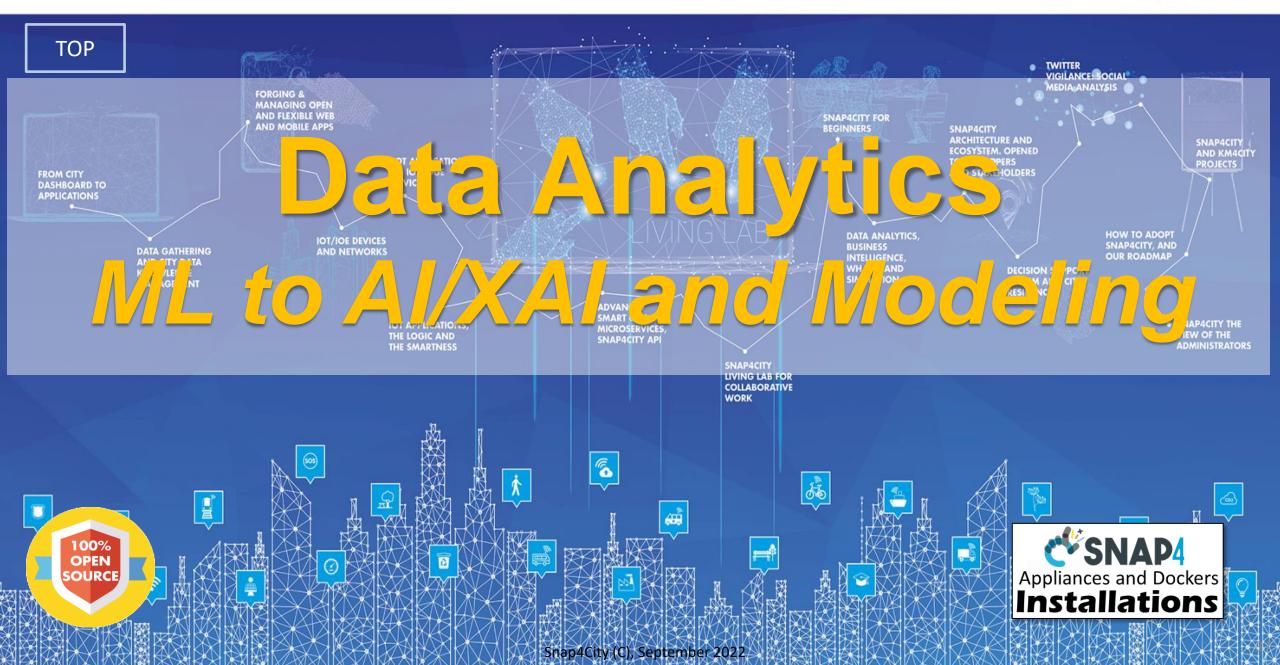
### **Smart Building**





#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**

























#### 15 Minute City Index:

13 differente subindexes



- Monitoring and Prediction of energy consumption
- Stimulating: Bike sharing, e-bikes, car charge, etc.



- Smart City infrastructure: monitoring and resilience
- Effective and Low cost smart solutions
- What-if analysis, Simulations





Monitoring and Predictions for

- NO2, NOX, CO2, Traffic flow, pollutant, landslide, etc.
- Traffic flow reconstruction



- Industry 4.0 integrated solutions
- **Decisions Support Systems**
- Process optimization
- Predictive maintenance



- Monitoring resource consuption,
- business intelligence tools for decision makers,
- Reduction production costs



- Shortening justice time
- Predictiction of mediation proneness
- Ethical Explainable Artificial Intelligence





# Mobility and Transport









## Available DATA ANALYTICS (1)

#### • Mobility and Transport

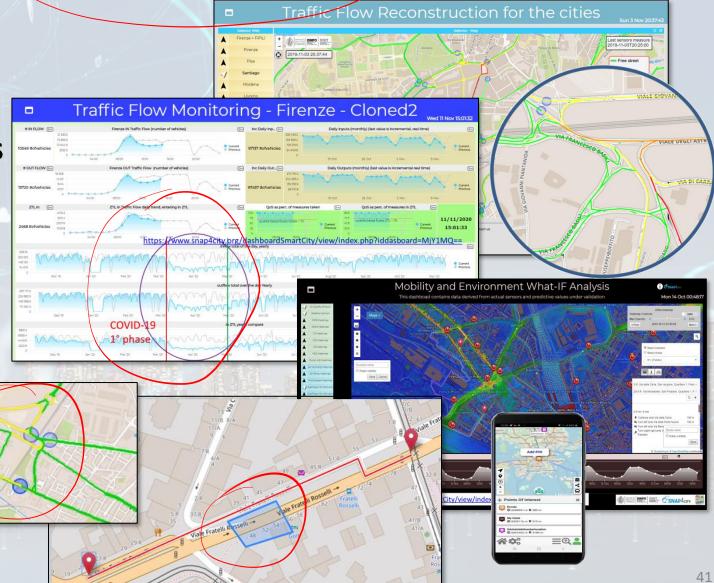
- What if analysis: routing, traffic flow, demand vs offer, pollutant, etc. (Simulation + ML)
- Traffic flow reconstruction from sensors and other sources (simulation + ML)
- **Predictions** for: traffic flow, smart parking, smart bike sharing, people flows, etc. (ML, DL)
- Public Transportation: Ingestion and modelling of GTFS and Transmodel
  - 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)
- Tracking fleets, people, via devices: OBU, OBD2, mobile apps, etc.
- Routing and multimodal routing (multistop travel planning), constrained routing, dynamic routing
- Computing Origin Destination Matrices from different kind of data (analysis)
- Computing typical trajectories on the basis of tracks (analysis, ML)
- Computing Messages for Connected drive
- Slow and Fast Mobility 15 Minute City Indexes (analysis, 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

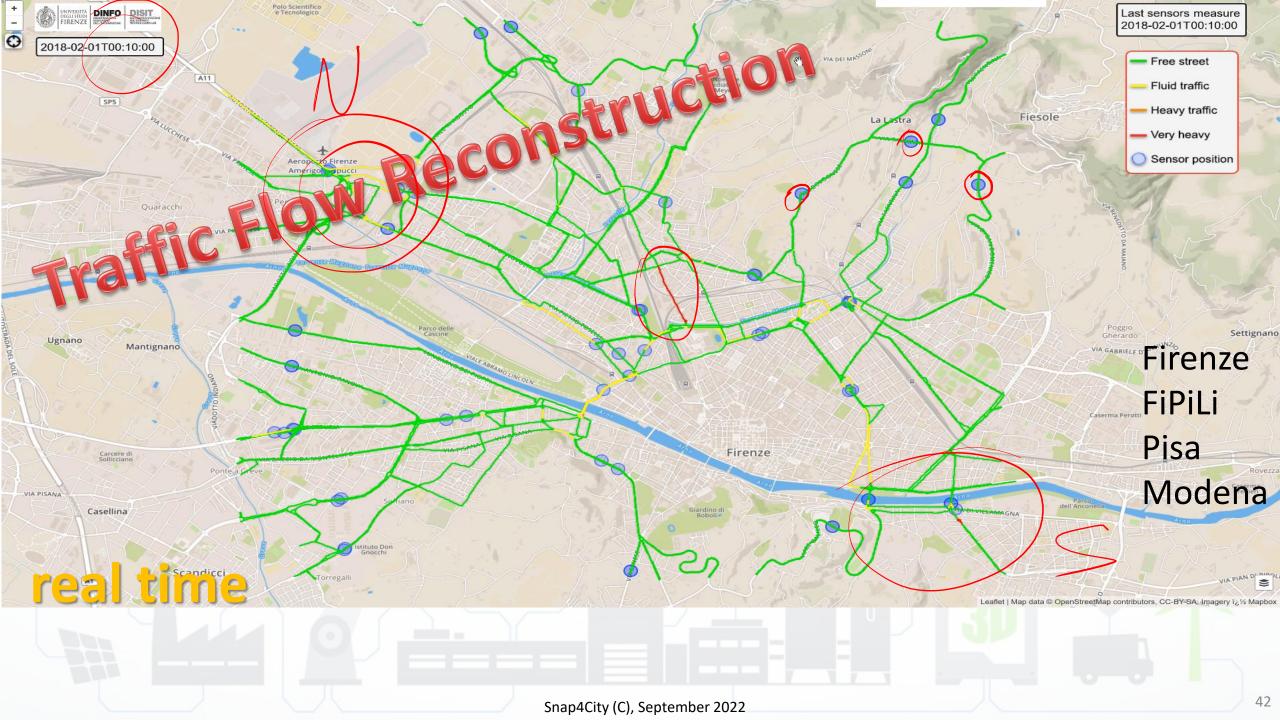
## Mobility and Transport Traffic Flow Analysis

- Multiple Domain Data
  - Traffic Flow sensors, city structure, weather
- Decision Makers Multiple Locations
  - Real time Monitoring, predictions
  - Traffic Flow Predictions,
  - Traffic Reconstructions, routing
  - Dashboards, What-IF analysis
  - Mobile App, people flows
- Historical and Real Time data
- Services Exploited on:
  - Dashboards, Mobile App
- Since 2017, 2019

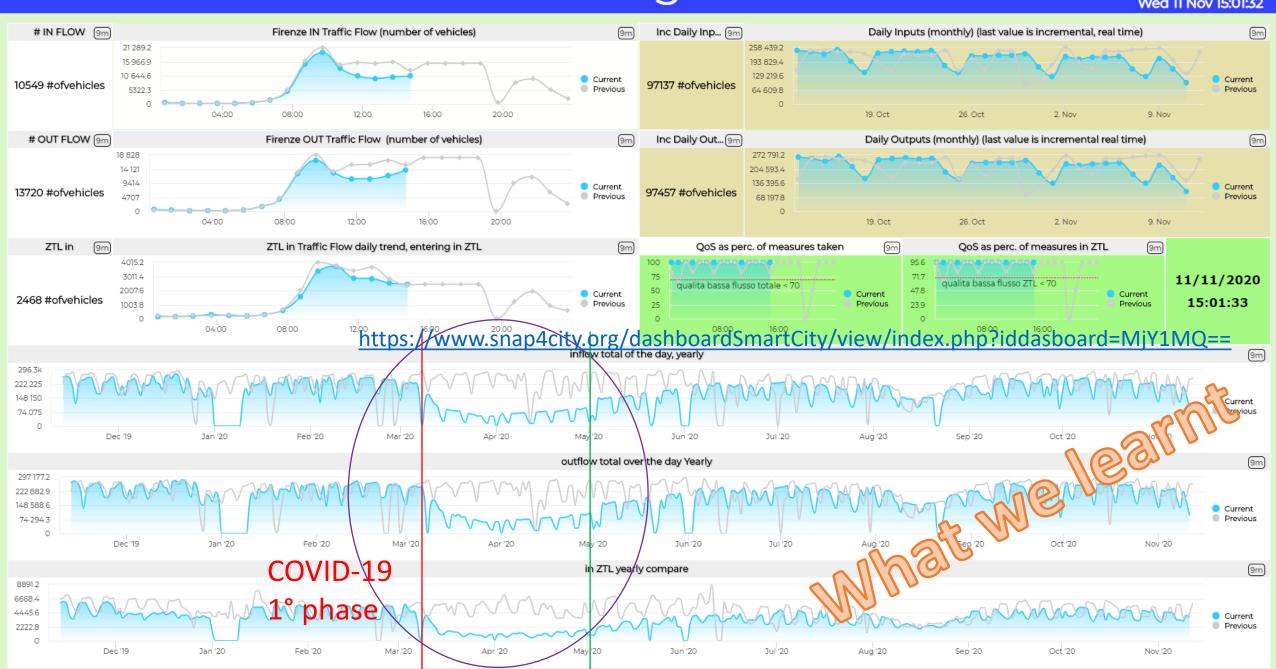
Cities: Firenze, Pisa, Livorno, Modena, Santiago di Compostela







## Traffic Flow Monitoring - Firenze - Cloned2

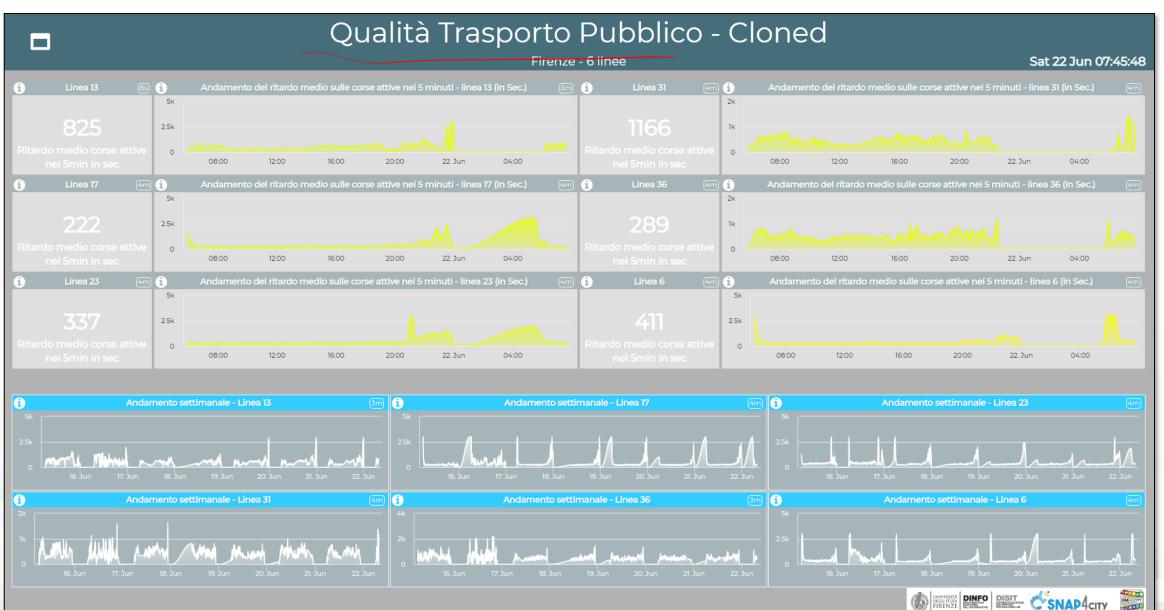














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Categ

ory

of free slot data

**Features** 

Free parking slots

Time

Month

Day

Day week

Weekend



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

Description of features variable

Real number of available slots recorded

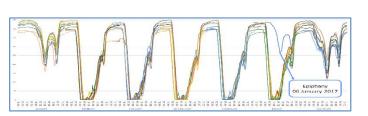
every 15 minutes

Hours and minutes

Month of the year (1-12)

Day of the month (1-31) Day of the week (0-6)

0 for working days, 1 else



	S	Previous	Difference between the number of free		
	■ ■	observation's	spaces at time $i$ and number of free		
	ළි	difference	spaces at time $(i - 15 \text{ minutes})$ recorde		
	Baseline features	(POD)	in the previous week		
	ill.	Subsequent	Difference between the number of free		
	ase	observation's	spaces at time $i$ , and the number of free		
	Ф	difference	spaces at time $(i + 15 \text{ minutes})$ recorded		
		(SOD)	in the previous week		
	Weather features	Temperature	City temperature measured one hour earlier than Time (°C)		
		Humidity	City humidity measured one hour earlier than Time (%)		
		Rainfall	City rainfall measured one hour earlier than Time (mm)		
	Fraffic Sensors features	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	Average of distance between vehicles,		
		Vehicle Time	over one-hour period		
	$\Box$	Vehicle	Number of vehicles per kilometer, over		
		Concentration	one-hour period		

Artificial Intelligence
Predictions

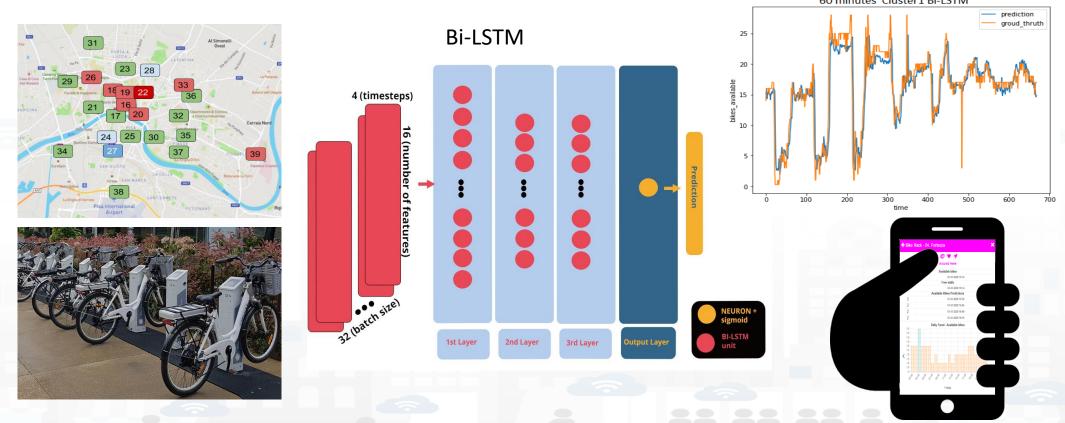
97% of precision







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



E. Collini, P. Nesi and G. Pantaleo, "Deep Learning for Short-Term Prediction of Available Bikes on Bike-Sharing Stations," in IEEE Access, vol. 9, pp. 124337-124347, 2021, doi: 10.1109/ACCESS.2021.3110794. https://ieeexplore.ieee.org/abstract/document/9530580

## Tuscany Region

### **SNAP4**CITY

Dashboards & Services:

 Mobility: public transport operators schedule and paths, traffic Fi-Pi-Li main road, parking status and predictions, traffic sensors, Origin Destination matrix, routing, multimodal routing, etc.

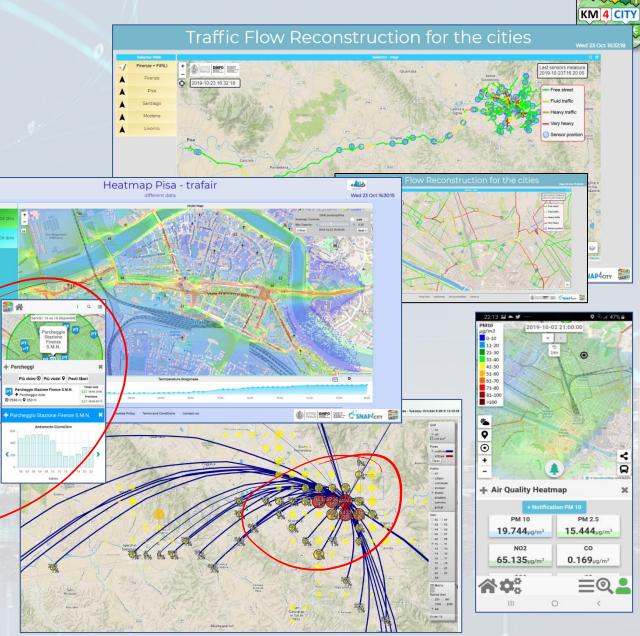
• Social: Hospitals and triage, etc.

• Environment: sensors, heatmaps,

alerting,

Pollution Forecast: NOX, NO2

- Weather Forecast,
- Culture and Tourisms
- Etc.
- Mobile App and MicroApplications:
  - Tuscany in a Snap (all stores)
  - Tuscany where what... km4city (all stores)
- Numbers: 1.5 M complex events per day Snap4City (C), September 2022









# City User behaviour analysis









## **Available DATA ANALYTICS (2)**

- City Users and Social
  - People detection and classification: persona, carts, bikes, etc. (ML, DL)
  - people counting and tracking (via thermal cameras, ML, DL)
  - People prediction: wifi, mobile, etc.
  - People counting via head counting (via thermal cameras, ML, DL)
  - People flows prediction and reconstruction, (ML, DL)
    - Wi-Fi data, mobile apps data, Mobile Data, etc.
  - User engagement and suggestions for sustainable mobility (Rule Based, ML)
  - User's behaviour analysis,
    - origin destination matrices, hot places, time schedule, Recency and frequency, permanence, typical trajectory, etc.
    - People flow analysis from PAX Counters and heterogenous data sources
  - 15 Minute City Index, etc. (modeling and computability)









## The App is a Bidirectional Device

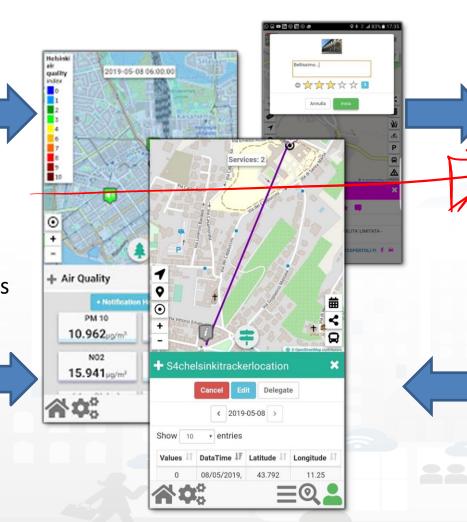
- GPS Positions
- Selections on menus
- Views of POI
- Access to Dashboards
- searched information
- Routing
- Ranks, votes
- \ Comments
- Images
- Subscriptions to notifications
- •

#### **Produced information**

- Accepted ?
- Performed?

•





#### **Derived** information

- Trajectories
- Hot Places by click and by move
- Origin destination matrices
- Most interested topics
- Most interested POI
- Delegation and relationships
- Accesses to Dashboards
- Cumulated Scores from Actions
  - Requested information
- Routing performed
- •

#### **Produced information**

- Suggestions
- Engagements
- Notifications











## To propose suggestions and Engage city user I need to know how they are moving



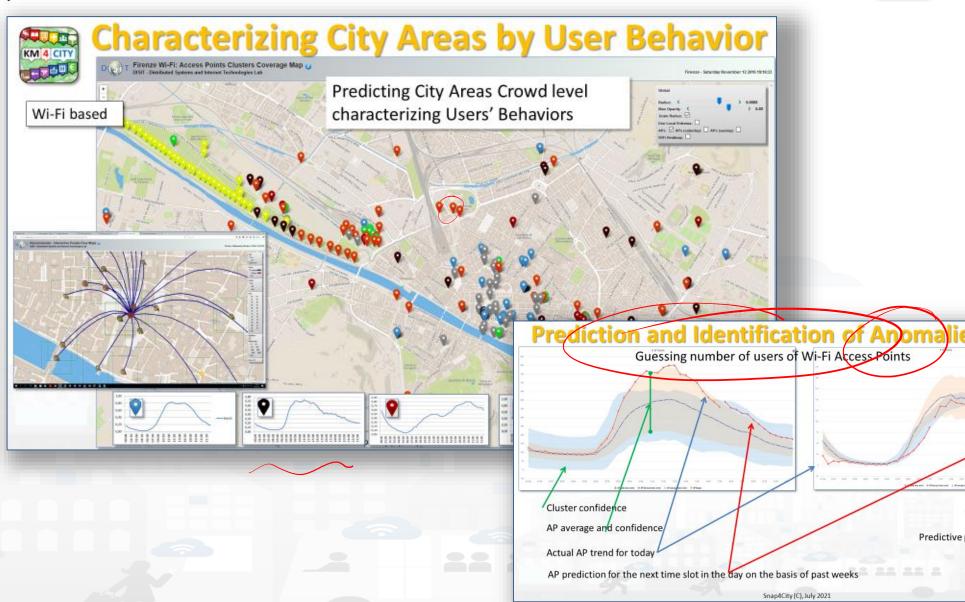




## **People Flows**



- Prediction of people flows on the basis of Wi-Fi data
- Anomaly detection
- Resolute H2020
- Classification of city areas



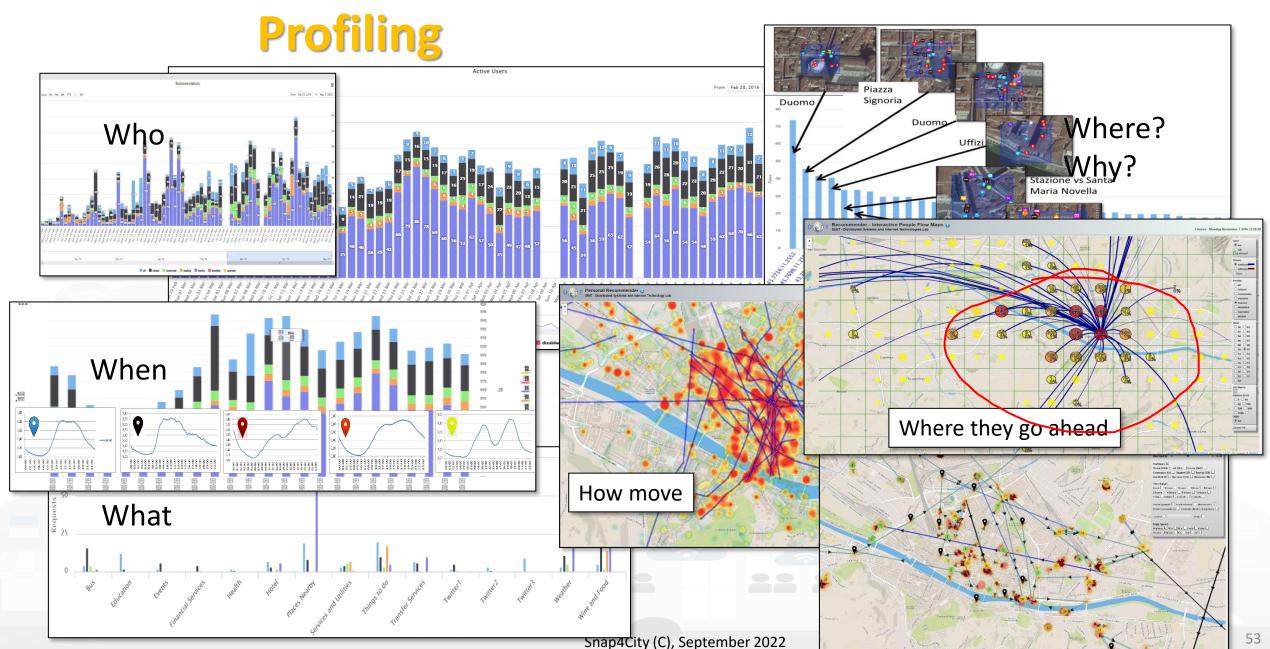






## **User Behavior Analyser for Collective**

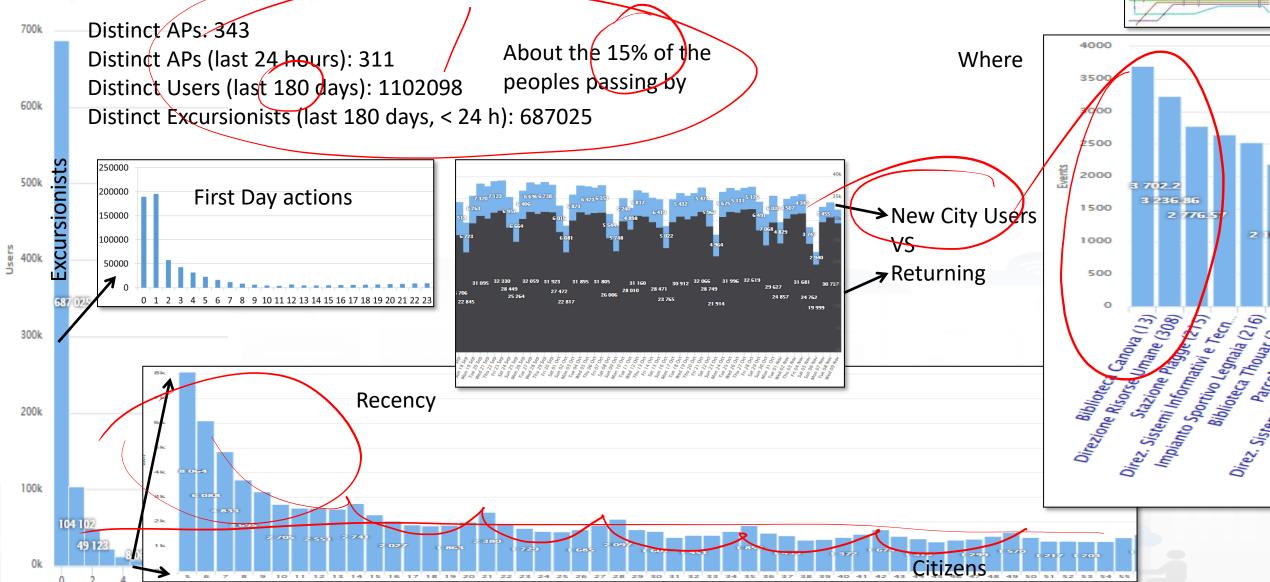






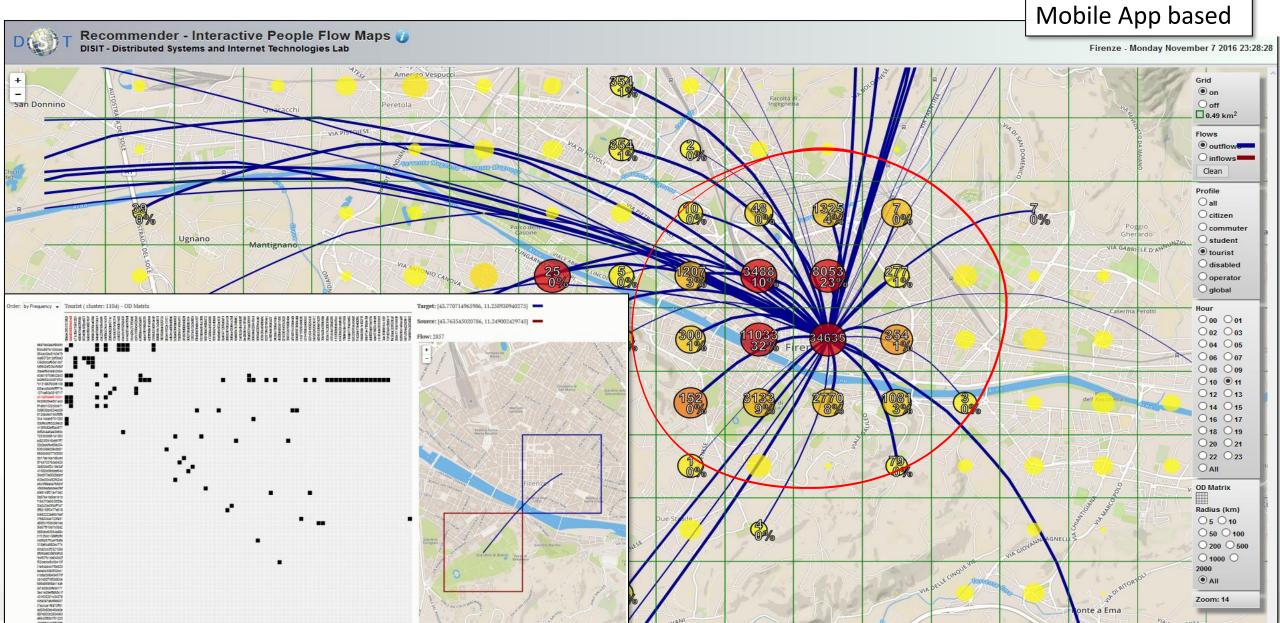
## **User Behaviour Analysis**

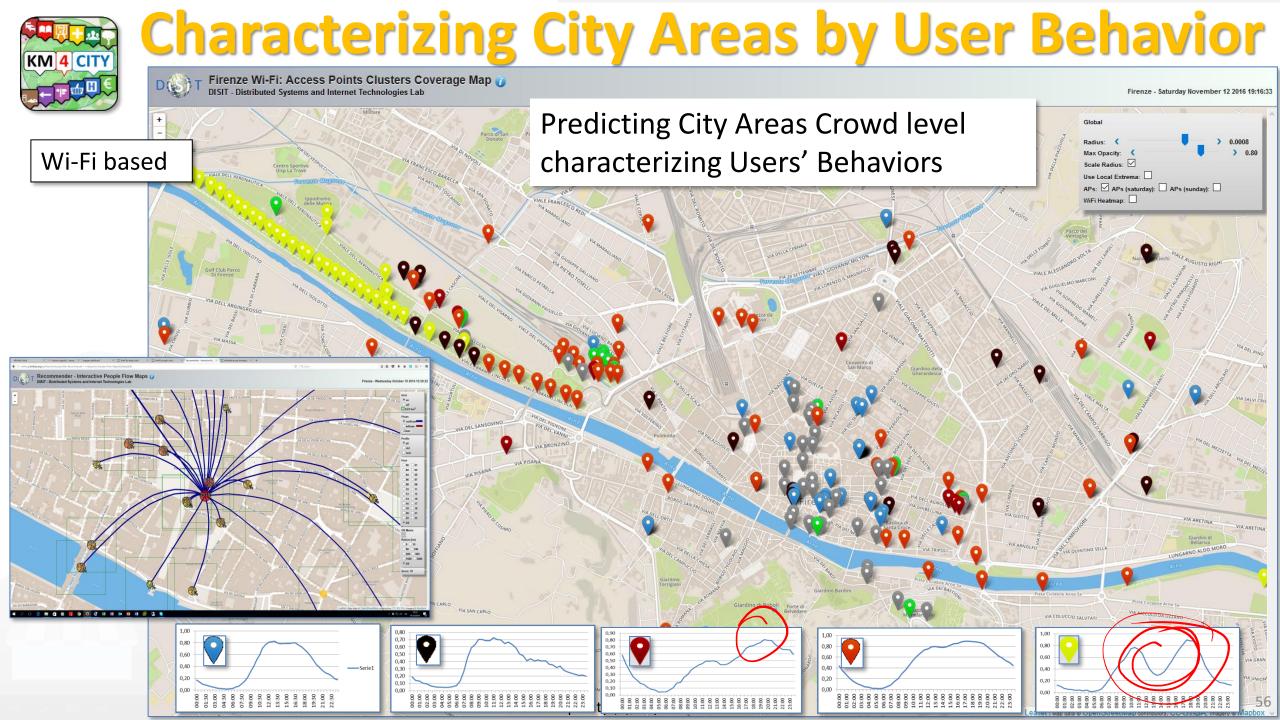




## Scalable multiresolution OD matrix





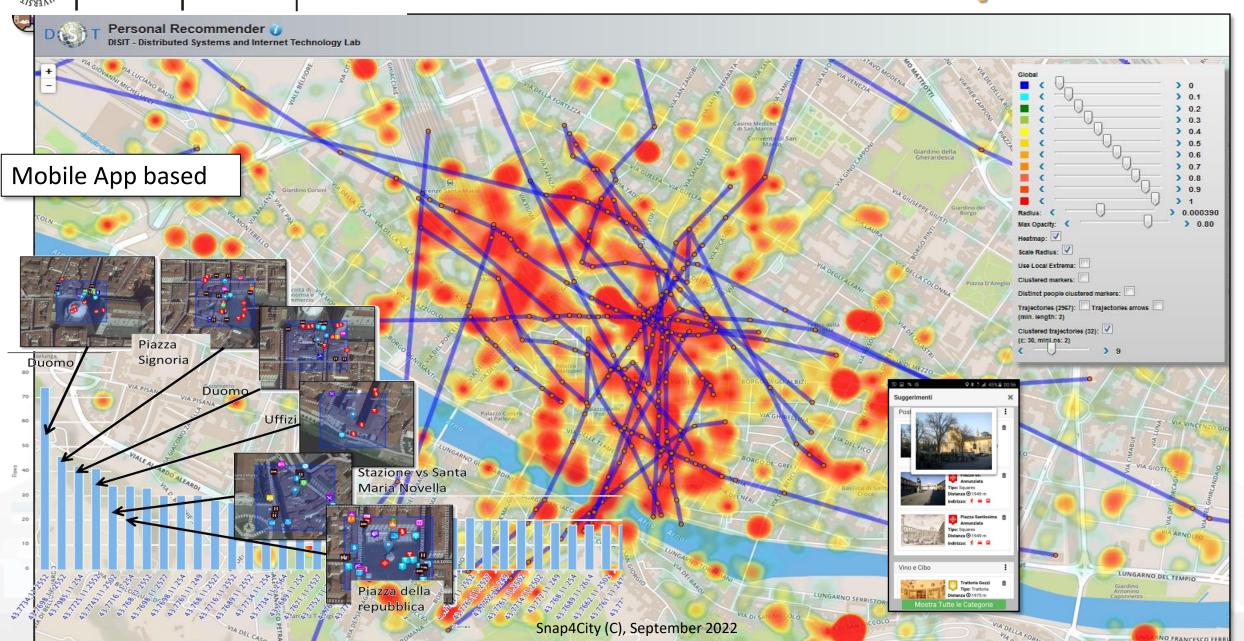






## **User Behavior Analyzer**











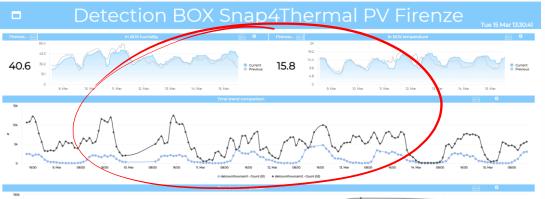








## A view and data from the Thermal Camera









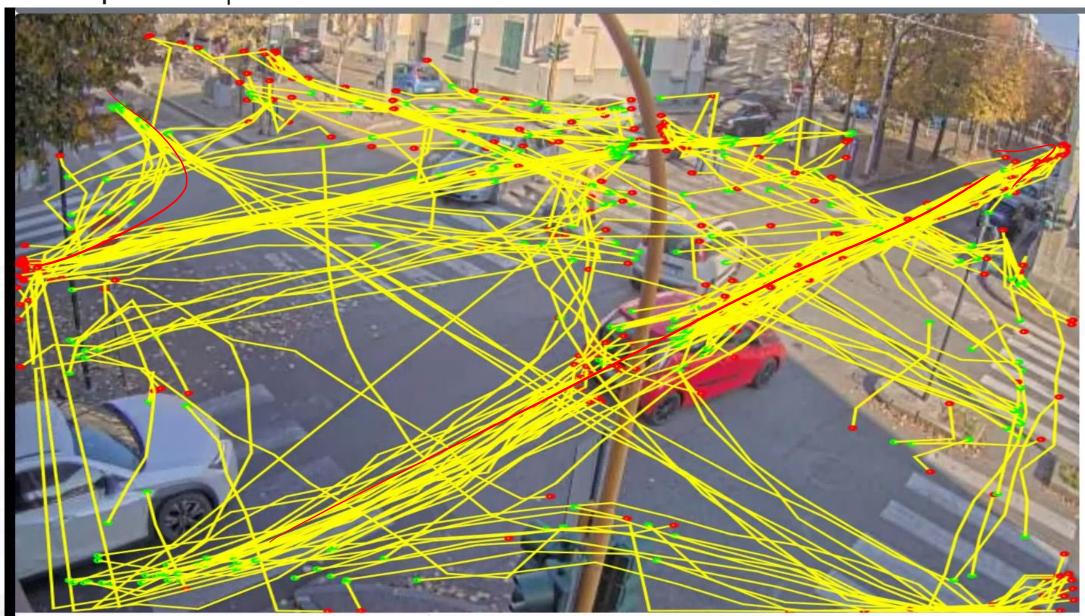








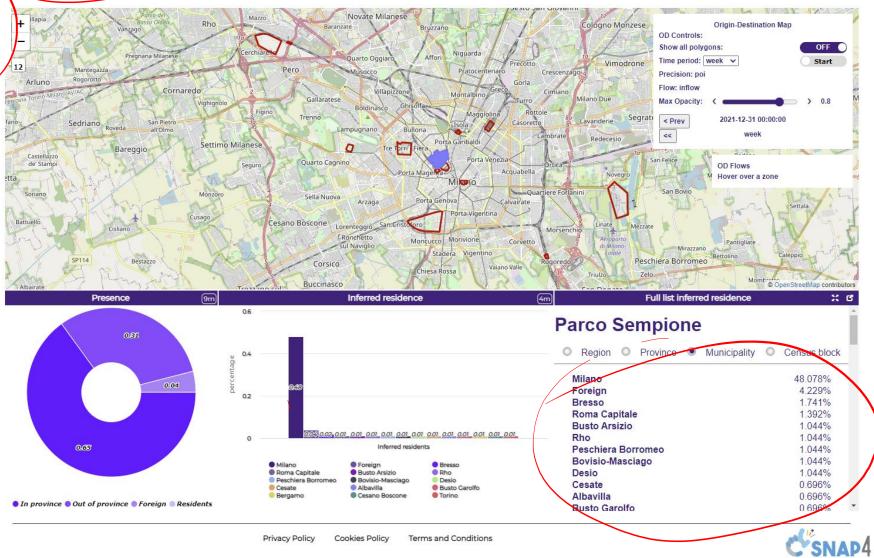






Visual Analytic on Milan Area













## Available DATA ANALYTICS (2b)

- City Users and Social
  - 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 and estimation of presences/tickets
    - Prediction/assessment of reputation as Trip Advisor
    - Anomaly detection

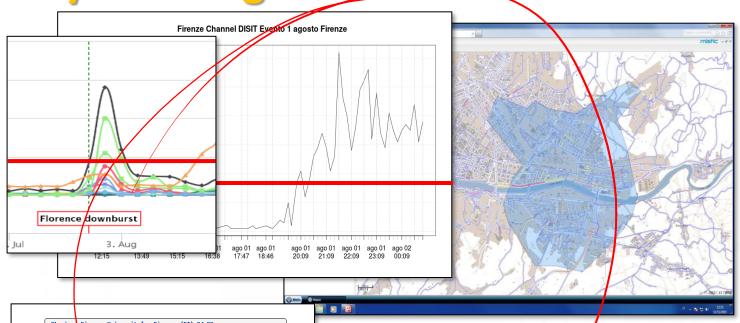


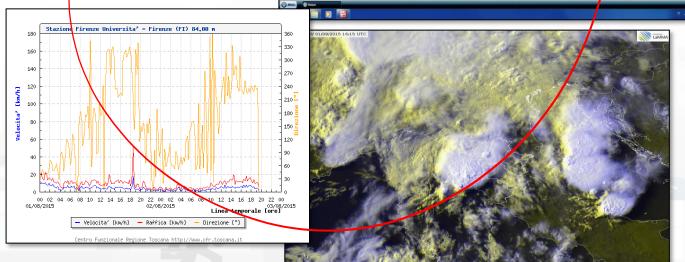
## Willer Vigilance SNAP4city





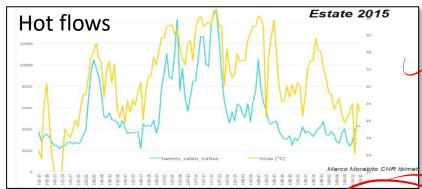
**Early Warning** 



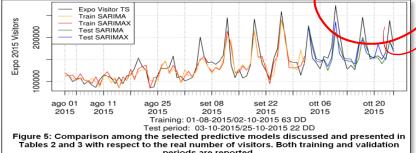


#### Snap4City (C), September 2022

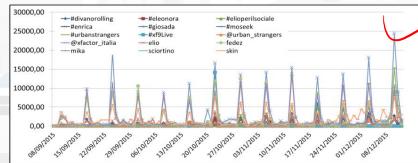
## **Predictive models**



#### Attendance at long lasting events: EXPO2015



#### Attendance at recurrent events: TV, footbal





#ethicaltourism

#grandinavi

#locazioni

#overtourism

Interreg

**M**editerranean

#4thIndustrialRevolution #affitti #affittibrevi #airbnb

#airbnbfication | #carbonfootprint | #chiantishire | #climatechange

#fairbnb | #gentrification | #gentrificazione

#locazionituristiche | #marketingTerritoriale | #Outlet

#greentourism | #home-sharing | #iperturismo

#responsibletravel | #sharingEconomy

#socialtourism | #SustainableDevelopmentGoals

#turismoEsperenziale #turismoetico #turismoSmart

#turismososteriioile | #turismoverde | #voluntourism

MERIT-DATA

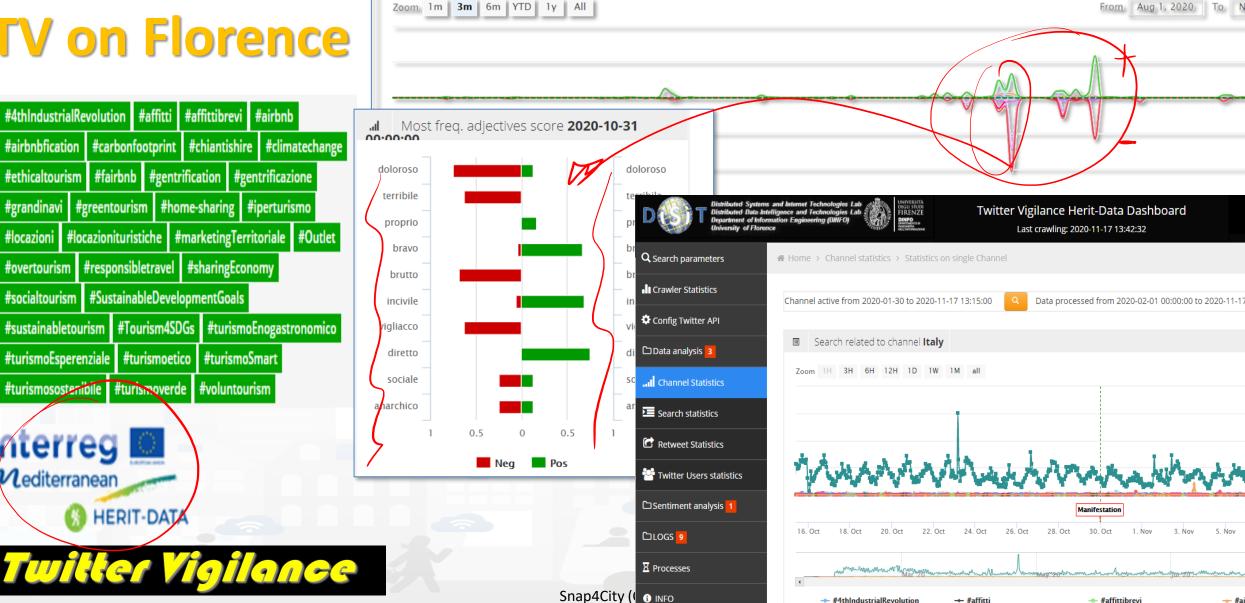


DISTRIBUTED SYSTEM AND INTERNET TECHNOLOGIES LAB















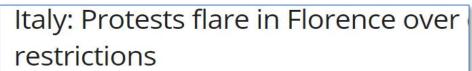




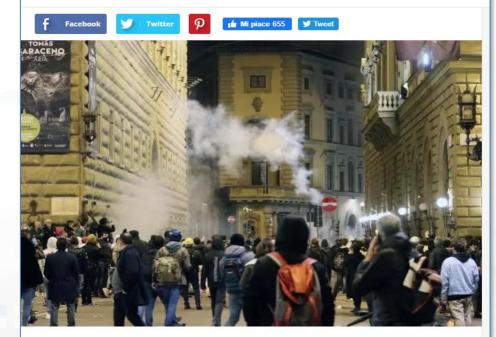




**Event:** clashes between police and demonstrators in Florence, during an unauthorized demonstration against the restrictions of Covid-19 government measures (on 30th October 2020 n)ght).



31 Oct, 2020



Florence mayor condemns violence and vandalism in historic centre in "surreal, terrible and painful night" for Tuscan city.

Protesters clashed with police in central Florence last night during an unauthorised protest over Italy's latest restrictions aimed at curbing the spread of covid-19.

#### Oct. 30 Protest Devastates Florence

10/31/2020 🖾 📇

In partnership with la Repubblica



During the unauthorized evening demonstration against recent social lockdown pandemic measures lasting four hours in downtown Florence on the evening of October 30 saw varticinante througha Maletov ecoktaile, emaching chan windows and equad ear windobjold

#### TOP NEWS STORIES

#### Oct. 30 Protest Devastates Florence

During the unauthorized evening demonstration against recent social lockdown pandemic measures lasting four hours in downtown Florence [...]

#### **COVID Demonstrations Hit Florence**

Unlike the U.S. protesters in Italy reserve public space for peaceful demonstrations which are organized to communicate a [...]

#### New Anti-Covid Regulations: Dos &

Given the upswing in Coronavirus infections and consequent hospital admissions since the beginning of October, Italian Prime Minister [...]

#### New Pandemic Regulations in Italy

In the face of the rise of Coronavirus infections, the Italian government has extended

#### LA COMPAGNIA

Original Language Movies International Film Festivals









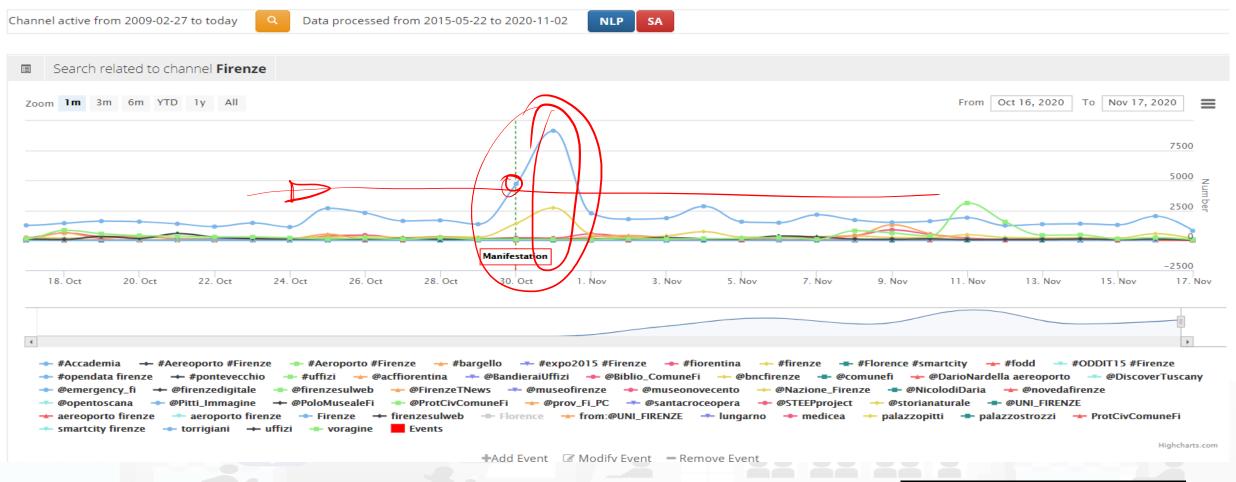


MERIT-DATA





## **30 October Manifestation**







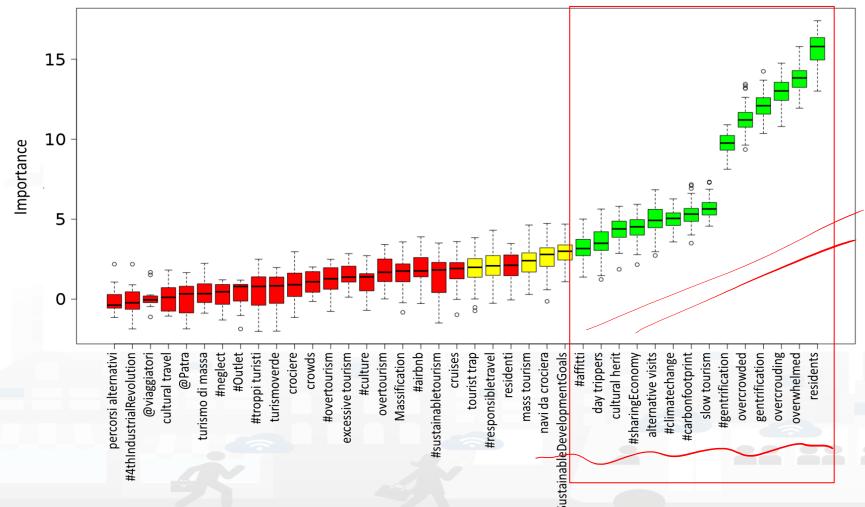








## **Relevant Feature Selection - Italy**



Random Forest Model Results expoiting the confirmed keys:

- Accuracy = 0.90
- Kappa index = 0.81

Twitter Vigilance

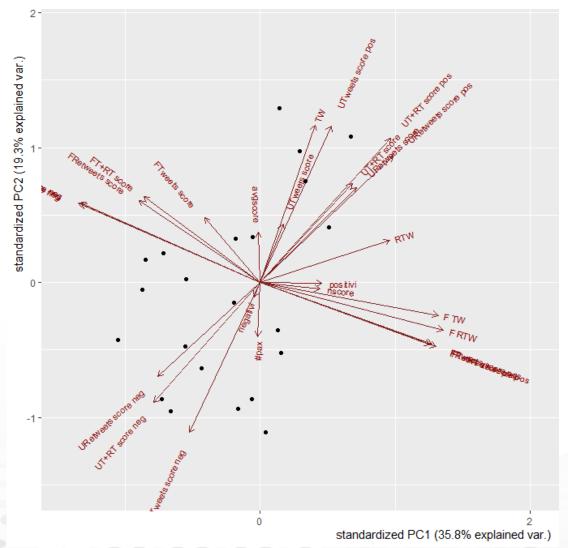




## Reputation



- Prediction/estimation of
   Average Score of Trip
   Advisor as a function of
   Twitter Vigilance Metrics +
   other information
- Prediction/estimation of
  Negative Scores on specific
  Museum or service as a
  function of Twitter Vigilance
  Metrics + other information









## Environmental









## Available DATA ANALYTICS (3)

- Environment and Weather
  - **Predictions** of pollution conditions for diffusion NOX, PM10, PM2.5, on the basis of traffic flow, 48 hours
  - Long term predictions of European Commission KPIs on
    - NO2 average value over the year
    - PM10 ......
  - Prediction of landslides, 24 hours in advance
  - Computation of CO2 on the basis of traffic flows
    - each road for each time slot of the day
  - Heatmaps production, dense data interpolation for
    - Weather conditions: temperature, humidity, wind, DEW
    - Pollutants and Aerosol: NO, NO2, CO2, PM10, PM2.5, etc.
  - Impact of COVID-19 on Environmental aspects





## AQI Indexes estimation via R studio and IOT App

### European Air Quality Index EAQI

http://airindex.eea.europa.eu/

Pollutant	Index level (based on pollutant concentrations in µg/m3)					
	Good	Fair	Moderate	Poor	Very poor	
Particles less than 2.5 µm (PM <sub>2.5</sub> )	0-10	10-20	20-25	25-50	50-800	
Particles less than 10 µm (PM <sub>10</sub> )	0-20	20-35	35-50	50-100	100-1200	
Nitrogen dioxide (NO <sub>2</sub> )	0-40	40-100	100-200	200-400	400-1000	
Ozone (O <sub>3</sub> )	0-80	80-120	120-180	180-240	240-600	
Sulphur dioxide (SO <sub>2</sub> )	0-100	100-200	200-350	350-500	500-1250	

Measurements of up to five key pollutants supported by modelled data determine the index level that describes the current air quality situation at each monitoring station.

The index corresponds to the poorest level for any of five pollutants according to the following scheme.

#### Legend of Environmental data:

https://www.snap4city.org/435

### Common Air Quality Index CAQI

http://www.airqualitynow.eu

Qualitative name	Index or sub-index	Pollutant (hourly) density in µg/m <sup>3</sup>				
		NO <sub>2</sub>	PM <sub>10</sub>	<b>O</b> <sub>3</sub>	PM <sub>2.5</sub> (optional)	
Very low	0–25	0–50	0–25	0–60	0–15	
Low	25–50	50–100	25–50	60–120	15–30	
Medium	50–75	100–200	50-90	120-180	30–55	
High	75–100	200–400	90–180	180–240	55–110	
Very high	>100	>400	>180	>240	>110	

The index is defined away from roads (a "background" index). **CAQI** is computed on the basis of  $NO_2$ ,  $PM_{2,5}$ ,  $PM_{10}$  and  $O_3$ .





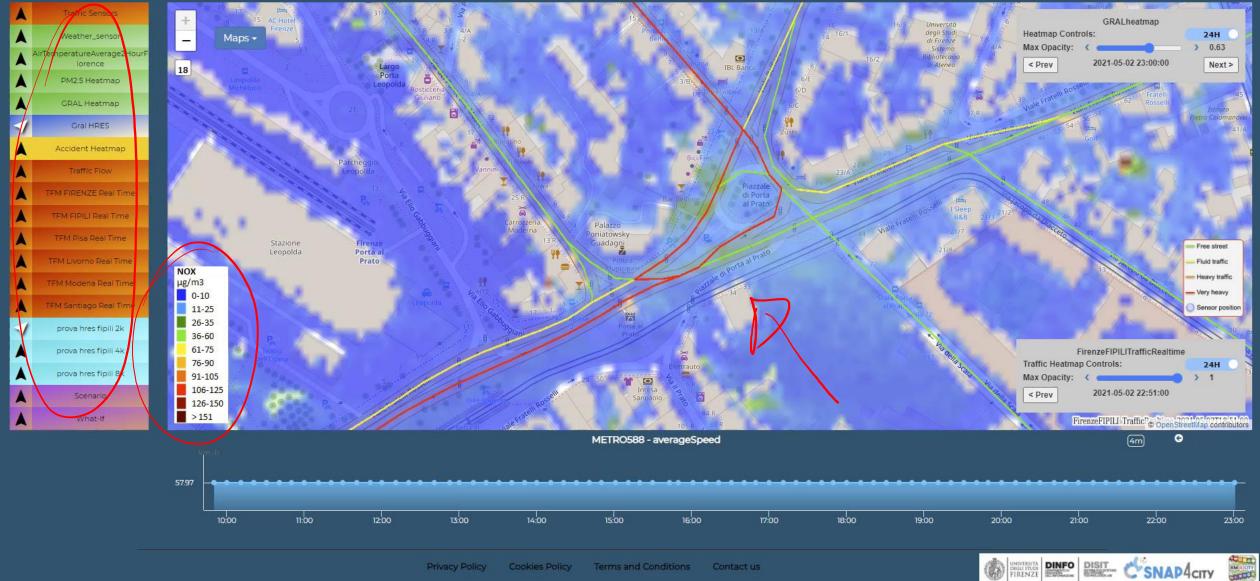
## NOX and PM10 prediction particles flows

- GRAL predictions: PM10, NOX, ....
  - Comparison wrt real time values in actual value of Sensors
  - Graz Lagrangian Model.
- GRAL model takes into account:
  - pollution sources (for example the vehicles, their distribution on the streets, the about of pollution they produce according to their distribution over time and space, etc.),
  - structure of the city (streets and shape
     3D of the buildings),
  - weather forecast (wind intensity and direction), etc.
- GRAL can be applied on NOX, PM10, PM2.5, ... or any other particles



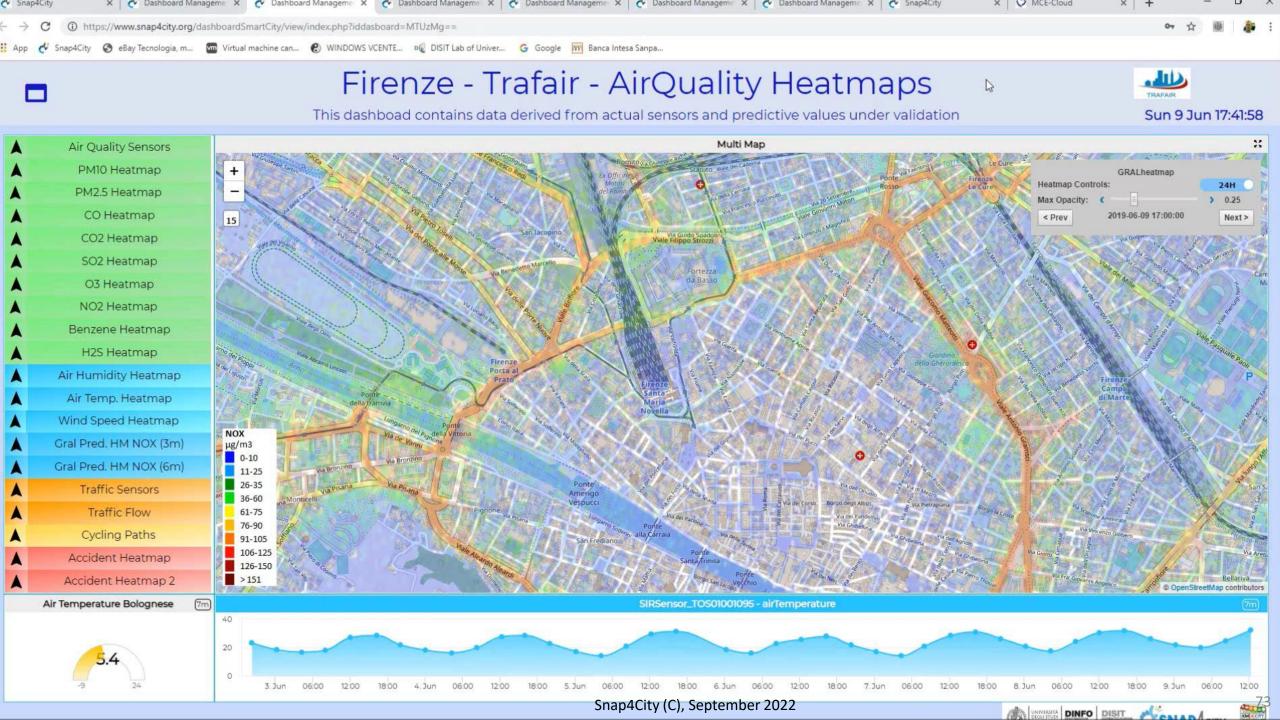
## Traffic Flow Manager on multiple cities

Sun 2 May 23:16:31



https://www.snap4city.org/dashboardSmartCity/view/index.php?iddasboard=MzEyNg==

Snap4City (C), September 2022



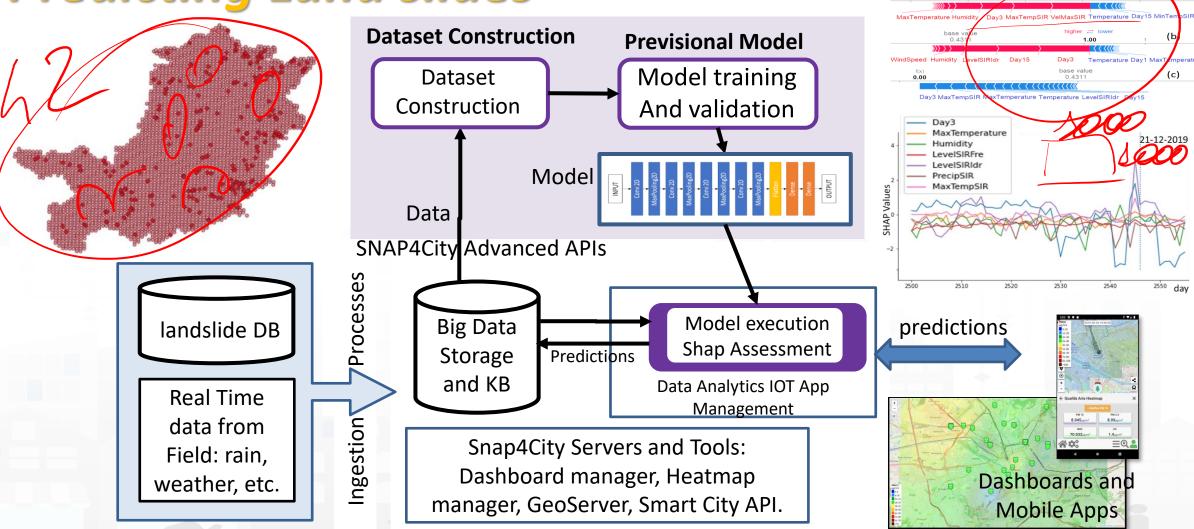








**Predicting Land slides** 



E. Collini, L. A. I. Palesi, P. Nesi, G. Pantaleo, N. Nocentini and A. Rosi, "Predicting and Understanding Landslide Events with Explainable AI," in *IEEE Access*, doi: 10.1109/ACCESS.2022.3158328.

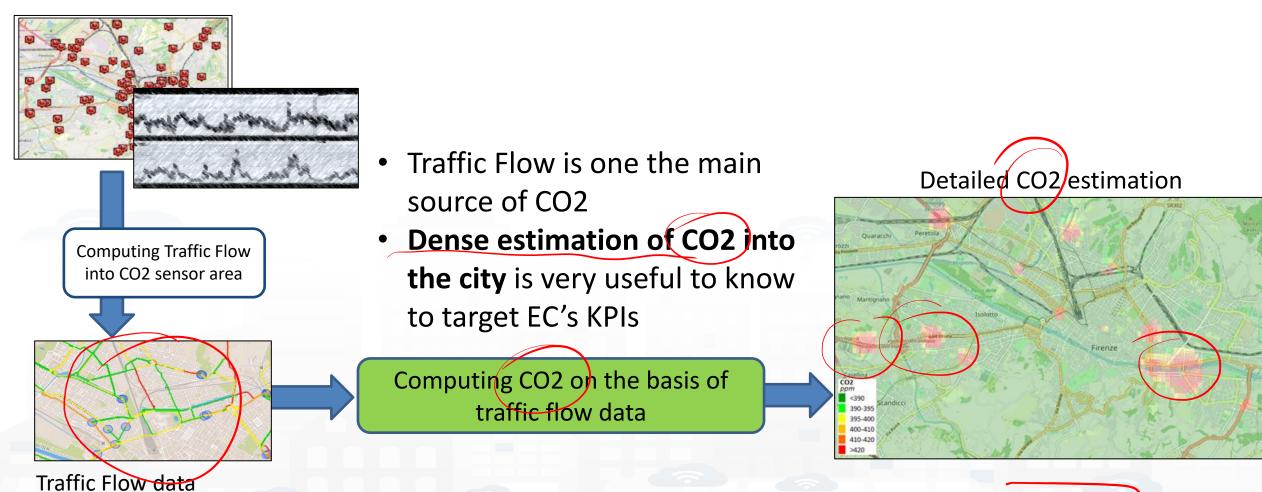








## UNIVERSITA DEGLI STUDI FORMAZIONE DIST DISTRIBUTED SYSTEMS AND STRIBUTED SYSTEMS AND STR



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/









Predicting EC's KPI on NO2 months in advance

Deep Learning Long Terms Predictions of NO2 mean values, From 30 to 180 days in advance

The features used as input for the predictive models are:

Month

dayOfTheYear

NO2

Tmean

Humidity

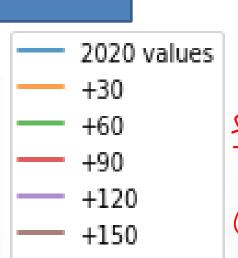
windMean

NoxDomestic numberOfVehicles

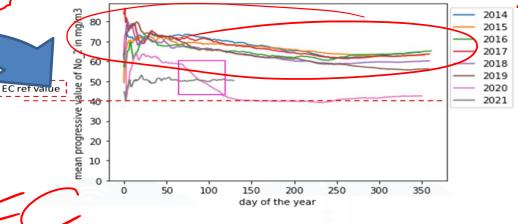
NO2cumulated

NO2progresseveMean

\ numberOfVehicles@umulated ₷



+180



		Air Quality	WHOguidelines		
Pollutant	Averaging period	Objective and legal nature and concentration	d 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 ug/m³	e target value has become a it value since 1 January 2015	10 μg/m³	
PM <sub>10</sub>	One day	Nο Limit value, 50 μg/m³	ot to 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		ot to be exceeded on more n 25 days per year, averaged over three years	100 μg/m³	
NO <sub>2</sub>	One hour	Limit value 200 ug/m³ (*)	t to be exceeded more than 18 times a calendar year	200 µg/m³ (*)	
NO <sub>2</sub>	Calendar year	Limit value, 40 μg/m³		40 μg/m³	





# Semantic reasoning, NLP









# **Available DATA ANALYTICS (4)**

- Time Series
  - Time Series Anomaly detection
  - Data quality assessment and control
  - short and long term prediction
  - **Interpolation** of Data on regular grid for calibrated beatmaps
- Semantic Reasoning
  - Ontology Modelling and integration, expert system construction
  - Knowledge modelling and reasoning on BFF stores: spatial, temporal, relational
  - Virtual Assistant for tourism guidance
- Matrices, Images, Maps and 3D Digital Models
  - Conversion of Satellite data images into regular ground images
  - Extraction information from Orthomaps, LIDAR, etc., regarding city structures
  - Digital Twin of Cities and Objects: pattern extraction, 3D reconstruction



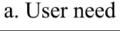


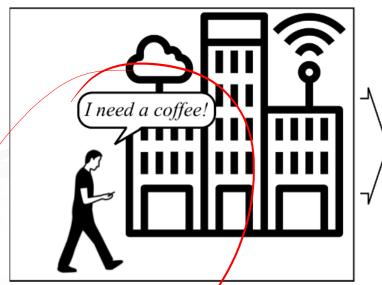






### **Location Aware Virtual Assistant**





What is the user expression of his/her need?

b. Personal assistant

- 1. Where is he/she?
- 2. What kind of service does he/she need?
- 3. Who can satisfy his/her need?

What information a human personal assistant needs to provide an answer?

#### c. Answer

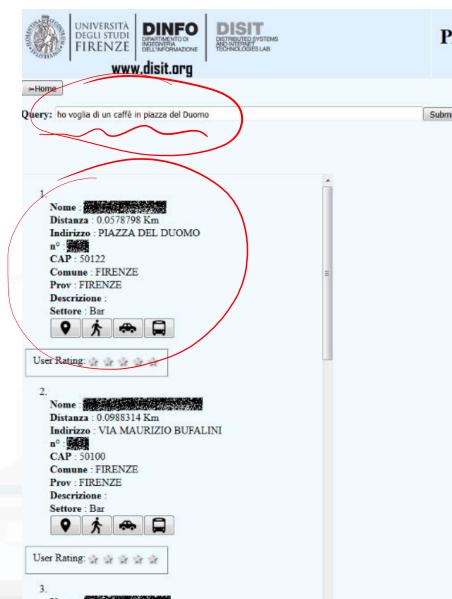


What would a human personal assistant answer?

Massai, P. Nesi, G. Pantaleo "PAVAL" A location-aware virtual personal assistant for re-trieving geolocated points of interest and location-based services", Journal Engineering Applications of Artificial Intelligence, Elsevier, https://www.sciencedirect.com/science/article/pii/S0952197618301994







#### PAVAL - Km4City Semantic Service Search





DISIT

www.disit.org

Lingua impostata: ITA

#### **Query Results:**



Dall'analisi della query risulta che sei interessato a trovare aziende e servizi relativi a: BAR nei dintorni della posizione rilevata.











Class	Definition of the class	Tot.
Type 1	Direct request of a full or partial Km4City service category label. i.e. "I need a restaurant", "Bed and breakfast around me"	25.60 %
Type 2	Direct request of the name of a precise local business i.e. "Take me to 'Da Mario'"	2.93 %
Type 3	Queries not exhibiting the service name as the user need i.e. "I want to eat spaghetti", "My stomach hurts"	74.40 %
Type 4	Presence of precise geographical reference i.e. "I need a restaurant in via dell'Oriuolo", "Hotels in Piazza del Carmine"	35.12 %
Type 5	Presence of partial or misspelled geographical reference i.e. "Eat in Piazza del Dpomo", "Bar near piazza P. Leopoldo"	16.32 %
Type 6	Presence of multiple geographical references i.e. "I want to read a newspaper near via Dante Alighieri in Pisa"	5.24 %
Type 7	Presence of geographical references i.e. "Find a place to eat near", "I want to drink something in via"	59.16 %
Type 8	Queries inside Florence municipality i.e. "Museums near piazza della Signoria"	69.71 %
Type 9	GPS localization allowed i.e. Given authorization from the browser: "I need a restaurant nearby"	100 %
Type 10	Not transactional queries i.e. "When did Garibaldi die?", "Find a video on YouTube"	2.31 %

160 —	
140 —	
120 —	
100 —	
80 —	
60 —	
40 —	<del>                                     </del>
20 —	
0	
\$3 x	Laurant todica tring the laur toda the food and robacco and robacco to the food and robacco to the food and robacco to the food to the food and robacco to the food to the food and robacco to the food to the foo
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	140.

#### ITA

PERSONAL ASSISTANT	PRECISION	RECALL	F-MEASURE			
Paval	79.28%	99.61%	88.74%			
Google Assistant	77.74%	28.34%	40.75%			
Apple Siri	60.07%	64.42%	62.29%			
Microsoft Cortana	63.51%	14.43%	22.92%			

#### **ENG**

	' /			
PERSONAL ASSISTANT	PRECISION	RECALL	F-MEASURE	
Paval	74.67%	96.24%	84.57%	
Google Assistant	75.24%	48.92%	58.77%	
Apple Siri	72.28%	70.65%	71.45%	
Microsoft Cortana	66.81%	27.53%	38.36%	







# Digital Twin Construction



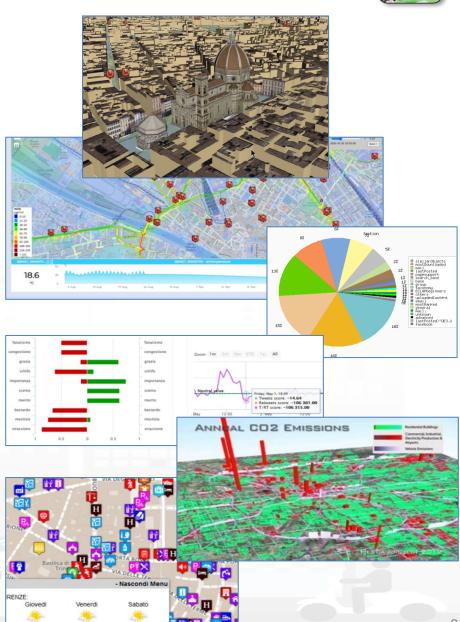


## **Digital Twin**



### Digital Twin

- Connected with real physical systems
- Modelling aspects: structural, visual, informative, real time data sensors (context), POI, functional, resource managements, etc.
- Integration of AI/XAI techniques with simulations and modelling
- Easier to understand the context, review from multiple points of view
- Useful to perform
  - Discussion with city users
  - Support decision makers
  - By Case Experiments for analysing
    - New solutions, impact of disaster (natural and provoked)
    - Reduction of costs in the analysis, in reduction of mistakes







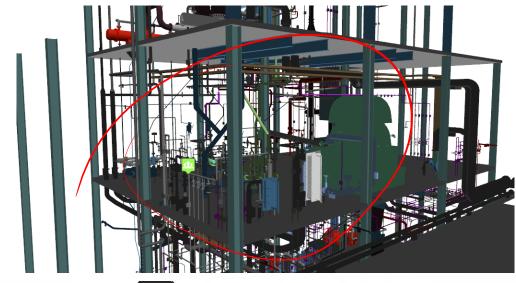


# **Digital Twin**

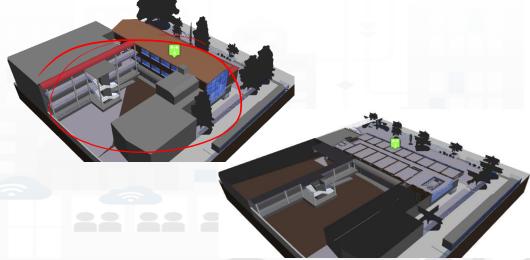














UNIVERSITÀ **DEGLI STUDI** FIRENZE

INGEGNERIA DELL'INFORMAZIONE

3D City Construction SNAP4city



Multiple images from several points of views



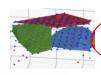
**Raw Facades** pictures



GIS orthomap



**GIS Building** plant shapes and positions Building heights (at the eaves)



Lidar Data

Sky Pattern

Terrain (DTM)

Heatmaps, Traffic flow, Pins, IOT, POI, RT data, .....

Events on back office

3D design of High Value Buildings, HVB

Facades patterns extraction

Roof patterns extraction

Create 3D building with flat roofs (by extrusion)

Create 3D buildings with 3D roofs' shapes

Integrated view of HVB + building with roofs and facades

Generate 3D city representation

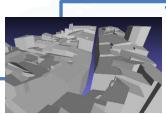
Facades' patterns

**HVB** model with patterns

**Roof patterns** 



Create 3D building with photorealistic texture



3D buildings with roof and facade patterns



Extruded building with picking functionality

3D City Digital Twin on Dash

Orientation, Position, Light, zoom, view point, etc.

### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**











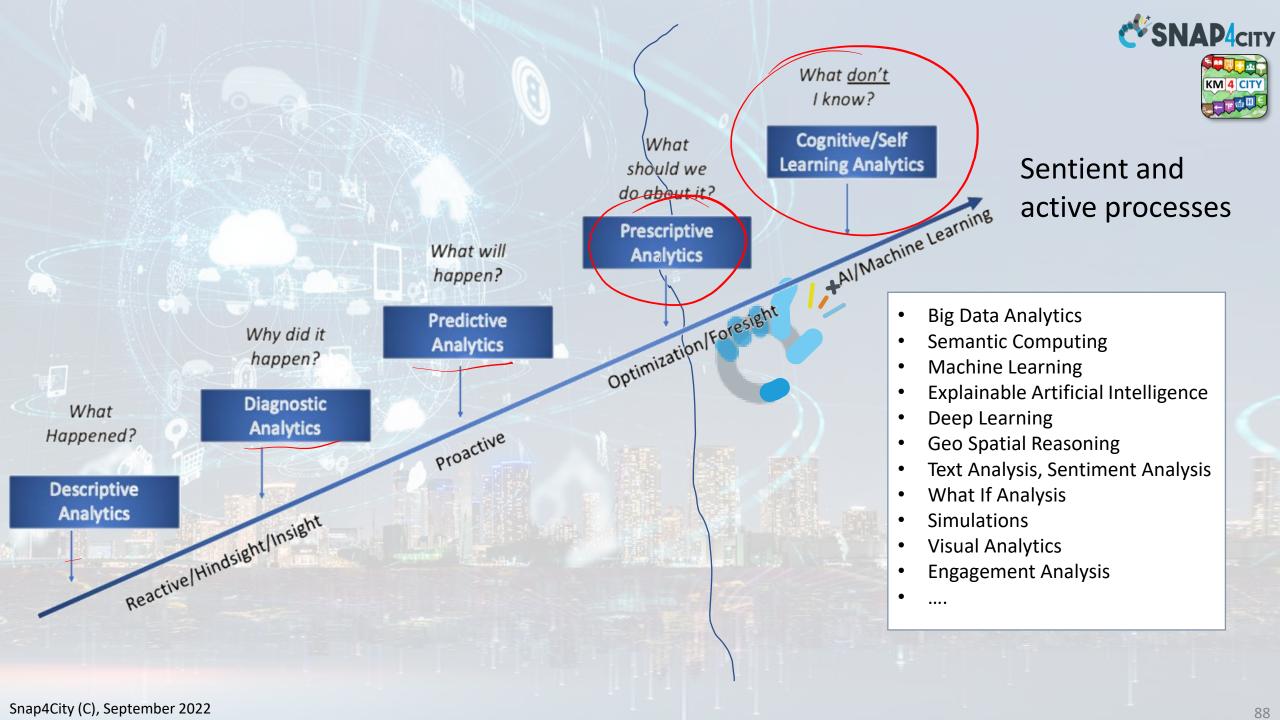
# Available DATA ANALYTICS (5)

### Management and strategies

- What-if analysis, dynamic routing, origin destination matrices production from a large range of sources
- Early warning computation
- Estimation of KPI and local indexes for: quality of life (15MinCityIndex)
- Production Optimization
- Planning and Monitoring renovation works via objective KPIs
- Managing Maintenance and teams
- Predictive Maintenance and costs predictions: chemical plant, vehicles, boats

### Resilience and Risks Analysis

- Resilience analysis wrt European Guidelines on Resilience of critical infrastructure, and transport systems
- Risk analysis: natural and non natural disaster



# 15MinCityIndex

What would support my neighborhood to become a 15-Minute City?

#### **Using the Open Data:**

We developed a data analytic tool based on municipal and national open data to assess services adequacy for people living in each 15 minutes areas of the city.

Good public transport services: bus, new tram line, train stations, cycle paths.



Careggi/Rifredi is a relevant district in Florence because of hosting the main Florence/Tuscany hospitals Careggi and Meyer, but also university headquarters and many other workplaces.



Signa

Lastra a Signa

15Min Index

Critical

Index

università degli studi FIRENZE

Badia a

DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB

Osmannoro







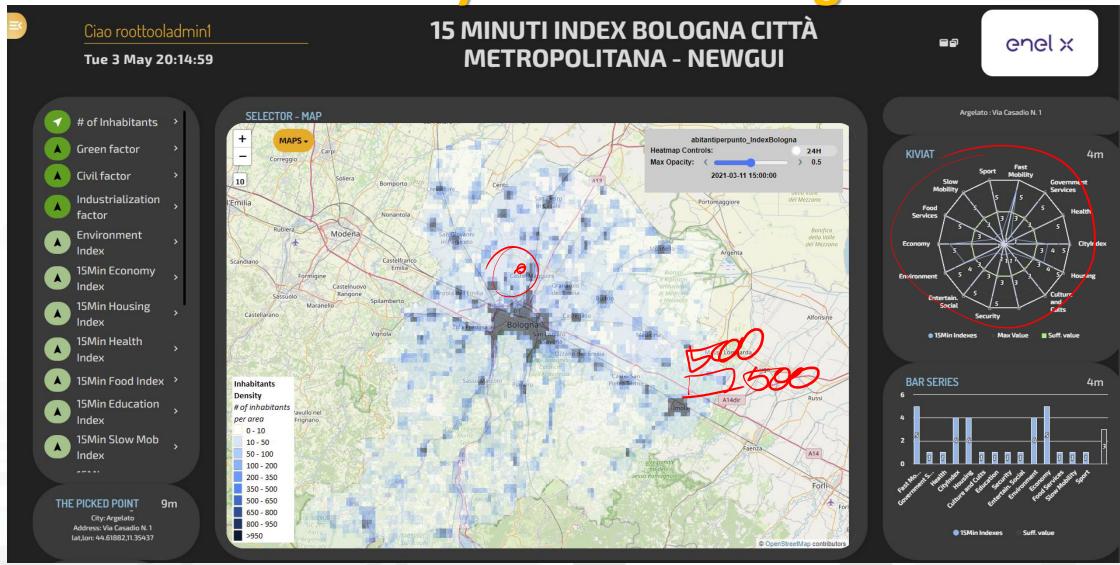








# 15MinCityIndex on Bologna





DISIT Lab, Distributed Data Intelligence and Technologies
Distributed Systems and Internet Technologies
Department of Information Engineering (DINFO)
http://www.disit.dinfo.unifi.it

# Early warning, detection

Issue:

- Detection of critical condition
- Not easily detected with other means
- Impact:
  - Early warning, faster reaction
  - Increased resilience
- Several metrics related to
  - Volume of retweets
  - Sentiment analysis

**P**repare

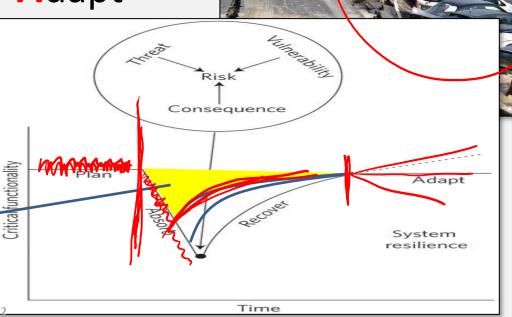
**A**bsorb

Recover

**A**dapt



resolute



damage

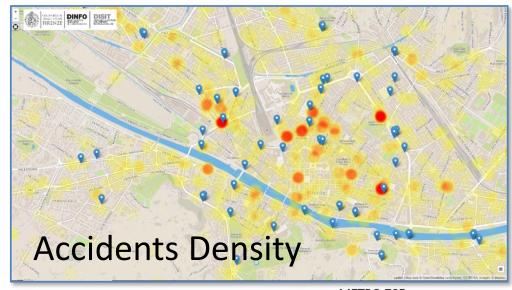


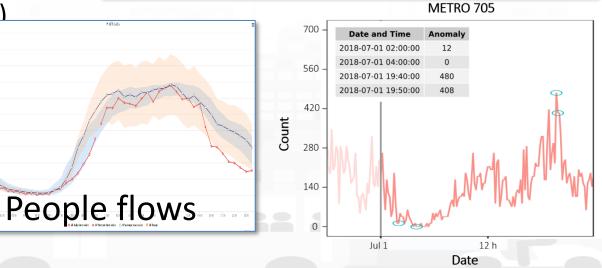




### **Anomaly Detections**

- About the IoT Devices status
  - Eventual problems on IoT Devices, connections, etc.
- About People Flows and Density
  - Early warning of the ineption of critical events
- About traffic flow
  - Early warning on eventual incidents, or on the inception of critical conditions on the traffic (e.g., a reduction in viability, a broken bus
- About....
  - Early warning, early detection of problem
- Recurrence analysis
- Causal Analysis





# What-if Analysis on Pub Transport







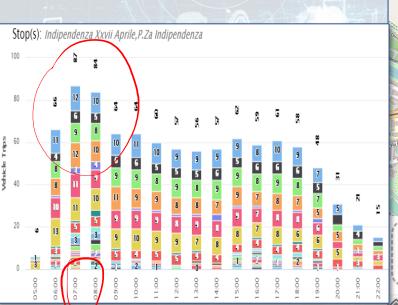


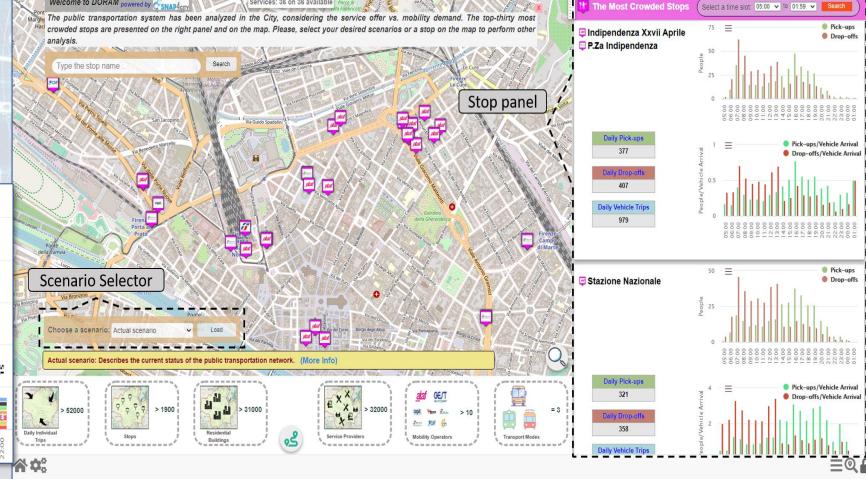


- Definition of scenarious impact on
  - Traffic, Pollutant, parking, public transport, private flows, etc.

KPI analysis

**Public Services** 













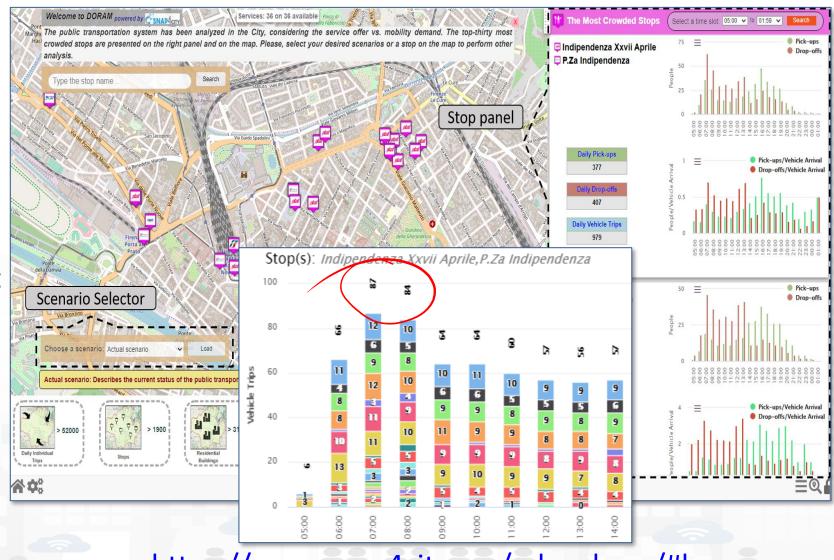
### **Analysis of**

- Demand of Mobility
  - Via OD matrices
  - POI, city structure, etc.

### With respect to

- Offert of Transportation:
  - Public services
  - Private services
  - Multiple agencies
  - GTFS

Critical Busses, bus-stops, paths, rides, etc.



https://www.snap4city.org/odanalyzer/#b

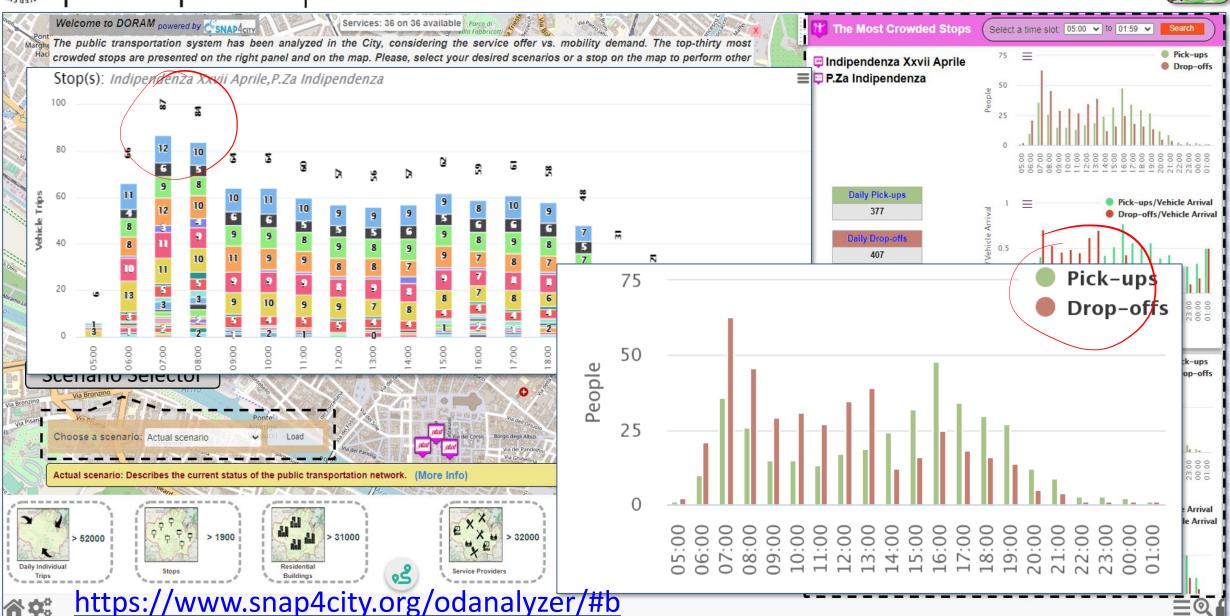


DINFO
DIPARTIMENTO DI
INGEGNERIA
DELL'INFORMAZIONE

DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB

### **DORAM**













# **Decision Support Systems, What-if**

### Event planning, via what-if analysis

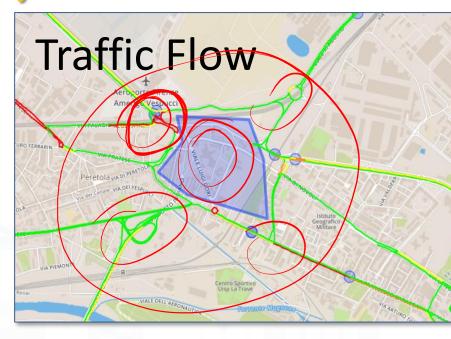
- Change in the graph structure of the city
- Impact on the flow of people and vehicles
- Adaptation: public transport, traffic, pedestrian management, etc.

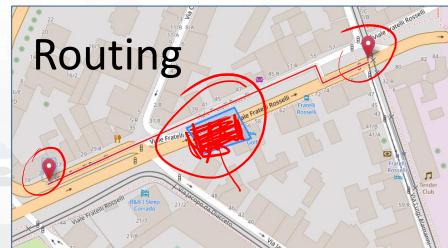
### Immediate reaction to natural events or not

- Everything is ready and updated in real time
- Each view is contextualized in terms of data: descriptive and prescriptive

### Digital Twin

- More detail in the context integrated data
- Greater realism in deductions and representations
- Less fragmentation and non-uniformity in the views to support decisions











# City Resilience CSNAP4city





# Early Warning, Detection

### Issue:

- Detection of critical condition
- Not easily detected with other means

### Impact:

- Early warning, faster reaction
- Increased resilience

Prepare

**A**bsorb

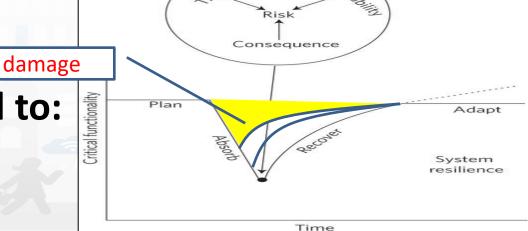
Recover

**A**dapt



### Several metrics related to:

- Volume of retweets
- Sentiment analysis







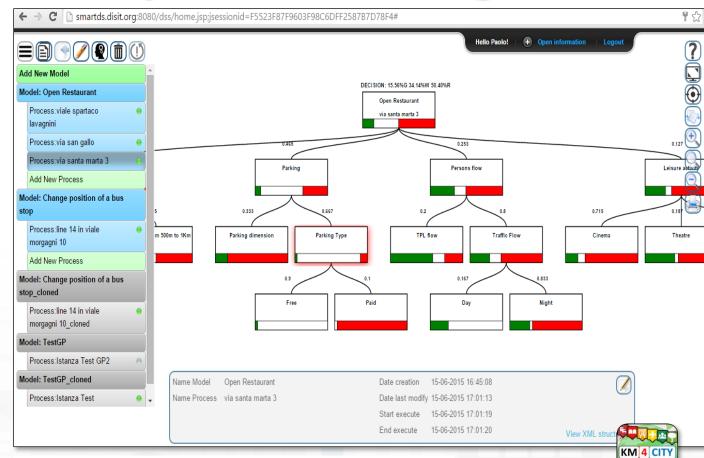






### **Smart Decision Support, system thinking**

- Smart Decision Support System based on System Thinking plus
- Actions to city reaction, resilience, smartness, ...
- **Enforcing Mathematical model** for propagation of decision confidence..
- Collaborative work, ...
- Processes connected to city data: DB, RDF Store, Twitter, etc.
- Production of alerts/alarms
- Data analytics process
- Twitter Processes
- reuse, copy past, ...

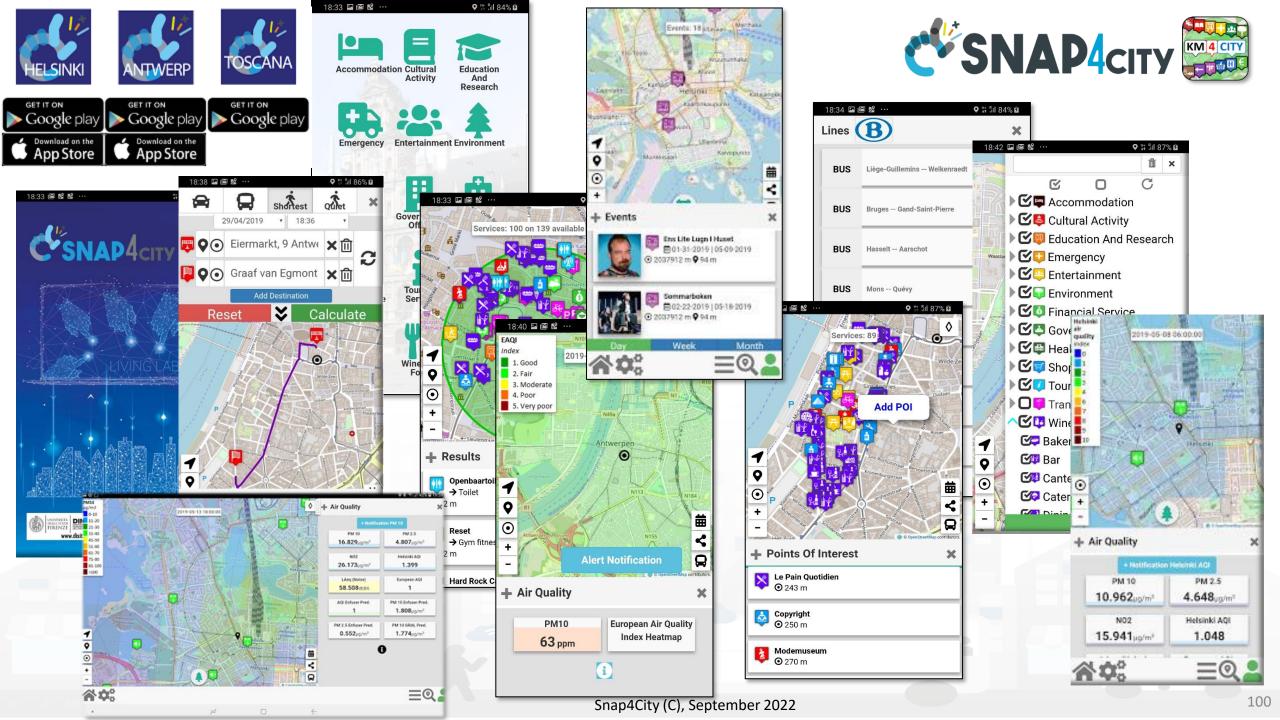


http://smartds.km4city.org

### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**

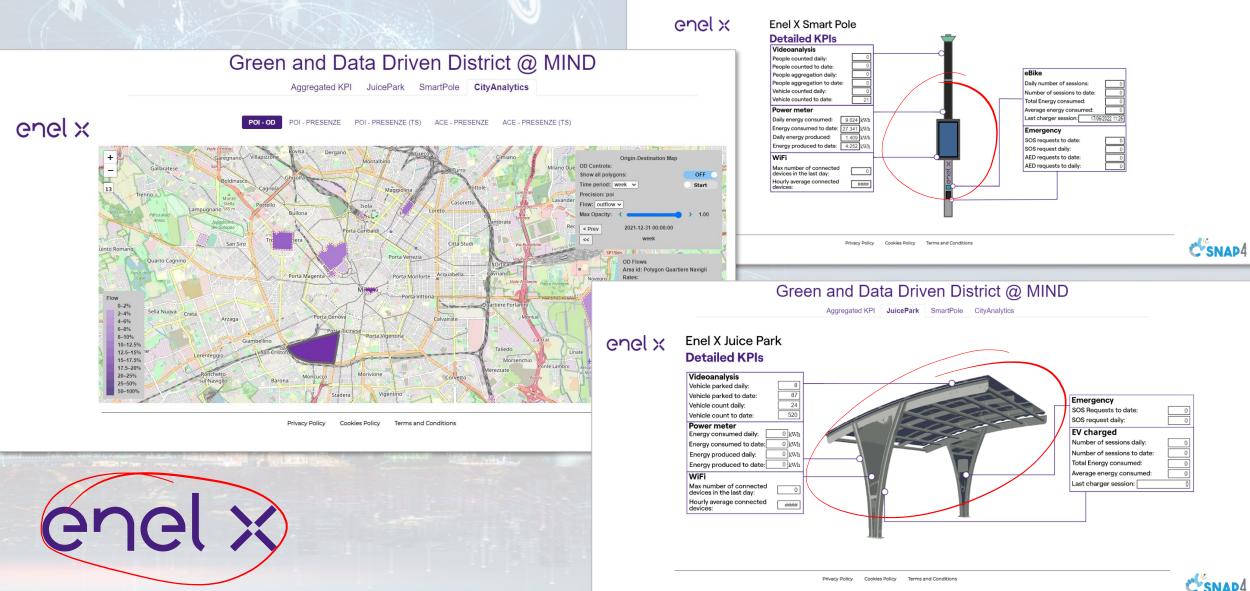








### Energy monitoring and business intelligence and Data Driven District @ MIND



# Karlstad Street Lights CAPELON





# Herit-Data Twitter Vigilance









https://www.snap4city.org/dashboardSmartCity/view/index.php?iddasboard=MzExNQ==

Snap4City (C), September 2022













### **Twitter Vigilance Herit-Data:** Some Numbers

Channel Name	Total Number of Collected TW+RTW	Number of Collected Tweets	Number of Collected Retweets	Twitter Volume Processing Time Range	NLP & Sentiment Analysis Processing Time Range	NLP & Sentiment Analysis Languages
Spain	113.7 Millions	40.99 Millions	72.49 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	English, Spanish
France	50,1 Millions	16.0 Millions	34.1 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	Italian, English, French
Greece	12.3 Millions	4.2 Millions	8.1 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	English
Italy	2.97 Millions	1.0 Million	1.9 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	Italian, English
Croatia	35.8 Thousands	15.5 Thousands	19,8 Thousands	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	English

Updated: Dec. 2021







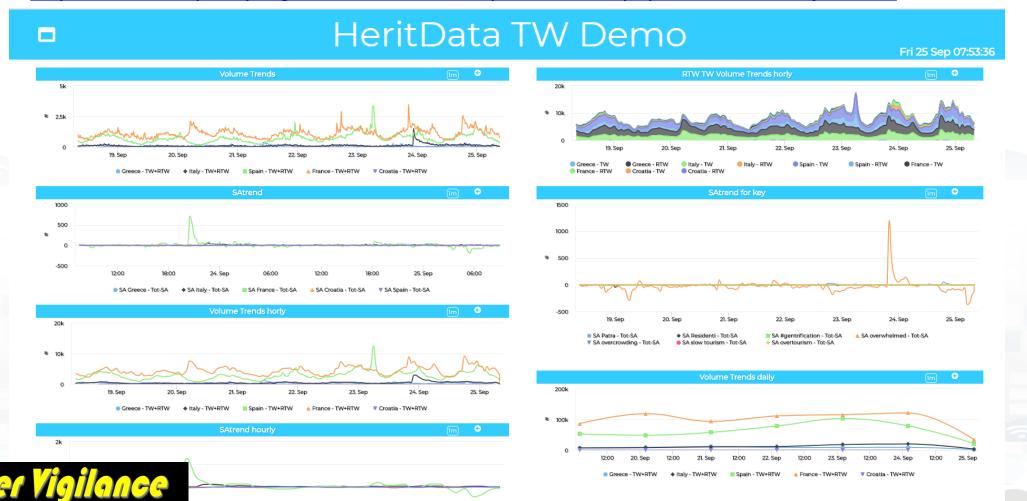






### Twitter Vigilance Data on Dashboards trought IoT-App

https://www.snap4city.org/dashboardSmartCity/view/index.php?iddasboard=Mjc3NQ==



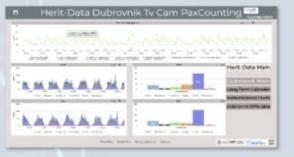
# Dubrovnik

### Tourism Domain

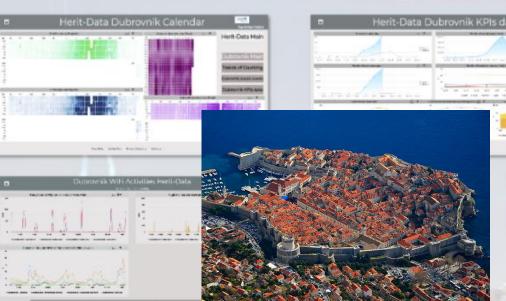
- Counting People
- TV Cameras and WiFi
- Social Media

### Dashboards

- Monitoring and real time control
- People flow
- Twitter Vigilance
- Historical and Real Time data
- Services Exploited on:
  - Dashboard
- Since 2020

















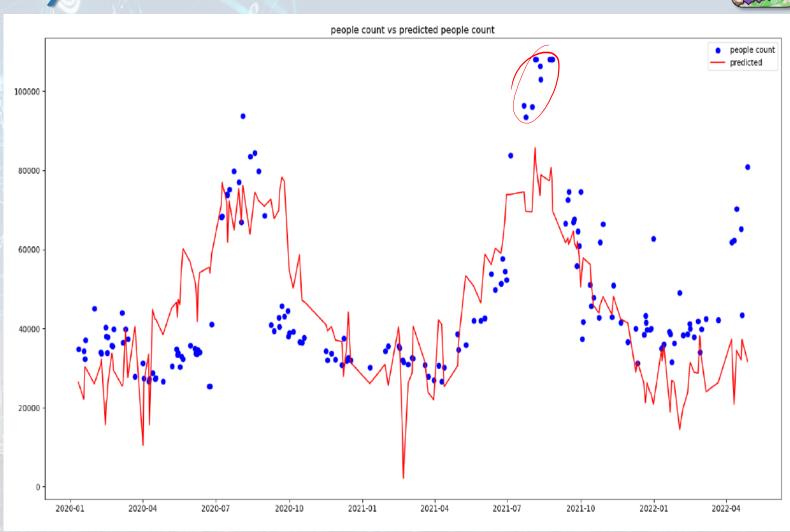




# **Dubrovnik: Data Analytics**

- Effects of advertising via Social Media

Effect of weather conditions



# Pont du Gard

- Tourism Domain
  - KPIs
  - Social Media
  - People Flows
  - Bike Flows
- Dashboards
  - Monitoring KPI
  - People and bikes flows
  - Twitter Vigilance
- Historical and updated data
- Services Exploited on:
  - Dashboard
- Since 2020

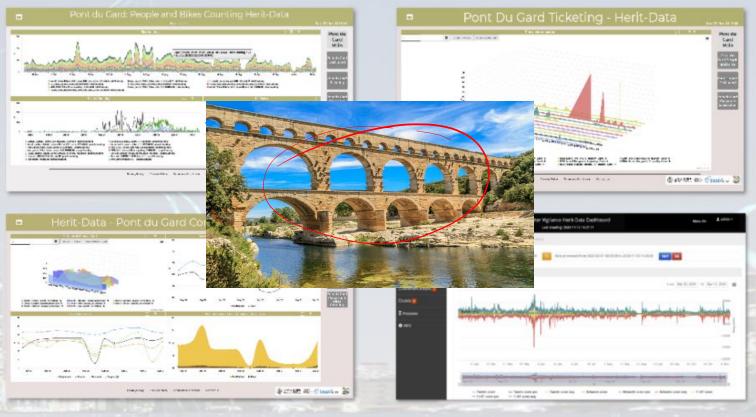












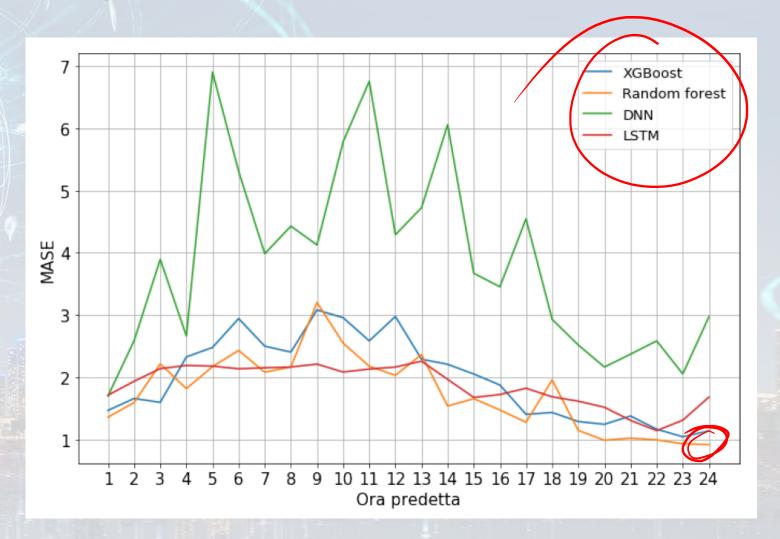






# Pont du Gard: data analytics

- Prediction of the number of sold tickets 24 hours in advance
- As a function of ......





# Valencia, FSMLR

### Tourism Domain

- Counting People
- Environmental data
- Social Media

### Dashboards

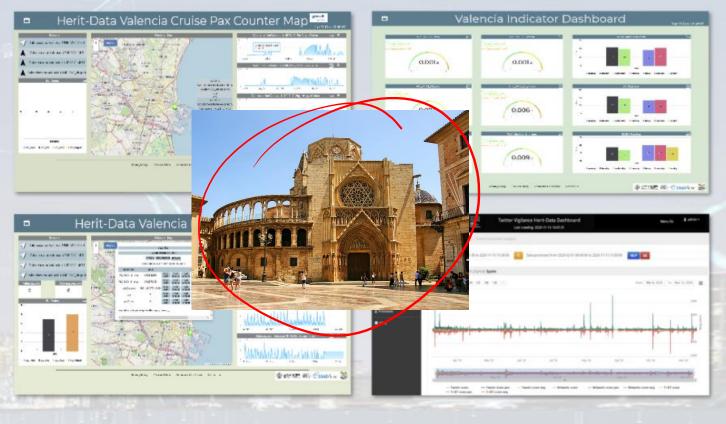
- Monitoring and real time control
- People flow
- Twitter Vigilance
- Historical and Real Time data
- Services Exploited on:
  - Dashboard
- Since 2020













# Valencia, FSMLR

### Tourism Domain

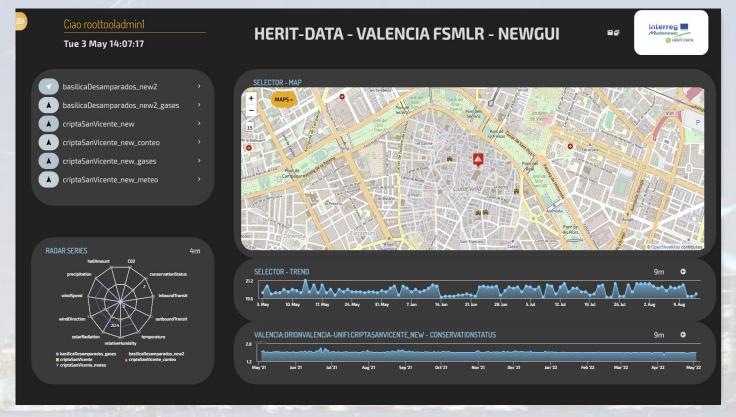
- Counting People
- Environmental data
- Social Media
- Dashboards
  - Monitoring and real time control
  - People flow
  - Twitter Vigilance
- Historical and Real Time data
- Services Exploited on:
  - Dashboard
- Since 2020











# Mostar, Bosnia Herzegovina







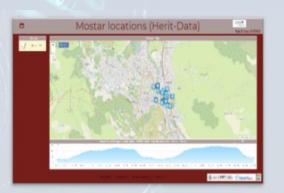
## Tourism Domain

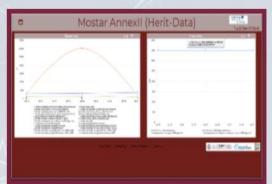
- KPI: business, house prices, investments, stay duration, etc.
- Profiled POI, Point of Interests
- People flows: arrivals, overhights

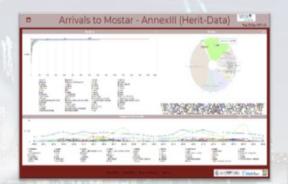
## Dashboards

- Monitoring KPI
- POI, flows
- Historical and updated data
- Services Exploited on:
  - Dashboard
- Since 2020

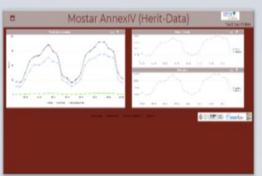
https://www.snap4city.org/dashboardSmartCity/v
iew/index.php?iddasboard=MzE0OQ==













# West Greece

- Tourism Domain
  - KPIs: ODM Flows, ...
  - Social Media
  - People Flows
- Dashboards
  - Monitoring KPI
  - People flows
  - Twitter Vigilance
- Historical and updated data
- Services Exploited on:
  - Dashboard
- Since 2020

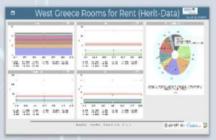
















Interregi







### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**













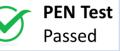
# **Using from Cloud or Installing on Premise**

- Cloud «as a service»: a number of installations are in place
  - The largest <a href="https://www.snap4city.org">https://www.snap4city.org</a>
    - 20 tenants/organizations, Billions of data
    - 1 hour deploy new organization, devices, data, dashboards
- Installations on public or private cloud, or on private servers
  - A number of ready to use configurations from 1VM to multiple scalable solutions: <a href="https://www.snap4city.org/471">https://www.snap4city.org/471</a>
    - VM: Appliances ready to use
    - Docker compose, Tool for generating and downloading the docker compose files
      - Micro X version can be installed and tested in 4 hours.



# https://www.Snap4City.org













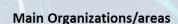
## • > 7 running installations

Toscana, Pisa, Sweden, ISPRA, Snap4.eu,

- Altair, Italmatic, ....
- 13 actions, 12 pilots on 10 Countries
  - >40 cities/areas

## Wide MULTI-tenant deploy, e.g.,

- 19 Tenants / Organizations
- > 7700 users on
- > 1400 Dashboards
- > 16 mobile Apps
- > 2 Million of structured data per day
- > 520 IoT Applications/node-RED
- > 700 web pages with training
- > 75 videos, training videos

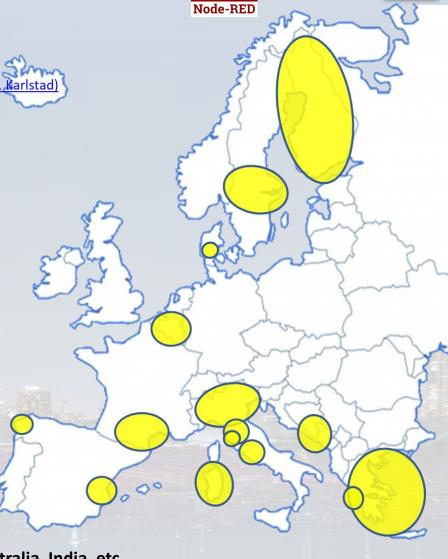


- Antwerp area (Be)
- Bologna (I)
- Capelon (Sweden: Västerås, Eskilstuna, Karlstad)
- DISIT demo (multiple)
- Dubrovnik, Croatia
- Firenze area (I)
- Garda Lake area (I)
- Greece (Gr)
- Helsinki area (Fin)
- Livorno area (I)
- Lonato del Garda (I)
- Modena (I)
- Mostar, Bosnia-Herzegovina
- Oslo & Padova (Impetus)
- Pisa area (I)
- Pistoia (I)
- Pont du Gard, Occitanie (Fr)
- Prato (I)
- Roma (I)
- Santiago de Compostela (S)
- Sardegna Region (I)
- Siena (I)
- SmartBed (multiple)
- Toscana Region (I), SM
- Valencia (S)

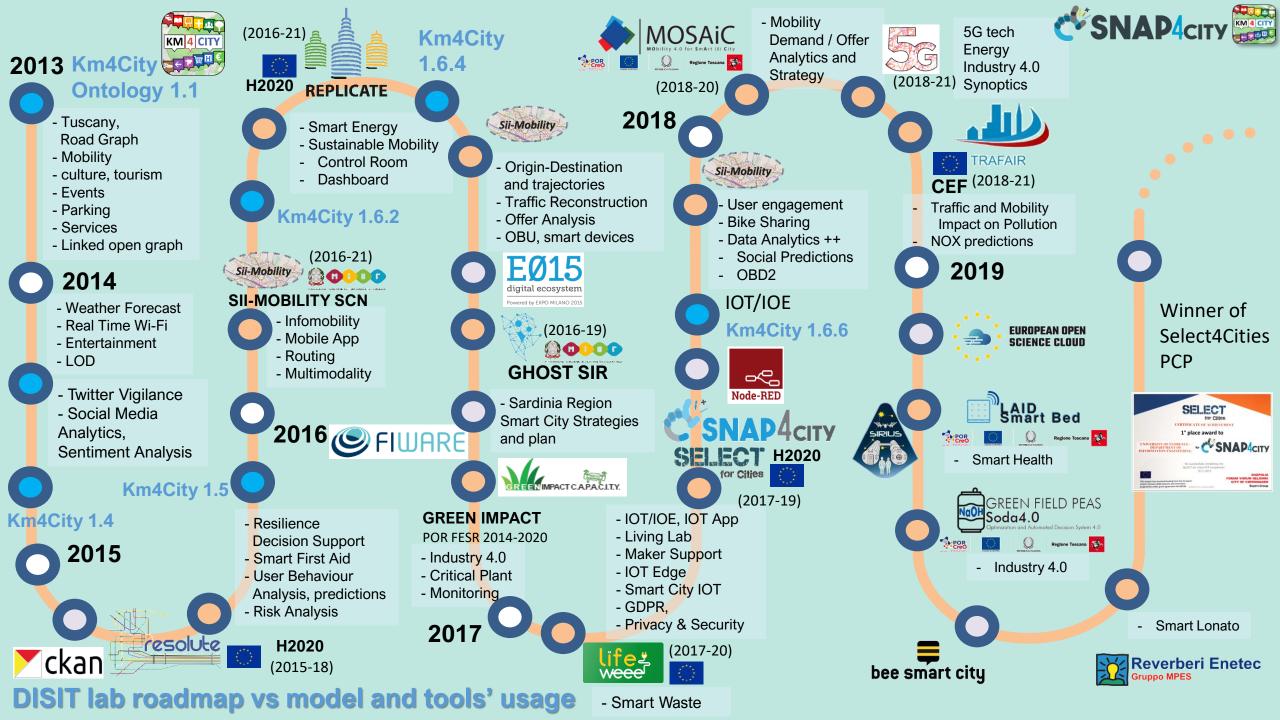
EUROPEAN OPEI

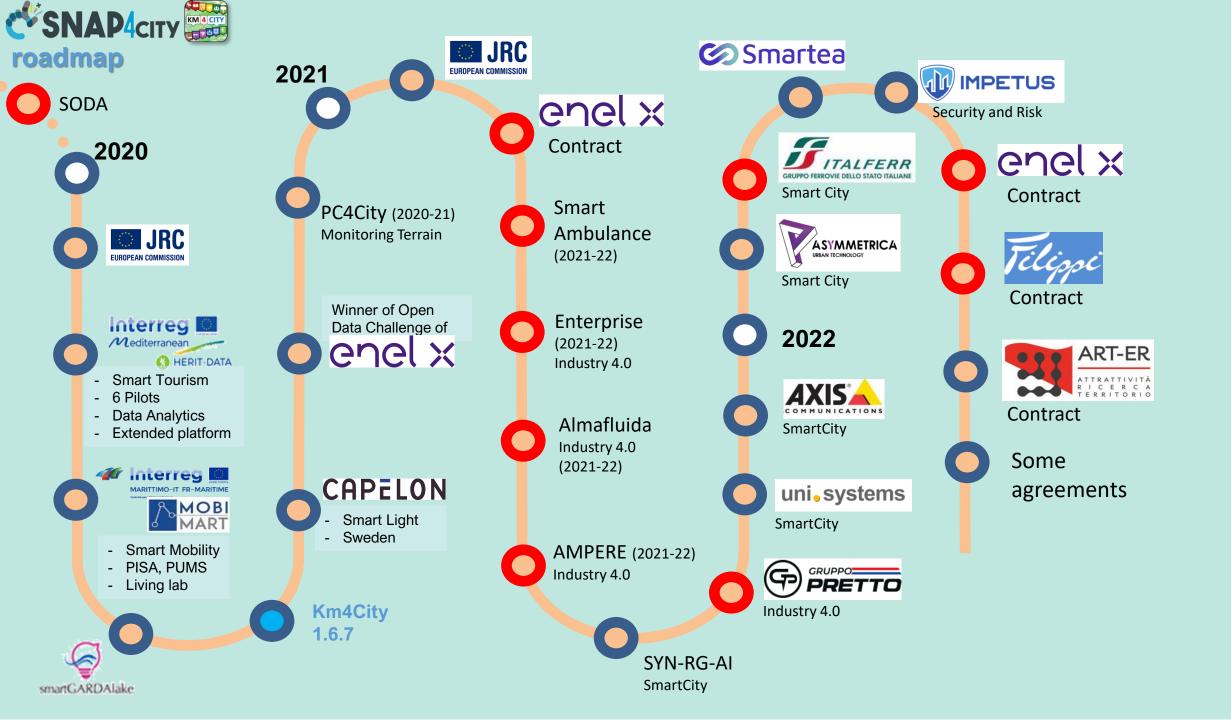
- Venezia area (I)
- WestGreece area (Gr)

Trials in Israel, Colombia, Australia, India, etc.....



0











# https://www.snap4city.org/4



- <u>Scenario: Copernicus Satellite Data</u>
- Scenario: SmartBed, Materasso Intelligente
- MicroServices Suite for Smart City Applications
- Scenario: MODBUS for Snap4Industry Snap4City Applications
- Scenario: MOBIMART Interreg: MOBilità Intelligente MARe Terra
- Scenario: City of Roma case, mobility and environmental data
- Scenario: Herit-Data video and aims
- Scenario: Control Room vs Video Wall
- Scenario: Snap4Home the case of: Alexa, Philips, Sonoff, TP-link, etc. (Italiano)
- Scenario: how to manage maintenance and accidents workflows
- <u>Scenario: Snap4Home, how to exploit Snap4City solution on home automation</u>
- Scenario: Energy Monitoring
- Scenario: Multipurpose User Engagement Tools
- Scenario: 5G Enabled Water Cleaning Control (smart city, industry 4.0)
- Scenario: High Level Control of Industrial Plant (industry 4.0)
- Scenario: Vehicle Monitoring via OBD2
- Scenario: Events and Museums Monitoring in Antwerp
- Scenario: High Resolution Prediction of Environmental Data
- Scenario: Mobility and Transport Analyses in multiple cities
- Scenario: People Flow Analysis via Wi-Fi
- Scenario: Antwerp Pilot on Environmental Data
- Scenario: Helsinki Pilot on Environmental Data
- Scenario: Firenze Smart City Control Room
- Scenario: Mobile & Web App: Toscana Where What ... Km4City, Toscana in a Snap
- Scenario: Helsinki Pilot on User Behaviour
- Scenario: Antwerp Pilot on User Behaviour





## Scenarious

- Data Analytic: Origin Destination Matrices, Algorithms and tools
- Data Analytic: Traffic Flow Reconstruction
- Data Analytic: in general, and the cases of Antwerp and Helsinki
- Data Analytic: Predicting Air Quality
- <u>Data Analytic: Analyzing Public</u>
   <u>Transportation Offer wrt Mobility Demand</u>

# https://www.snap4city.org/577



### On Line Training Material (free of charge)

	1st part (*)	2nd part (*)	3rd part (*)	4th part (*)	5th part (*)	6th part (*)	7th part (*)
what	General	Dashboards	IOT App, IOT Network	Data Analytics	Data Ingestion processes	System and Deploy Install	Smart City API: Web & Mob. App
PDF	CAMANAGE CONTROL OF THE PROPERTY OF THE PROPER	C SMAN 4 COT TO SMAN A COT TO	COMMISSION DESCRIPTION OF THE PROPERTY OF THE	C'SNAP4orr Comment is a DRAFF	CONAL ACTY  CONAL ACTY  CONTROL OF  CONTRO	C SMAS for E	C DARF for E
Inter active	C SNAP4ory Service of State of	CANALAGON BE STANKED OF THE STANKED	C SNAP for Emergina PROF	CAMADACIO COMPANIA DE LA CONTROL DE LA CONTR	CEMANAGE STREET TO STREET	CEMAPACITE STATE OF S	CANADAGY COMMING OF THE PROPERTY OF THE PROPER
Videol	You	You Tube	You Tube	You Tube	You Tube	You	You
Video2	You	You	You	You	You	You	You
Video3	You	You	You	You	You Tube	You	You
Video4	You	You	You	none	You Tube	none	none
duration	2:55	3:16	3:41	2:00	2:48	2:35	1:47

## 2022 booklets



Snap4City



https://www.snap4city.org/download/video/DPL SNAP4CITY 2022-v02.pdf

Snap4Industry



https://www.snap4city.org/downloa d/video/DPL SNAP4INDUSTRY 2022v03.pdf

Snap4City (C), September 2022







# Overview





SMART CITIES AND SMART INDUSTRY

Snap4City: FIWARE powered smart app builder for sentient cities

Vith the contribution of





https://www.snap4city.org/d rupal/sites/default/files/files/FF ImpactStories Snap4Cit y.pdf







# Overview

















## **Snap4City Platform**

### **Technical Overview**

From: DINFO dept of University of Florence, with its

DISIT Lab, Https://www.disit.org with its Snap4City solution

#### Snap4City:

- Web page: <a href="https://www.snap4city.org"><u>Https://www.snap4city.org</u></a>
- https://twitter.com/snap4city
- https://www.facebook.com/snap4city

#### Contact Person: Paolo Nesi, Paolo.nesi@unifi.it

- Phone: +39-335-5668674
- o Linkedin: https://www.linkedin.com/in/paolo-nesi-849ba51/
- Twitter: <a href="https://twitter.com/paolonesi">https://twitter.com/paolonesi</a>
- o FaceBook: https://www.facebook.com/paolo.nesi2

Access Level: Public

Date: 05-04-2021

Version: 5.3

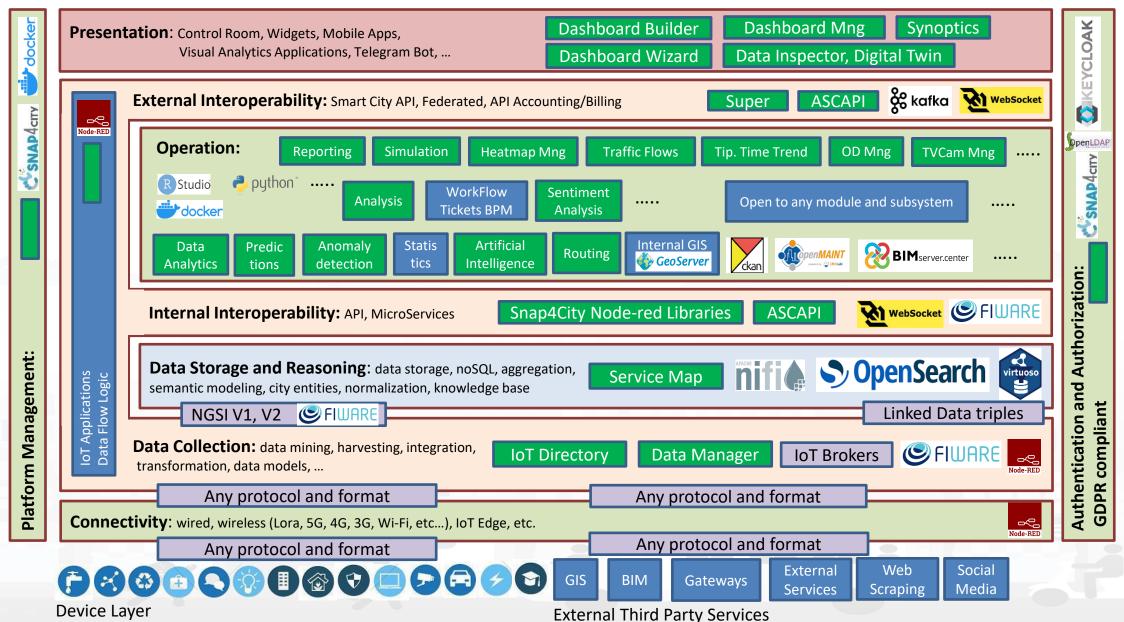
https://www.snap4city. org/drupal/sites/default /files/files/Snap4City-PlatformOverview.pdf











TOP









### CONTACT

DISIT Lab, DINFO: Department of Information Engineering Università degli Studi di Firenze - School of Engineering

Via S. Marta, 3 - 50139 Firenze, ITALY https://www.disit.org









Email: snap4city@disit.org

Office: +39-055-2758-515 / 517

Cell: +39-335-566-86-74 Fax.: +39-055-2758570