

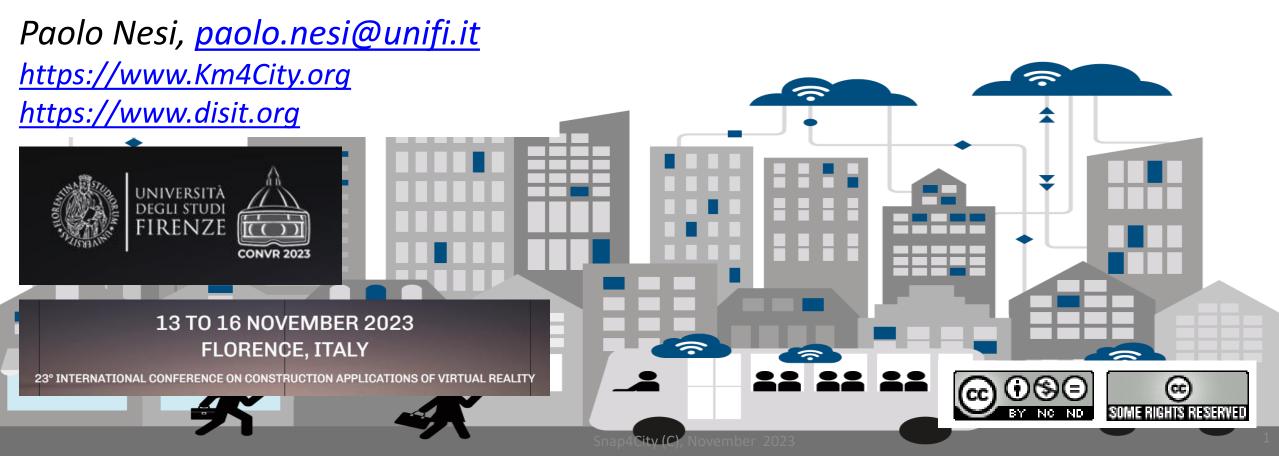








### Artificial Intelligence for Large Infrastructures: from monitoring to what-if analysis



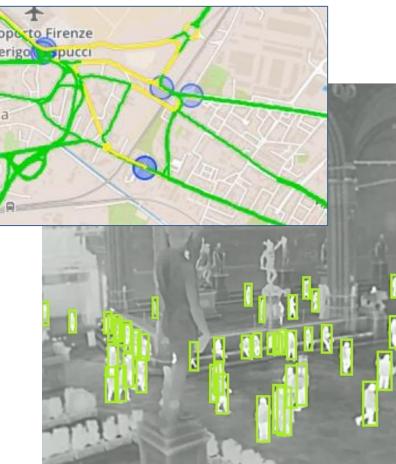






### **Public Spaces as Critical Infrastructures**

- The City is a system of systems for city users
  - Cascading effects
- Transport networks
  - Main means for rescue teams, food, water, etc.
- Communication, ICT infrastructure
  - TV cam, switches, cyber,
- Energy networks
  - power supply for health, cyber systems, etc.
- Hospitals networks
- Aggregation areas



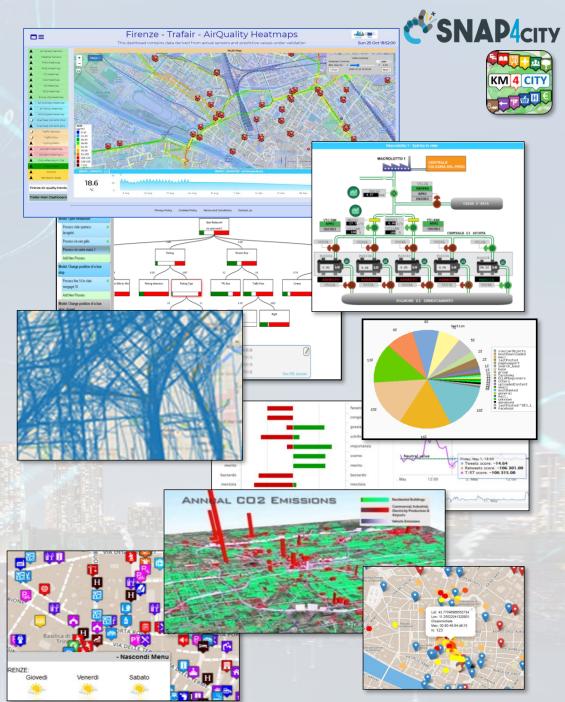
https://www.snap4city.org/download/video/DPL\_SNAP4SOLU.pdf

### **Data Driven Decision Support**

- Decision Support system
  - Assessment / Strategies
  - Data Rendering,
    - visual analytics, business intel..
  - Data Analytics, ML, Al
  - Data aggregation, Storage, indexing
  - Data Ingestion



Snap4City (C), November 2023





### Powered by



PEN Test Passed





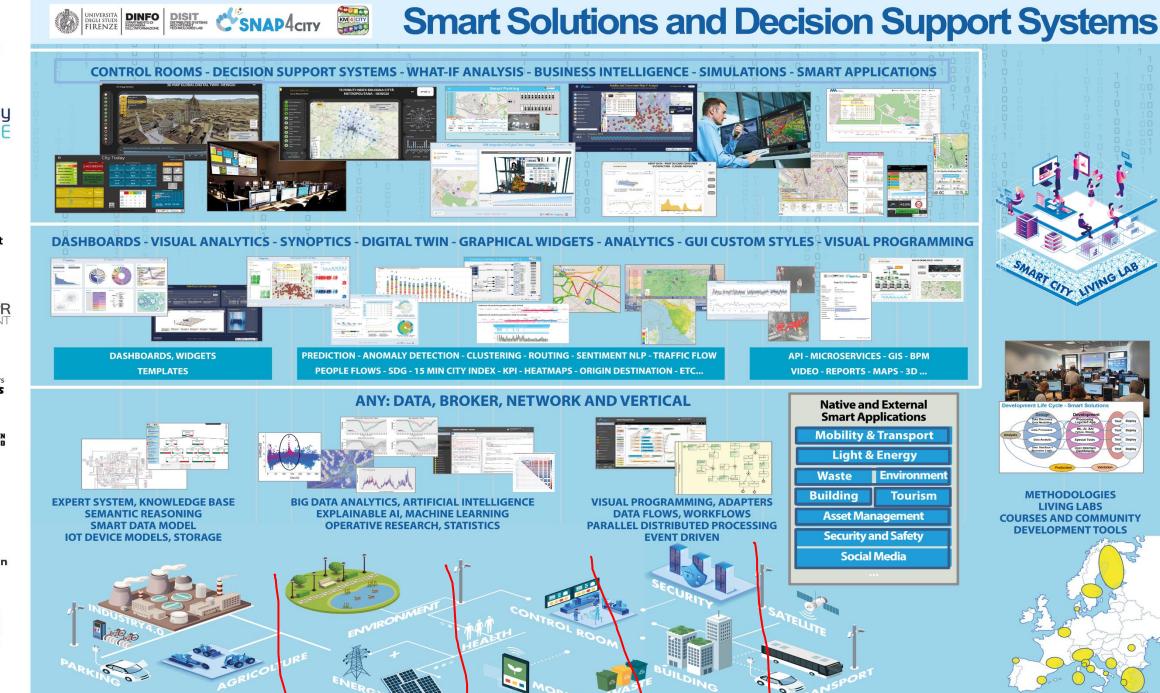












### **High Level Types**

• POI, IOT Devices, shapes,..

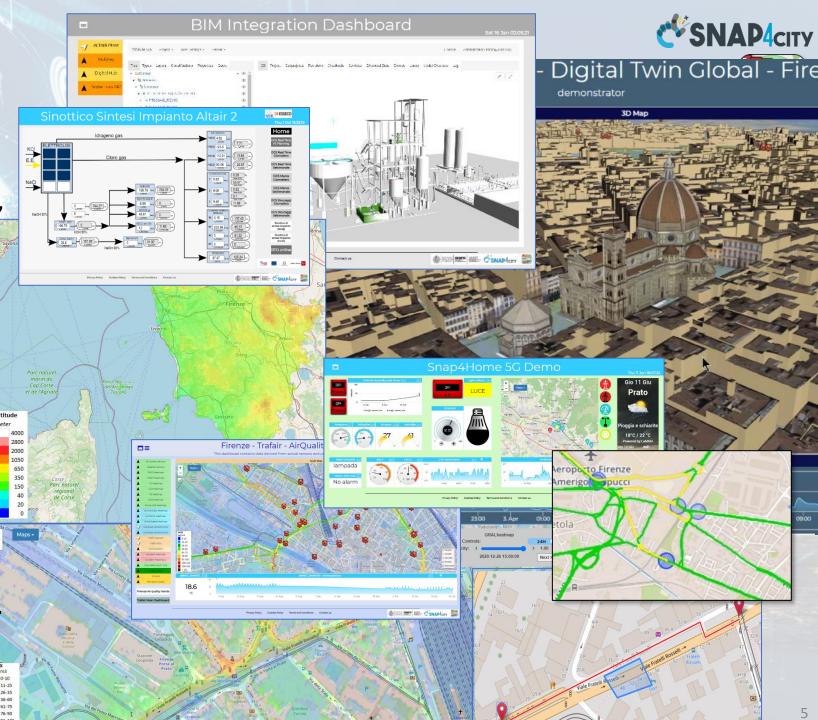
- FIWARE Smart Data Models,
- IoT Device Models
- GIS, maps, orthomaps, WFS/WMS, GeoTiff, calibrated heatmaps, ...
- Satellite data, ..
- traffic flow, typical trends, ..
- trajectories, events, Workflow, ..
- 3D Models, BIM, Digital Twins, ..
- OD Matrices of several kinds, ..
- Dynamic icons/pins,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

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• decision scenarios, ....

etc.

10/22



## Standards and Interoperability (6/2023)

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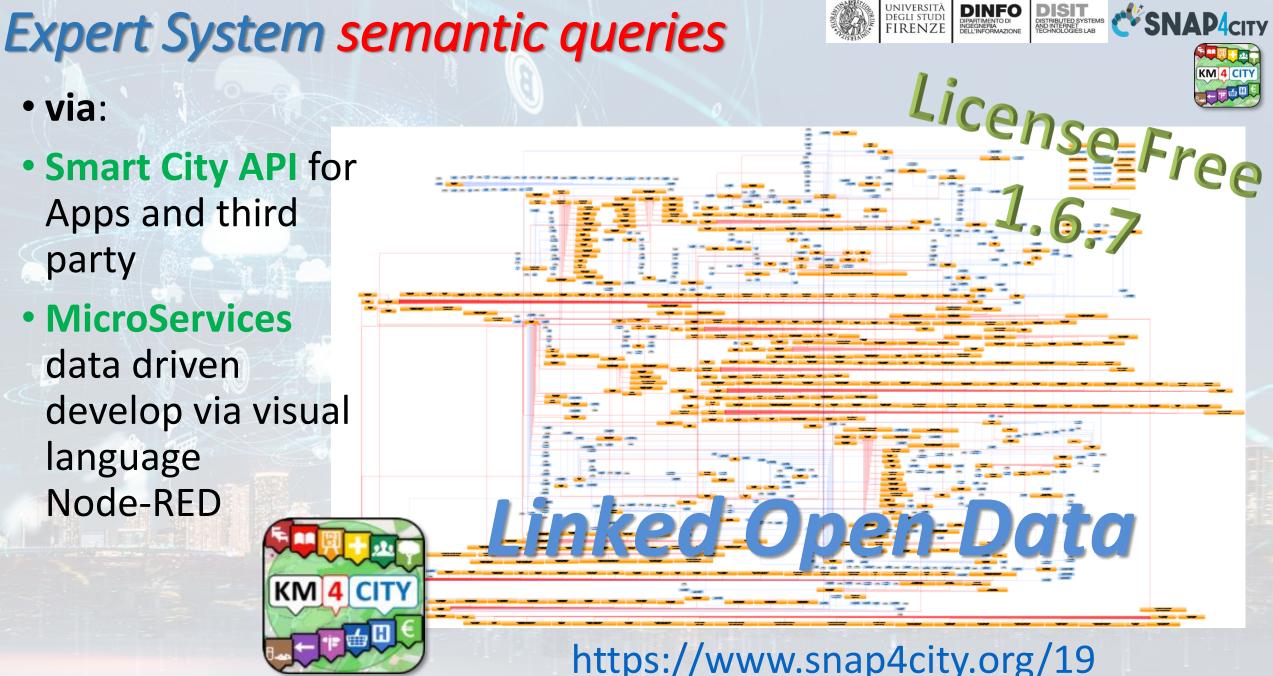
**Compliant with:** 

- 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, VMS, ....
- 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, NeTEx, ..
- Social:Twitter, FaceBook, Telegram, ..
- Events: SMS, EMAIL, CAP, RSS Feed, ..
- OS: Linux, Windows, Android, Raspberry Pi, Local File System, AXIS, ESP32, etc.





https://www.snap4city.org/65



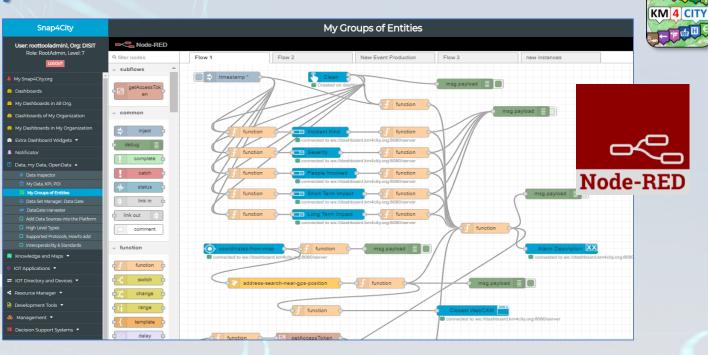
## • via:

- Smart City API for Apps and third party
- MicroServices data driven develop via visual language Node-RED

### 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



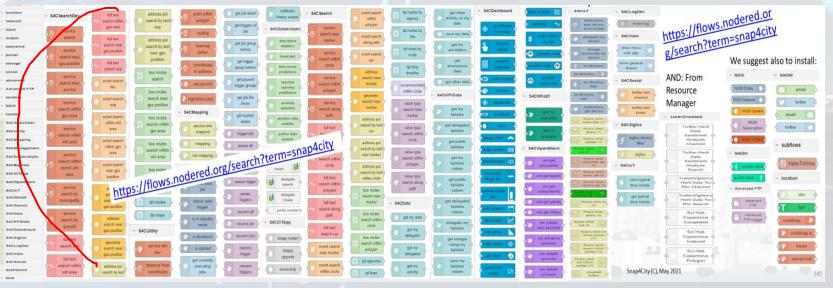
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### Available Al Solutions on Snap4City

- Mobility and Transport
- Environment, Weather, Waste, Water
- City Users Behaviour and Social analysis
- Energy and Control, Security, .....
- Tourism and People
- Security and Safety
- High Level Decision Support Solutions
  - Asset management
  - Resilience and Risks Analysis
- Low level Techniques

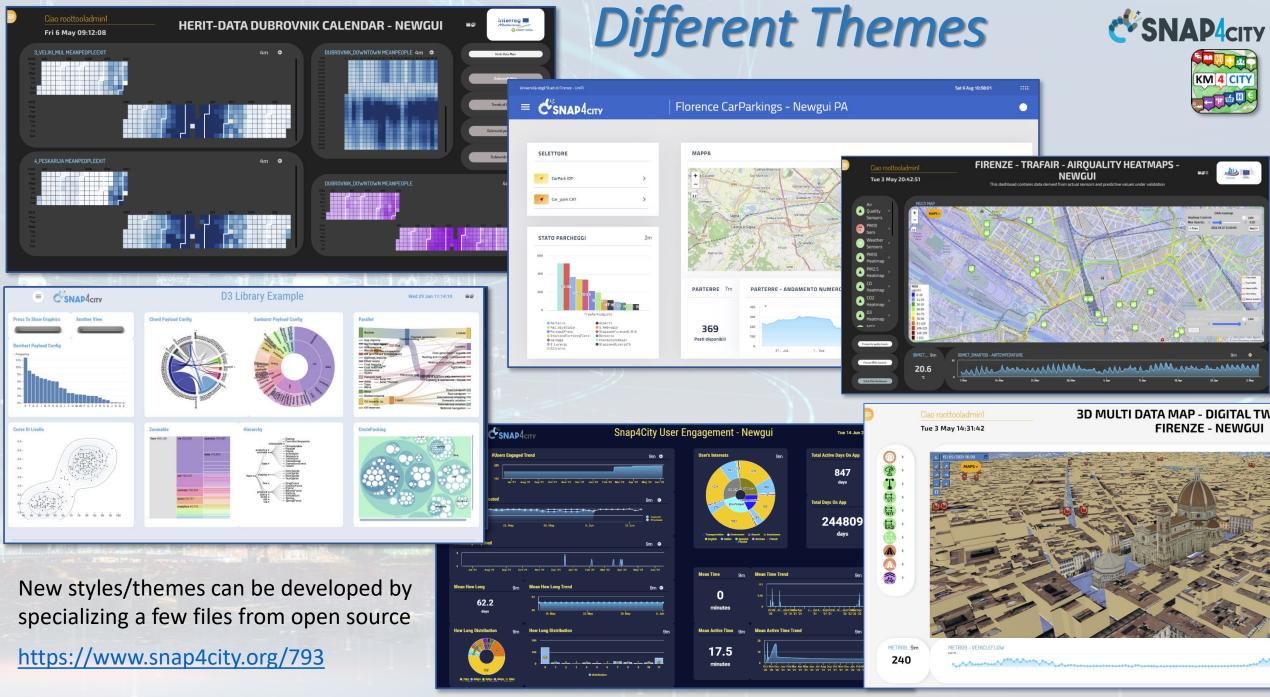
https://www.snap4city.org/download/video/course/p4/





https://www.snap4city.o rg/download/video/DPL SNAP4SOLU.pdf







## Dashboards and GUI Purposes

- Real Time: control room, monitoring, acting
  - H24 Video Wall representation of the status:
- Quasi Real Time, short term monitoring and management/acting
  - Situation Rooms: interactive data representation with visual analytics and business intelligence, What-if analysis by scenario
  - Operational management, real time What-if analysis by scenario
- Mid and Long term, for tactic and strategic planning/restructuring
  - Visual Analytics and in deep Business Intelligence
  - Long term What-If analysis

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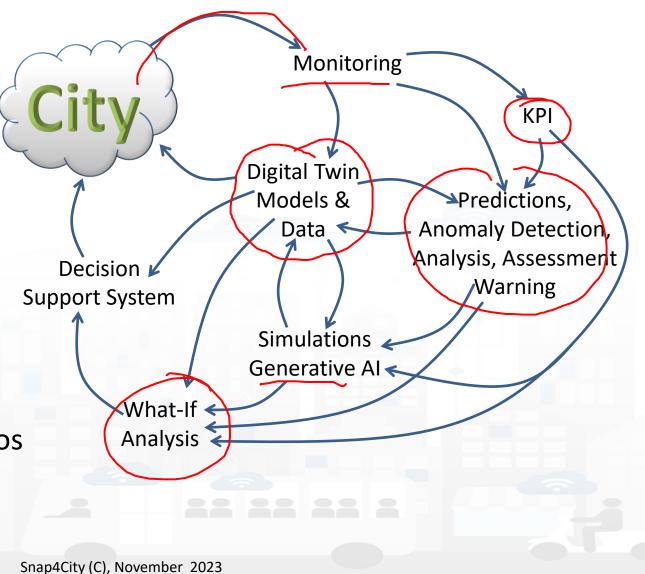






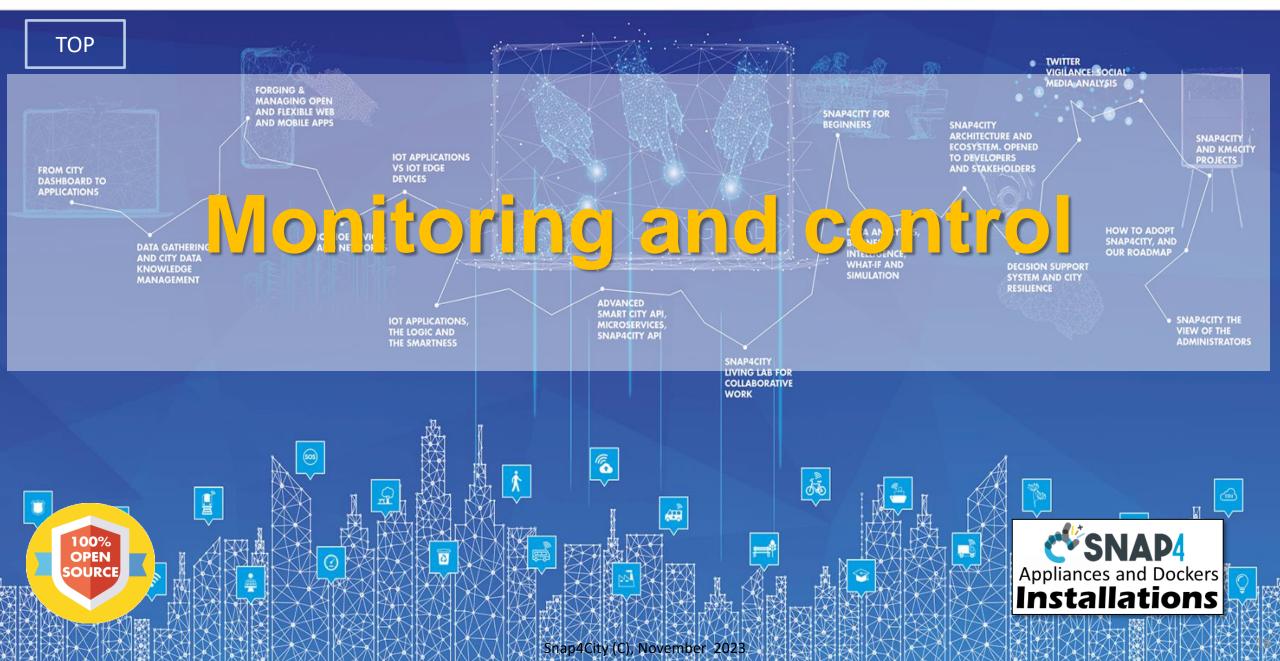


- Controlling Status: management, and operational.
  - Monitoring via KPI
  - $\circ$  Computing predictions vs KPI
  - $\,\circ\,$  Anomaly detection
  - Neuro-Symbolic analysis
  - Risk assessment
  - $\,\circ\,$  Early warning on critical conditions
- Making plan: tactic and strategic, medium and long range, micro/macro
  - Simulation & predictions
  - Generative AI Prescriptions, scenarios
  - Resilience to Unexpected unknows
  - What-if analysis wrt scenarios



#### **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:

https://www.snap4city.org/74

- Multiple Levels, Mobile Apps, API
- Since 2017

Snap4City (C), November 2023









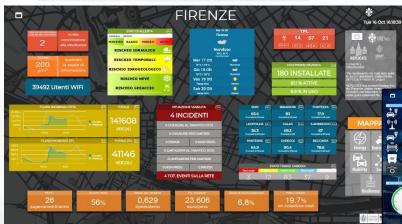






- **Smart City Control Room**
- **Dashboards and Services**
- Mobile App: Firenze Where What





- 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
- Weather
  - Forecast and actual

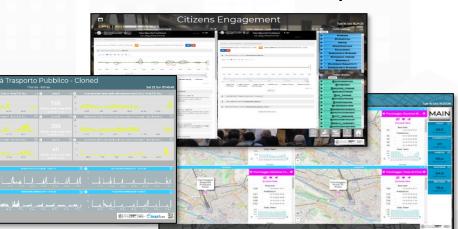


Florence Case **SNAP4**city

- 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











Slow

Mobility

Food

Services

Economy

diutant

Averaging peri

One day

One day

Calendar yea

Calendar year

Maximum daily

8-hour mean

One hou

Calendar year

Environment

Ent

15Min Ir

Weight

Average



Government

Health

Average

Housing

READ IN

<=>

WHOguidelines

99<sup>th</sup> percenti

(3 days/year

99<sup>th</sup> percentil

(3 days/year

ncentration

25 µg/m³ (\*

10 µg/m<sup>3</sup>

50 µg/m³ (\*)

20 µg/m<sup>3</sup>

100 µg/m<sup>i</sup>

200 µg/m³ (\*)

40 µg/m3

 $\infty$ 

GOALS

PRENERSON IN THE ROAD

搿

Services

Fast

Mobility

3

8 DESKI HORAND

**Air Quality Directive** 

Target value, 25 µg/m<sup>3</sup>

Limit value, 50 ug/m

Limit value, 40 µg/m<sup>3</sup> (\*)

Target value, 120 µg/m<sup>3</sup>

Limit value, 200 µg/m3 (\*)

Limit value, 40 µg/m

The target value has become a

limit value since 1 January 2015

Not to be exceeded on more

than 35 days per year

Not to be exceeded on more

than 25 days per year, averaged

over three years Not to be exceeded more that

18 times a calendar year

Sport



## **Key performance 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: NO2, PM10, PM2.5

(https://environment.ec.europa.eu/topics/air en);

- PUMS: mobility and transport vs wnv
- SUMI: mobility and transport vs env
- ISO indicators: city smartness, digitization. **Tech level**

Global VS Local

**Real Time** 

VS **Sporadic** 



### Traffic Flow Monitoring - Firenze - Cloned2

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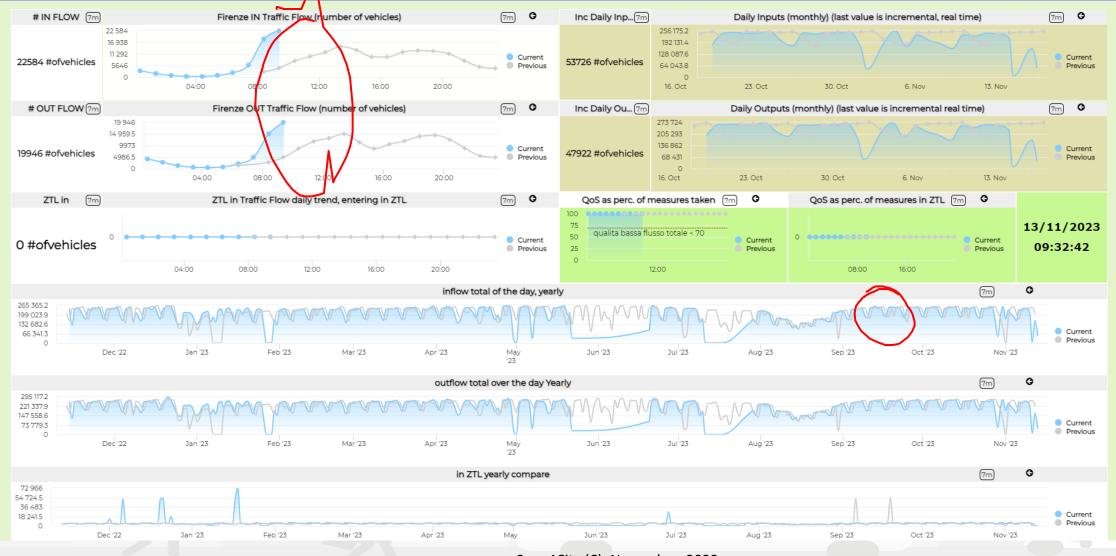
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DELL'INFORMAZIONE

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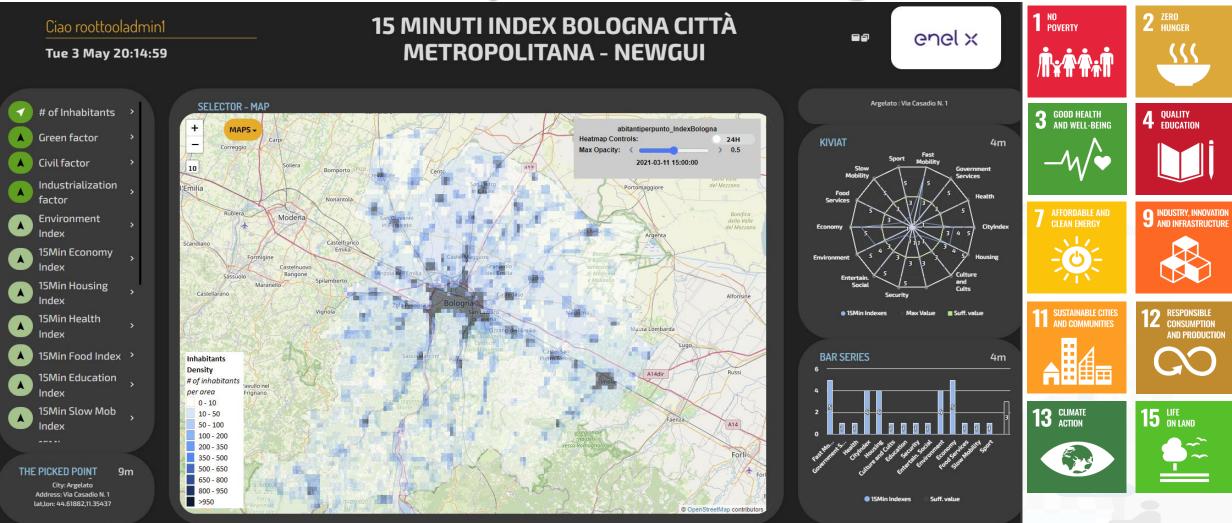








### **15MinCityIndex on Bologna**



DISIT

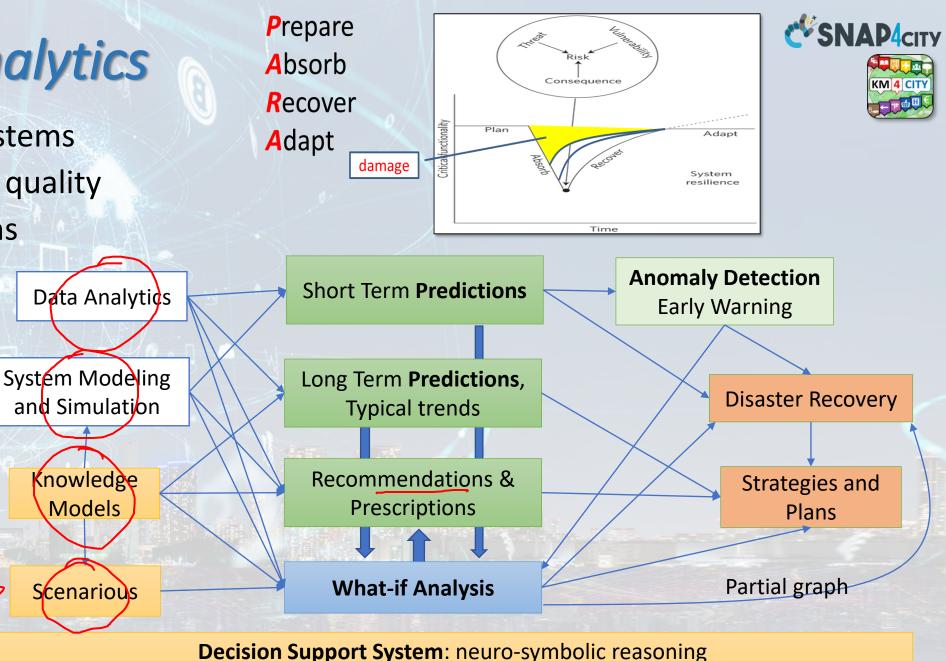
### SNAP4city

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



## **Snap4City Analytics**

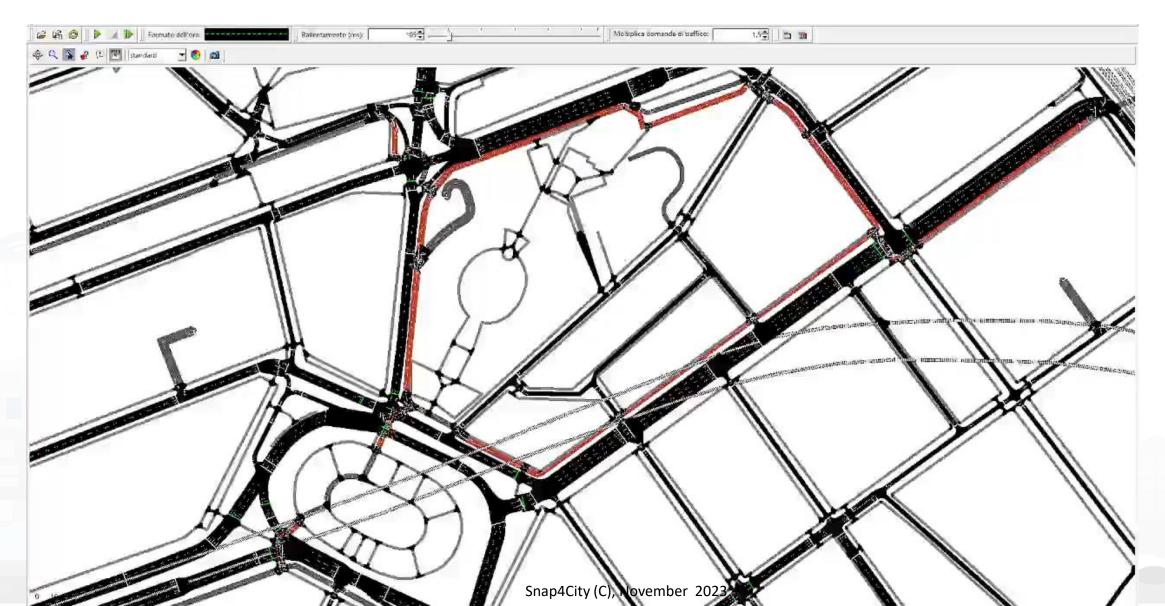
- Decision support systems
- Improvement of life quality
- Sustainable Solutions
- Reduction of costs
- Risk Assessment
- Resilience



targeting Indicators: Quality of Life, PUMS, SUMI, KPI, SDG, 15MinIndex,...

Snap4City (C), November 2023













## Smart City Digital Twin City Digital Model with...

- Intuitive platform
- Any Data TYPE, any data source, any protocol
- Data storage seamless
- ightarrow Data analytics ightarrow artificial intelligence, AI/XAI
- • Data Ethics, AI Ethics, GDPR
- Data Representation, any kind
- Key Performance Indicators, any kind
- What-IF analysis Simulation, prediction, 2D/3D
- Micro, Meso e macro scales
- Operation, planning tactic and strategic
- Collaborative and shared representation
- Sustainable, shared, open source 100%

#### **Complex and heterogeneous information, interoperability**

- o GIS, ITS, AVM, IoT, BIM, CKAN, etc.
- Satellite services
- o MaaS, last-mile delivery HUBs
- o etc.

merigo

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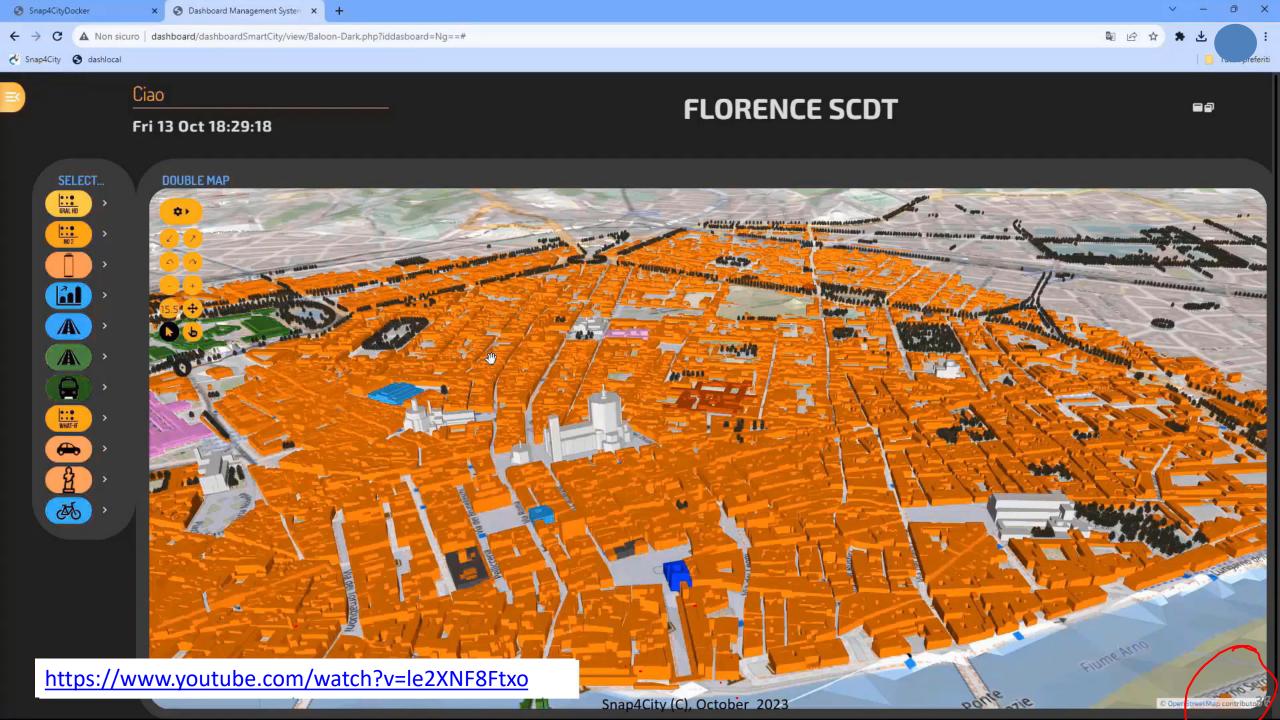












#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**





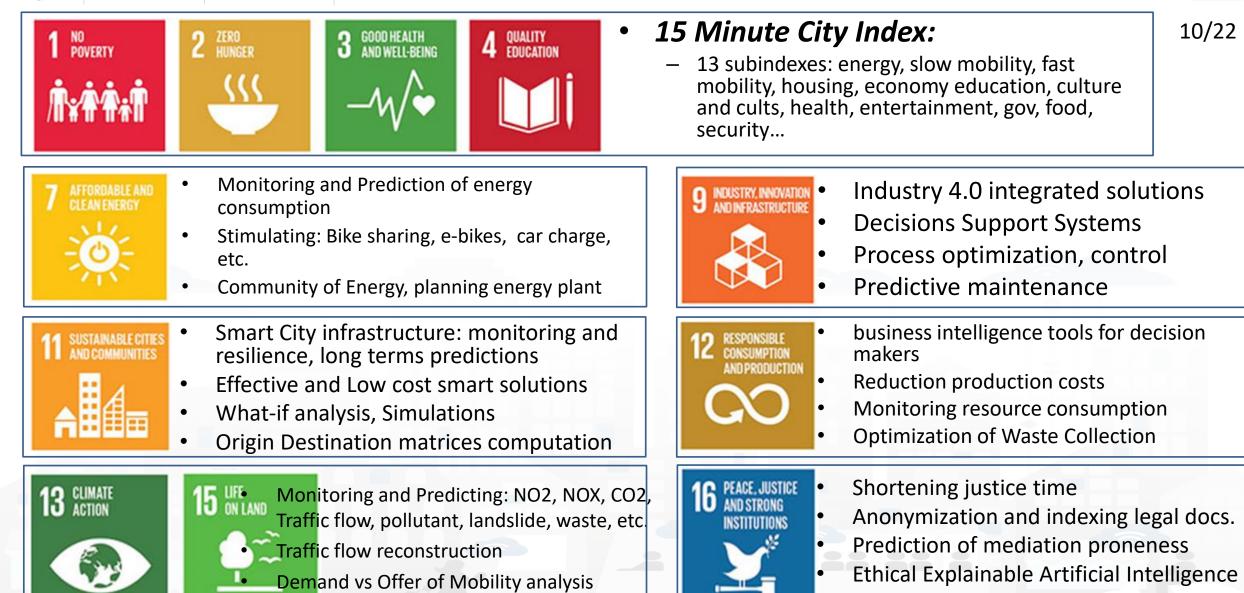






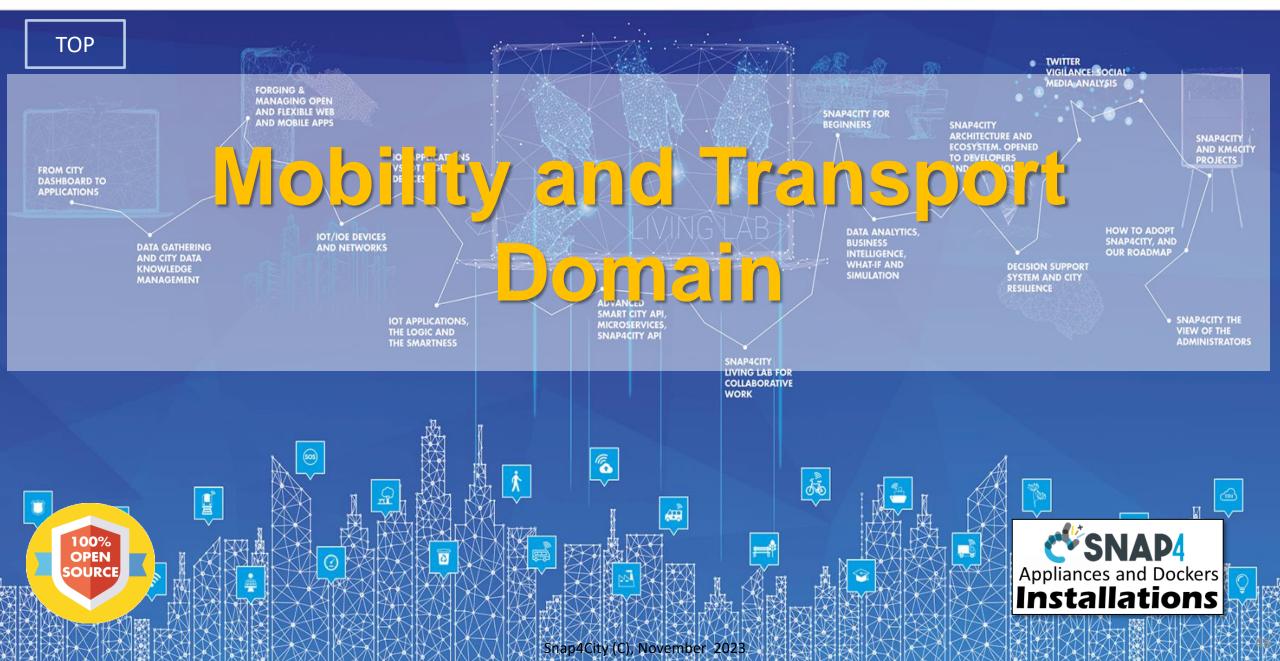






#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**









## **Mobility and Transport**

- 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)
- **Public Transportation**: Ingestion and modelling of GTFS, Transmodel, NeTEx, 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)
- 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)
- Definition of Scenarios: traffic, road graph, conditions, etc.
- Etc

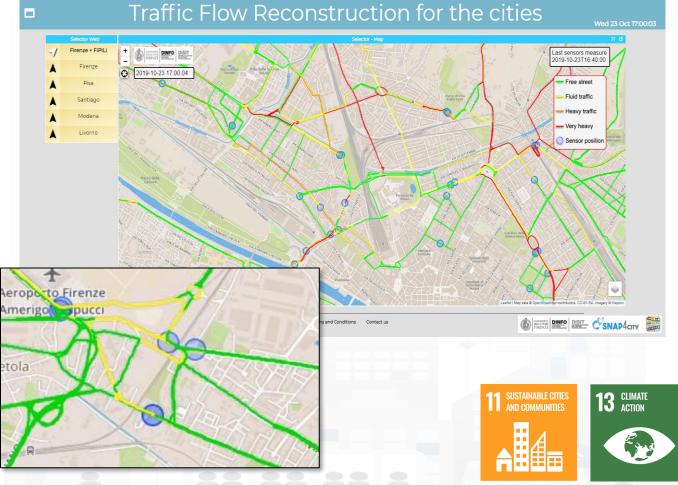


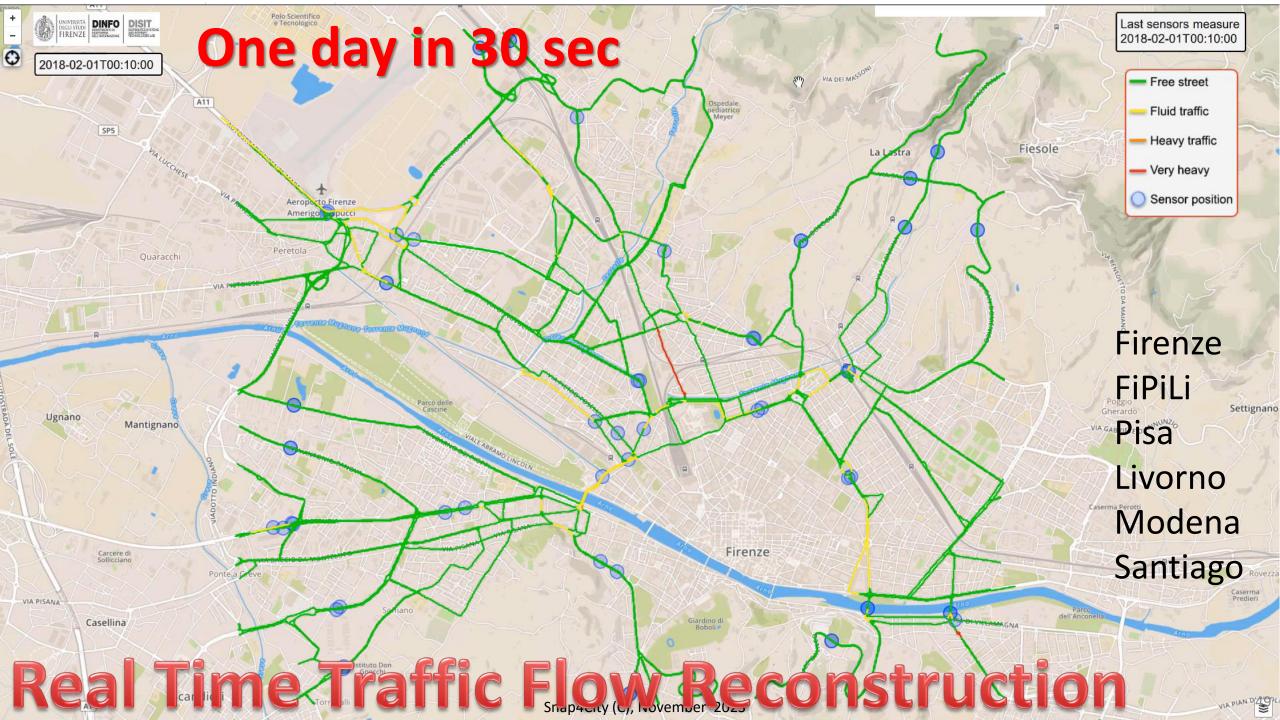


### Why Dense Traffic Flow Reconstruction ?

- Making decision on mobility and transport solutions → what if analysis
- Controlling pollution
- Dynamic Routing for Firebrigade, Ambulances, general public
- Planning Public
  Transportation routing

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







13 CLIMATE ACTION

SUSTAINABLE CITIES

AND COMMUNITIES

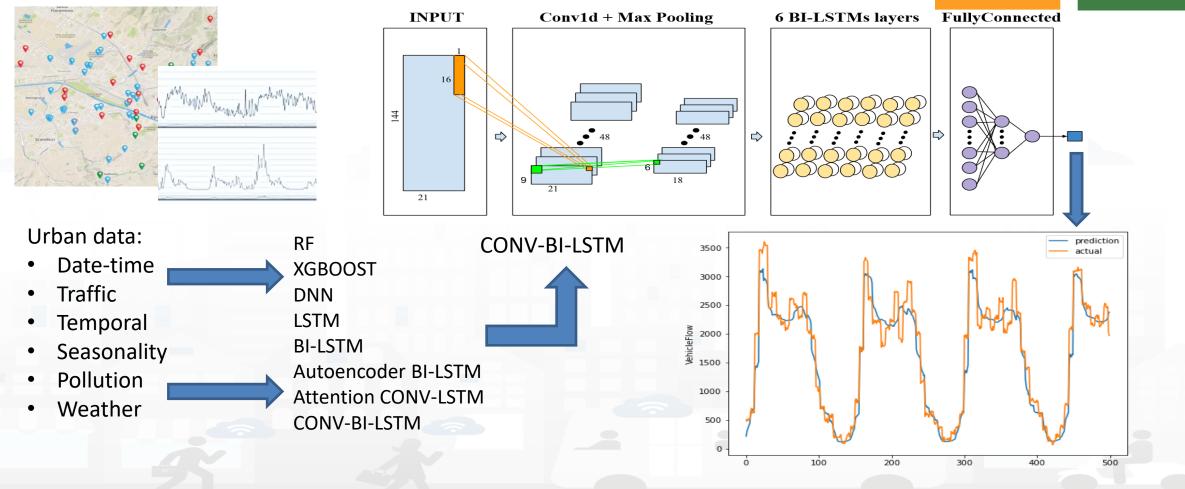
# Short-Term Prediction of City Traffic Flow via Convolutional Deep Learning

AND INTERNET TECHNOLOGIES LAB

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INGEGNERIA DELL'INFORMAZIONE





51

**Analysing Features vs ML/AI Models** 

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### Chose the best model and/or the best compromise

	Featur	es adopte	ed in the	model			Median value of MAPE for prediction results by technique min										
ID	Date time	Traf plus	Temp oral	Season ality	Air poll	weath er	RF	XGBO OST	DNN	LSTM	BI-LSTM	Autoenco der BI- LSTM	Attention CONV- LSTM	CONV- BI- LSTM			
C1	Y	Y	Y	Y	Y	Y	29.342	34.552	42.754	49.407	34.865	34,708	37,059	31.365	29.342		
C2	Y	Y	Y	Y	Y	Ν	29.682	35.545	43.400	49.832	35.870	35,707	39,506	35.613	29.682		
C3	Y	Y	Y	Y	N	Y	28.782	34.441	35.465	36.824	31.555	32,998	33,179	30.894	28.782		
C4	Υ	Y	Y	Y	N	Ν	30.935	35.373	38.942	35.383	30.564	32,969	35,713	32.485	30.564		
C5	Y	Y	Y	Ν	Y	Y	29.776	34.469	33.425	42.301	39.865	37,167	35,161	36.897	29.776		
Có	Y	Y	Y	N	Y	Ν	29.598	35.547	33.865	36.792	35.097	35,322	29,923	25.981	25.981		
C7	Y	Y	Y	N	N	Y	29.421	33.711	31.377	34.736	40.510	37,110	30,741	30.106	29.421		
C8	Y	Y	Y	N	N	N	31.245	34.414	32.026	37.823	40.662	37,538	31,263	30.500	30.500		
C9	Y	Y	N	Y	Y	Y	29.626	36.919	42.187	37.068 [38]	34.297	35,608	36,651	31.115	29.626		
C10	Y	Y	N	Y	Y	Ν	29.964	35.802	47.201	41.334	34.743	35,272	40,658	34.116	29.964	-	Quite good
C11	Y	Y	N	Y	N	Y	29.785	35.976	45.451	44.756	41.620	38,798	37,345	29.240	29.240		Quite good
C12	Y	Y	N	Y	N	Ν	31.262	35.792	36.040	37.228	32.727	34,259	32,701	29.363	29.363		model, RF
C13	Y	Y	N	N	Y	Y	29.431	35.935	34.448	35.829	34.619	35,277	32,287	30.126	29.431		model, M
C14	Y	Y	N	N	Y	N	29.764	36.374	36.203	43.510	35.744	36,059	33,015	29.827	29.764		1 data source
C15	Y	Y	N	N	N	Y	29.972	35.423	31.526	46.201	37.209	36,316	32,919	34.313	29.972		I data source
C16	Y	Y	N	N	N	N	30.960 [14]	34.235	30.338	37.068 [23]	38.082 [39]	34,235[45]	29,455[46]	28.573	28.573		Easy to compute
C17	Y	N	Y	Y	Y	Y	29.281	34.503	72.909	64.557	48.685	41,594	51,026	29.144	29.144		Lusy to compute
C18	Y Y	N	Y Y	Y Y	Y	N	30.184	35.350	59.458	68.127	46.874	41,112	44,810	30 1 03	30.163		and manage
C19	Y	N	Y	Y	N	Y	28.711	34.316	45.679	46.211	33.404	33,86	37,125 40,862	28.571	28.571		and manage
C20 C21	Y	N N	Y	N N	N Y	N Y	31.211 30.689	34.784 35.774	51.603 36.428	45.188 48.608	48.643 40.092	41,713 37,933	34,801	30.122 33.175	30.122 30.689		
C21	Y	N	Y	N	Y	N	30.505	36.165	37.337	61.168	34.420	37,933	34,801	31.434	30.505		
C22	Y	N	Y	N	N	Y	30.036	34.779	37.583	64.341	51.063	42,921	33,455	29.328	29.328		
C24	Y	N	Y	N	N	N	32.629	34.312	36.849	53.854	41.912	38,112	33,257	29.665	29.665		
C25	Y	N	N	Y	Y	Y	28.766	35.906	71.829	65.565	54.403	45,154	52,023	32.218	28.766		Best model
C26	Y	N	N	Y	Y	N	30.008	37.317	67.870	49 3 00	46.880	42,098	53,256	38.642	30.008		
C27	Y	N	N	Y	N	Y	28.986	35.218	57.938	50.333	59.419	47,318	43,298	28.658	28.658		- 1 data source
C28	Y	N	N	Y	N	N	31.068	35.878	66.634	50.957	55.096	45,487	47,097	27.561	27.561		
C29	Y	N	N	N	Y	Y	29.301	37.532	38.325	40.677	50.303	43,917	35,554	32.784	29,301		CONV-BI-LSTM
C30	Y	N	N	N	Y	N	29.323	37.284	37.149	48.801	55.064	46,174	34,721	32.294	29.323		
C31	Y	N	N	N	N	Y	20.064	36.331	34.638	56.157	45.016	40,673	35,293	35 0 19	29.964		
C32	Y	N	N	N	N	N	29.281	34.574	33.028	57.961	44.977	39,775	29,320	25.612	25.612		
										Shap+Ch	y (c), nover	1001 2025					







## Decision Support Systems, What-if

Snap4City (C), November

#### Event planning, via what-if analysis

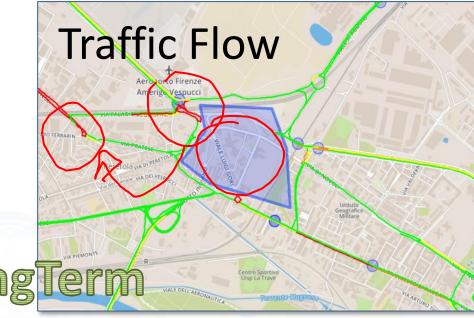
- $\circ~$  Change in the graph structure of the city
- $\circ~$  Impact on the flow of people and vehicles
- Adaptation: public transport, traffic, pedestrian management, etc.

#### $\odot$ Immediate reaction to natural events or not

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

### Digital Twin

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



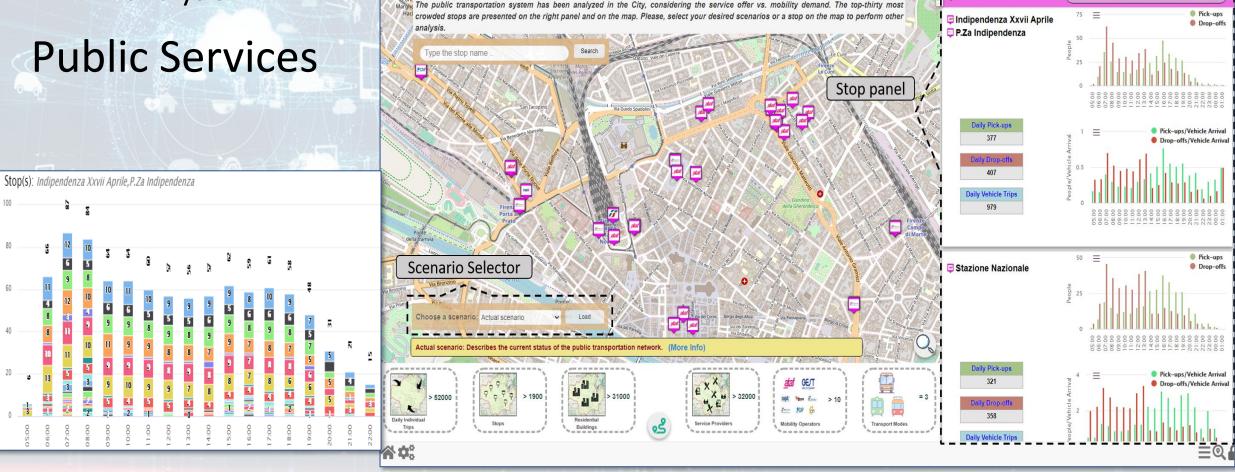
Routing

### What-if Analysis on Pub Transport

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

Welcome to DORAM

• KPI analysis



Services: 36 on 36 available

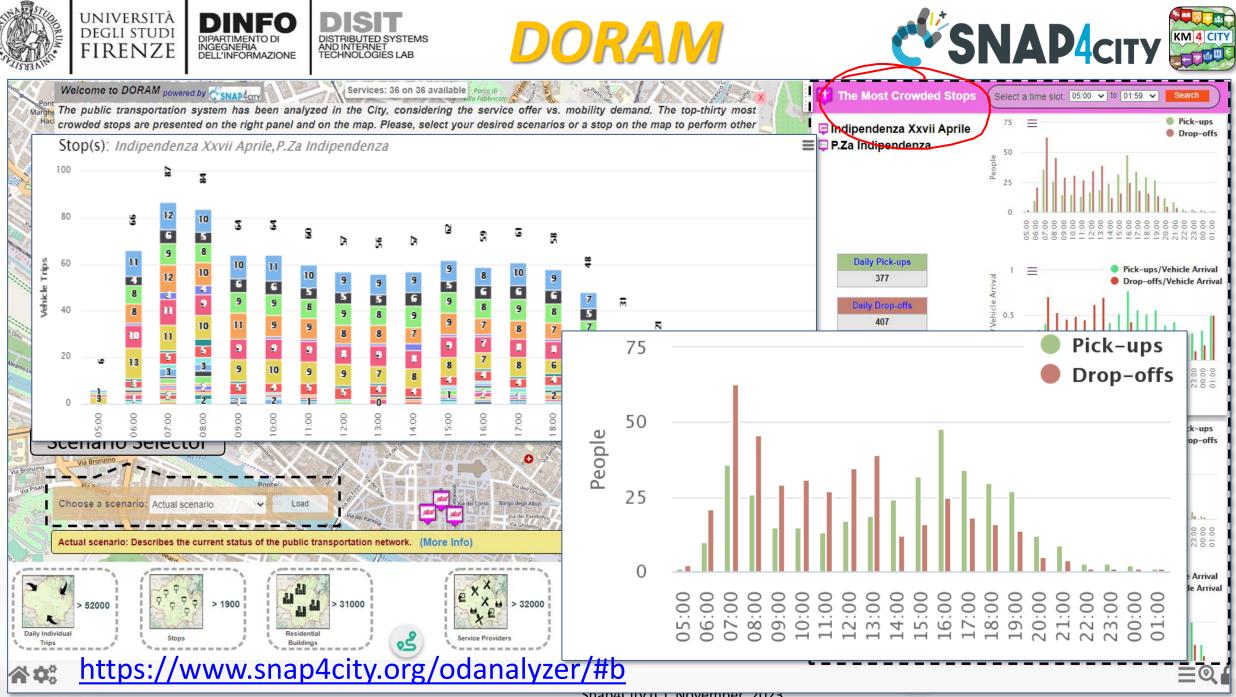


Select a time slot: 05:00 v to 01:59 v

università degli studi FIRENZE DIPARTIMENTO DI INGEGNERIA DELL'INFORMAZIONE

ne Most Crowded Stops

#### Snap4City (C), May 2022

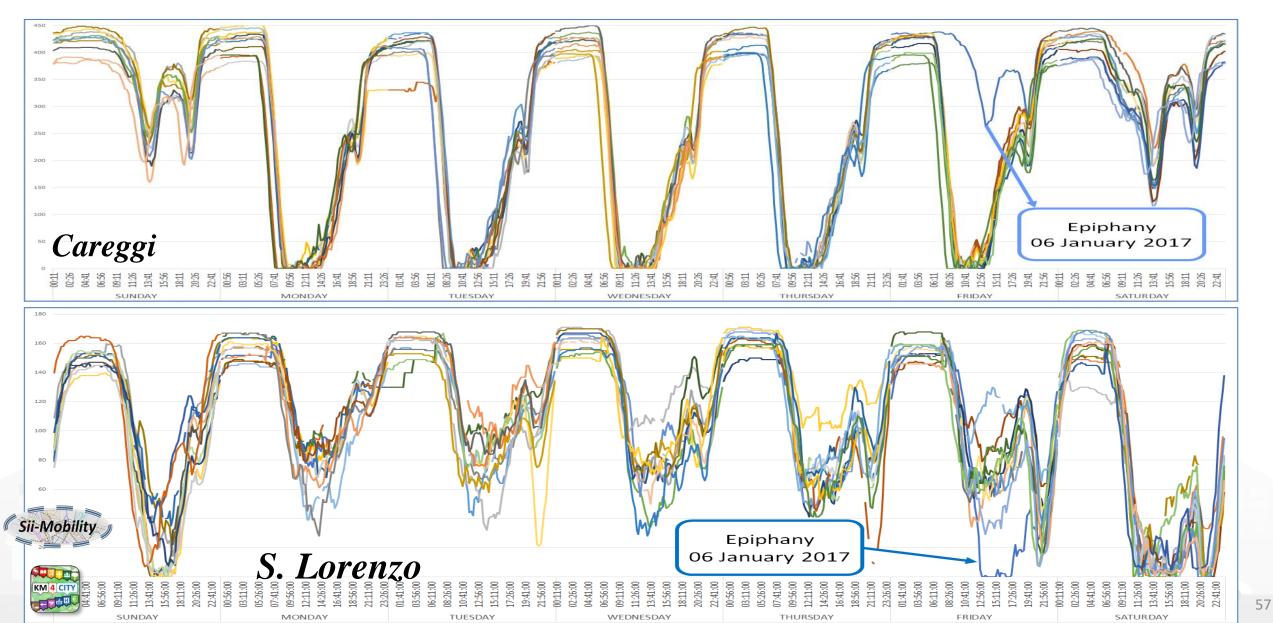


Snap4City (C), November 2023





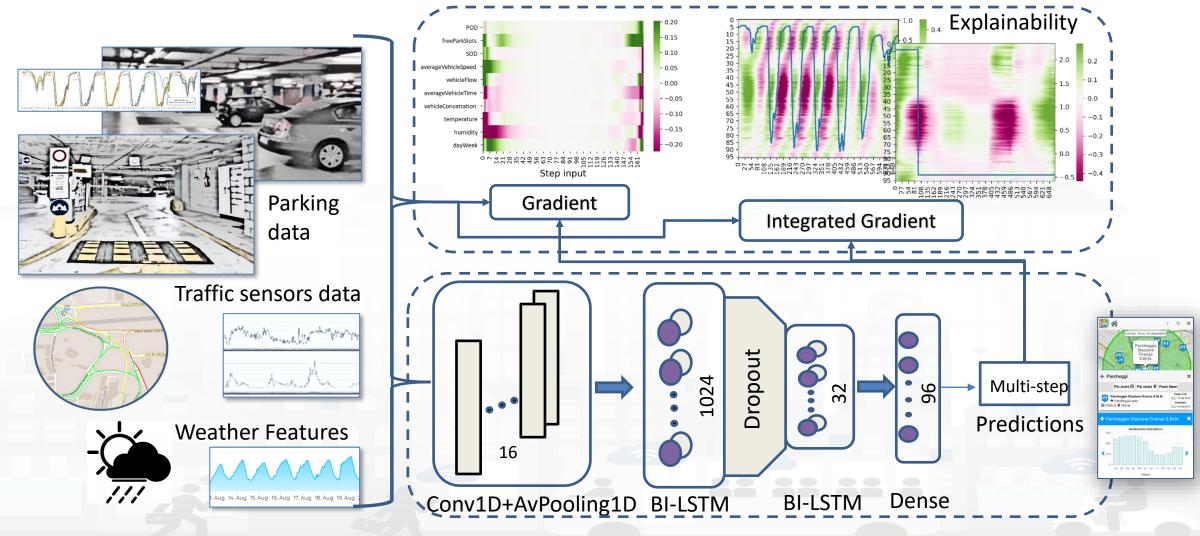
# **Free Parking space trends**







## **Deep Learning AI to surely Park!**

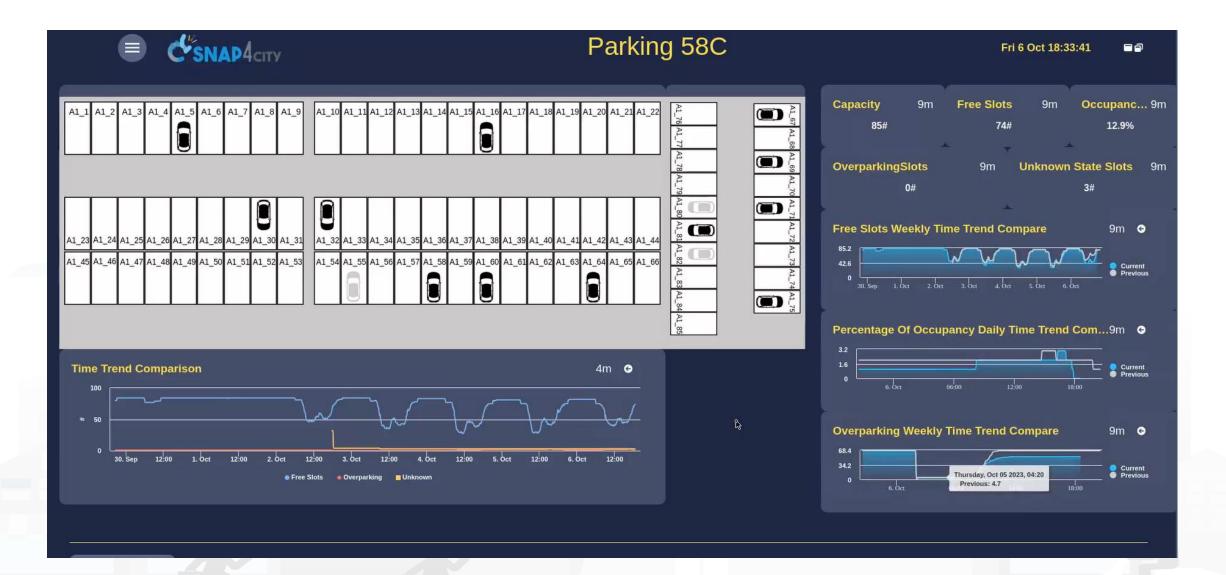


Snap4City (C), November 2023



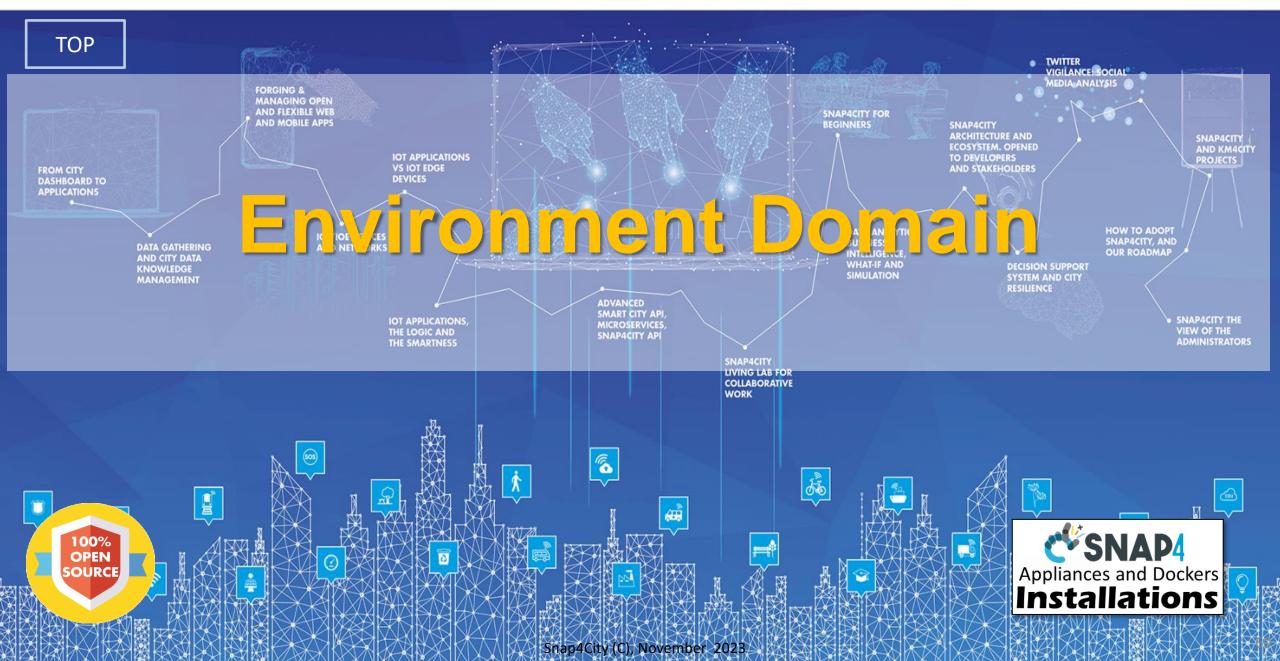






#### **SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES**





### **Environment and Quality of Life** <u>്രീപ്പിയം</u> എം **Air Quality Predictions**

 $\odot$ 

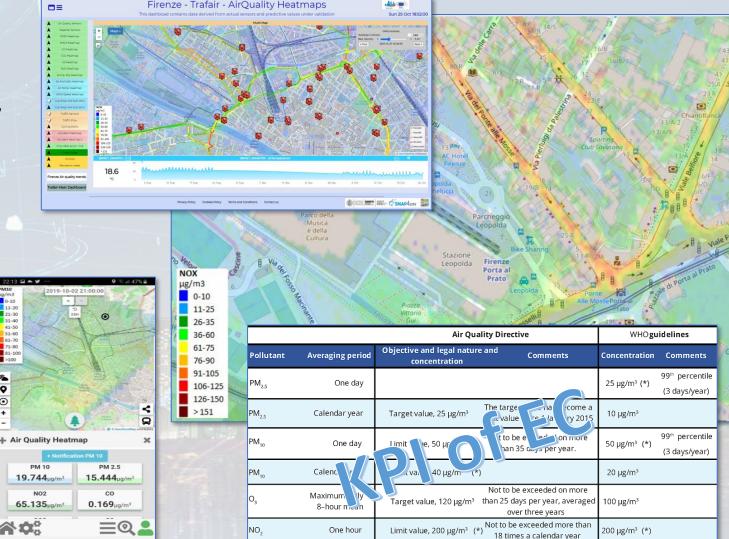
19.744µg/m

65.135µ

**D** 

- 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

•		_
///	Firenze, Pisa,	Livorna



Limit value, 40 µg/m<sup>3</sup>

Calendar vear

40 µg/m<sup>3</sup>





# **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.







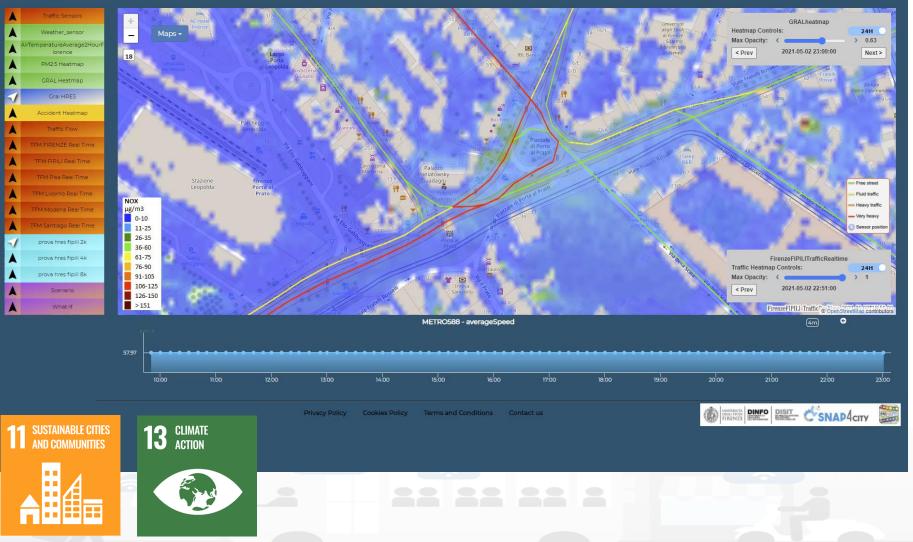
# Environment **C<sup>C</sup>SNAP4**city

Traffic Flow Manager on multiple cities



Sun 2 May 23:16:31

- **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



Snap4City (C), November 2023



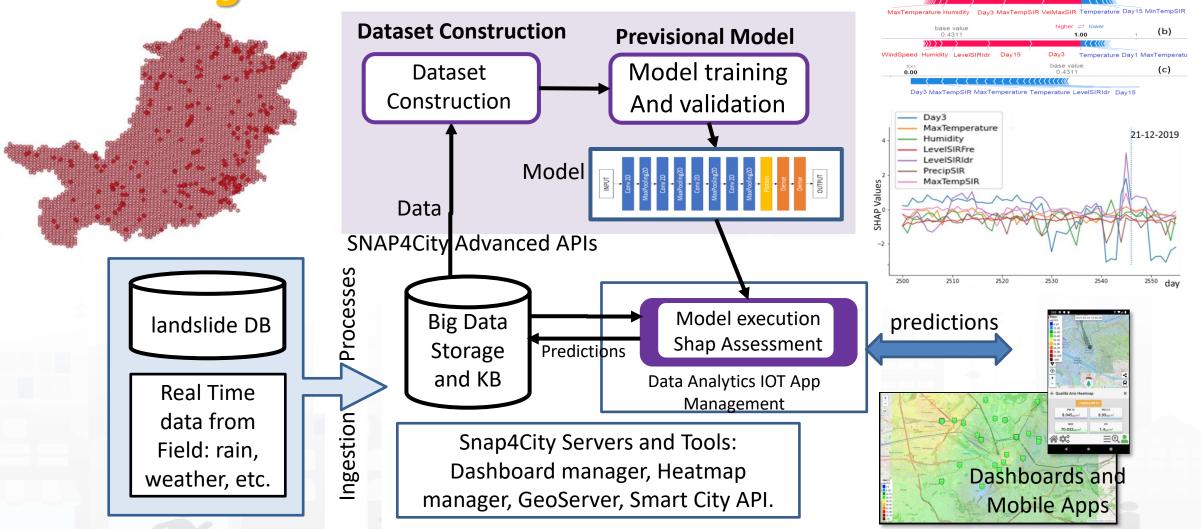
## **Predicting Land slides**





base value

0.4311



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. <u>https://ieeexplore.ieee.org/abstract/document/9732490</u> Snap4City (C), November 2023 (a)



## **Comparing Predictive Model/architectures**

UNIVERSITÀ Degli studi

FIRENZE

Model

MAE MSE RMSE Accuracy Sensitivit Specificit TSS PfA Precision F1 score MCC OA Kappa AUC INGEGNERIA DELL'INFORMAZIONE DISTRIBUTED SYSTEM AND INTERNET TECHNOLOGIES LAB

	XGBoost	RF	CNN	Auto	SIGMA	10000			
				encoder		Day3	Day		High
	0.000173	0.000334	0.000600	0.009218	0.004169	MaxTempSIR	MaxTempS		
	0.000173	0.000334	0.000259	0.009218	0.004169	LevelSIRIdr	LevelSIR		
	0.0131	0.0182	0.0160	0.0960	0.064572	Latitude	Latituc		
c <b>y</b>	0.99	0.99	0.99	0.99	0.99	Humidity	Humidit		
/ity	0.79	0.36	0.24	0.19	0.06	MaxTemperature	MaxTemperatu	re	
ity	0.99	0.99	0.99	0.99	0.99	PrecipSIR	PrecipSI	R	
	0.78	0.35	0.23	0.18	0.05	LevelSIRFre	LevelSIRFr	e	Φ
	0.01%	0.02%	0.01%	0.11%	0.39%	Day15	Day1	5	Value
on	0.63	0.35	0.33	0.64	0.003	Day1	Day	1	
е	<mark>0.70</mark>	<mark>0.36</mark>	<mark>0.27</mark>	<mark>0.29</mark>	<mark>0.007</mark>	Longitude	Longitud	e • • • • •	Feature
	0.70	0.36	0.28	0.35	0.01	Temprerature	Tempreratur	e	atı
	2.40	1.72	1.55	1.64	1.02	Day30	Day3	0	Ъ
	0.70	0.36	0.27	0.29	0.01	VelMedSIR	VelMedSIF	· · · · · · · · · · · · · · · · · · ·	
	0.89	0.68	0.99	0.92	0.53	VelMaxSIR	VelMaxSIF		
						WindSpeed	WindSpee		
						MinTempSIR	MinTempSI	R	
						Altitude	Altitud	le <b>en la constant</b>	
						Vegetation	Vegetatio	in ••••••••••••••••••••••••••••••••••••	
Global Explainable AI				le Al		MinTemperature	MinTemperatur	e	Low
- Feature relevance						0.0	0.2 0.4 0.6 0.8 1.00	-6 -4 -2 0 2 4 6	
							Mean( SHAP value )	SHAP value (impact on model output)	
								Red: positive, blue: negativ	/e;
							- vs intensity and impact		
								is intensity and impact	65





## Local Explainable AI - understanding the single event

- The local explanation puts in evidence the features which provided major contribution to the prediction
- For example considering Figure10a, the value of VelMaxSIR, MaxTempSIR, Day3 and Humidity contributed significantly to the classification of the observation as a landslide event

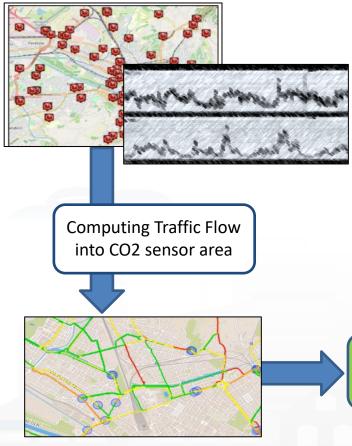


Day3 MaxTempSIR MaxTemperature Temperature LevelSIRIdr Day15

FIGURE 10. Local feature relevance via SHAP, as interpretation of events in terms of feature values: (a) and (b) are events with predictions of landslide, (c) a no landslide event.



# **Estimating City Local CO2 from Traffic Flow Data**



UNIVERSITÀ

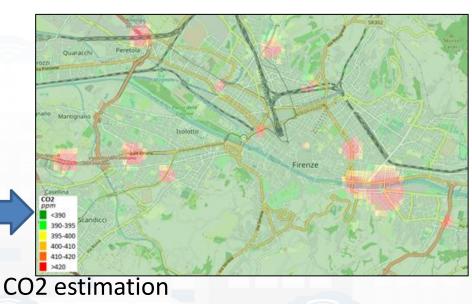
degli studi FIRENZE

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





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

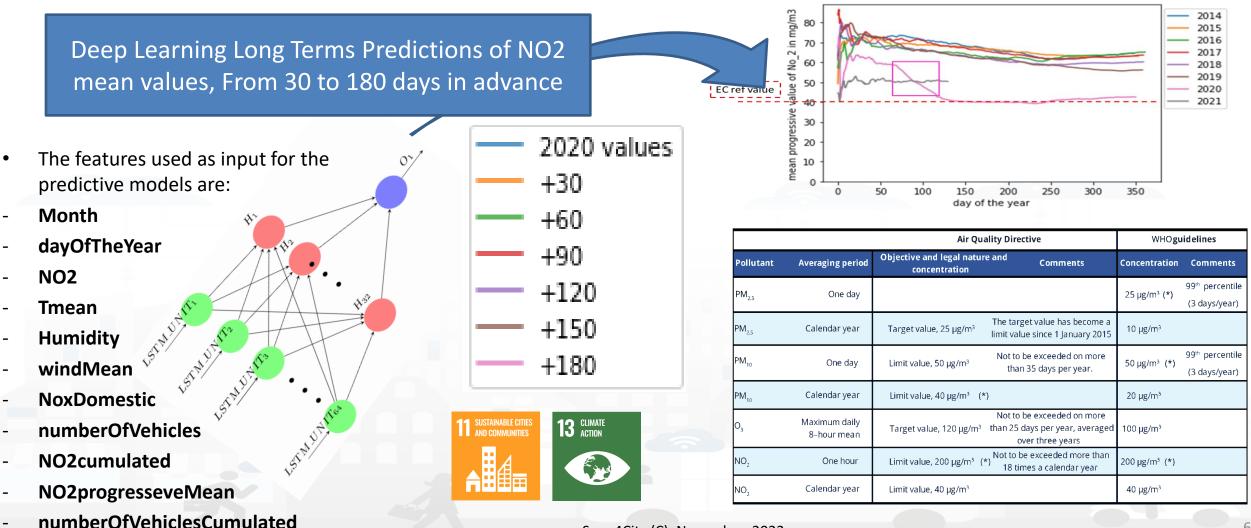
Snap4City (C), November 2023







# Predicting EC's KPI on NO2 months in advance



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



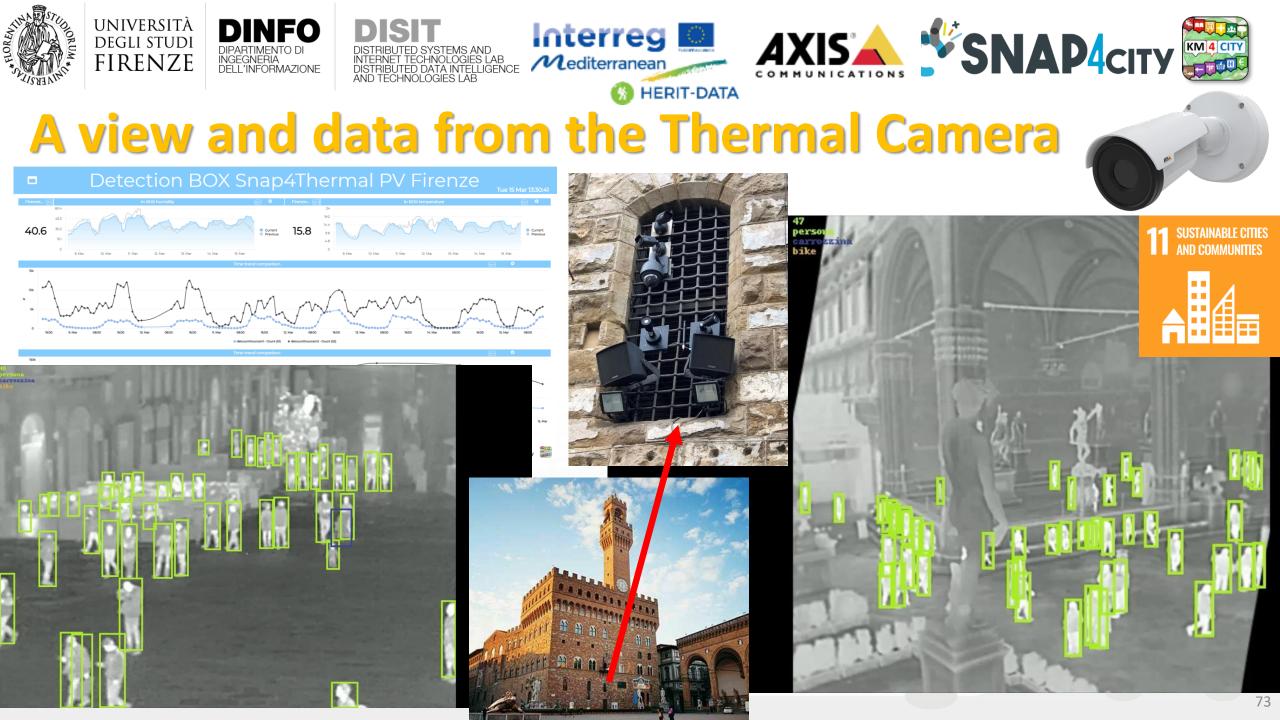






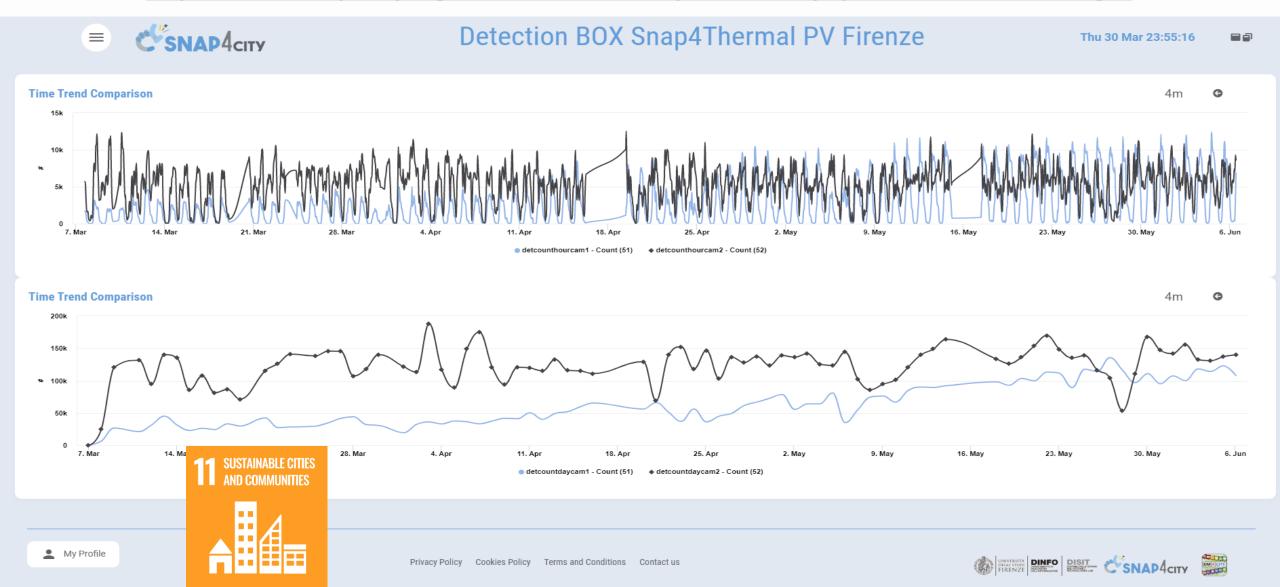
## **City Users Behaviour, Safety, Security and Social Analysis**

- People detection and classification: persona, strollers, bikes, etc. (ML, DL)
- people counting and tracking, head counting, people trajectories (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)
- Video management System integration for security
- **15 Minute City Index** , etc. (modeling and computability)
- Computing SDG, etc., (DP)
- Ftc.





#### https://www.snap4city.org/dashboardSmartCity/view/Gea.php?iddasboard=MzM3Ng==









## **People Counting and Tracking**

DINFO DIPARTIMENTO DI INGEGNERIA DELL'INFORMAZIONE



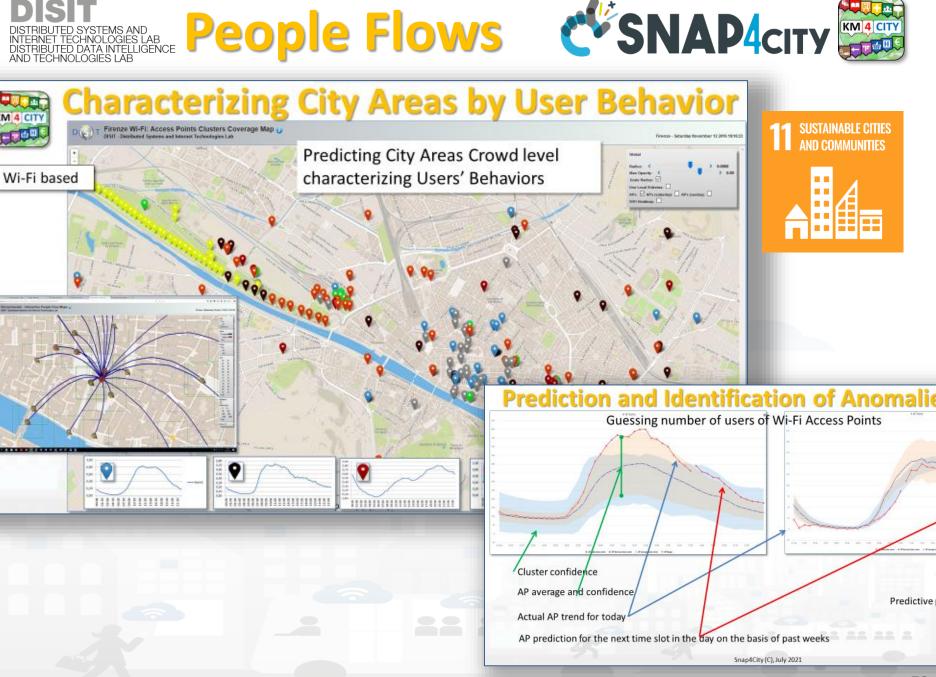
DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB DISTRIBUTED DATA INTELLIGENCE AND TECHNOLOGIES LAB



- UNIVERSITÀ DEGLI STUDI FIRENZE INGEGNERIA DFI I 'INFORMAZIONE Prediction of
  - people flows on the basis of Wi-Fi data

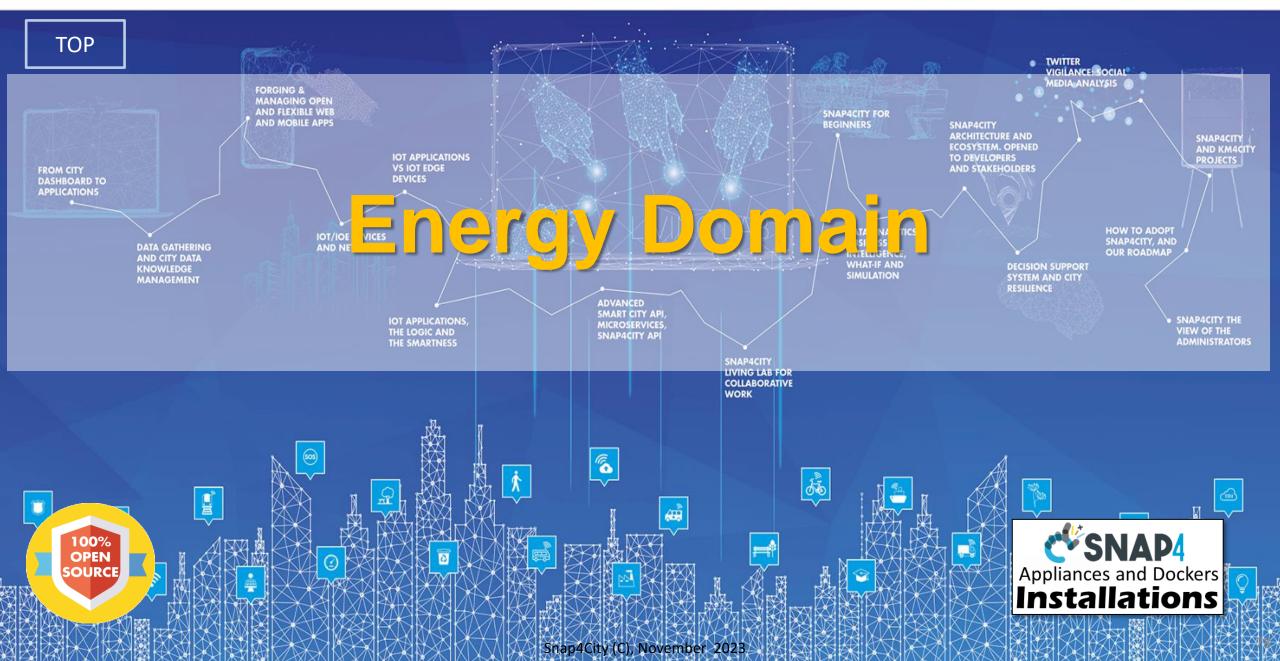
KM 4 CITY

- Anomaly detection
- Resolute H2020
- Classification of city areas



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









# Energy

- Monitoring Energy Consumption in single building, area and per zone
- Matching Energy consumption with respect to the actual usage
- Computing Roof orientation for Photovoltaic installations
- Simulation of Photovoltaicc installations to identify the best parameters of size and storage
- Smart Light management, unicast and multi cast management, smart light controlled by traffic flow data
- Collecting and managing Communities of Energy
- Monitoring Energy provisioning on **recharging station**
- Optimization of battery life
- Computing KPI
- Etc.









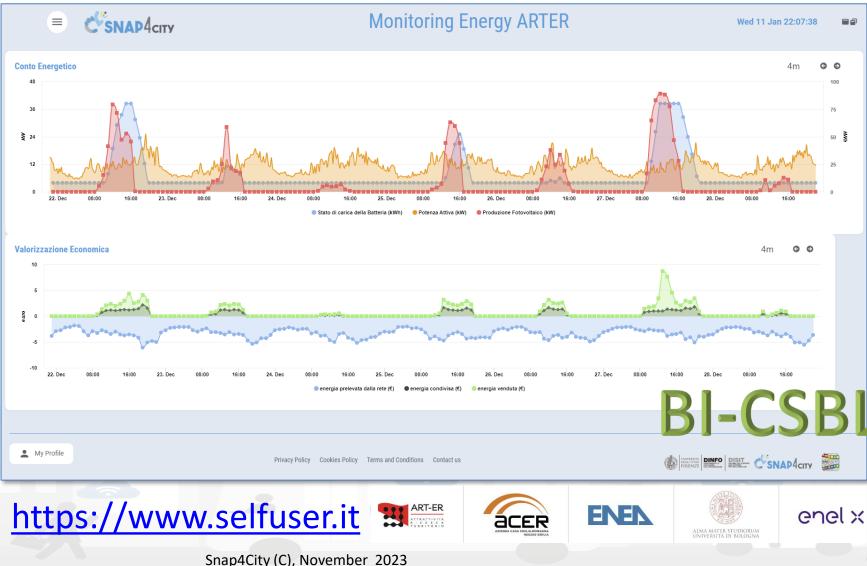






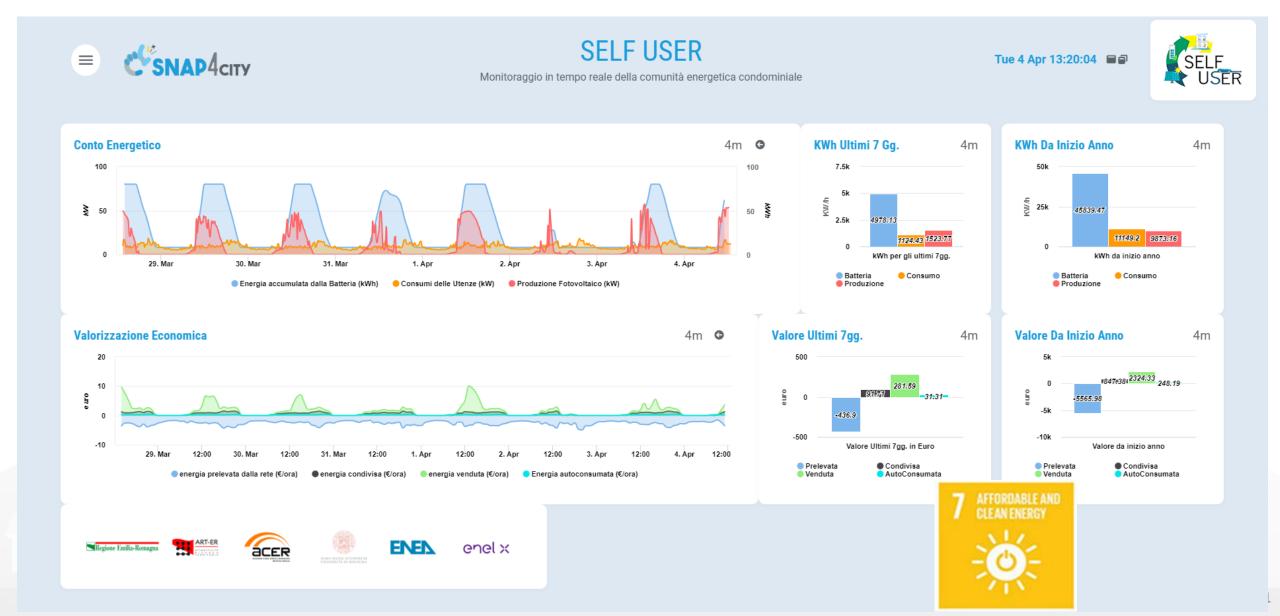
#### Field-tested energy community: the selfconsumer condominium

The Self User project creates in the pilot condominium, through the collection and analysis of data, a model for calculating and enhancing the impact of an energy community on a community of people, with a view to actions to combat energy poverty











-5k 2024

- no PV

🛕 - PV + battery 10kWh

2025

- with PV

PV + battery 15kWh

2026

2027

- PV + battery 2,4 kWh

2028

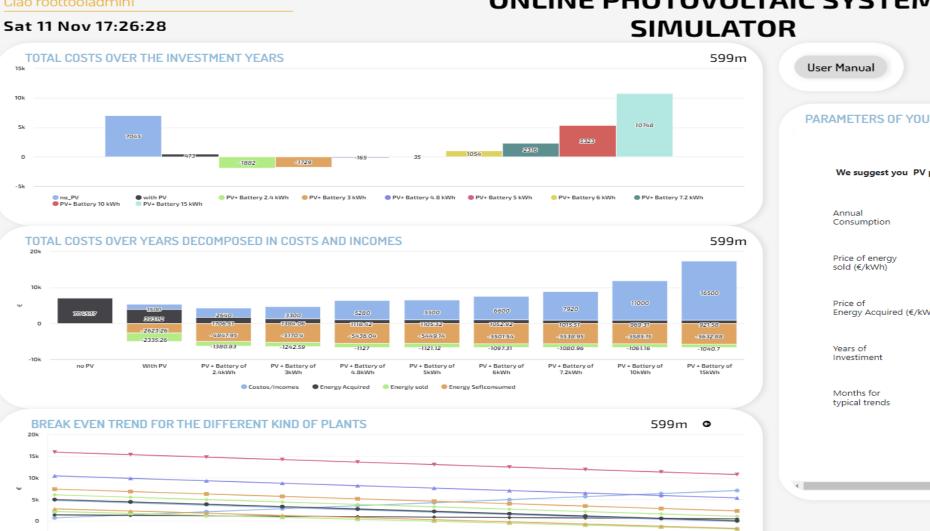
🔺 - PV + battery 3kWh

2029

PV + battery 4,8kWh



https://www.snap4city.org/dashboardSmartCity/view/Baloon.php?iddasboard=MzczNg==



2030

- PV + battery 5kWh

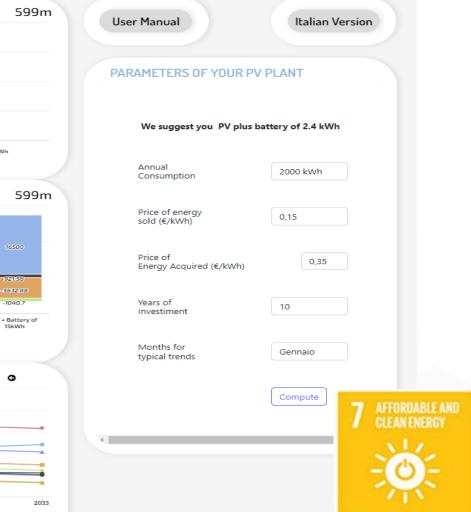
2031

- PV + battery 6kWł

2032

PV + battery 7,2kWh

**ONLINE PHOTOVOLTAIC SYSTEM** 



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



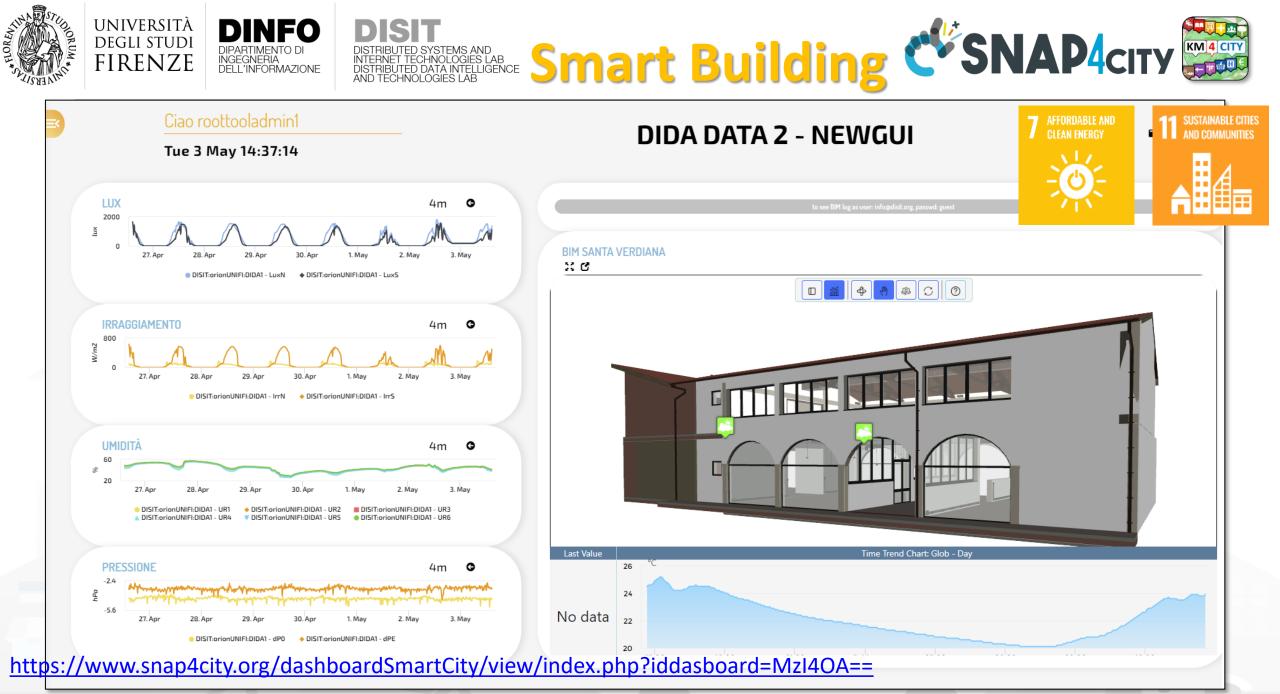






# **Smart Buildings, Snap4Building**

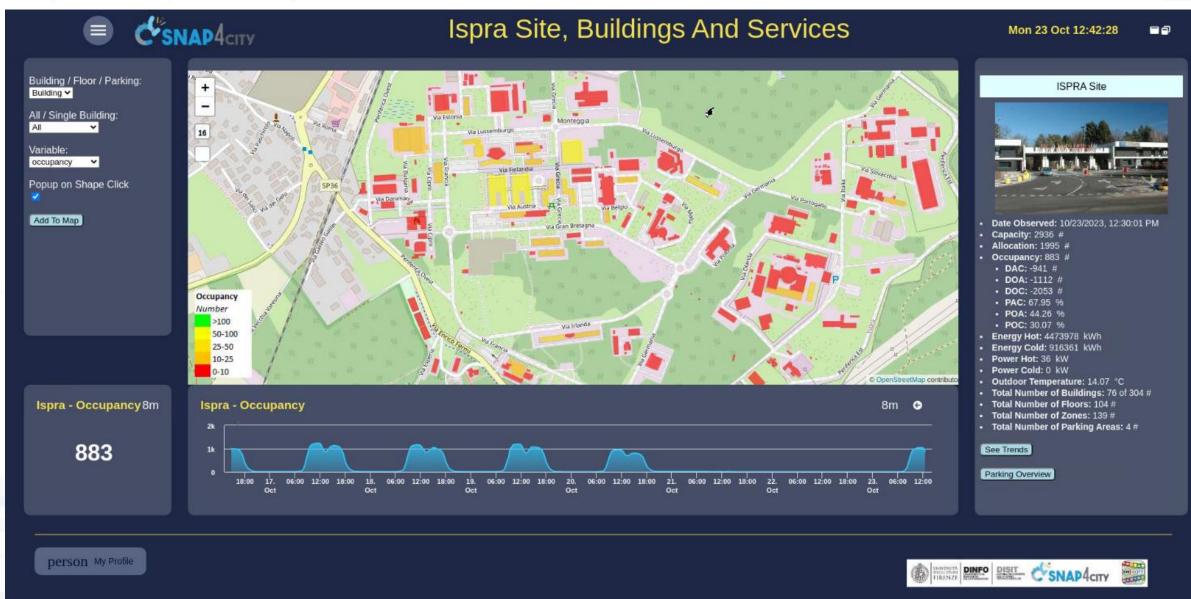
- Digital Twin for monitor, control and manage distributed infrastructures
  - 2D/3D representations of the whole set of buildings, BIM modeling
  - Entities (building, floors, rooms, parking, charging stations, gates, etc.) with their shapes and descriptors, and data monitoring the allocation to office, meeting, cafeteria, storage, stairs, elevator, etc.
- Monitoring and computing KPI on real time for
  - energy consumed or produced (hot/cold), parking, logistic, presences, cleaning, air quality, departments, subareas, maintenance, etc.
  - allocation/designation, dispositions, heating, cooling, temperature, equipment, etc.
  - grouped in Zones











#### 

STATES IL

#### **Building 27B Trends**





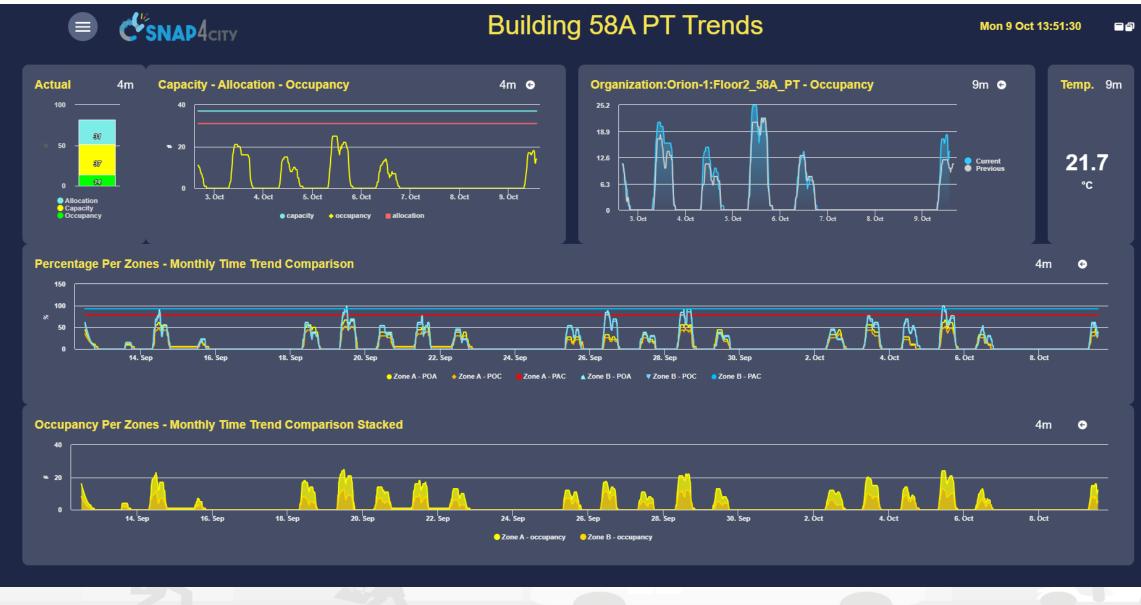


















### **Be smart in a SNAP!**



7-9 November 2023, Barcelona, Spain

**SMARTCITY EXPO WORLD CONGRESS** 

Visit Snap4City in Hall 1



#### CONTACT

TOP

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