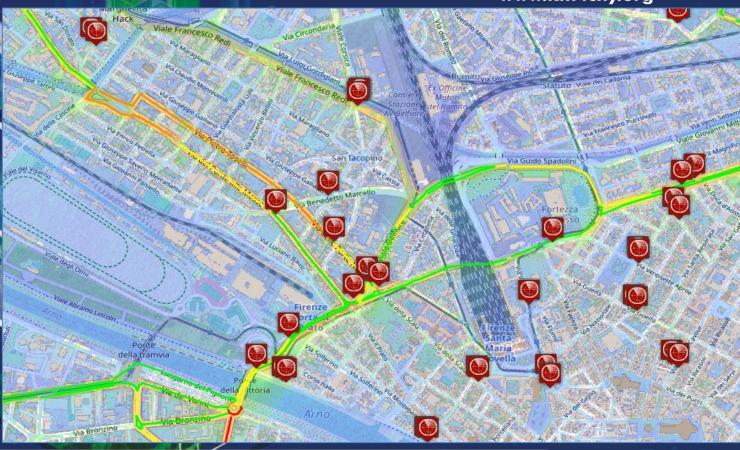


www.km4city.org

www.snap4city.org www.snap4solutions.org

Environmental overview

101010010



DIGITAL TWIN SOLUTIONS TO SETUP SUSTAINABLE DECISON SUPPORT SYSTEMS AND BUSINESS INTELLIGENCE







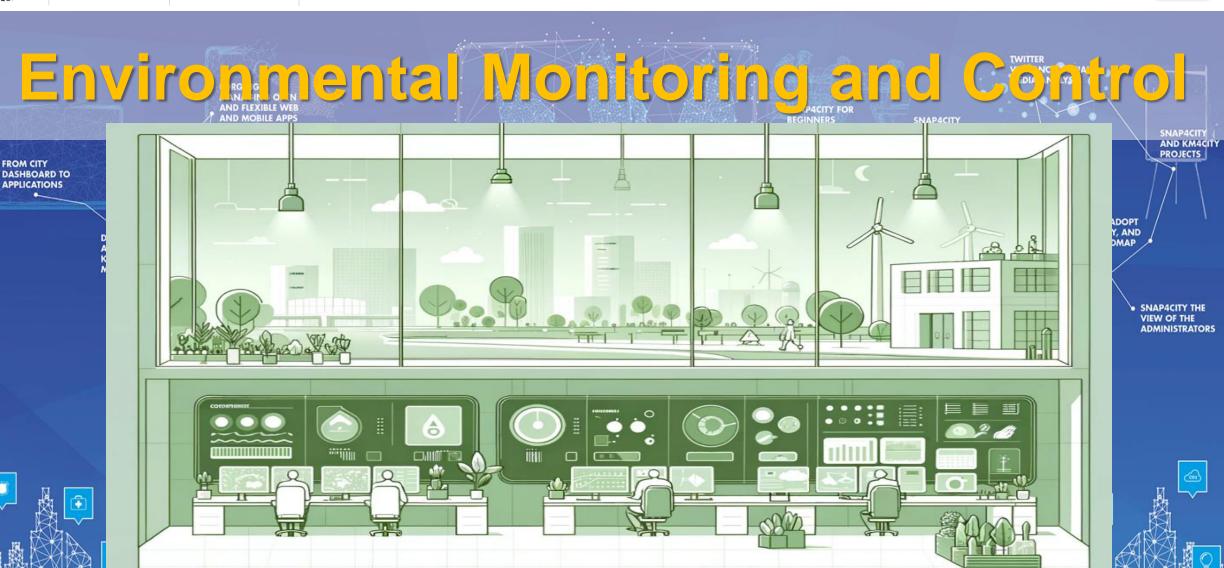






DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB DISTRIBUTED DATA INTELLIGENCE AND TECHNOLOGIES LAB





Key Performance Indicators, KPI

Housing

Culture





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);



- SUMI: mobility and transport vs env
 - https://www.snap4city.org/951
- SUMP/PUMS: mobility and transport vs env.
- **ISO indicators:** city smartness, digitization, tech level.
- Low Level/Real Time: global traffic, quality of service, betweenness, centrality, queue, time to travel, etc.

		Air Qu	WHOguidelines		
Pollutant	Averaging period	Objective and legal natur concentration	e and Comments	Concentration	Comments
PM _{2.5}	One day			25 μg/m³ (*)	99 th percentile (3 days/year)
PM _{zs}	Calendar year	Target value, 25 µg/m³	The target value has become a limit value since 1 January 2015	10 μg/m³	
PM ₁₀	One day	Limit value, 50 µg/m³	Not to be exceeded on more than 35 days per year.	50 μg/m³ (*)	99th percentile (3 days/year)
PM ₁₀	Calendar year	Limit value, 40 μg/m³ (*	r)	20 μg/m³	
0,	Maximum daily 8–hour mean	Target value, 120 μg/m³	Not to be exceeded on more than 25 days per year, averaged over three years	100 µg/m³	
NO _z	One hour	Limit value, 200 μg/m³ (*) Not to be exceeded more than 18 times a calendar year	200 μg/m³ (*)	
NO ₂	Calendar year	Limit value, 40 μg/m³		40 μg/m³	

Mobility

Food

Services

Economy

Environment

15Min



Realtime



DEGLI STUDI FIRENZE











10/22









15 Minute City Index:

13 subindexes: energy, slow mobility, fast mobility, housing, economy education, culture and cults, health, entertainment, gov, food, security...



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



- Smart City infrastructure: monitoring and resilience, long terms predictions
- Effective and Low cost smart solutions
- What-if analysis, Simulations
- Origin Destination matrices computation



Monitoring and Predicting: NO2, NOX, CO2, Traffic flow, pollutant, landslide, waste, etc. Traffic flow reconstruction

Demand vs Offer of Mobility analysis



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



- business intelligence tools for decision makers
- Reduction production costs
- Monitoring resource consumption
- **Optimization of Waste Collection**

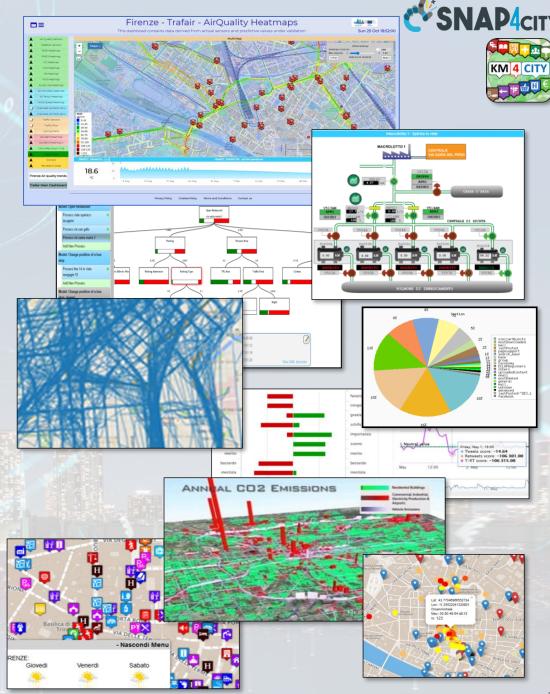


- Shortening justice time
- Anonymization and indexing legal docs.
- Prediction of mediation proneness
- Ethical Explainable Artificial Intelligence

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













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

15MinCityIndex

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

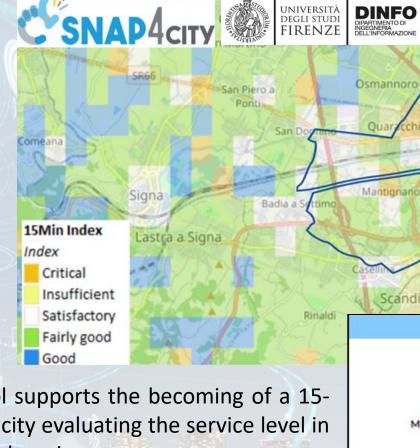
Using the Open Data:

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.

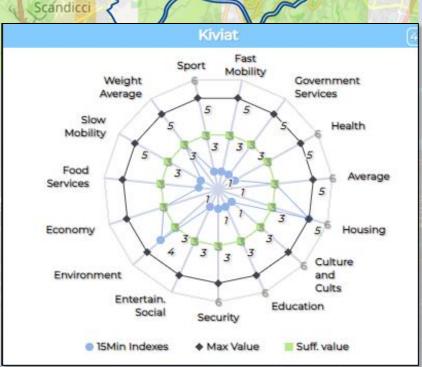


Careggi/Rifredi is a relevant district in Florence because of hosting the main Florence/Tuscany hospitals Careggi and Meyer, but also university headquarters and many other workplaces.



The tool supports the becoming of a 15-Minute city evaluating the service level in various domains.





DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB

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

Snap4City (C), June 2024



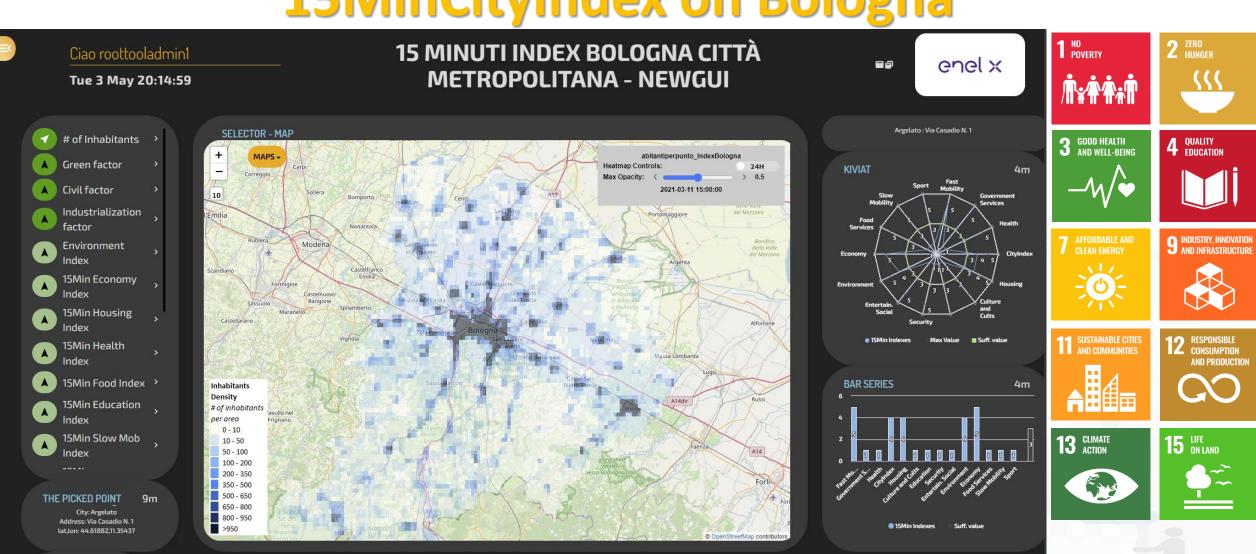








15MinCityIndex on Bologna





Powered by **SET STATE**

> **FREE** TRIAL

















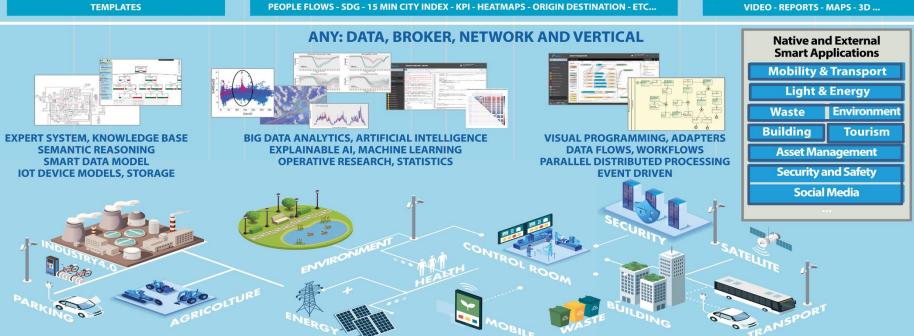




Smart Solutions and Decision Support Systems











METHODOLOGIES LIVING LABS COURSES AND COMMUNITY DEVELOPMENT TOOLS



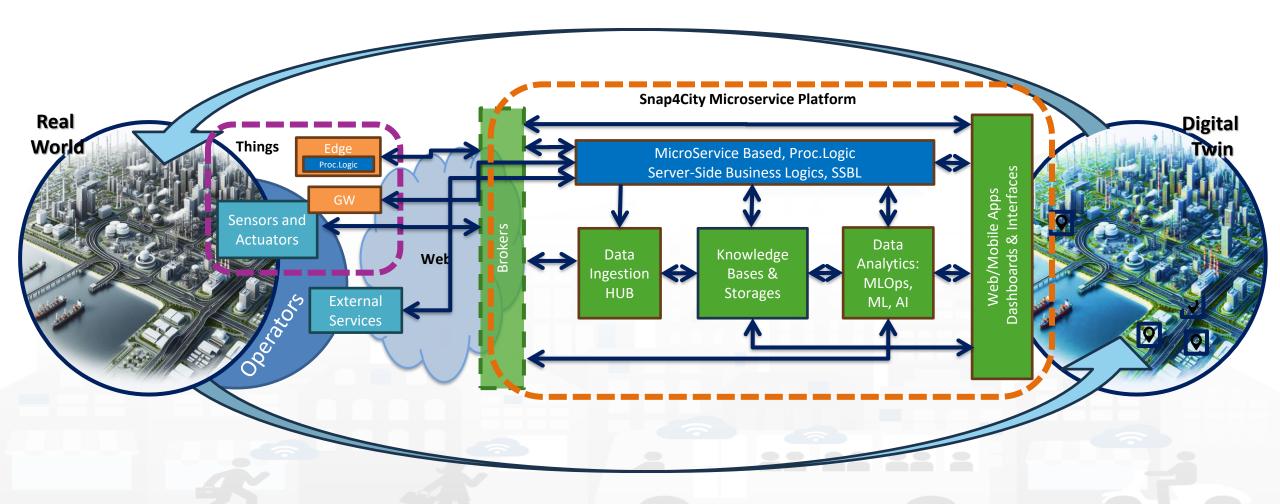








Digital Twin Development Platform



Standards and Interoperability (6/2023)

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, 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.



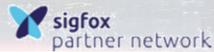






















https://www.snap4city.org/65







High Level Types

Snap4City (C), June 2024

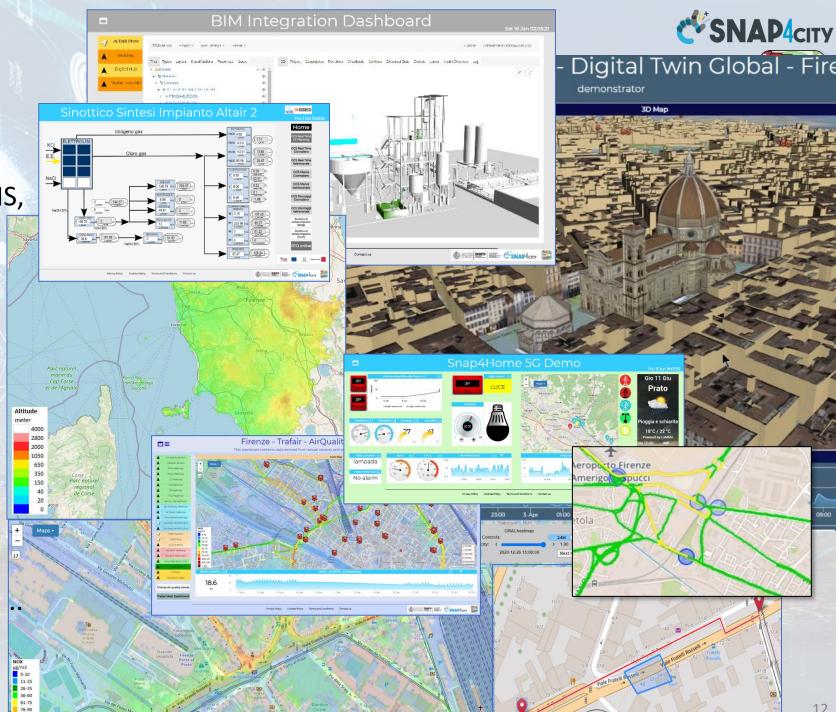
- POI, IOT Devices, shapes,...
 - FIWARE Smart Data Models,
 - IoT Device Models
- GIS, maps, orthomaps, WFS/WMS, GeoTiff, calibrated heatmaps, ...
- Satellite data, ..
- traffic flow, typical trends, ...
- trajectories, events, Workflow, ...
- 3D Models, BIM, Digital Twins, ...
- OD Matrices of several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

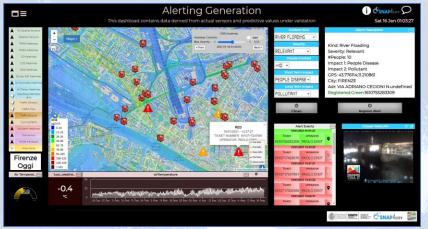


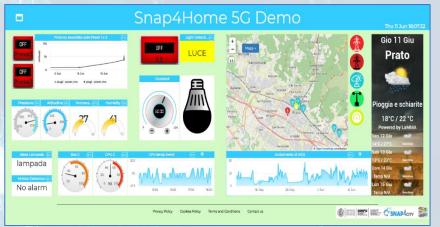


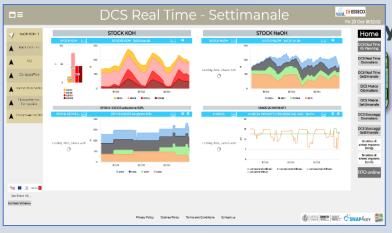


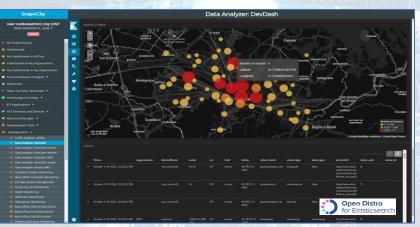




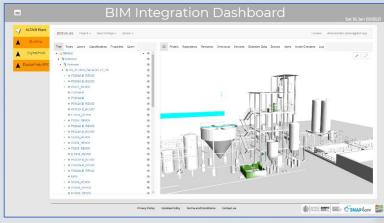




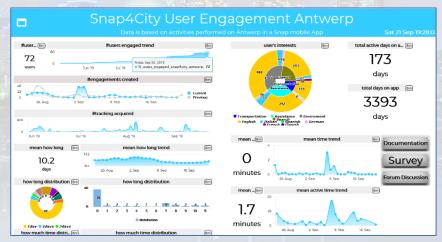


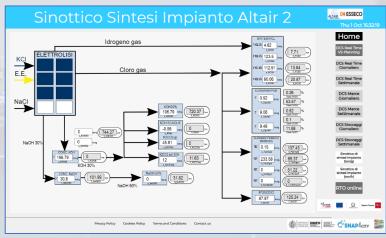


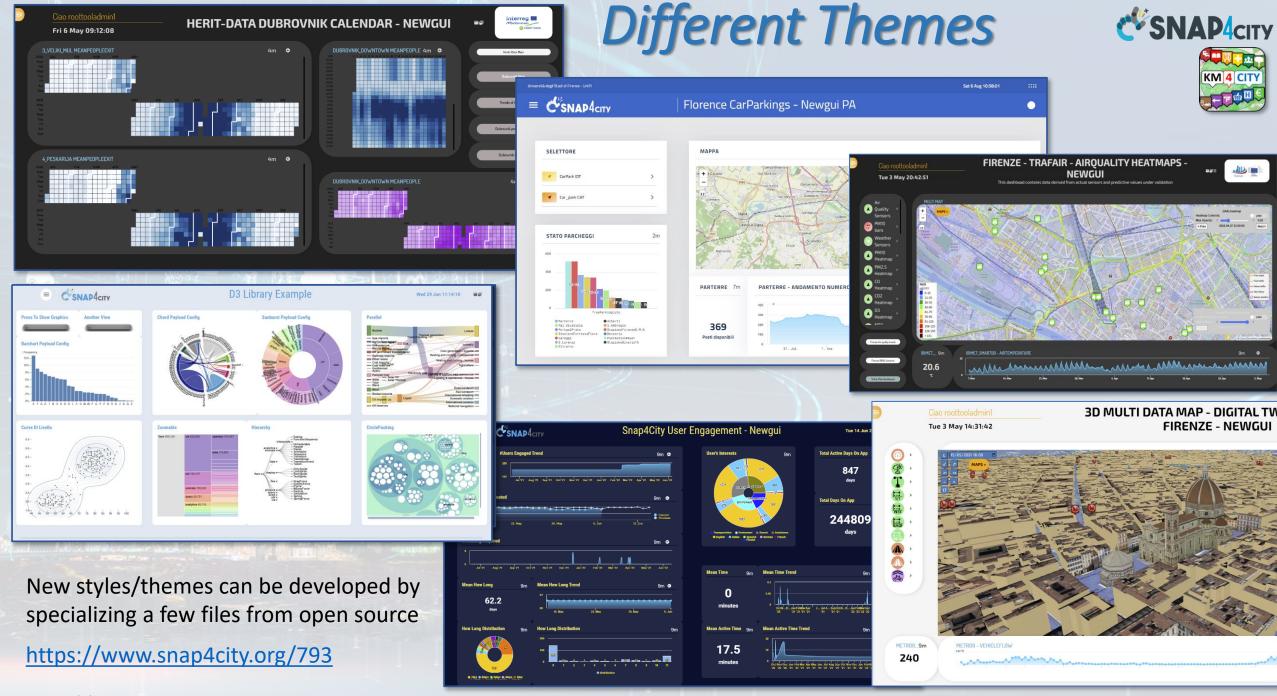












Snap4City (C), June 2024



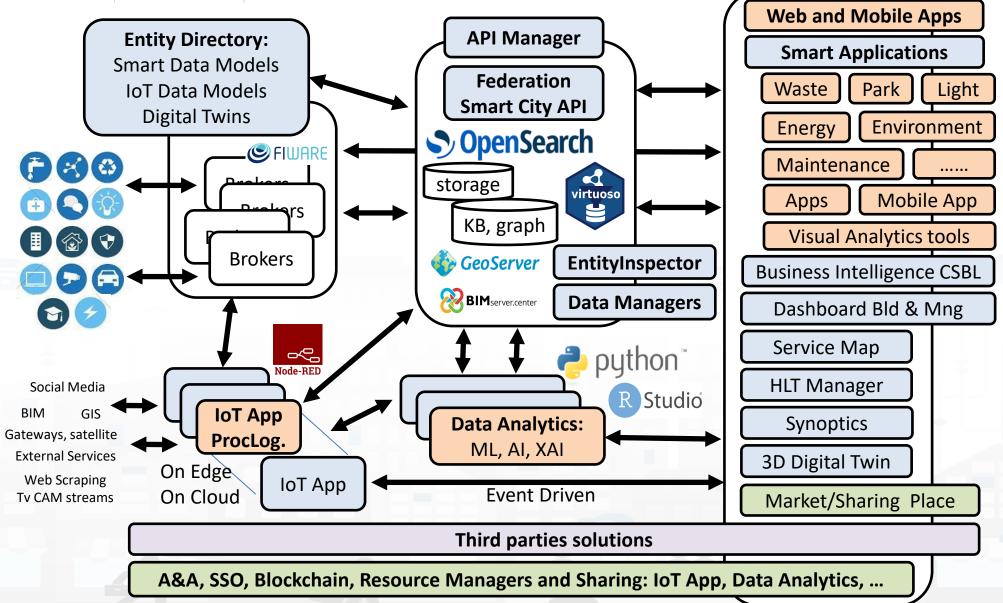


DISTRIBUTED SYSTEMS AND INTERNET TECHNOLOGIES LAB DISTRIBUTED DATA INTELLIGENCE AND TECHNOLOGIES LAB









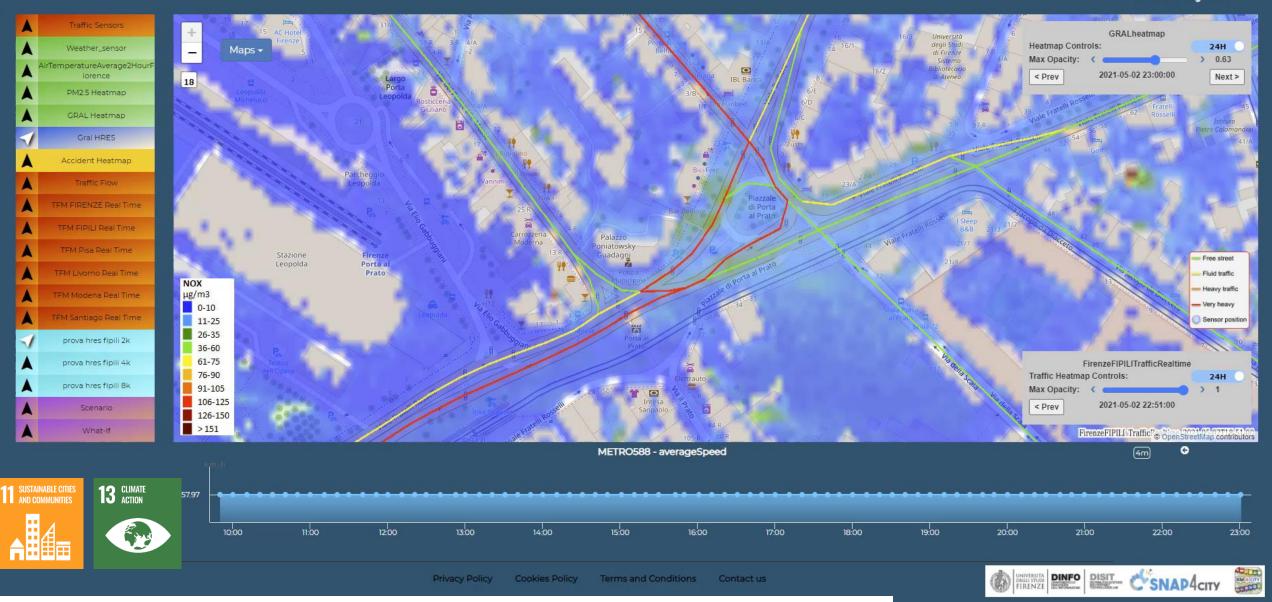
SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES





Traffic Flow Manager on multiple cities

Sun 2 May 23:16:31



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

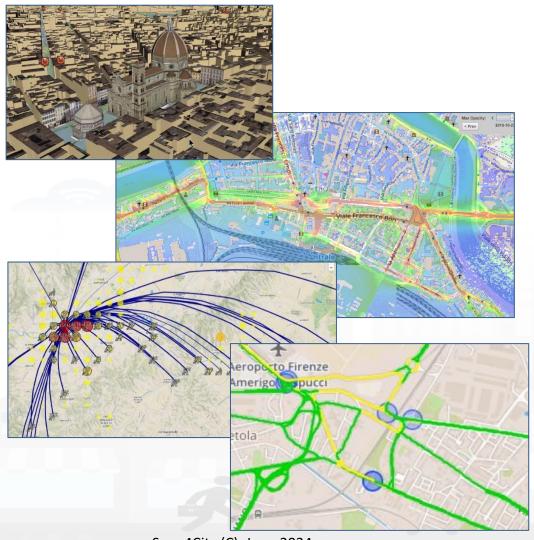








Smart City Digital Twin City Digital Model with...



- Intuitive platform
- Any Data TYPE, any data source, any protocol
- Data storage seamless
- Data analytics \rightarrow artificial intelligence, AI/XAI
- Data Ethics, Al 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%

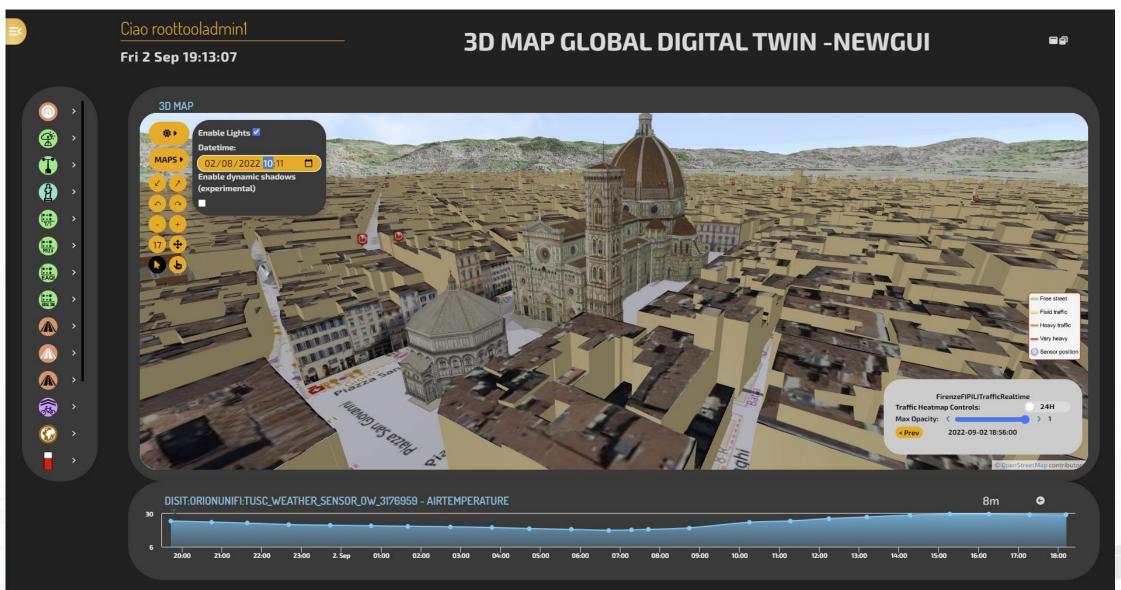
Complex and heterogeneous information, interoperability

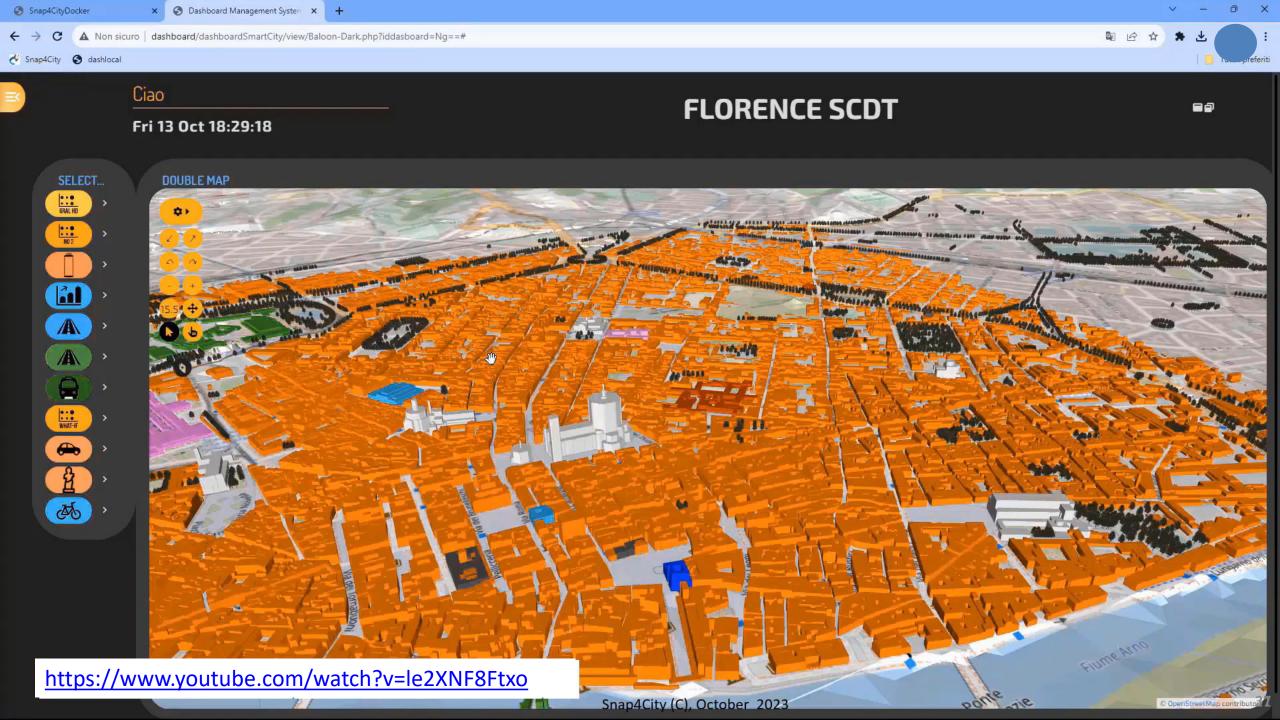
- o GIS, ITS, AVM, IoT, BIM, CKAN, etc.
- Satellite services
- o MaaS, last-mile delivery HUBs
- etc.











SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES





Available AI Solutions on Snap4City

SNAP4city

KM 4 city

- 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

Snap4City (C), June 2024

Snap4City What-If

- Decision support systems
- Improvement of life quality

Data Analytics

System Modeling

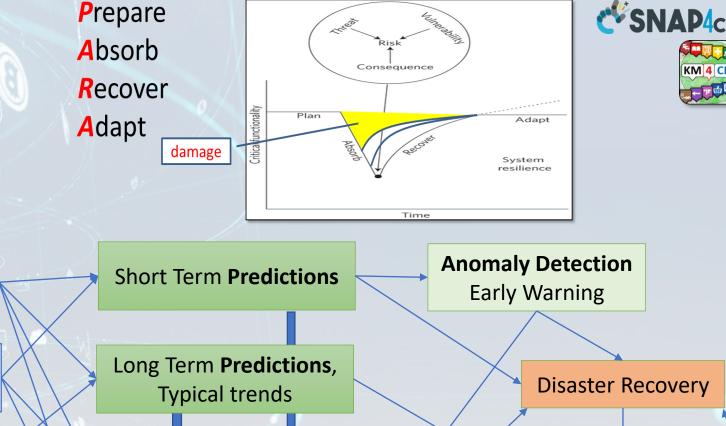
and Simulation

Knowledge

Models

Scenarious

- Sustainable Solutions
- Reduction of costs
- Risk Assessment
- Resilience



Strategies and

Plans

Partial graph

Decision Support System: neuro-symbolic reasoning targeting Indicators: Quality of Life, PUMS, SUMI, KPI, SDG, 15MinIndex,...

Recommendations &

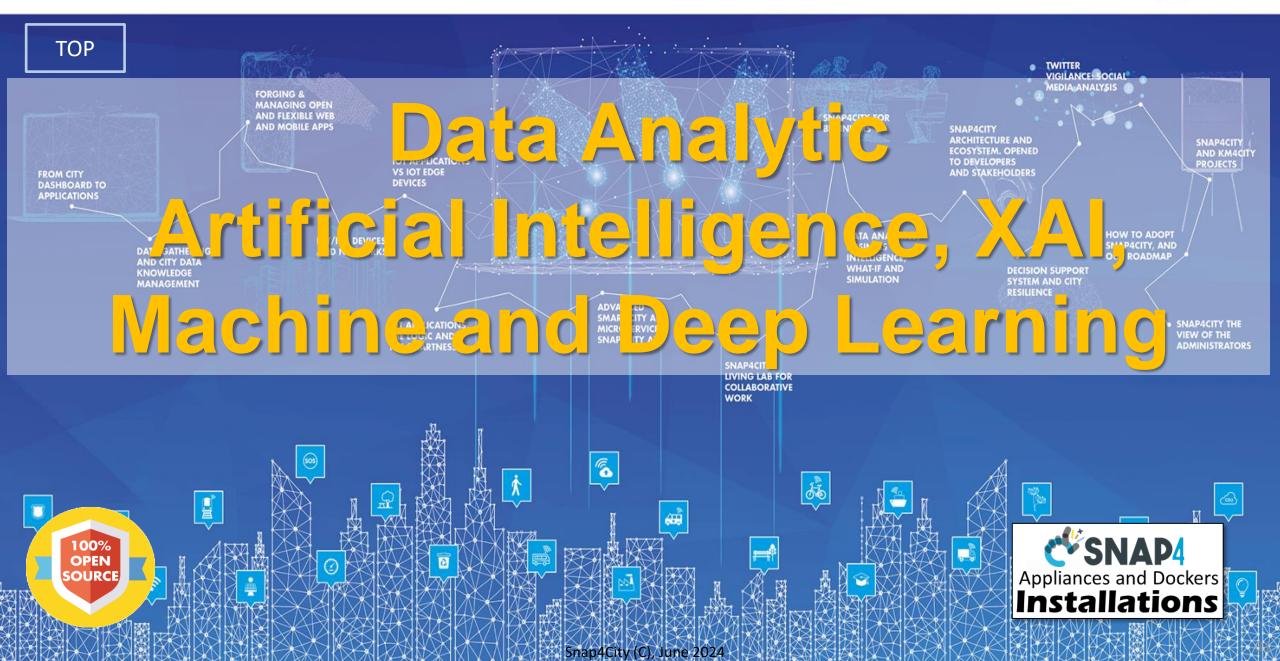
Prescriptions

What-if Analysis

Snap4City (C), June 2024

SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES









Environment, waste, land, etc., Domain (2024)

- Goals:
 - Reduction of pollutant emissions and EC taxations
 - Cost Reduction for waste collection, reduction of waste collection impact on mobility
- Solutions for Operation (monitoring, managing, mobile apps, digital signages, control rooms)
 - Monitoring emissions, weather, waste, water, etc.: sensors, traffic, flows,
 - Early detection/warning of critical conditions on emissions, weather, waste, water, fire, animals, ...
 - Early detection/warning of critical conditions for landslides, water flooding, beach
 - Managing Smart Waste: bins/lockers, waste collection daily plan, pay as you throw, PAYT, etc.
 - Short terms prediction of emissions: CO2, NO2, etc.
 - Production of suggestions, nudging
 - Computing and predicting long terms KPI indicators of the European Commission
- Solutions for Planning (optimization and what-if analysis)
 - Identification of main CO2/NO2 emission locations in the city, total production from traffic
 - Reduction of Pollutant Emissions, via optimization: semaphore cycles, viability
- Algorithms and computational solutions, see next slide





Tools: Environment and Weather (2024)

- 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.



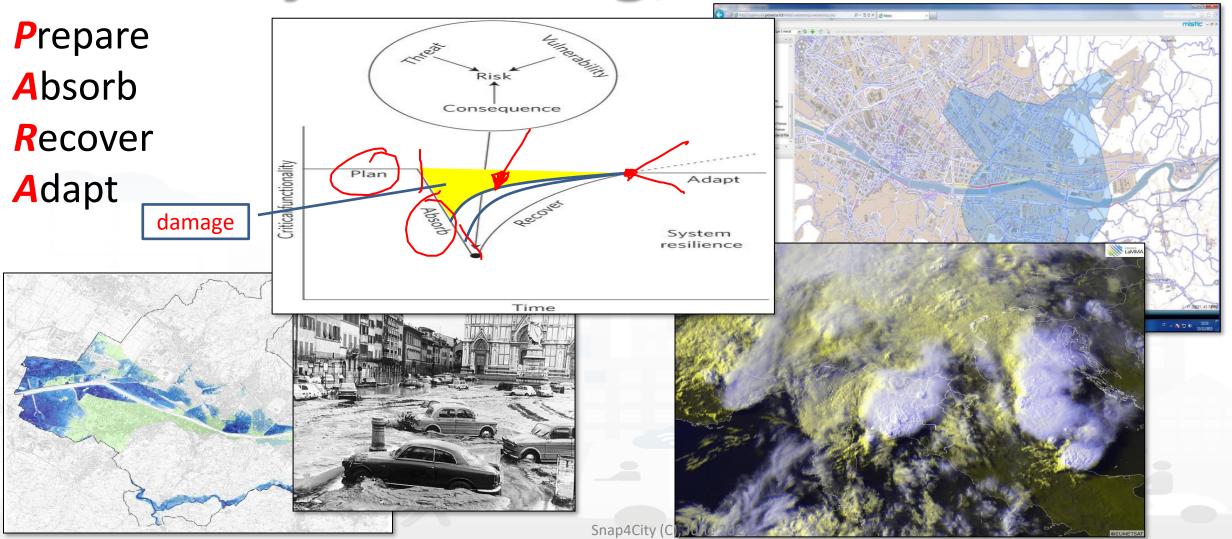








Early Warning, Detection



Environment and Quality of Life

Air Quality Predictions

Multiple Domain Data

- Traffic Flow data, Pollutant: NOX, CO2, PM10, PM2.5, O3,
- 3D City structure, weather, ...

Multiple Decision Makers

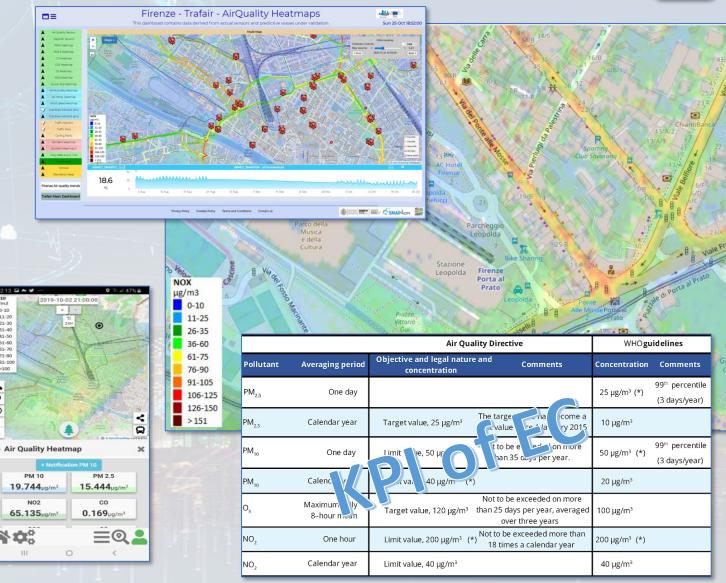
- Pollutant Predictions: NOX, NO2, ...
- City officers, energy industries
- Dashboards, What-IF analysis
- Traffic Flow Reconstruction

Historical and Real Time data

- Billions of Data
- Services Exploited on:
 - Dashboards, Mobile App
- Since 2020





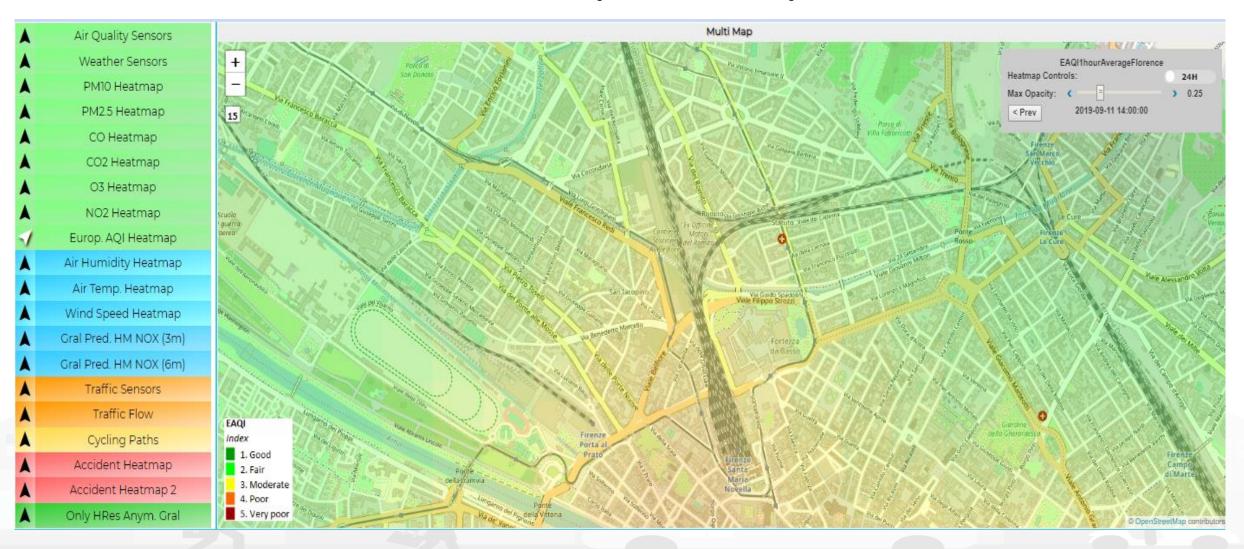








EAQI Heatmap and sequence

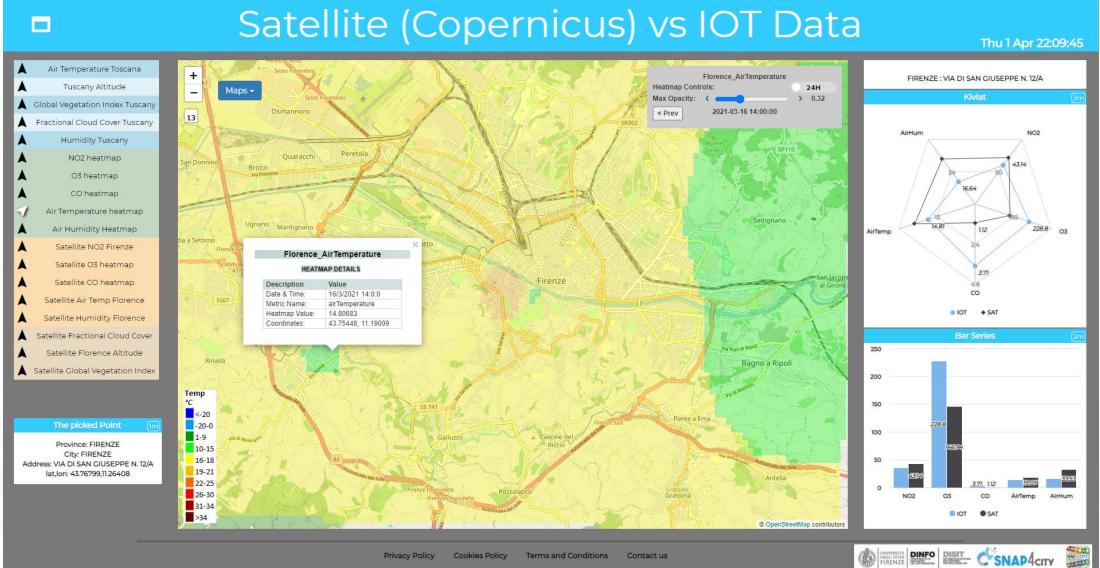


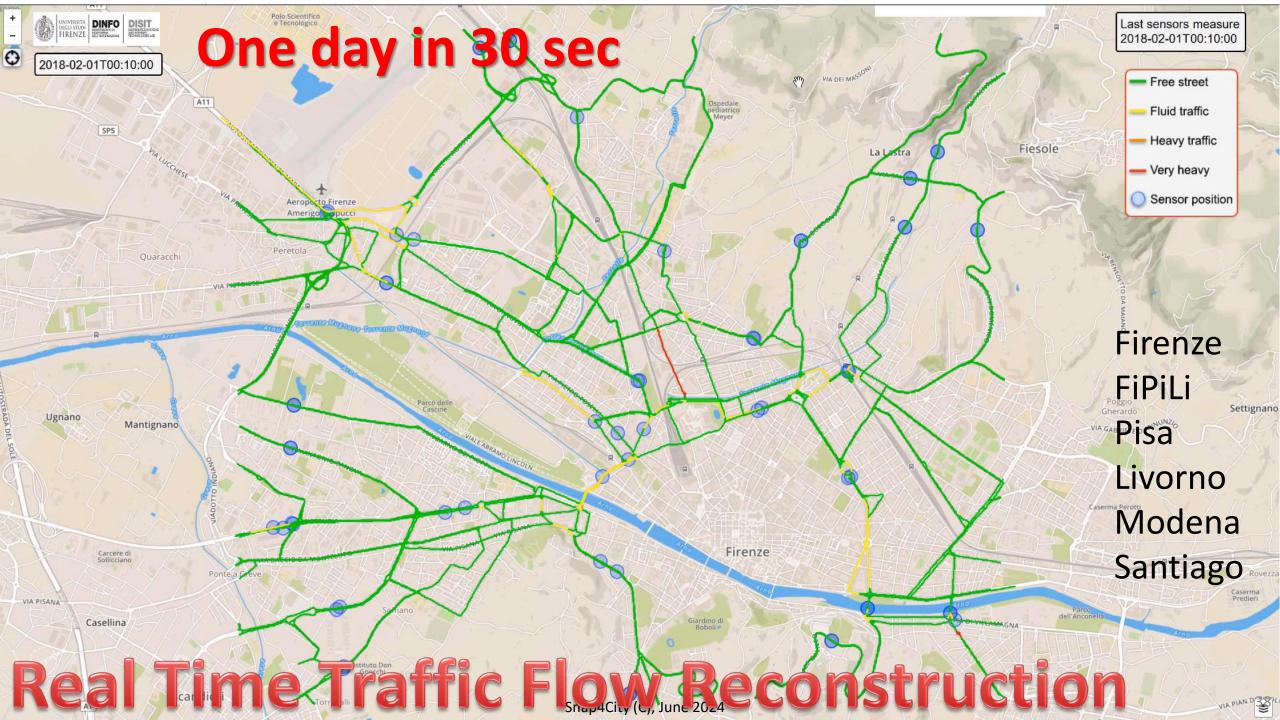














TOP

1-48 Hour prediction of NOx













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 Al



Impact of COVID-19

Multiple Domains Data

 Traffic, environment, People, parking, stock options, Twitter, tc.

Decision Makers Multiple Locations

- NO2 long term predictions
- Twitter analysis

Historical and Real Time data

Services Exploited on:

- Dashboards
- Social media,
- Sentiment Analysis
- Since 2019, 2020

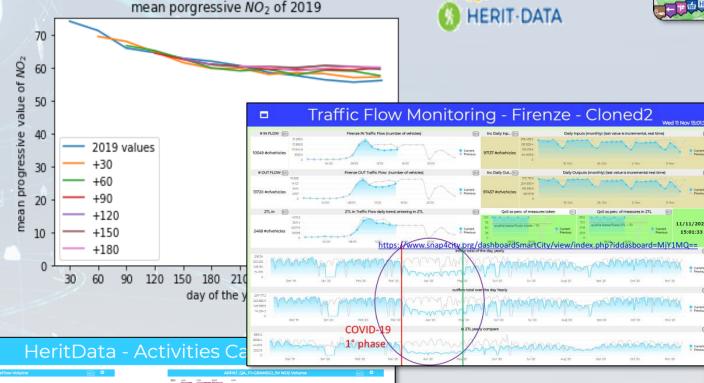
Cities: Firenze, Pisa, Livorno, Toscana











	HeritData	- Activities C	6 Cer 79 3er 20 Fee 27 Mar
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metric	model30	model60	model90	model120	model150	model180			
MAE	1.21	1.31	1.52	2.04	2.31	2.37			
RMSE	2.16	2.61	4.18	6.77	7.83	7.93			
MAPE	1.99	2.20	2.65	3.57	4.07	4.18			
R2	0.91	0.83	0.80	0.54	0.45	0.14			
Table 4. Assessment of the predictive models with respect to the									
actual v	alues of t	he 2019.							

Snap4City (C), June 2024









TOP

Computing CO2 from traffic Data



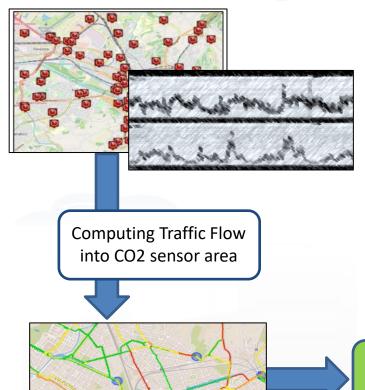








Estimating City Local CO2 from Traffic Flow Data



Traffic Flow is one the main source of CO2 (ton of CO2 x Km x Vehicle)





K1: Fluid Flow

K2: Stop and Go

Dense estimation of CO2 into the city is very useful to know to target EC's KPIs

Computing CO2 on the basis of traffic flow data



CO₂ estimation

Traffic Flow data

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/









TOP

Long Term Prediction of Annual Mean of NO2 index of EC







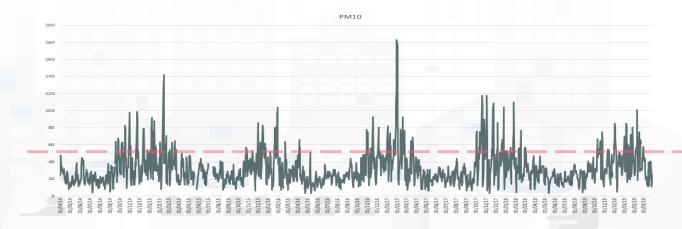




Predicting Air Quality

- European Air Quality Directive
- Predicting critical days
 - PM10 with an accuracy of more than 90% and precision of 85%;
 - PM2.5 with an accuracy of 90% and precision greater than the 95%.
- Simulating Long terms values
 - For long terms predictions

		Air Quality Directive		WHOguidelines	
Pollutant	Averaging period	Objective and legal nature concentration	and Comments	Concentration	Comments
PM _{2.5}	One day			25 μg/m³ (*)	99 th percentile (3 days/year)
PM _{2.5}	Calendar year	Target value, 25 μg/m³	The target value has become a limit value since 1 January 2015	10 μg/m³	
PM ₁₀	One day	Limit value, 50 μg/m³	Not to be exceeded on more than 35 days per year.	50 μg/m³ (*)	99 th percentile (3 days/year)
PM ₁₀	Calendar year	Limit value, 40 μg/m³ (*)		20 μg/m³	
O ₃	Maximum daily 8–hour mean	Target value, 120 µg/m³	Not to be exceeded on more than 25 days per year, averaged over three years	100 μg/m³	
NO ₂	One hour	Limit value, 200 µg/m³ (*	Not to be exceeded more than 18 times a calendar year	200 μg/m³ (*)	
NO ₂	Calendar year	Limit value, 40 μg/m³		40 μg/m³	







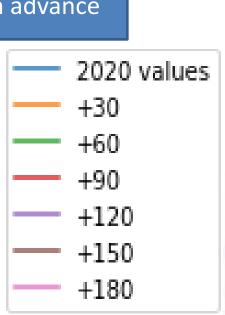




Predicting EC's KPI on NO2 months in advance

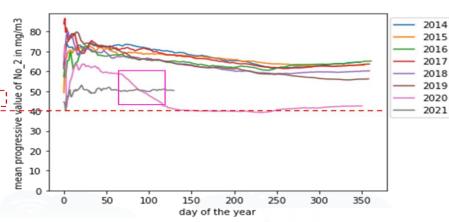
Deep Learning Long Terms Predictions of NO2 mean values, From 30 to 180 days in advance

- The features used as input for the predictive models are:
- Month
- dayOfTheYear
- NO2
- Tmean
- Humidity
- windMean ^{نځت}
- NoxDomestic
- numberOfVehicles
- NO2cumulated
- NO2progresseveMean
- numberOfVehiclesCumulated









		Air Qual	WHOguidelines		
Pollutant	Averaging period	Objective and legal nature concentration	and Comments	Concentration	Comments
PM _{2.5}	One day			25 μg/m³ (*)	99 th percentile (3 days/year)
PM _{2.5}	Calendar year	Target value, 25 µg/m³	The target value has become a limit value since 1 January 2015	10 μg/m³	
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NO ₂	Calendar year	Limit value, 40 μg/m³		40 μg/m³	

Predicting Land sliding









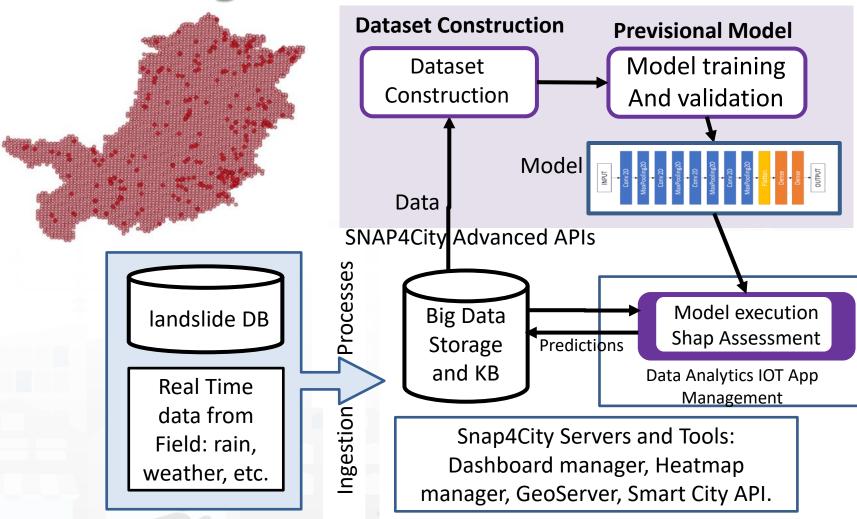


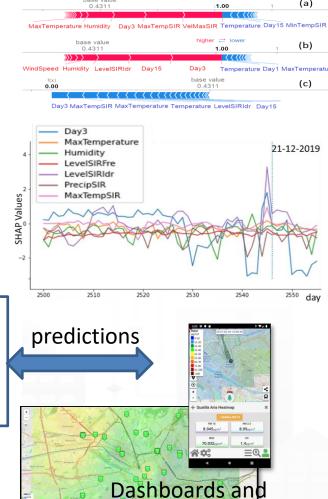






Predicting Land slides





Mobile Apps

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.





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
 Figure 10a, the value of
 VelMaxSIR, MaxTempSIR, Day3
 and Humidity contributed
 significantly to the classification of
 the observation as a landslide
 event

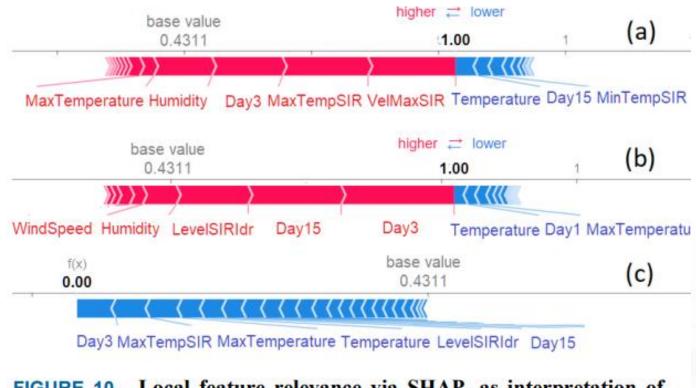


FIGURE 10. Local feature relevance via SHAP, as interpretation of events in terms of feature values: (a) and (b) are events with predictions of landslide, (c) a no landslide event.











Vaste value Outimisation







Smart Waste – Map view



- Reduction of costs for waste collection
 - Optimization of waste collection for the next day, forecast
 - Production of rides and paths for the drivers on waste collection
- Operator:
 - · Refine a search by using the filters on the left side
 - Click on a waste bin pin on the map:
 - A popup with real time data is shown
 - The fullness status of the selected group of bins is shown in the synoptic below the map
 - Specific fullness weekly trends are shown below the map
 - Chick on the «Table view» button to access the other dashboard





Search bins on map by filtering per:

- Kind (All, generic, plastic, paper, glass, metal, organic)
- Status (Active, Not Active)
- Fullness (Full, Half-full, Empty)
- Address
- Group of bins (by GroupID)





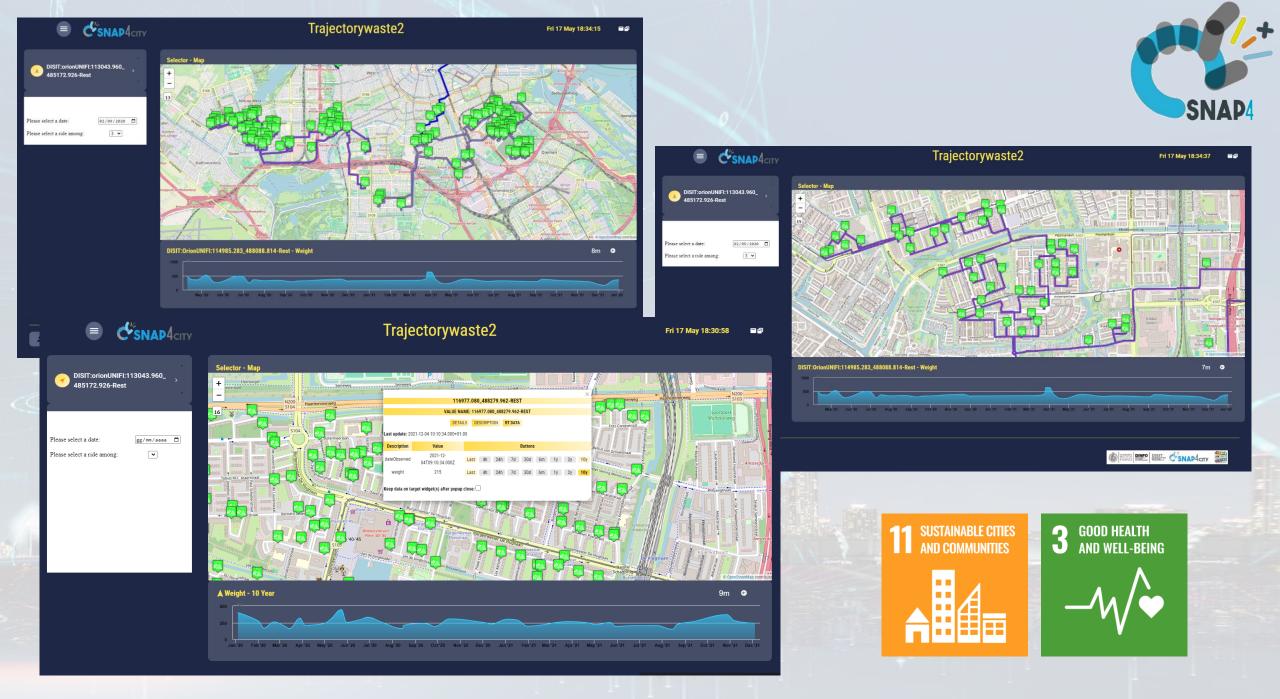












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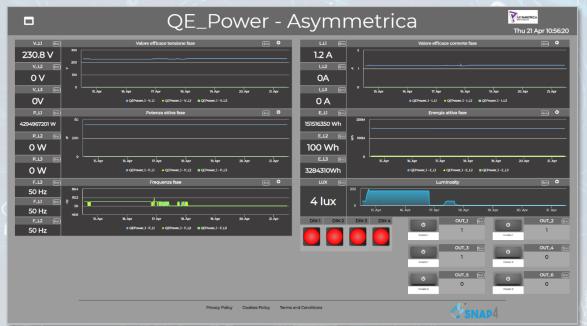




ГОР

others









- Power meter Data
- Smart Light data are coming (in collaboration with a multinational company)









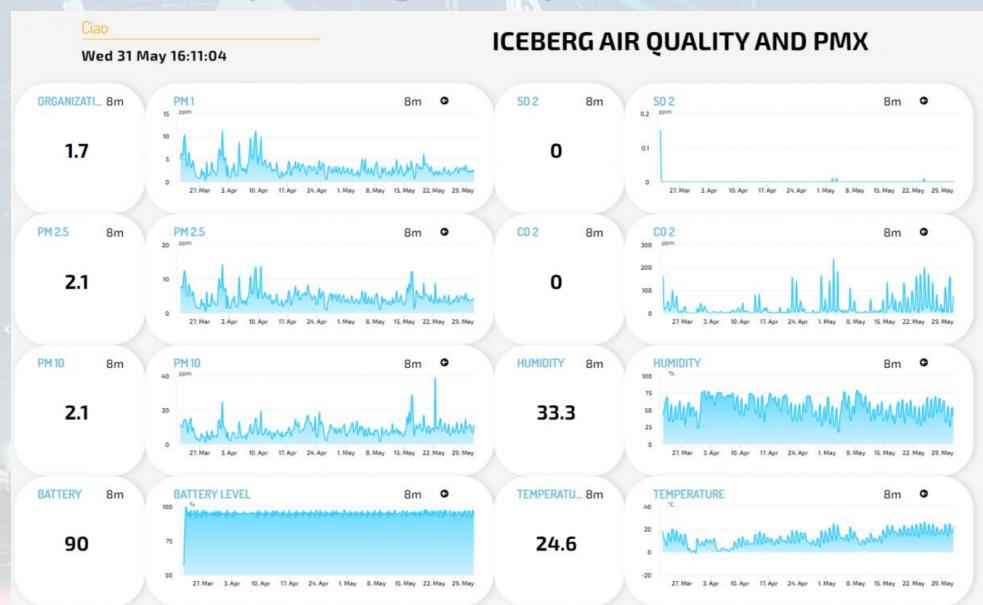




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TheLab.City LivingLab by ICEBERG, Romania





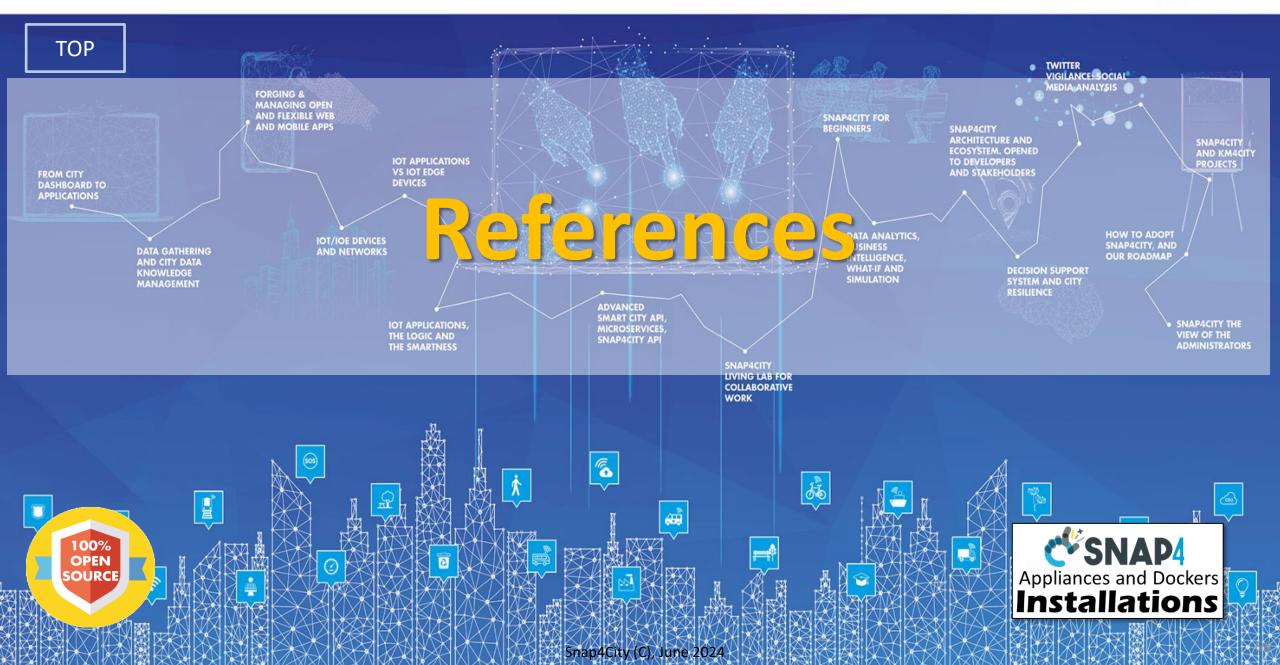
- Airquality
- Urban planning
- Parking
- Waste
- Etc.

https://thelab.city/

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SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES





2023 booklets

Smart City





https://www.snap4city.org /download/video/DPL SN AP4CITY.pdf Industry





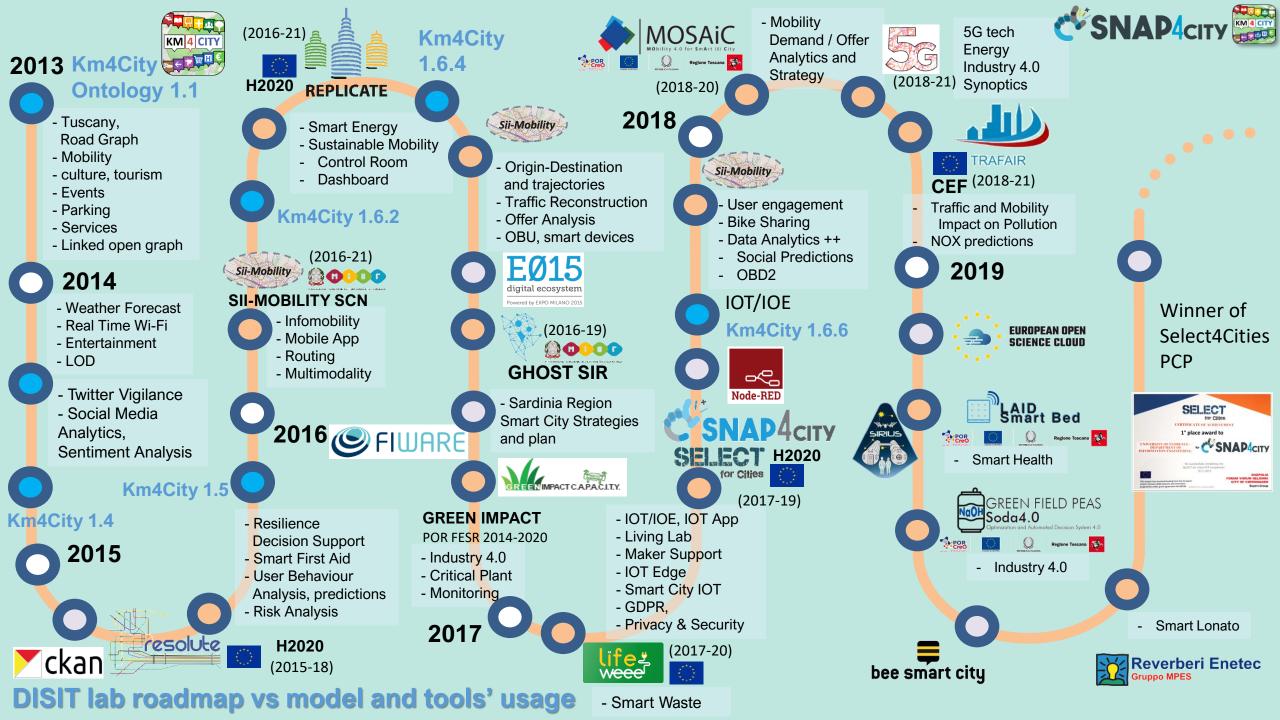
https://www.snap4city.org/download/video/DPL SNAP4INDUSTRY.pdf

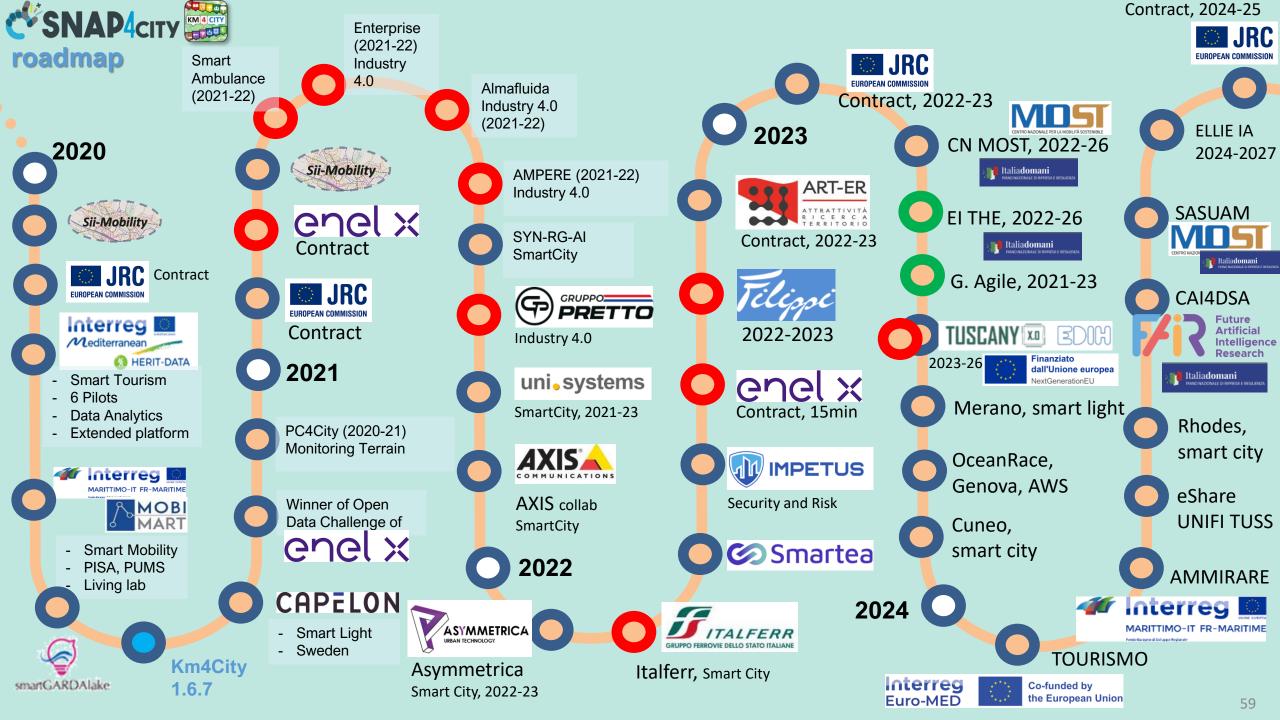
Artificial Intelligence





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





TOP







Be smart in a SNAP!





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

www.snap4city.org



Email: snap4city@disit.org

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

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