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Alcune nuove tecniche dell'intelligenza artificiale possono risolvere i problemi della mobilità e dei trasporti e quali dati sono necessari

new artificial intelligence techniques can solve mobility and transportation problems and what data is needed

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Why mobility is an AI problem now

Mobility systems are networks with feedback loops: travel demand, road conditions, signals, parking, emissions, public transport, and shared mobility all change together.

Recent AI solutions are not based on single model!

A decision pipeline is needed combining sensing, reconstruction, prediction, simulation, optimization, and explainability.

Congestion
waiting time
stops

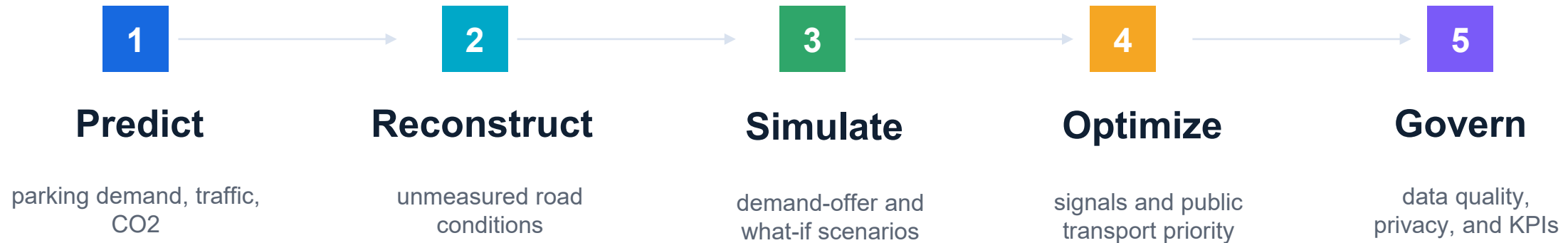
Parking
search traffic
emissions

Public transport
reliability
accessibility

AI turns sparse, messy data into three capabilities:

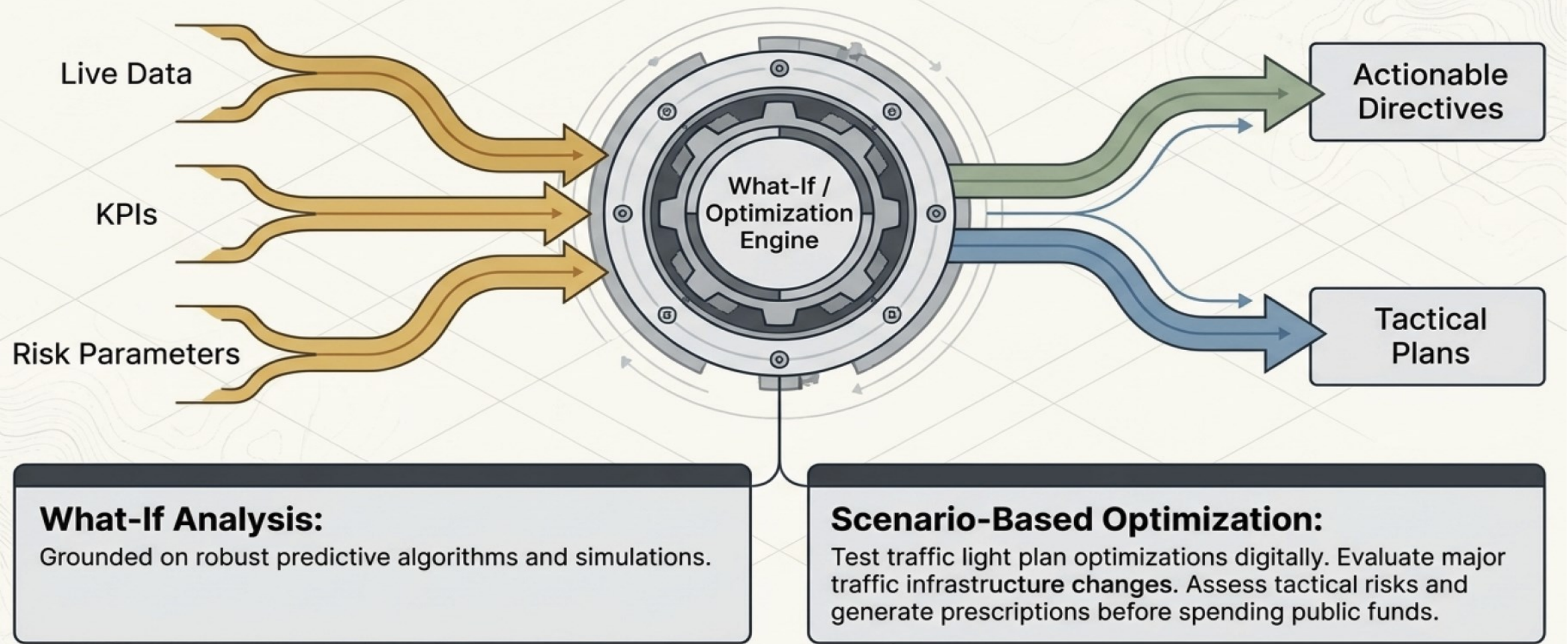
- 1. See the whole network**
- 2. Predict what happens next**
- 3. Optimize services under constraints**

Why mobility is an AI problem now



I am going to provide a concrete data checklist and implementation roadmap for a city or mobility operator.

Decision Support Systems





THE POWER OF ARTIFICIAL INTELLIGENCE AT THE SERVICE OF YOUR OPERATION AND PLAN

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FIWARE

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TRIAL**



**PEN Test
Passed**



**EU GDPR
COMPLIANT**

SNAP4
Appliances and Dockers
Installations

**EUROPEAN OPEN
SCIENCE CLOUD**



Node-RED

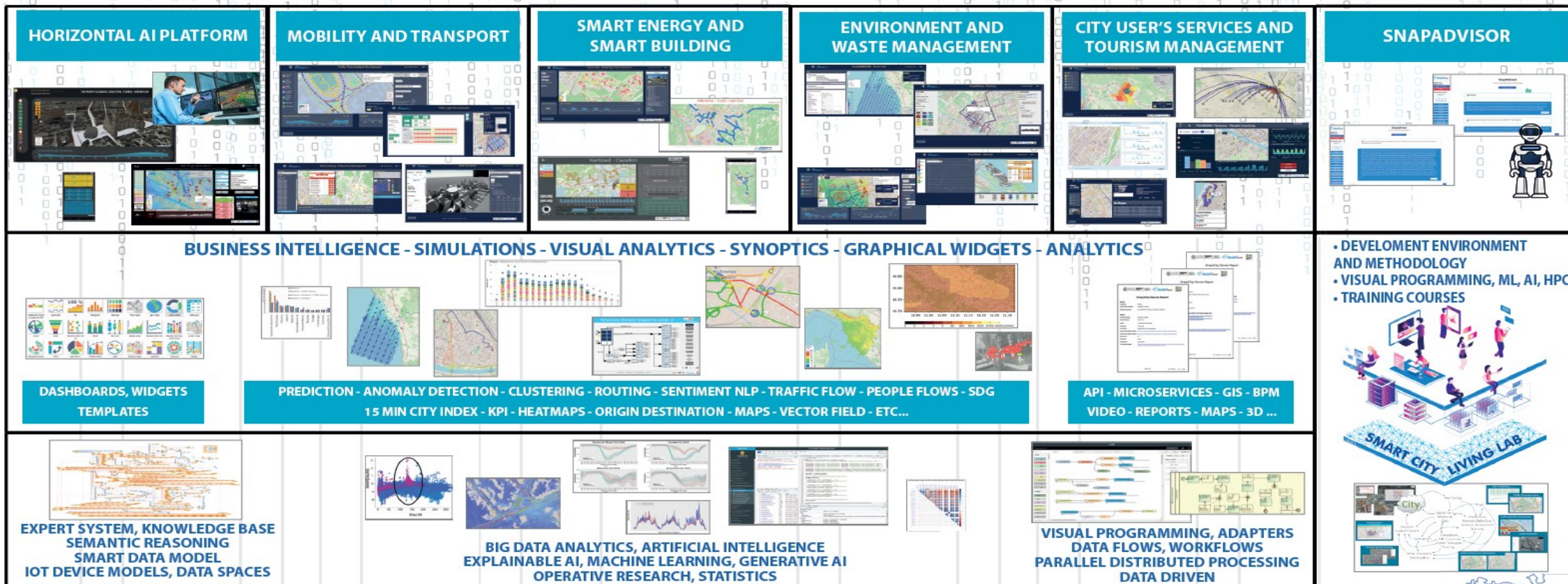
JS Foundation

E015
digital ecosystem



NVIDIA

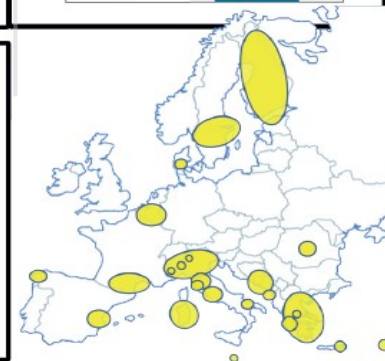
OPERATION AND PLAN - CONTROL ROOMS - DECISION SUPPORT SYSTEMS - WHAT-IF ANALYSIS - OPTIMIZATION - APPLICATIONS



FULL INTEROPERABILITY, ANY: DATA, BROKERS, NETWORKS AND VERTICALS

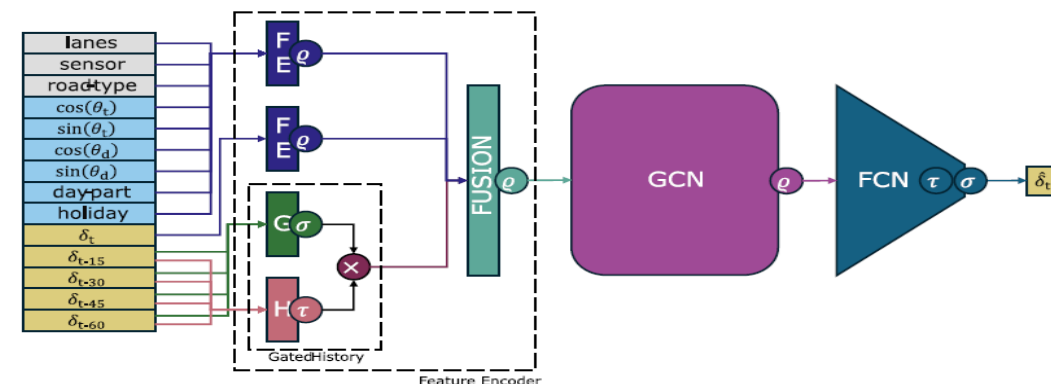
NATIVE AND EXTERNAL APPLICATIONS

Smart Parking
Smart Light
Smart Waste
Smart Energy
Smart Building
Smart Tourism
...

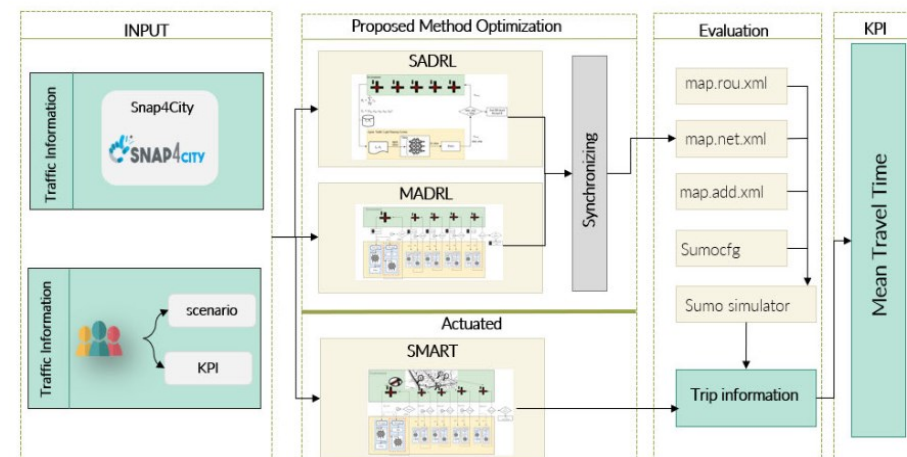


Technique families and the problems they solve

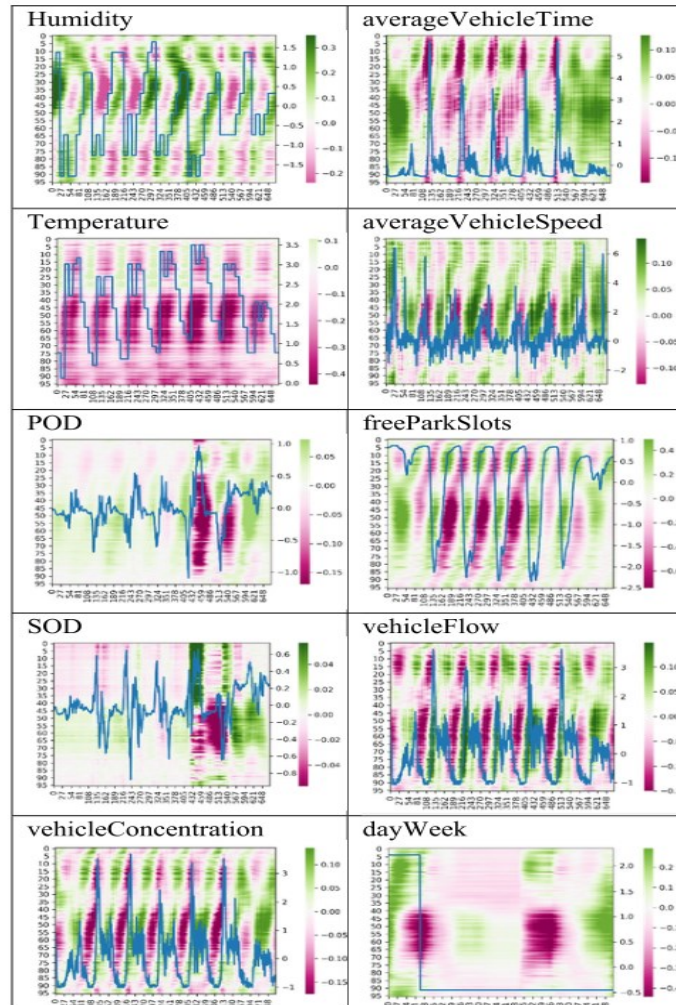
Technique	Best for	Mobility application
Deep learning on Time Series	Forecast temporal patterns	Free parking slots and short/mid-term demand
Hybrid physics + ML Neuro-Symbolic	Fill gaps in sparse sensing	Dense traffic flow reconstruction
Graph neural networks, GNN	Learn on road topology	Continuous density on unmonitored roads, traffic flow reconstruction
GA and DRL optimization	Choose actions under constraints. Optimized viability	1) Traffic light plans for tramway/BRT priority and compromise 2) Traffic infrastructure optimisation
Agentic AI based digital twins + XAI	Test scenarios and explain drivers	Demand-offer matching and CO2 levers



Road graphs + sensor data make modern AI spatial



Parking predict slot availability before drivers start searching



AI technique

Convolutional Bidirectional LSTM forecasts free spaces 24 hours ahead at 15-minute resolution.

Explainability of the effects of variables over time to the predictions

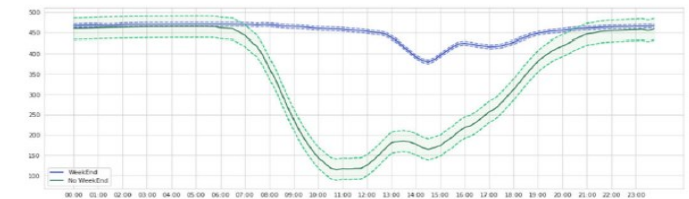
What it solves

Reduces cruising for parking and lets users plan trips one day ahead; the articles stress critical conditions when free slots approach zero.

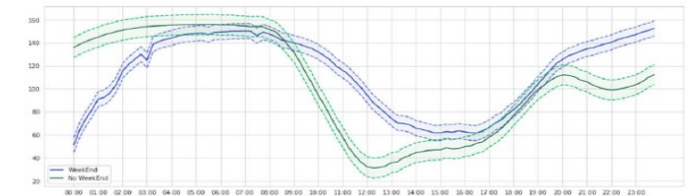
Reduce costs for predictions.

Mean Average Error 9.6-14.8

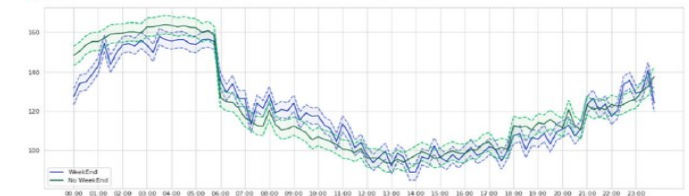
In predicting free slots for 1-hour forecast
Careggi / Beccaria / S. Lorenzo



(a) Careggi car park (hospital), capacity: 514



(b). Beccaria car park (downtown access), capacity: 203



(c) S. Lorenzo car park (old town access), capacity: 179

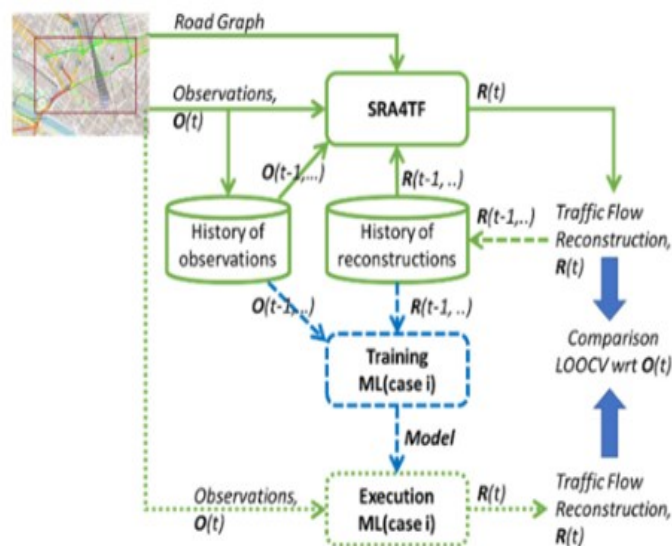
Traffic flow reconstruction: infers the roads without sensors

The core data problem: cities can measure a few links, but mobility decisions need every road segment.

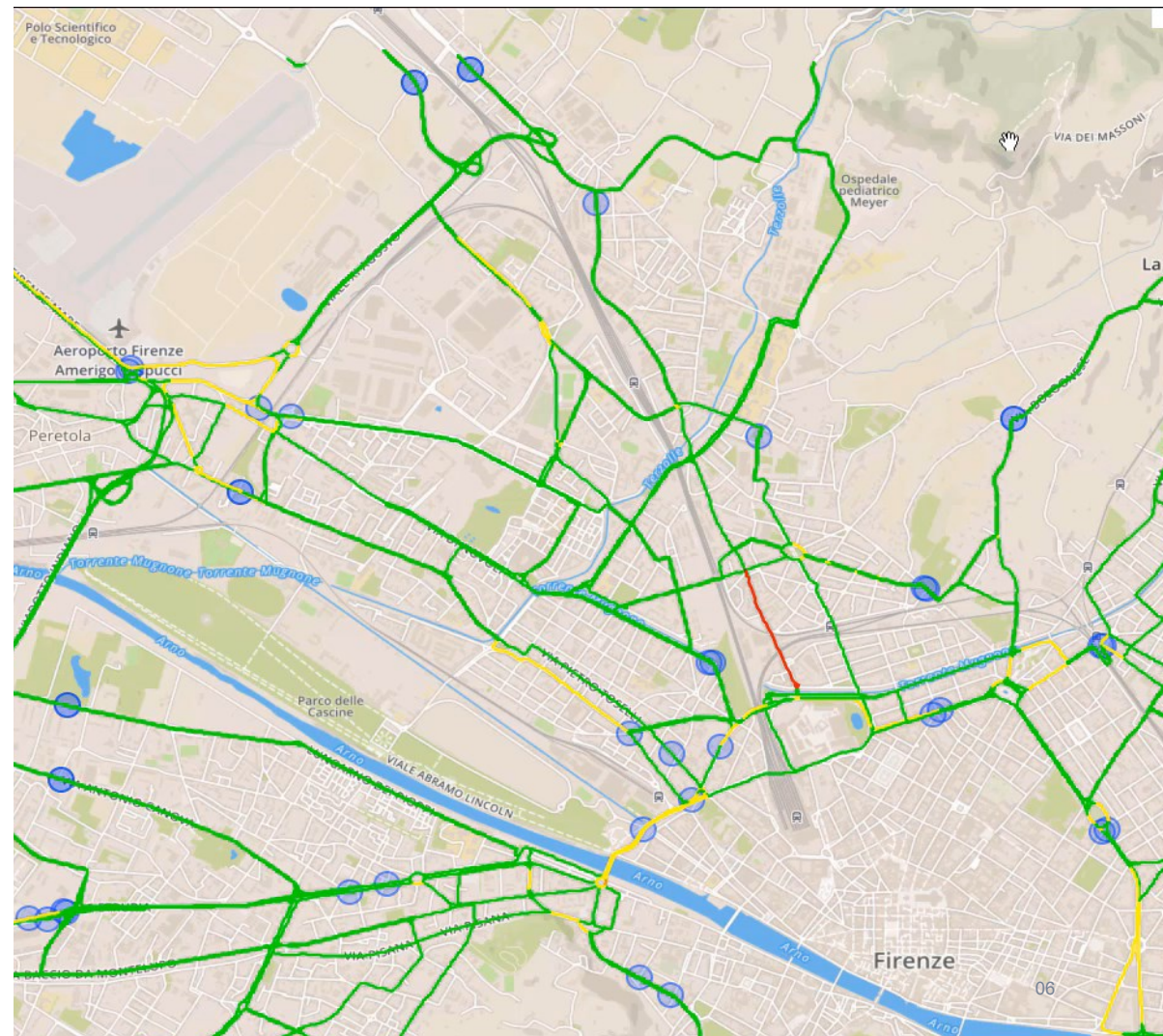
Few fixed sensors
counts + density +
speed

Hybrid PDE + ML
or lightweight GNN

Dense network state
for operations



Evidence point: one GNN reconstructs a 20775-segment road network from 96 sensors, with a RMSE 0.33 #veh/20m at sensor measurements and mean of RMSE 0.05 #veh/20m network-wide.



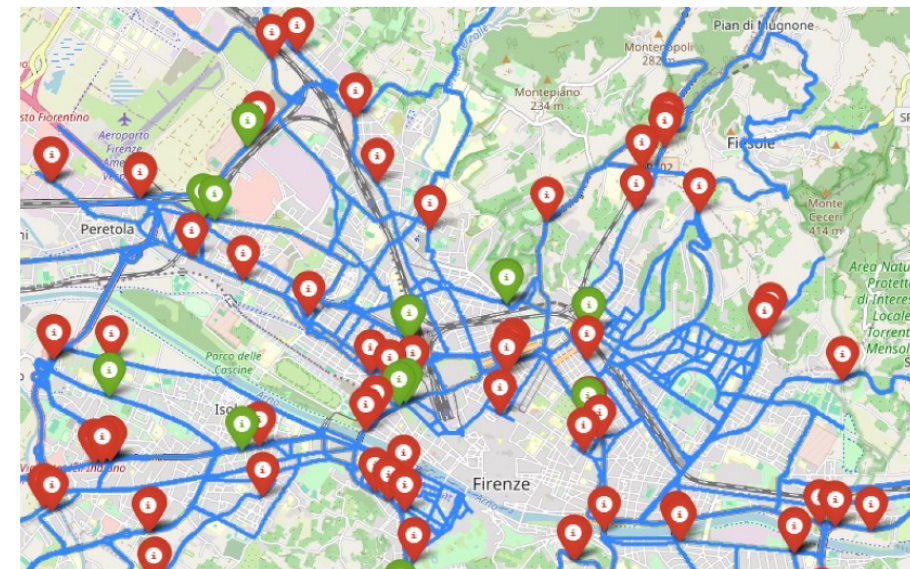
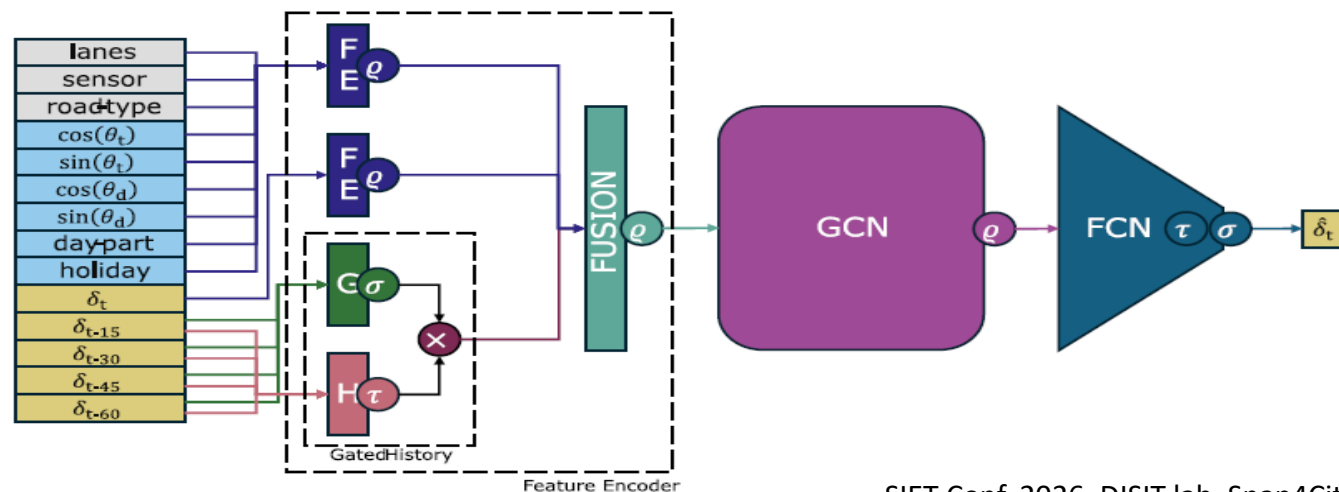
Graph neural networks, GNN, make road topology part of the model

Why it matters

Roads are not independent time series. Congestion propagates through junctions, and it can also back-propagate from bottlenecks.

What data is needed

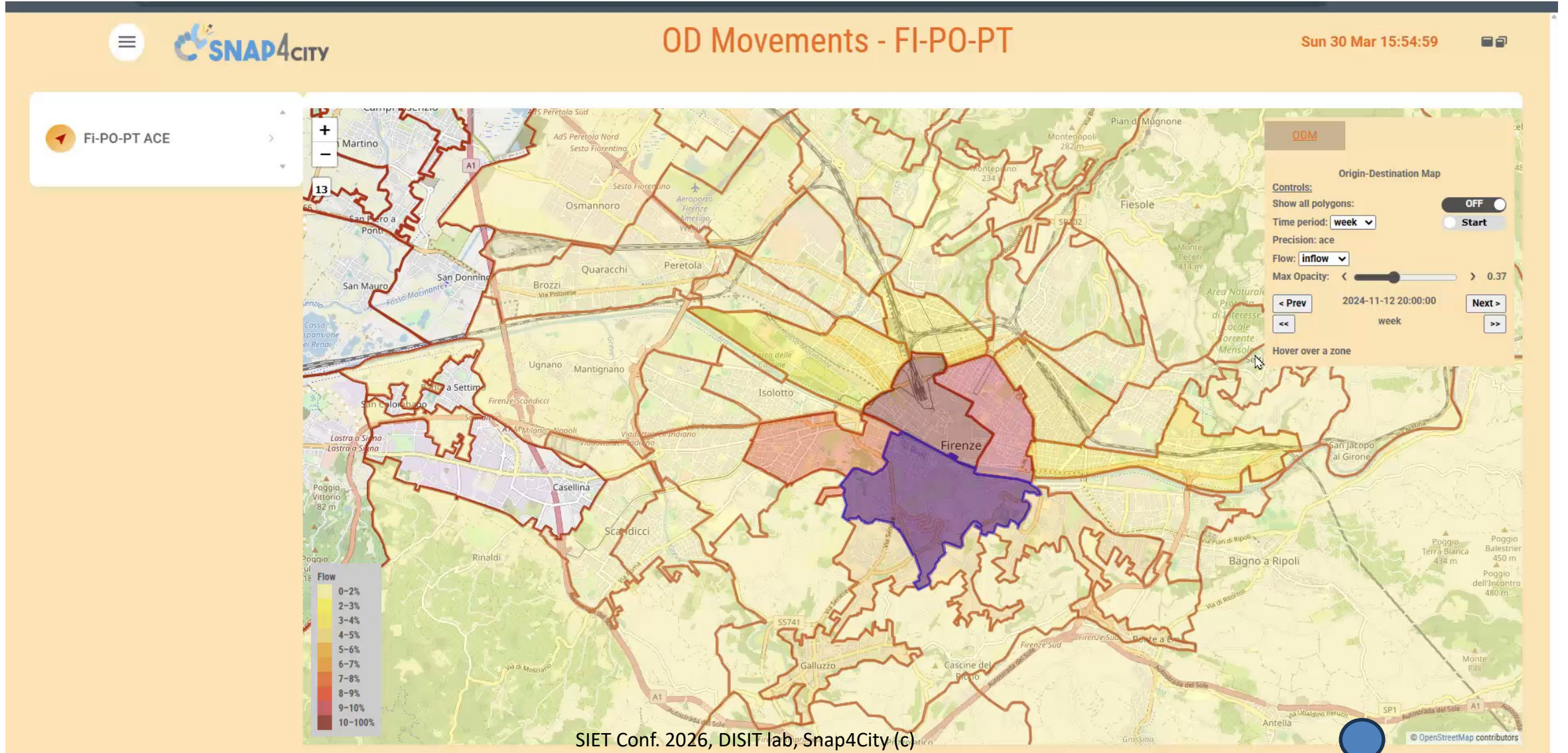
Road graph with lanes and road type; time-of-day and holiday features; current and lagged density estimates; sensor status; typical trends to cover missing observations.



Producing in real time:

- Predictions of traffic on dense network
RMSE 0.3-0.4 #Veh/20m
- Traffic distribution at junctions, TDM

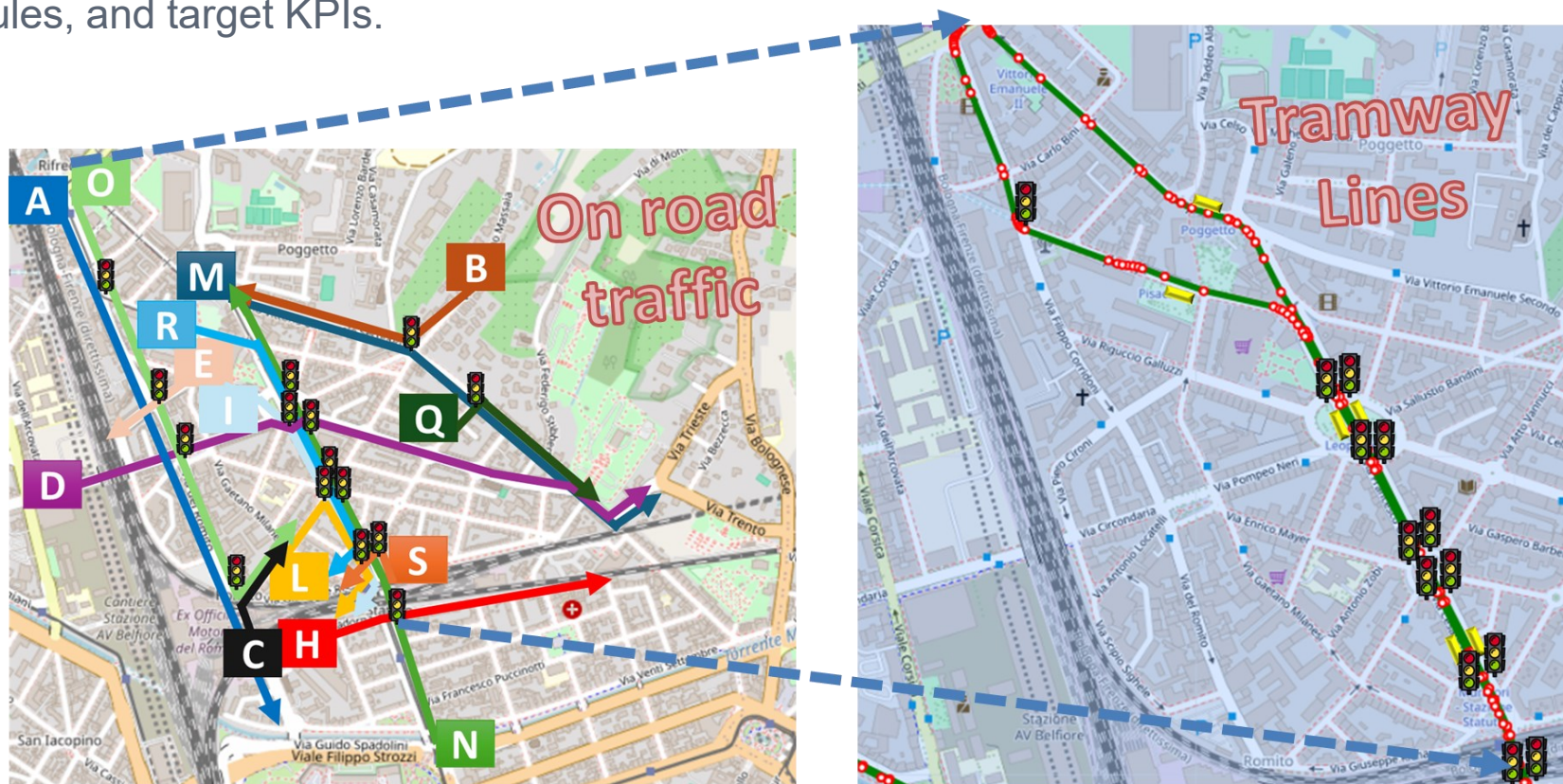
Origin Destination Matrices: Mobility Demand



Traffic Lights, TLP production: *optimize under multiple objectives*

Uses GA or Multi Agent DRL to balance travel time, waiting time, stops, queue constraints, and tramway priority at a macroscopic urban scale.

Data needed: junction topology, signal phases, legal constraints, traffic flow/density/speed, queue lengths, tram/BRT routes and schedules, and target KPIs.





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12

Mean-Travel-Time for Multi Agent DRL for TLP

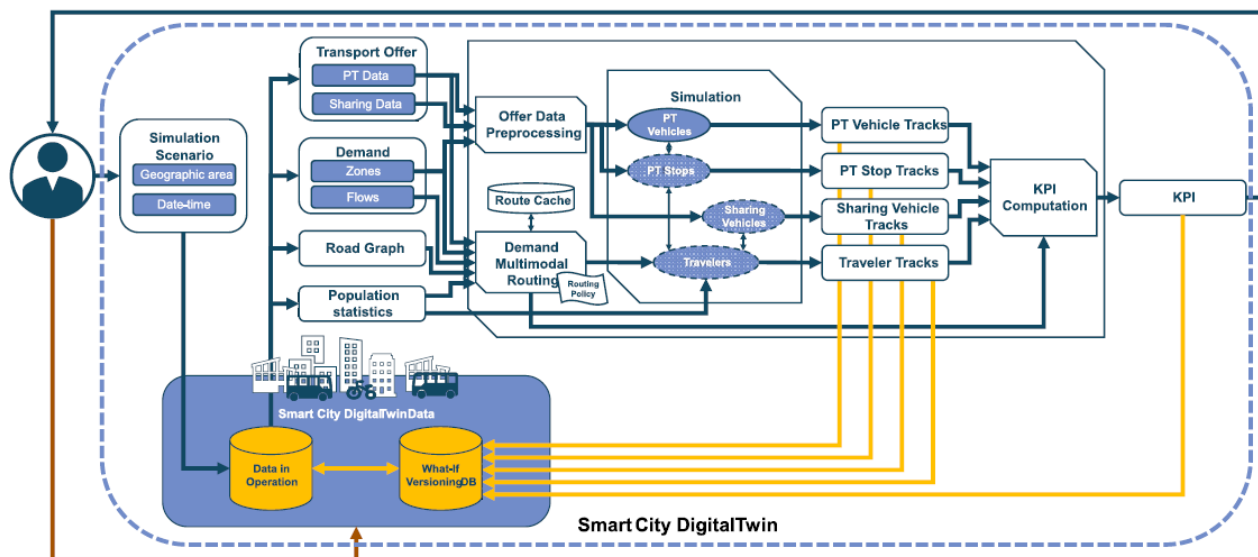
Model	TL	all	dir_N	dir_M	dir_A	dir_D	Careggi	Costanza
4TWD-NTNS-MWD-A	1	3013.85	176.93	233.75	194.46	237.65	436.00	427.00
4TWD-NTNS-MWD-P-A	1	3013.85	176.93	233.75	194.46	237.65	436.00	427.00
SUMO Actuated	1	2935.41	249.60	209.77	202.42	270.86	486.73	478.36
Webster	1	5188.87	211.66	242.32	205.46	562.31	984.00	427.00
Webster A	1	2968.90	183.50	242.67	201.27	251.76	482.27	427.00
SARL-FC DQN	Single-agent DRL	2834.93	206.68	244.78	199.21	243.26	486.72	485.00
SARL-FC DQN A		2760.12	206.35	244.63	198.00	244.76	436.00	427.00
MADRL-FC DQN	Multi-agent DRL	3089.20	188.29	220.91	205.07	248.06	485.00	445.00
MADRL-FC DQN A		2983.69	189.11	220.70	187.00	248.79	436.00	427.00
MARL-FC PPO		2910.76	200.38	235.41	198.14	237.78	547.00	445.00
MARL-FC PPO A		2855.12	200.93	235.53	196.39	237.97	436.00	427.00
SMART A	Actuated SMART	2599.13	182.14	200	188.28	235.11	436.00	427.00

Evaluation uses SUMO trip information - travel distance, wait time, CO2 emissions - with mean travel time as a key KPI.

What data is needed: real-time traffic observations, scenario files, signal definitions, road network, public transport movements, and reward/KPI definitions.

Demand vs Offer Matching: simulate multimodal alternatives, KPIs

MODOM evaluates how public transport, sharing services, and micro-mobility satisfy city users' demand. It is built for what-if planning inside a smart city digital twin.



~4 min

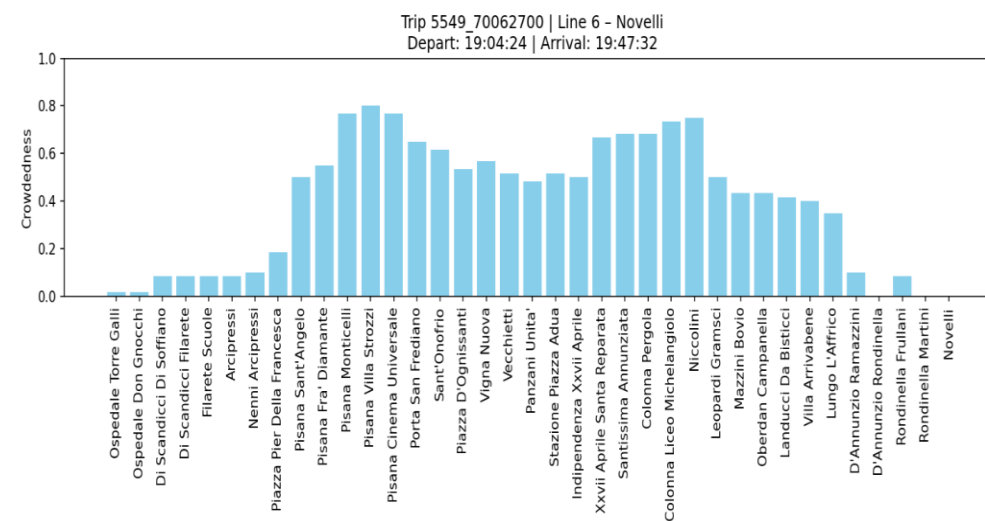
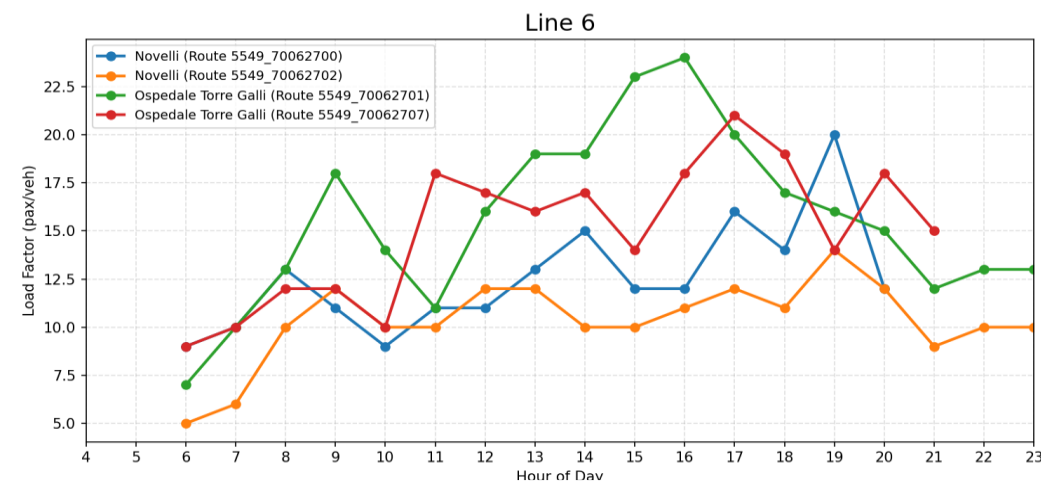
city-wide full-day
simulation

GTFS

directly instantiates
public transport offer

KPIs

tracks PT, stops, sharing,
and traveler events





Traffic Simulation-K8S

Sun 22 Jun 11:53:30



Ext

STOP

PAUSE

HELP

slow

fast

Delay: 30.0 ms

Stats

time: 0.000 s
payload: 0.0 KB
simulate: 0.00 ms
snapshot: 0.00 ms

Vehicle Summary

Quick Find

ID Edge / Lat, Long (float, float) /
X,Y (int, int)

SEARCH

CAR

BIKE

TRAIN

TRAM

PERSON

BUS

LIGHT

26 FPS (17-26)

▸ Lights

▸ Effects

▸ SSAO

▸ Scene

Close Controls

Wid

Prepare Simulation

Execute Simulation

KPI Simulation

Simulation:

firenzeodbus

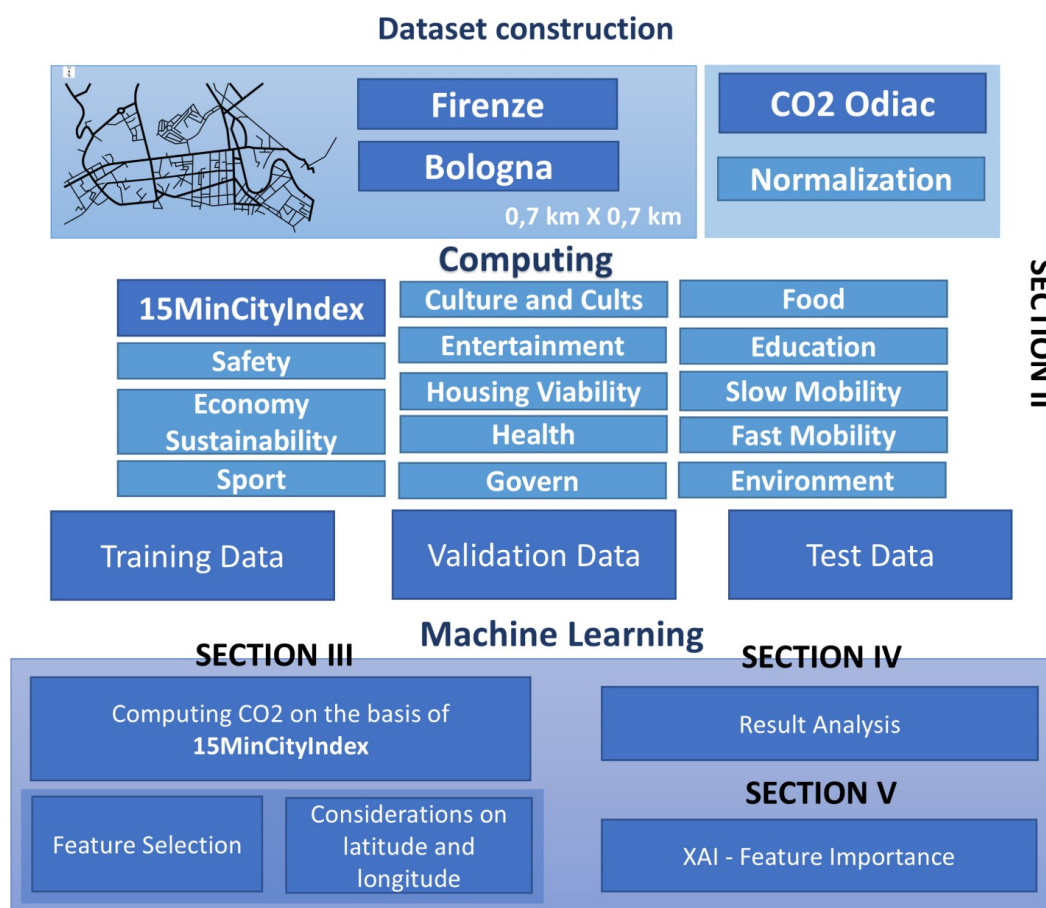
Execute

Simulation: 2025/06/22 11:53:27



CO2 and policy: explainable AI turns data into measures

The CO2 study used open urban data and satellite CO2 reference data to estimate emissions at fine spatial scale, then used XAI to reveal influential predictors.



Data needed: 15MinCityIndex, spatial grid, satellite CO2 target data, urban service locations, socio-economic indicators, mobility/environment features, and coordinates or graph/proximity data.

7.79%

Bologna XGBoost
MAPE with GPS

10.72%

Florence XGBoost
MAPE with GPS

0.96

R² for best XGBoost
models

SnapAdvisor Agentic AI support for decision makers

- **Colloquial interface** for Decision Makers with **Agentic AI expert** on Mobility and Transport, which knows:
 - Domain research innovations
 - historic data evolutions and conditions/events for the city
 - Real time data of any kind
 - Traffic, services, events, etc.
 - To represent data
 - To make What-if analysis
 - To make optimisations





• 15 Minute City Index:

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



- Optimization of car sharing/pooling
- Monitoring and Prediction of energy consumption
- Stimulating: Bike sharing, e-bikes, car charge, etc.
- Sizing energy plants, Community of energy



- Predictive maintenance
- Decisions Support Systems
- Process optimization, control
- Industry 4.0 integrated solutions
- AI assistant for commercial activities



- Reduction of emissions, reduction of congestions
- Smart City infrastructure: monitoring and resilience, long terms predictions, optim. operation and plan
- Effective and Low cost smart solutions
- What-if analysis, Simulations, optimization
- Origin Destination matrices computation



- Optimization of Waste Collection
- business intelligence tools for decision makers
- Reduction production costs
- Monitoring resource consumption
- Advisor for documentation, generative AI



- Reduction of emissions, reduction of congestions
- Monitoring and Predicting: NO₂, NO_x, CO₂, Traffic flow, pollutant, landslide, waste, etc.
- Traffic flow reconstruction, optimisation
- Demand vs Offer of Mobility analysis



- Shortening justice time
- Prediction of mediation proneness
- Assisting institution is taking legal decisions
- Anonymization and indexing legal docs.
- Ethical Explainable Artificial Intelligence
- Advisor for legal documentation, generative AI

15MinCityIndex on Bologna

enel x



Ciao roottooladmin!

Tue 3 May 20:14:59

15 MINUTI INDEX BOLOGNA CITTÀ METROPOLITANA - NEWGUI



enel x

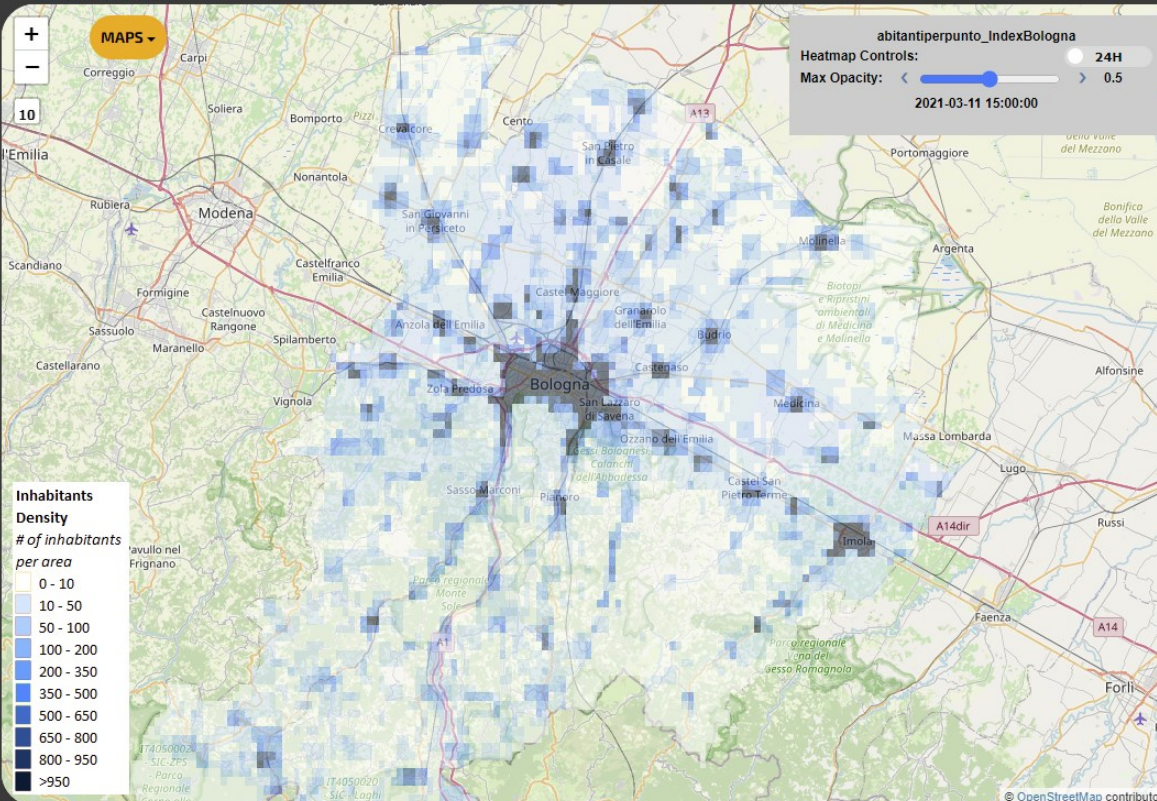
- # of Inhabitants
- Green factor
- Civil factor
- Industrialization factor
- Environment Index
- 15Min Economy Index
- 15Min Housing Index
- 15Min Health Index
- 15Min Food Index
- 15Min Education Index
- 15Min Slow Mob Index

THE PICKED POINT

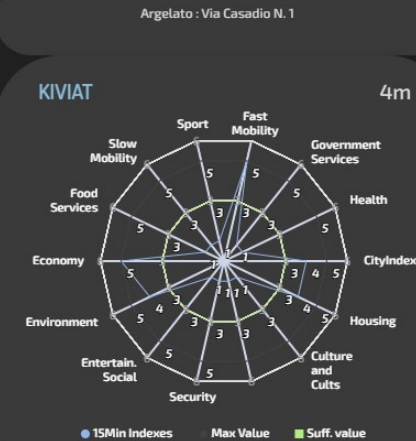
9m

City: Argelato
Address: Via Casadio N. 1
lat,lon: 44.61882,11.35437

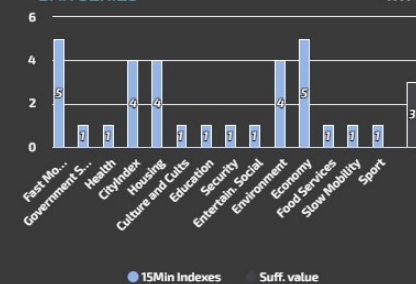
SELECTOR - MAP



KIVIAT



BAR SERIES



1 NO POVERTY



2 ZERO HUNGER



3 GOOD HEALTH AND WELL-BEING



4 QUALITY EDUCATION



7 AFFORDABLE AND CLEAN ENERGY



9 INDUSTRY, INNOVATION AND INFRASTRUCTURE



11 SUSTAINABLE CITIES AND COMMUNITIES



12 RESPONSIBLE CONSUMPTION AND PRODUCTION



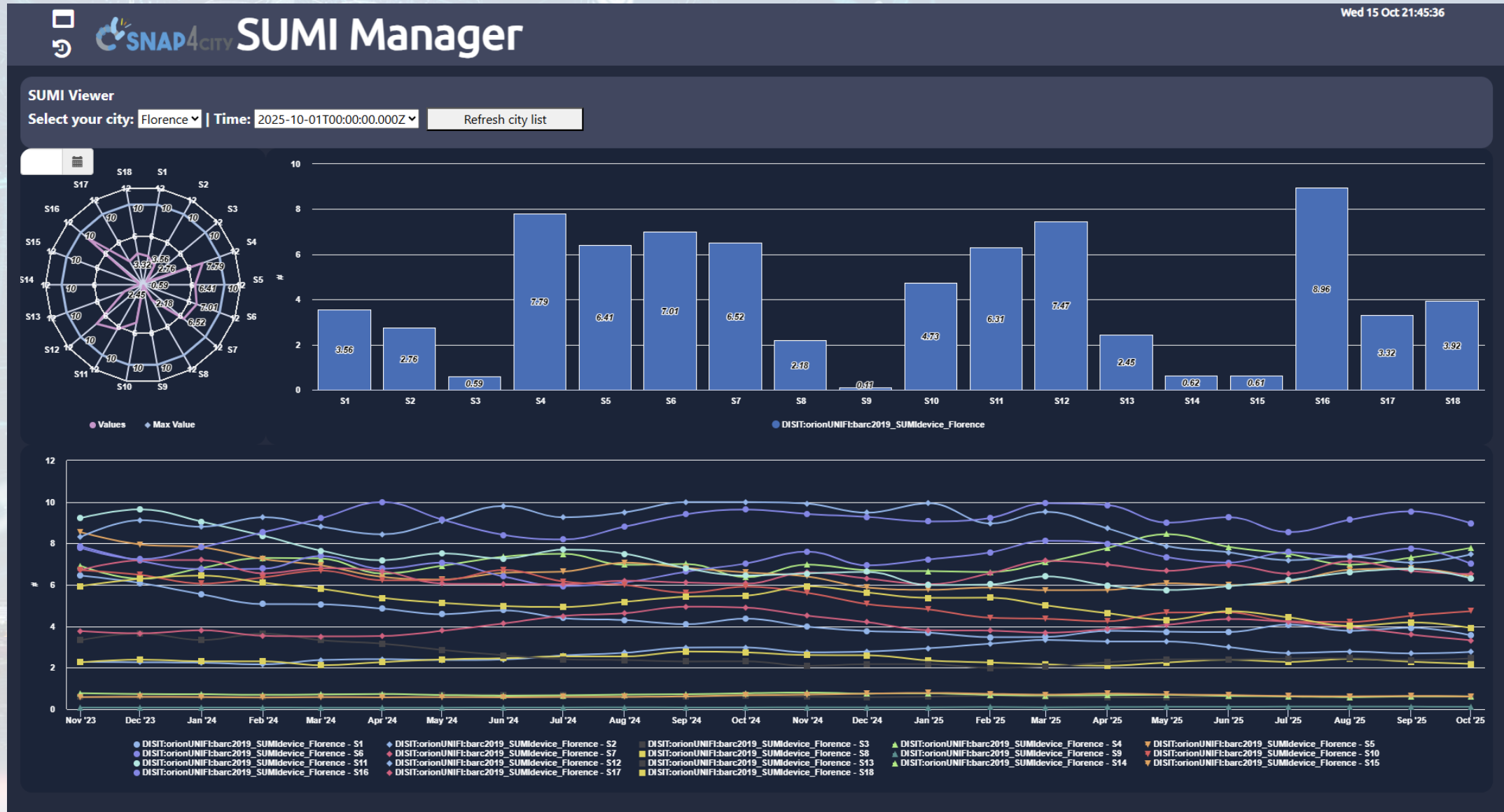
13 CLIMATE ACTION



15 LIFE ON LAND



SUMI: Sustainable Urban Mobility Indicators



More details can be obtained from DISIT lab articles appeared on top international Journals

- "Graph Neural Network for Continuous Traffic Density Estimation on Unmonitored Roads from Very Few Scattered Measurements", **Expert Systems With Applications, Elsevier, ESWA**, 2026. <https://www.sciencedirect.com/science/article/pii/S095741742601626X>
- "Data-Driven Mobility Demand-versus-Offer Matching by Dynamic Integration of Public and Sharing Services", **IEEE Access**, 2026. <https://ieeexplore.ieee.org/document/11373175> <https://doi.org/10.1109/ACCESS.2026.3661850>
- "Exploiting Open Data for CO₂ Estimation via Artificial Intelligence and eXplainable AI", **Expert Systems With Applications, Elsevier, ESWA**, 2025 <https://doi.org/10.1016/j.eswa.2025.128598>
- "Multi Agent Optimizing Traffic Light Signals Using Deep Reinforcement Learning", **IEEE Access**, 2025. <https://doi.org/10.1109/ACCESS.2025.3578518>
- "Macroscopic GA-based Multi-Objective Traffic Light Optimization Prioritizing Tramways", **Applied Soft Computing Journal, Elsevier**, 2025. <https://doi.org/10.1016/j.asoc.2025.113269>
- "High Precision Traffic Flow Reconstruction via Hybrid Method", **IEEE Transactions on Intelligent Transportation Systems**, 2023, <https://doi.org/10.1109/TITS.2023.3329544>
- "Predicting free parking slots via deep learning in short-mid terms explaining temporal impact of features", **IEEE Access**, 2023. <https://ieeexplore.ieee.org/abstract/document/10247516>, [10.1109/ACCESS.2023.3314660](https://doi.org/10.1109/ACCESS.2023.3314660) <https://www.ngen-ai.org>
- "An LLM-Based Multi-Agent Framework for Natural Language Interrogation and Exploitation of Smart City Services", Int. Conf. on Next Generation AI Systems, NGEN-AI 2026, Trento, <https://www.ngen-ai.org>, 1-4 September 2026.

A five-layer data blueprint fuels the AI algorithms solving smart-city transportation problems

DATA LAYER	SPECIFIC INPUT VARIABLES	ENABLED AI / ML TECHNIQUES	PRIMARY SOURCES
1. ROAD GRAPH	Directed nodes and arcs, lane counts, travel directions, road class (primary, secondary, residential), junction phases.	GNN Node Labeling, MaMoTLO (NSGA-II), SMART Actuated DRL, Map-Matching.	OpenStreetMap (OSM), Snap4City Semantic KB
2. SENSOR TELEMETRY	Real-time vehicle counts, concentration, velocities, historical typical time trends (TTT), garage ticket entries.	GNN-TFR, Hybrid PDE-ML (Random Forest, MLP), Conv-Bi-LSTM Parking Forecast.	Inductive loops, traffic cams, off-street garage gates
3. MOBILITY DEMAND	Origin-Destination Matrices (ODMs), hourly telecom cell phone presence logs, passenger transit schedules.	MODOM Agent-Based Simulation, Probabilistic Grid Resampling (200m MGRS).	Telecom operator logs, Census data, survey records
4. TRANSIT OFFER	Public bus, tram, and train timetables, vehicle capacities, stops, GBFS free-floating sharing fleets.	MODOM ABM, MaMoTLO Tramway Priority, Webster Coordination, SMART.	GTFS schedules, GBFS parking feeds, transit operators
5. SOCIO-ECONOMIC	Point-of-Interest densities (schools, health, food, govern), average taxable income, taxpayers, housing cost per m2.	XGBoost CO2 Estimation, GCN Proximity Modeling, SHAP Explainability.	Open Data Portals, Chamber of Commerce, satellite ODIAC
6. Agentic Operative KNOWLEDGE	Multidomain Knowledge, real time data, historical data, road graph, web info, etc.	Agentic AI, LLM, AI, Digital Twin, Simulation, Predictions	A large kind of knowledge on all research activities in the domain, on AI, and on data analytics

Conclusions: Agentic AI Engines are the new Kings

1 Data is the incipit, KPI are the driver

Road graphs, feeds, sensors, open data, and governance determine impact. One data can be in most cases surrogated by others. KPI are mandatory

2 AI can make sparse data operational.

Traffic reconstruction fills unmonitored roads without sensor saturation.

3 Prediction enables prevention.

Parking, traffic, critical conditions, and emissions can be anticipated rather than only observed.

4 Digital twins convert models into planning tools.

AI + simulations allow many what-if scenarios before street-level changes.

5 Optimization must encode policy.

Signals should balance efficiency, public transport priority, and fairness. *Also infrastructure can be optimized to better structured to sustain the traffic and reduce travel time, costs, consumption,...*

Bottom line: new AI techniques do not remove the need for transport expertise - they multiply its reach



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Alcune nuove tecniche dell'intelligenza artificiale possono risolvere i problemi della mobilità e dei trasporti e quali dati sono necessari

new artificial intelligence techniques can solve mobility and transportation problems and what data is needed

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