

The infographic depicts a smart city with various technology icons and labels. The central text reads "SMART CITIES: SYNERGY OF TECHNOLOGIES FOR A BETTER URBAN FUTURE". The background features a stylized cityscape with buildings and a river. Various icons and labels are scattered around the city, including: "Traffic Management" (top left), "Education" (top center), "Air Pollution" (top right), "Open Data" (top right), "Electromagnetic Emissions" (top right), "Internet of Things" (middle left), "Smart Health" (middle left), "Intelligent Shopping" (middle left), "Smart Environment" (middle right), "Smart Buildings" (middle right), "Smart Home" (bottom right), "Smart Street Lights" (bottom right), "Public Safety" (bottom left), and "Gas & Water Leak Detection" (bottom left). The city itself includes a river, a bridge, a hospital, a shopping center, a school, a factory, and various buildings and cars.

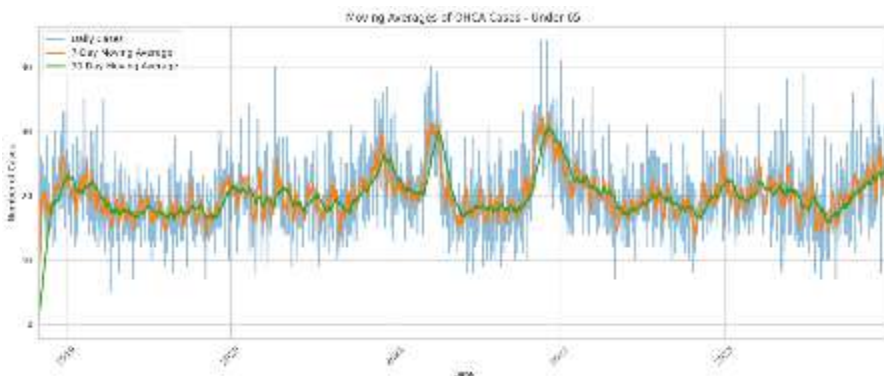
SMART CITIES: SYNERGY OF TECHNOLOGIES FOR A BETTER URBAN FUTURE



Introduction

What is a smart city?

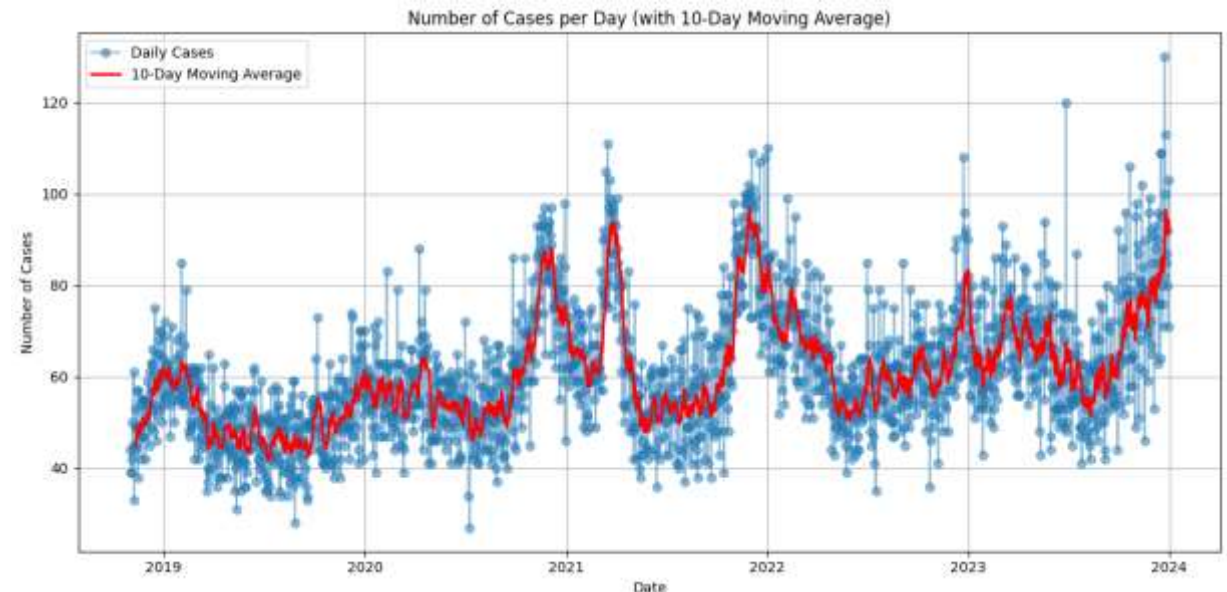
- Urban area that uses digital technology to collect data and operate services on many levels.
- Not defined by a single product or innovation
- Composite of cutting-edge technologies --> technological improvements
 - Data driven infrastructures
 - Smart healthcare systems and case prediction
 - **Optimized traffic control**
 - Optimized energy sustenance
 - Smart products



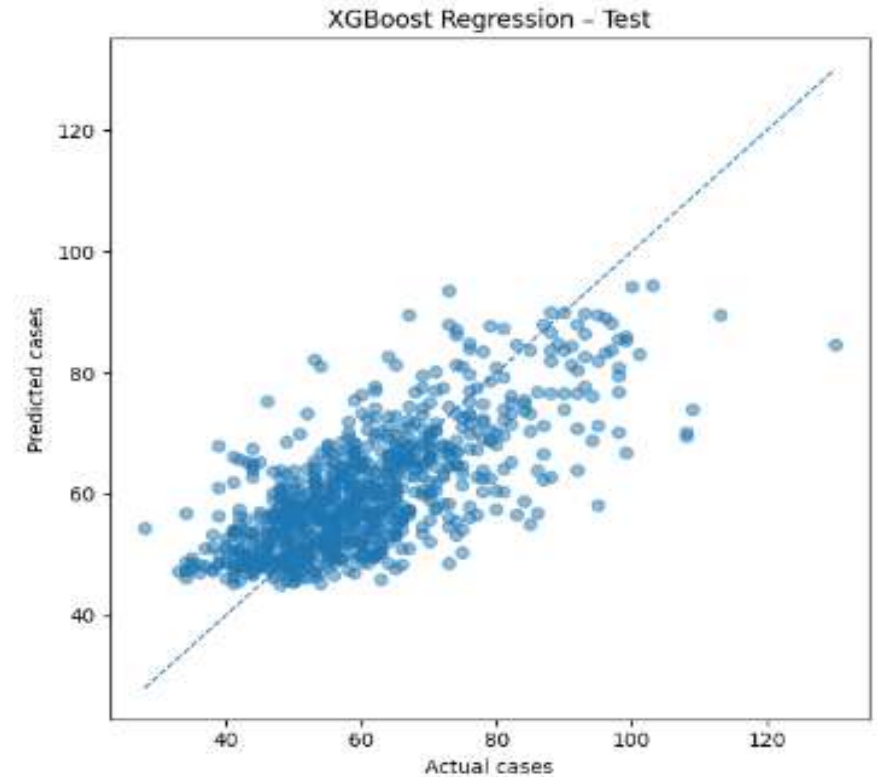
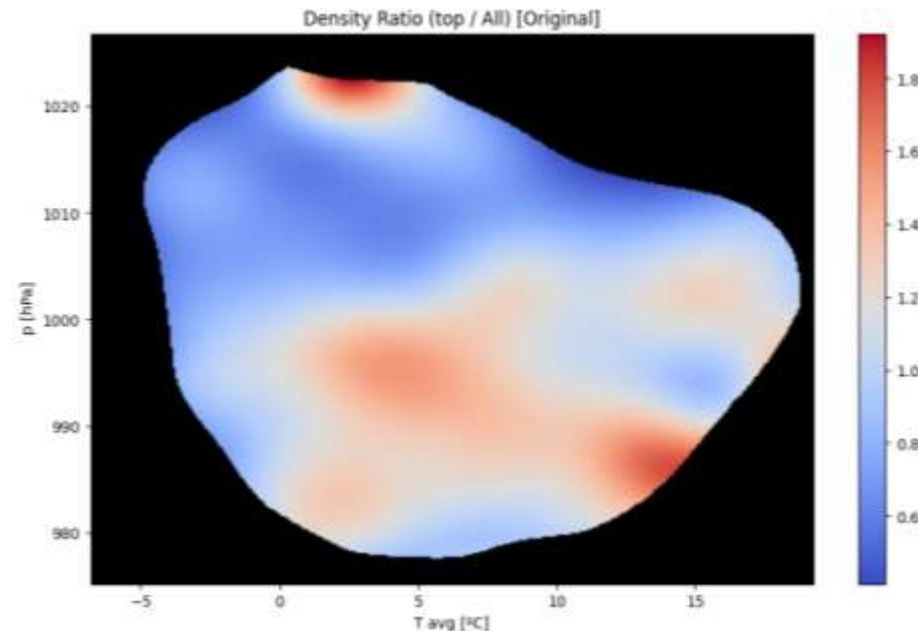


Data for detection and prediction

- Understanding and knowing when and how health hazards occurs is imperative
- **Goal:**
 - Predict daily incidences - with global parameters like weather
 - → Optimal hospital resource allocation
 - Identifying risk factors and high risk days
- **Algorithms tested on:**
 - Hungarian OHCA database, meteorological database
 - Known risk factors:
 - Global: weather, social background, seasonal changes
 - Individual: age, sex, genetics, recreational drug consumption (e.g. tobacco, alcohol), obesity

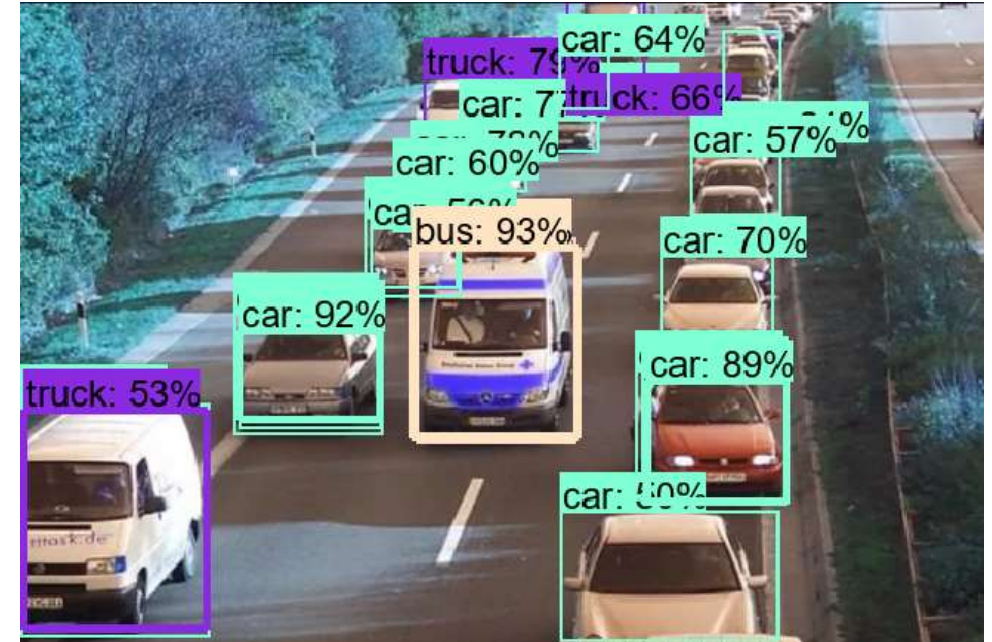


- General regressions usually fail
- Machine learning tool : XGBoost - decision trees
- XGBoost with all parameters
 - With previous days case numbers:
 - $R^2 = 0.5$
 - With out :
 - $R^2 = 0.24$
- Good fit for weakly correlating datasets
- Ways to enhance:
 - More information
 - Reliable healthcare
 - Outlier analysis



Importance of Data – Traffic Detection

- Information about current and exact traffic information is extremely relevant for optimization
 - Enhancing navigation
 - Identifying accidents
 - Optimal traffic light control
- Known solutions for car detection using convolutional networks
- NOW: maritime surveillance with satellite imagery



- Masked R-CNN
- ResNet50 backbone
- 384^2 RGB input
- 384^2 grayscale
- Optimizer: Adam
- Learning rate: $10e-4$
- Loss: BCE
- Sigmoid activation

- Accurate detection (Val_loss = 0,0056), but crude outline
- **Post processing** for exact outline:
 - Ocen removal
 - Canny edge detection (gradient change)
 - Filling inside edges

Layer (type)	Output Shape	Param #
input_image (InputLayer)	(None, 384, 384, 3)	0
functional_3 (Functional)	[(None, 96, 96, 256), (None, 48, 48, 512), (None, 24, 24, 1024)]	8,589,184
conv2d_3 (Conv2D)	(None, 24, 24, 256)	2,359,552
conv2d_transpose_8 (Conv2DTranspose)	(None, 48, 48, 256)	590,080
conv2d_transpose_9 (Conv2DTranspose)	(None, 96, 96, 256)	590,080
conv2d_transpose_10 (Conv2DTranspose)	(None, 192, 192, 256)	590,080
conv2d_transpose_11 (Conv2DTranspose)	(None, 384, 384, 256)	590,080
mask_output (Conv2D)	(None, 384, 384, 1)	257

Total params: 13,309,313 (50.77 MB)
Trainable params: 13,278,721 (50.65 MB)
Non-trainable params: 30,592 (119.50 KB)





Traffic optimization

Smart traffic lights



Sensors and
cameras



Intelligent and
automated



Image detection,
machine learning,
prediction





Busy intersection, jammed on peak hours

A lot of time lost

Stressed people going to work can lead do dangerous maneuvers

A crossroads in the east of Madrid



Benefits of smart traffic lights

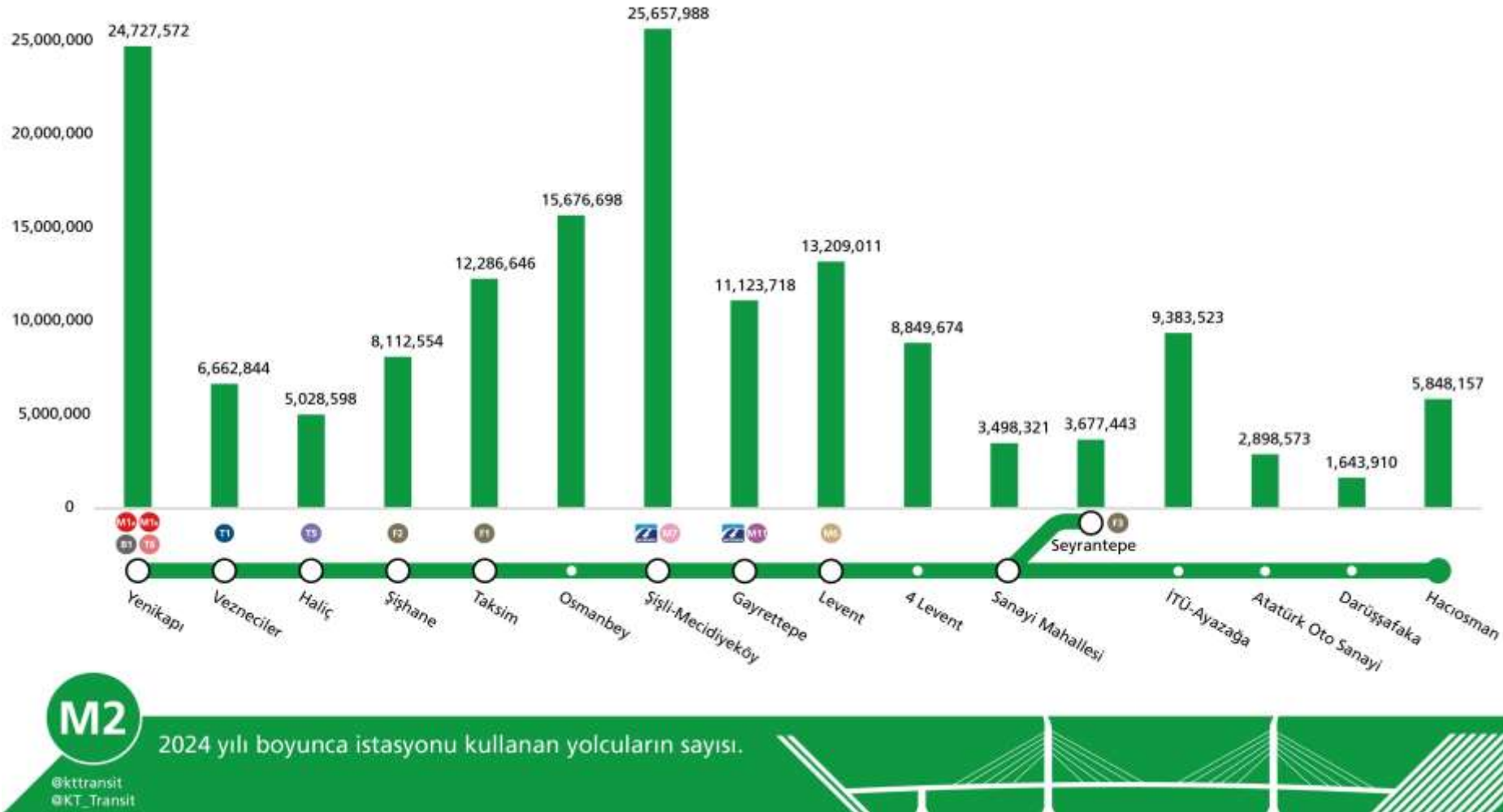
- Reduced journeys times
- Less CO2 emissions
- Avoiding dangerous and stressed behaviors
- Less accidents
- Flow management adapted to the reality of human behavior: flexible and accessible



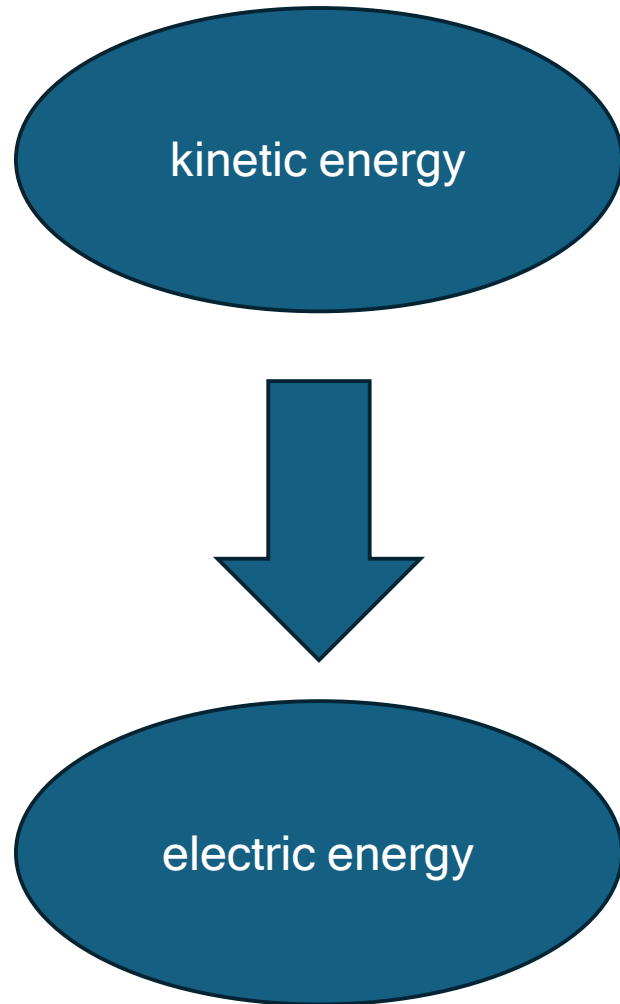
Smart energy generation



37000 people on average/day



Station-Based Passenger Usage on the M2 Metro Line in 2024



small amounts of energy for LED signs/traffic
lights
(10 bulbs for 20s)

Benefits

- renewable energy
- data collection
- sustainability
- no limitations

Risks

- high cost (110€ each)
- maintenance
- energy storage
- low power output
- space for wiring

Most optimal placements

-busy metro stations
-concerts/festivals

-schools/universities
-playgrounds

Japan (Tokio) Shibuya station

-since 2008

-400000 commuters pass
daily

-140 Wh/day

-150 €/m² (installed)





Interaction of the people and the city

Modular Kinetic Tile + Display Unit

Core Part:

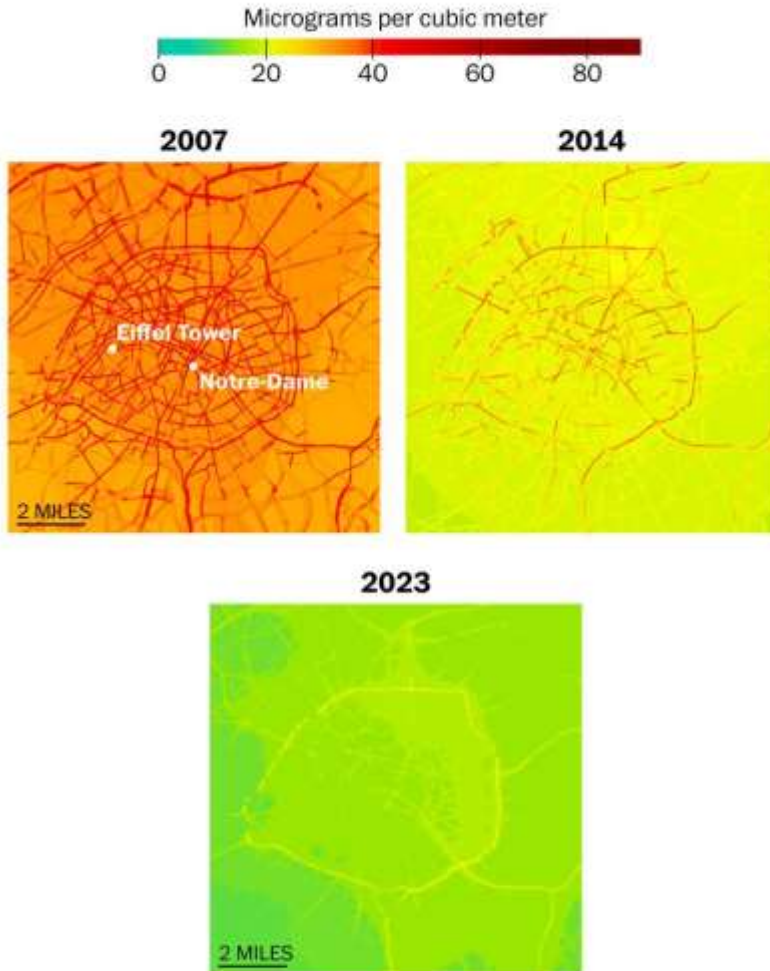
- **Pressure-sensitive kinetic tile** (like Pavegen): captures kinetic energy from footsteps, converts it into electricity.
- **Modular and replaceable** – tiles can be laid in rows in metro entrances, parks, or walking corridors.

Detachable/Changeable Add-ons:

- **Swappable display modules** (LED/OLED) with solar backup.
- **Customizable shell casing** or art-skin: community can redesign the housing.



Average PM 2.5 concentration in Paris



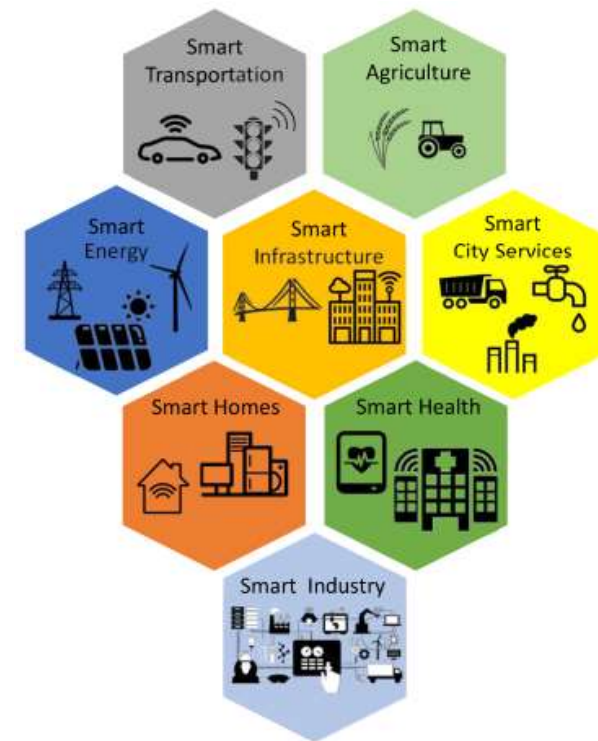
Source: Airparif

Interaction Logic

- Every step generates a small amount of electricity (measured in Wh).
- That energy is tracked *per tile* and *per user* (optional with *card/sensor use*).
- The display shows:
- 🌱 “You generated X watt-hours today!”
- 🚇 “By using M2 Line instead of driving, you saved Y kWh.”
- 🔋 “Your steps powered 2 minutes of metro lighting today.”

Can also show:

- **Leaderboard**
- **Monthly collective community goals**
- **Gamified messages** (e.g. “You’ve just powered a streetlamp for 15 seconds 💡”)



Q&A

