

BIG DATA EUROPE

Empowering Communities
with Data Technologies

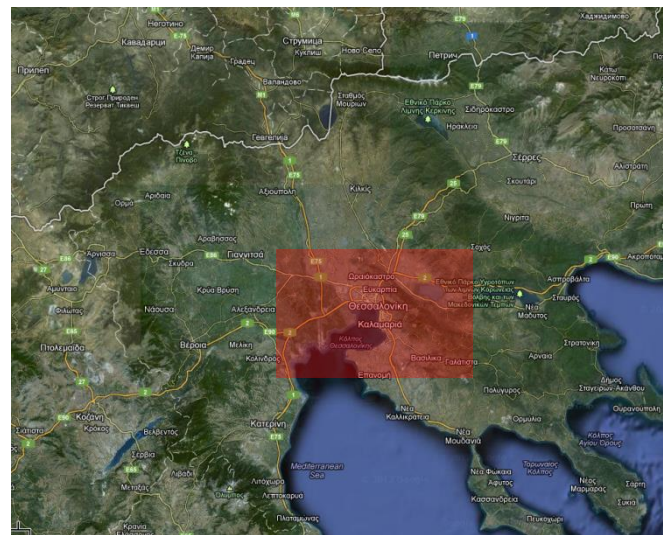
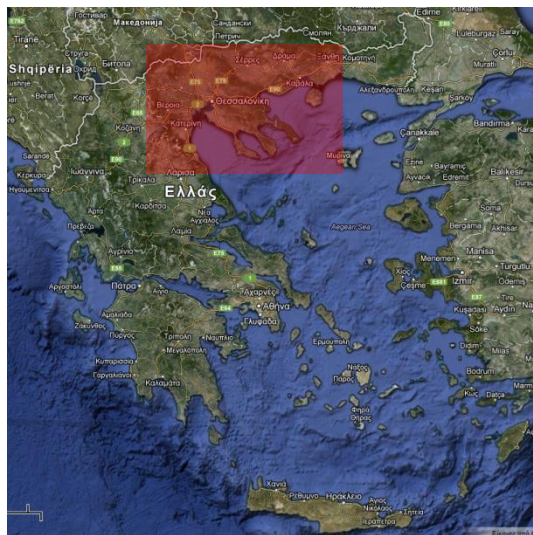


BIG DATA EUROPE TRANSPORT PILOT: INTRODUCING THESSALONIKI

Josep Maria Salanova Grau CERTH-HIT



Thessaloniki on the map



~ 1.400.000 inhabitants & ~ 1.300.000 daily trips

~450.000 private cars & ~ 20.000 motorcycles

1 (+1) public transport operator for urban trips & 1.950 taxis

~35 public transport operators for extra-urban trips

6.433 kms of streets - 8,8 kms of dedicated bus lanes - 89,4 kms of ring road

197.696 parking places



Probe data in Thessaloniki

- ⊙ Static sensors network: Point to point tracking of MAC ids along the network through Bluetooth detectors (43 devices).
- ⊙ Dynamic sensors fleet: Floating Car Data provided in real time by a professional fleet (more than 1.200 vehicles).
- ⊙ Cooperative technologies (COMPASS4D and COGISTICS): RSU is a static sensor and OBU is a dynamic sensors (CAM message).
- ⊙ Social media (Twitter & Facebook)

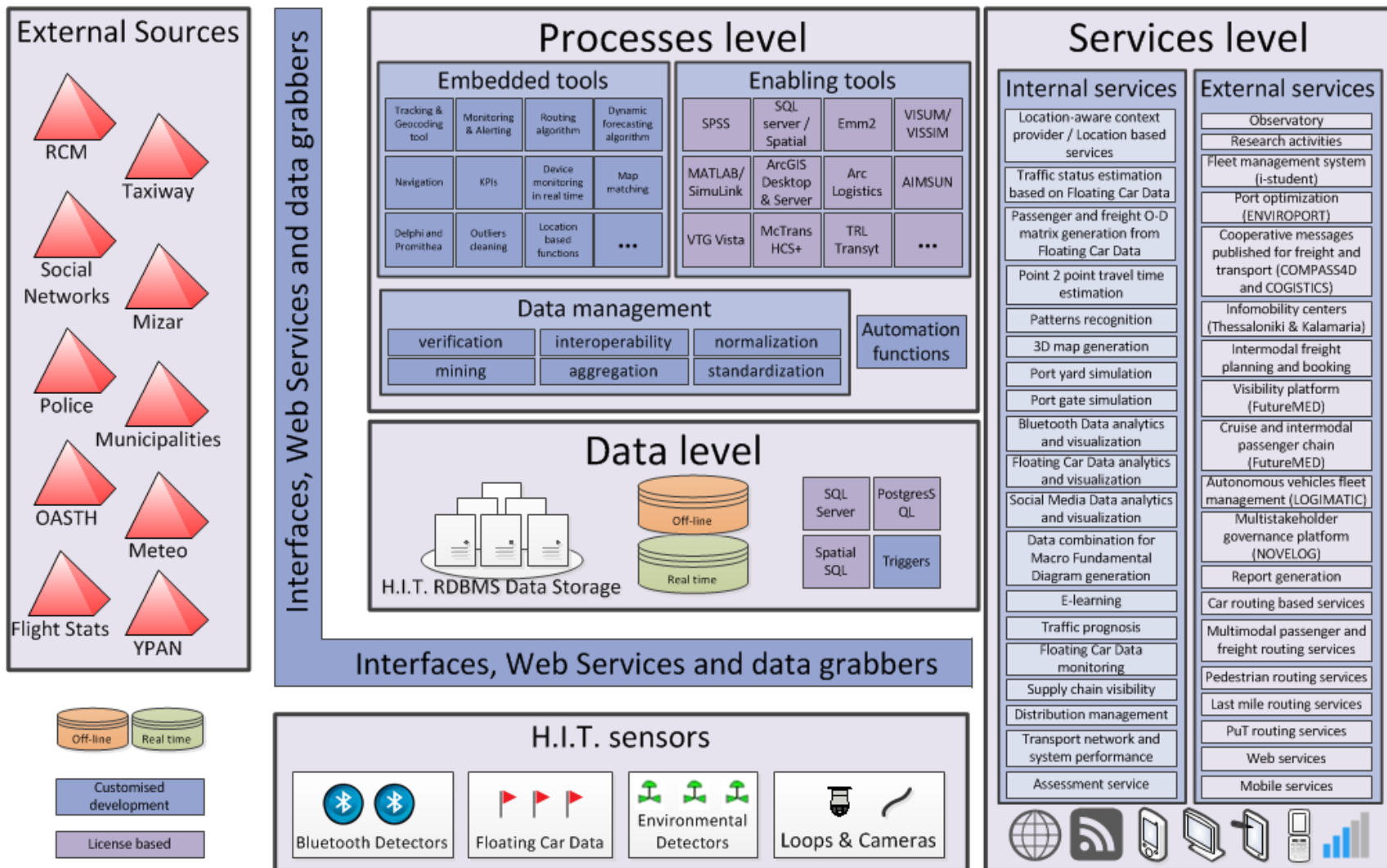


How do we use Probe Data?

- ⊙ Stationary sensors network: Point to point tracking of MAC ids along the network through 43 Bluetooth device detectors.
 - Travel time estimation
 - Route choice model calibration
 - Origin – Destination matrix estimation / **Mobility patterns estimation**
 - Traffic flow extrapolation
- ⊙ Dynamic sensors fleet: Floating Car Data provided in real time by a professional fleets composed of 1.200 taxis and 600 buses
 - Traffic status estimation (average speed)
 - Origin – Destination matrix estimation / **Mobility patterns estimation**
 - Taxi/bus performance indicators
- ⊙ Social media (geolocated tweets & Facebook check-in events)
 - Activity patterns estimation
 - Events / incidents detection
 - Attraction models estimation



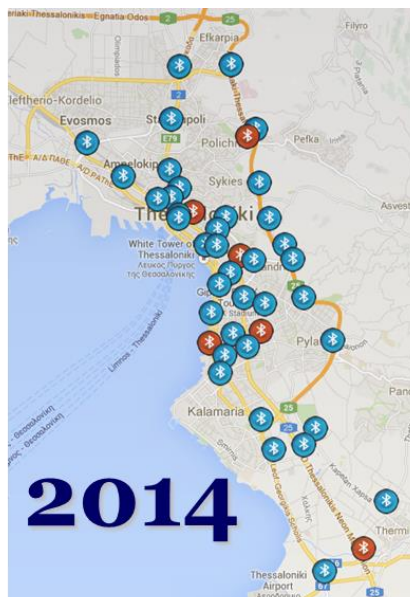
Thessaloniki Mobility lab





Point to point BT network

- ⊙ 43 detectors (EEA, SEE-ITS & EASYTRIP)
 - 4 million detections per week (peak period)
 - 25.000 unique devices detected per day (one intersection)
 - 1 million “tracked” trips per week
 - 20.000 “tracked” trips per day (one path)
- ⊙ More detectors installed in other cities and in Bulgaria (SEE-ITS & EASYTRIP)

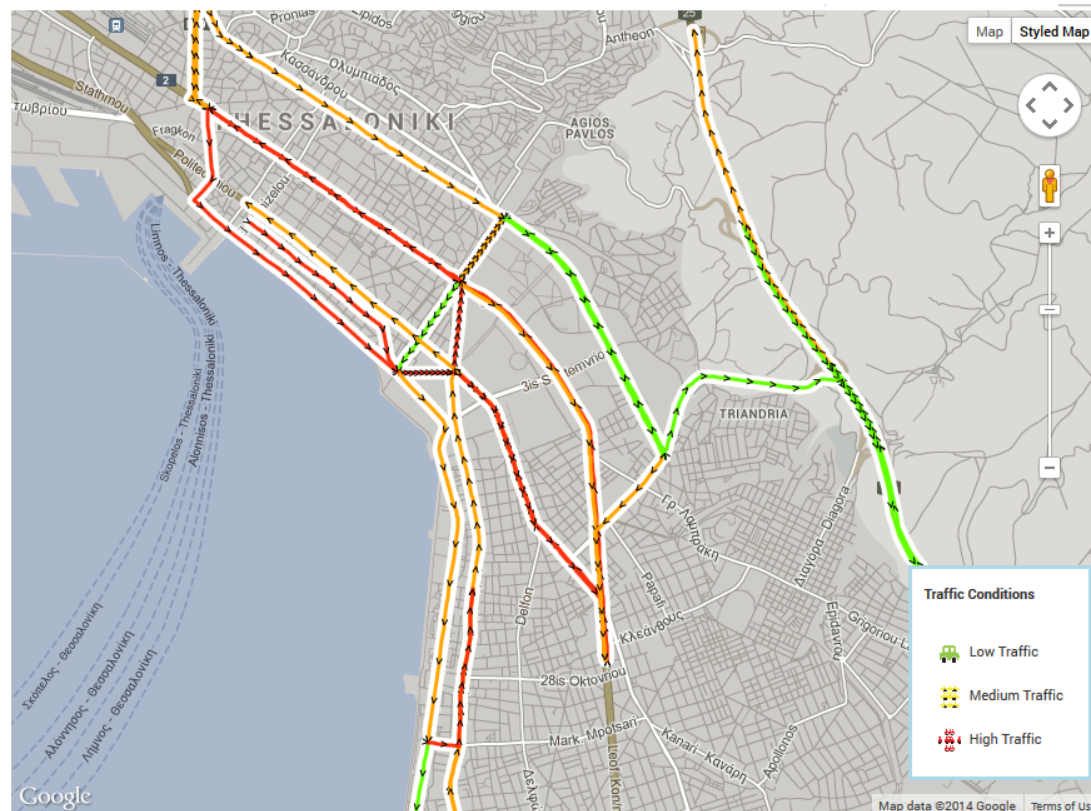




Point to point BT network

- Real time travel time provision to drivers (VMS, internet, smart device)

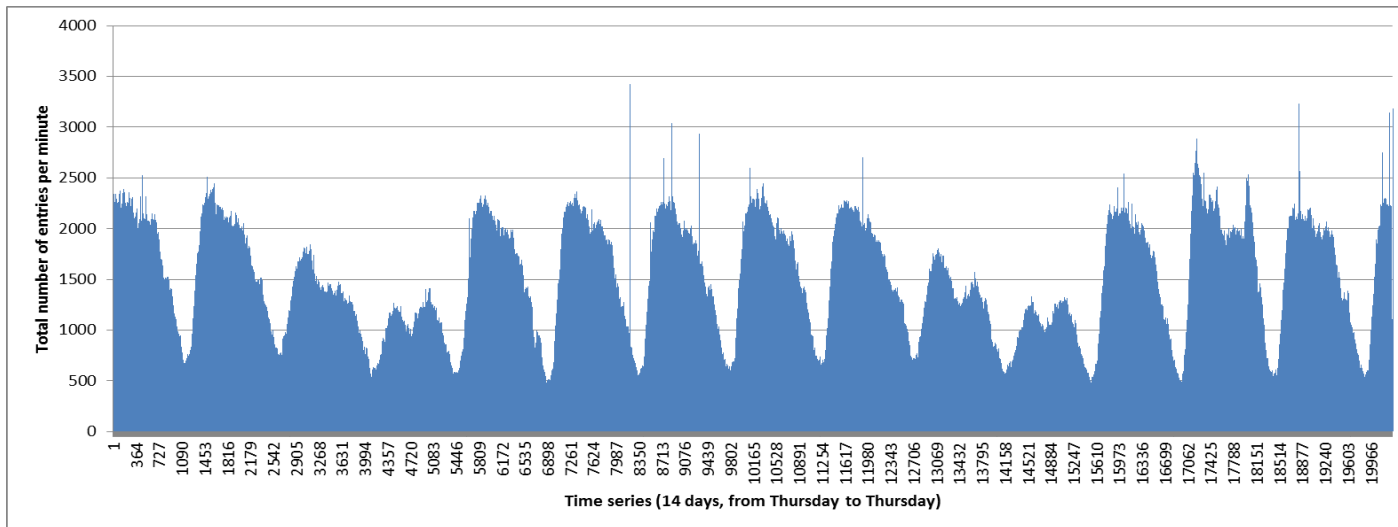
	K. Karamanli (Psaron/Kleanthous) - Platia CHANTH (CHANTH)	20'
	K. Karamanli (Psaron/Kleanthous) - Platia Sintrivaniou (Sintrivani)	17'
	Lagada (Ag. Pantou) - Platia Demokratias (Vardaris)	03'
	Platia Sintrivaniou (Sintrivani) - Lefkos Pyrgos	02'
	V. Olgas (Arch. Mousiou) - Platia CHANTH (CHANTH)	03'
	Platia CHANTH (CHANTH) - Platia Sintrivaniou (Sintrivani)	04'
	Platia Demokratias (Vardaris) - Platia Sintrivaniou (Sintrivani)	07'
	Evangelistria (Ag. Dimitriou/Ethn. Aminis) - Platia Sintrivaniou (Sintrivani)	02'
	Lefkos Pyrgos - Platia CHANTH (CHANTH)	03'
	Platia CHANTH (CHANTH) - K. Karamanli (Psaron/Kleanthous)	12'
	Platia Sintrivaniou (Sintrivani) - K. Karamanli (Psaron/Kleanthous)	05'
	Platia Sintrivaniou (Sintrivani) - Evangelistria (Ag. Dimitriou/Ethn. Aminis)	02'
	Platia Sintrivaniou (Sintrivani) - Platia Demokratias (Vardaris)	08'
	Platia Demokratias (Vardaris) - Lagada (Ag. Pantou)	03'





Floating Car Data

- ⊙ More than 1.200 vehicles (one taxi fleet)
 - Circulating 16-24 hours per day
 - Pulse generated each 100 meters (10-12 seconds)
 - 500-2.500 pulses per minute

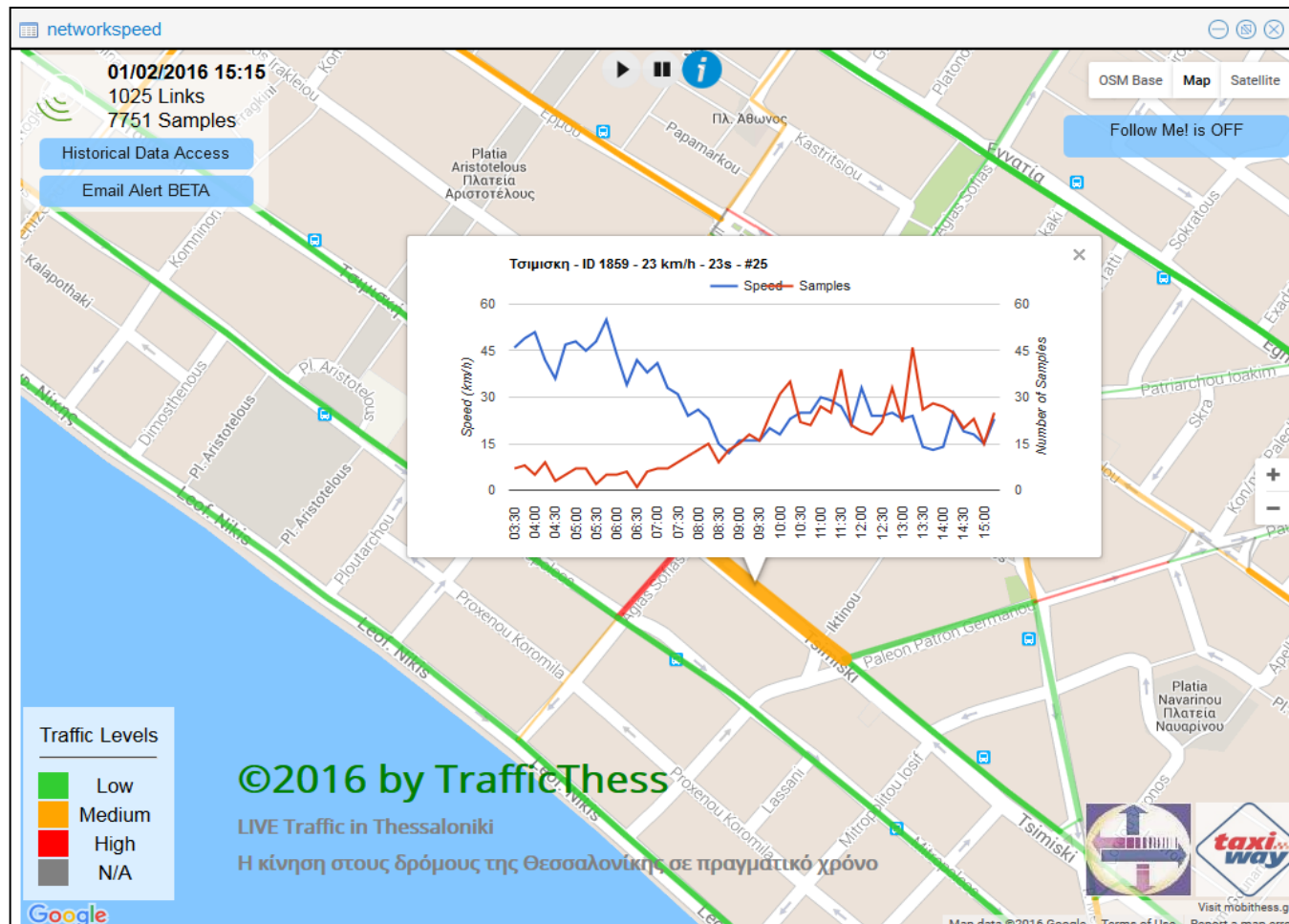


- ⊙ 600 vehicles generating CAM each second



Floating Car Data

- Real time traffic conditions information (speed)





Social media

44.000 check-in events per week (750 locations)

Up to

50 check-in events per minute (in the 136 locations tagged as bar)

17 check-in events per minute (in the 150 locations tagged as restaurant)

12 check-in events per minute (in the 32 locations tagged as outdoor)

10 check-in events per minute (in the 125 locations tagged as cafe)

10 check-in events per minute (in the 55 locations tagged as nightlife)

Up to

1265 check-in events during the “peak hour”

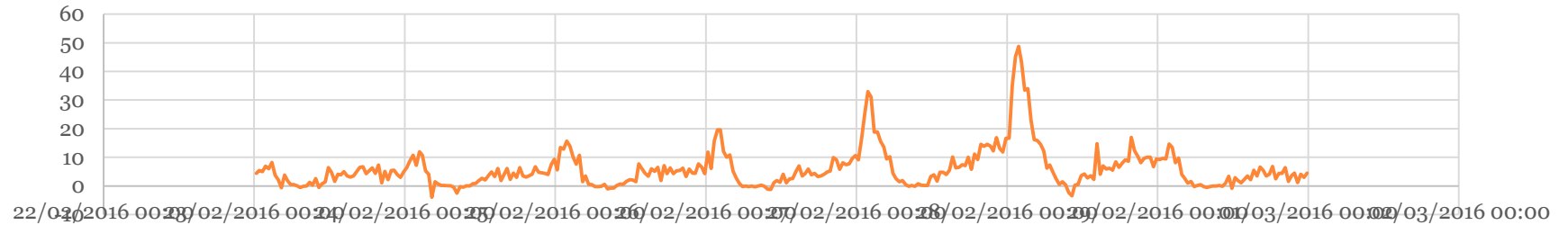
920 check-in events in bars (Sunday 01.00)

300 check-in events in restaurants (Saturday 22.00)

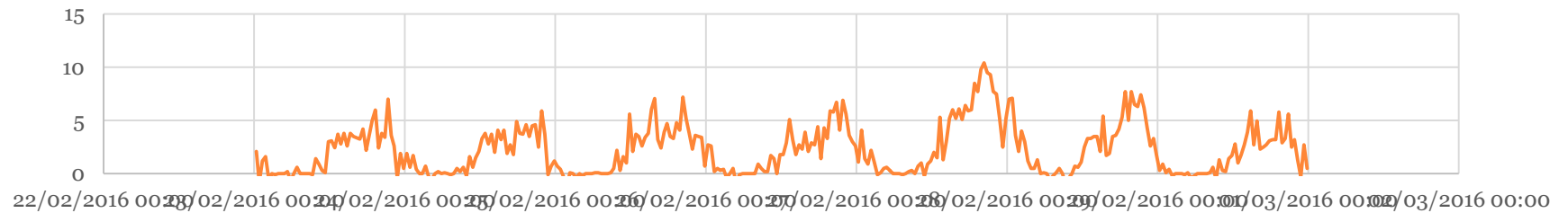


Social media

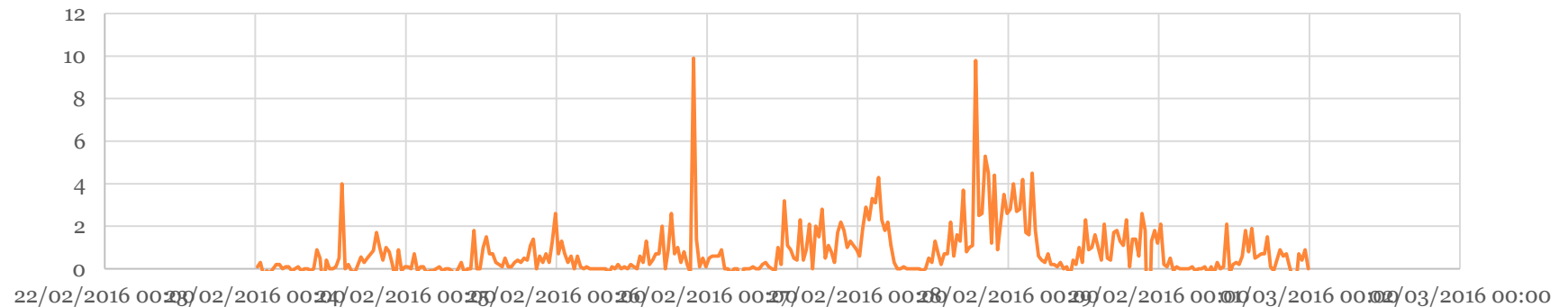
BAR



CAFE

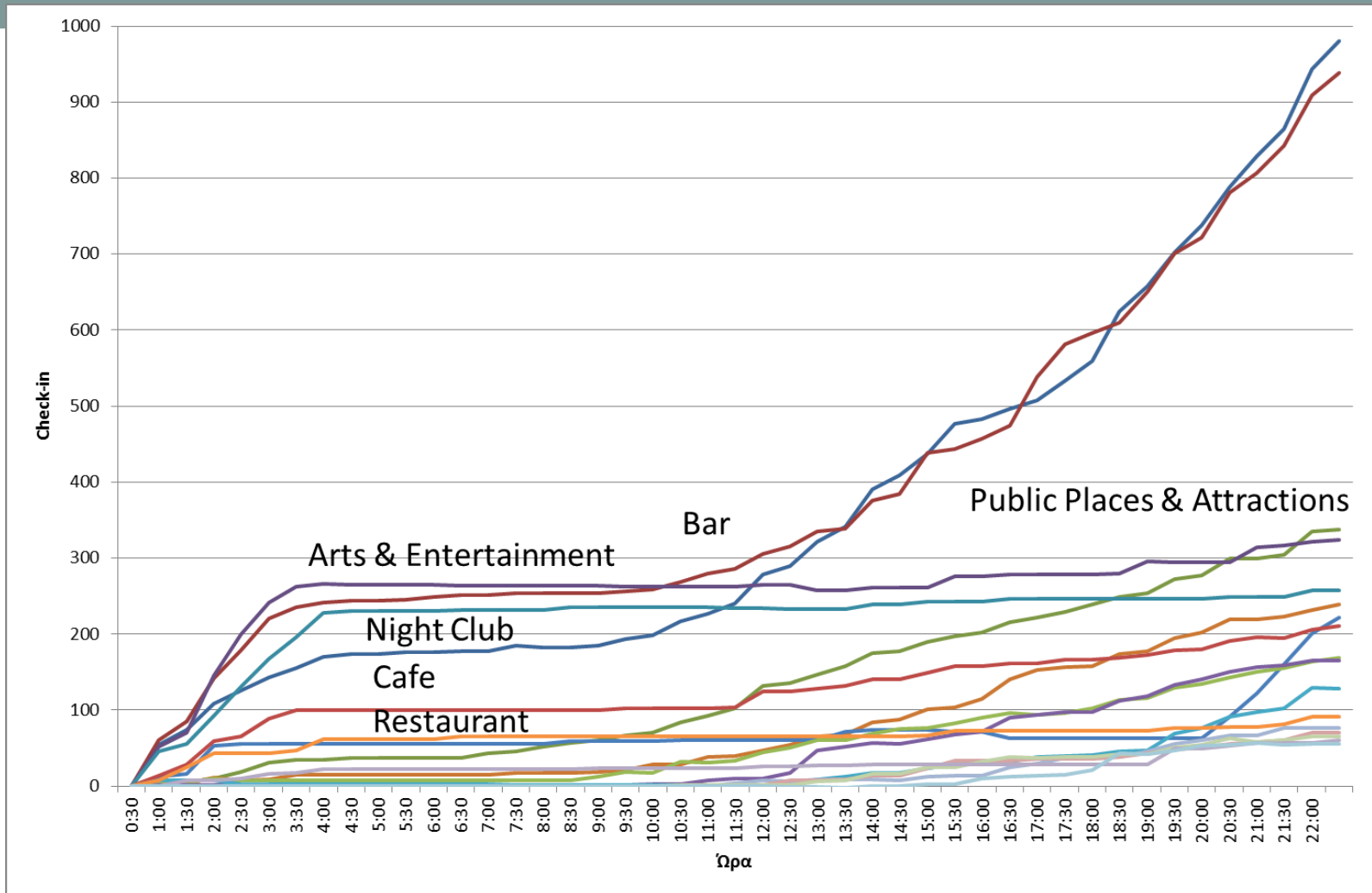


NIGHTLIFE



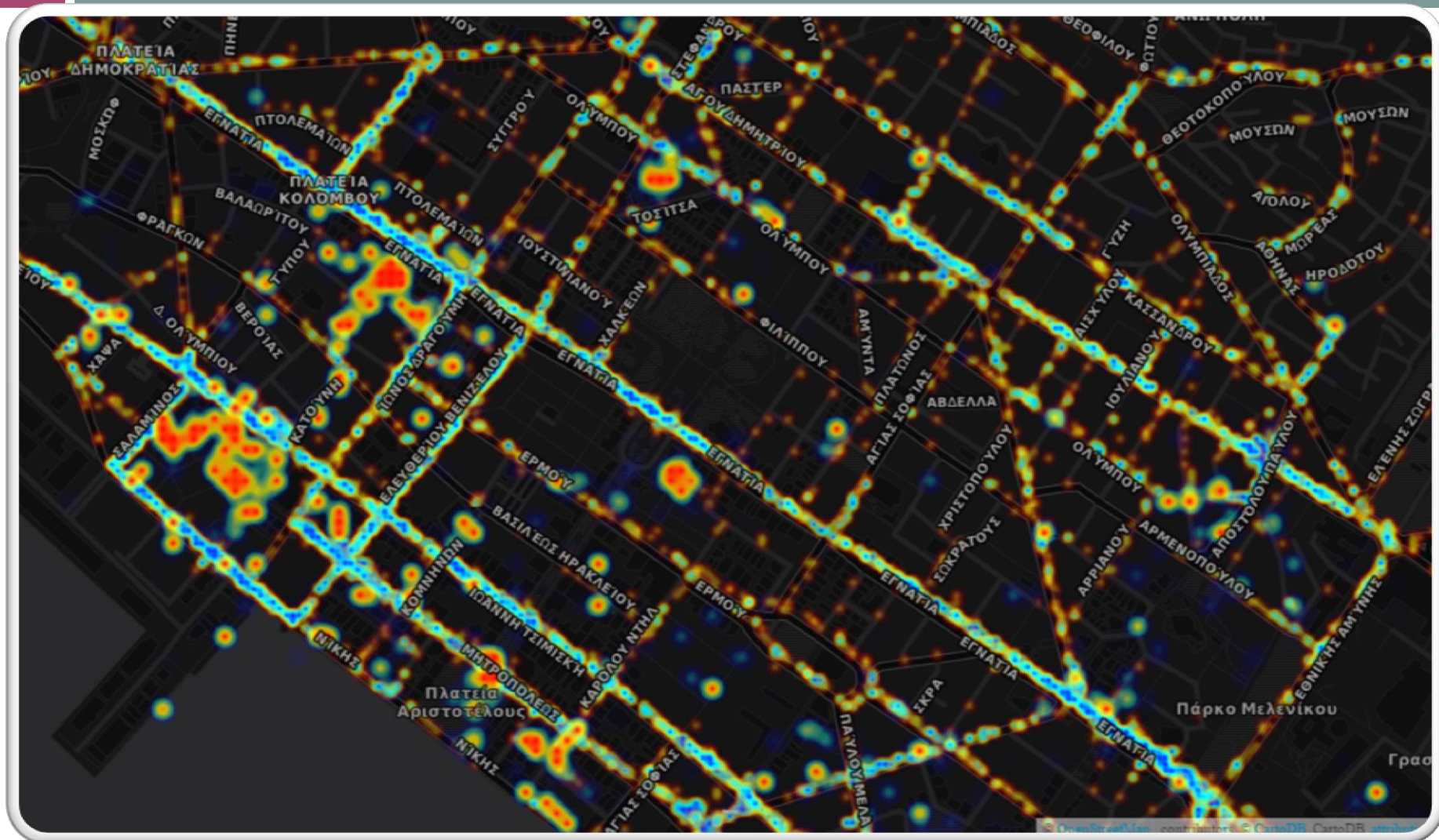


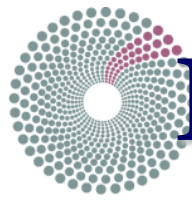
Social media



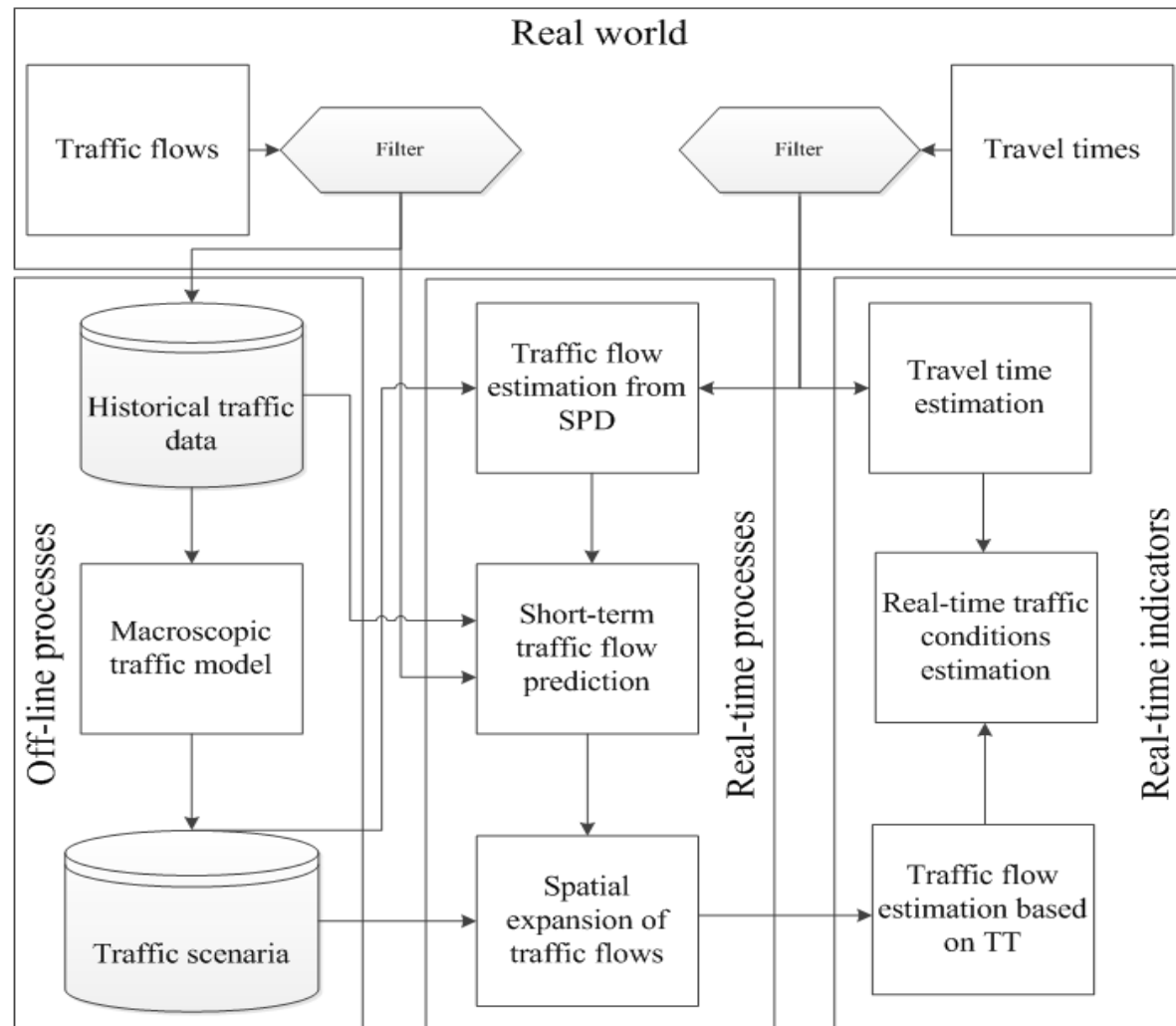


Social media





Real-time traffic conditions estimation





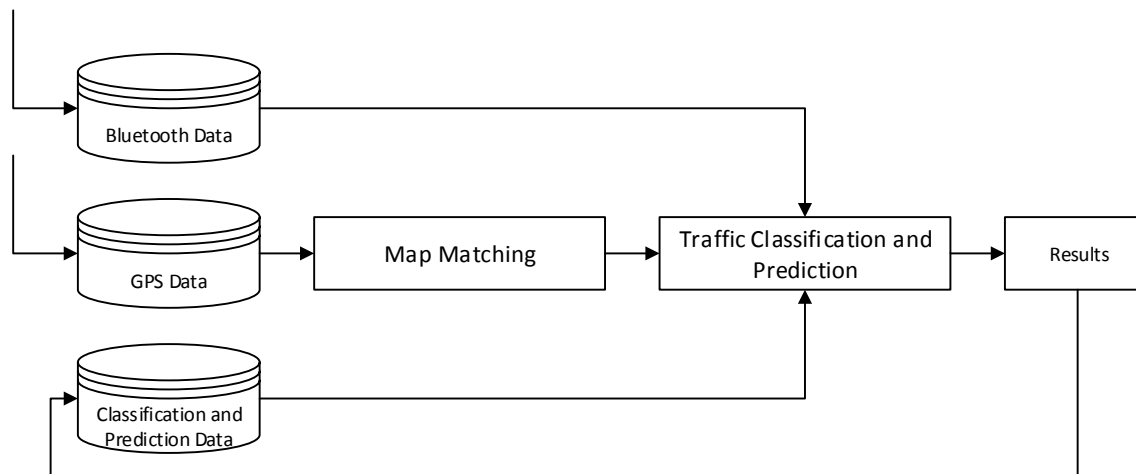
Processing of big data in Thessaloniki

- ⊙ Traffic flow estimation from stationary probe data
- ⊙ Travel time estimation using stationary probe data
- ⊙ Travel time estimation using floating probe data
- ⊙ Traffic flow estimation based on travel time
- ⊙ Short-term traffic flow prediction
- ⊙ Spatial expansion of traffic flows
- ⊙ Real-time traffic conditions estimation



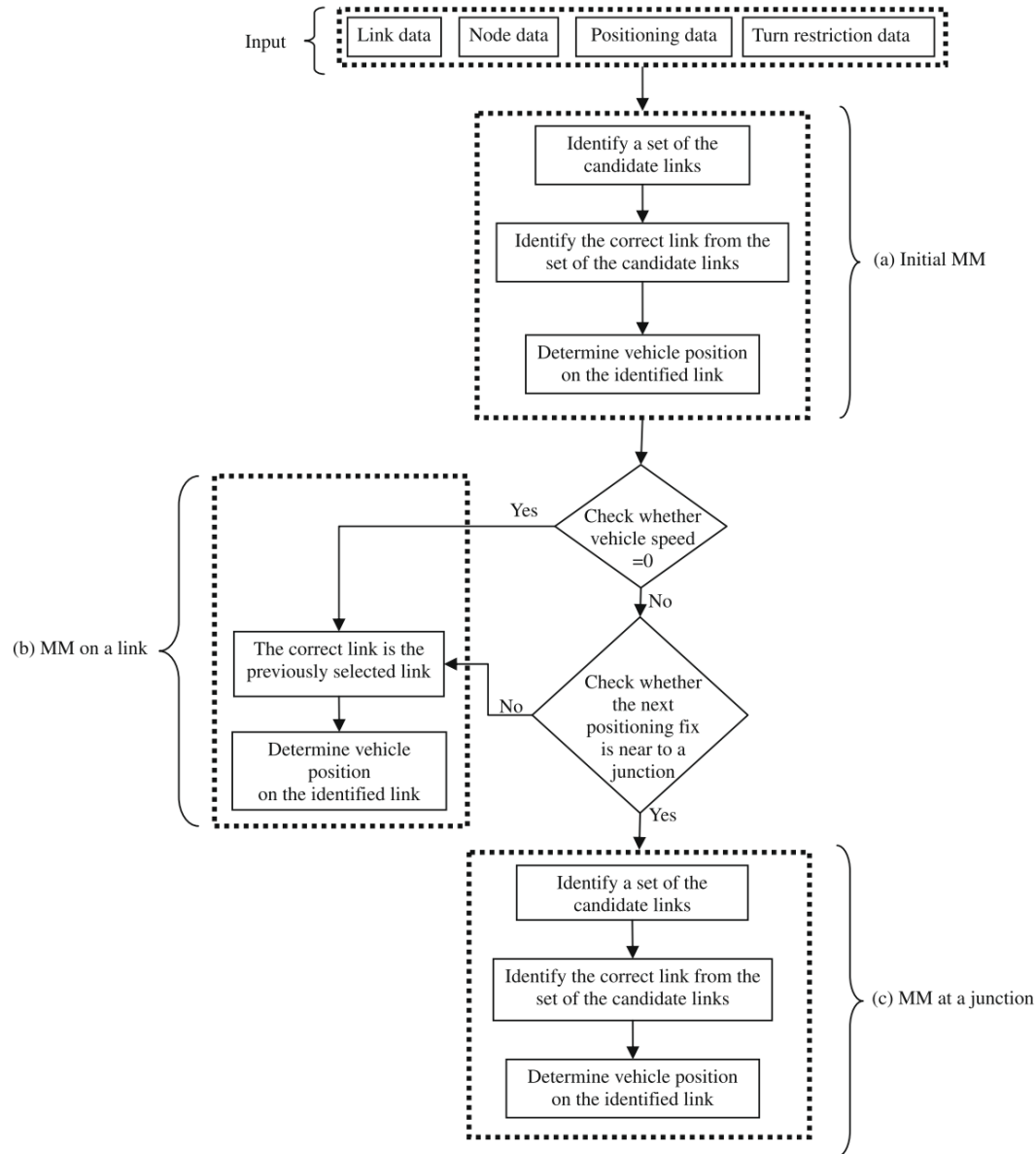
BDE pilot in Thessaloniki

- ◎ Probe data that is used
 - Floating Car Data (500-2.500 locations per minute)
 - Bluetooth detections (millions of daily detections in 43 locations)
- ◎ Services that are being implemented
 - Improved topology-based map matching
 - Mobility patterns recognition and forecasting



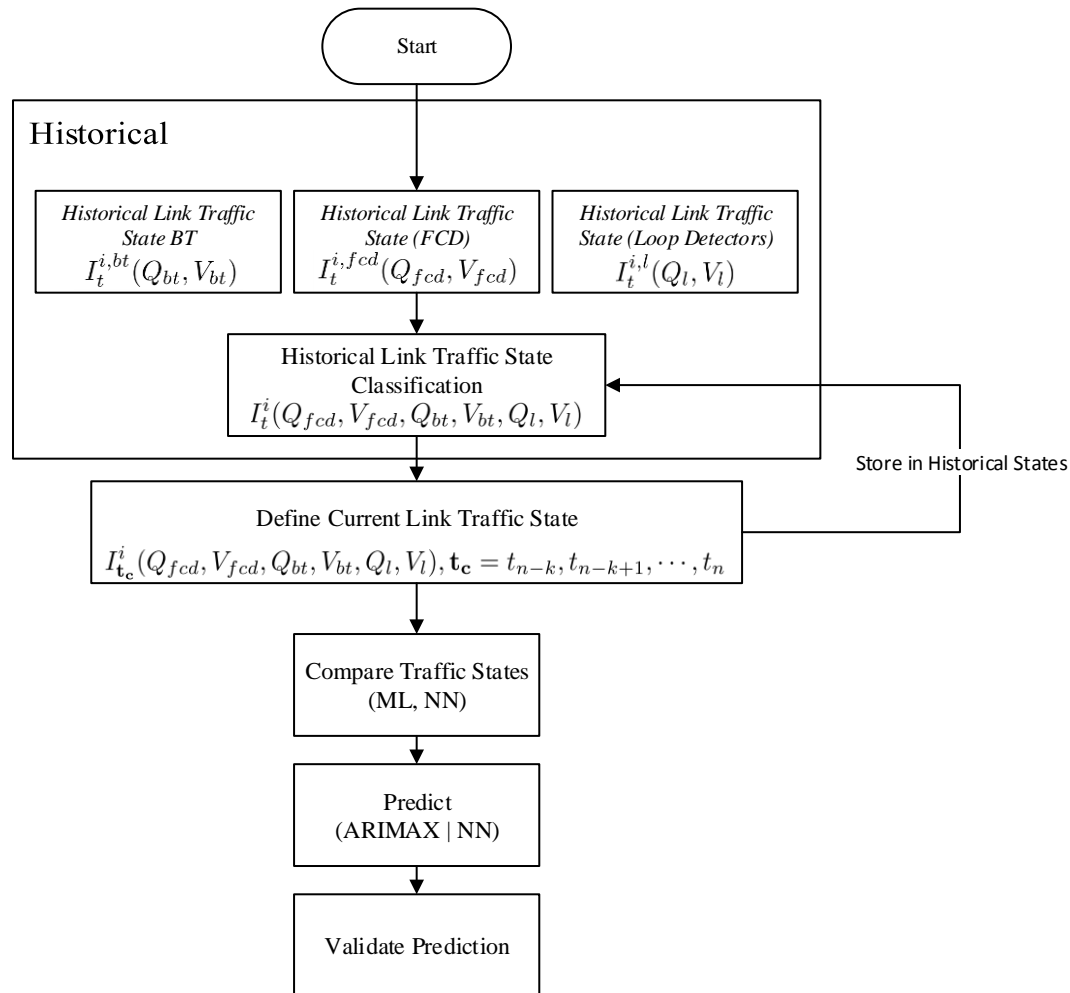


BDE pilot in Thessaloniki





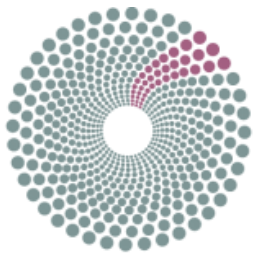
BDE pilot in Thessaloniki





BDE pilot in Thessaloniki

- ◎ Future plans (next 2 pilots)
 - Improve the 2 algorithms (historical data)
 - Replace the R components
 - Add the BT data source
 - Add other data sources (conventional and SM)
 - Include more datasets (PuT)
 - Use OSM data
 - Improve other processes (travel time estimation from BT)



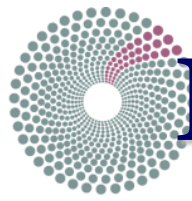
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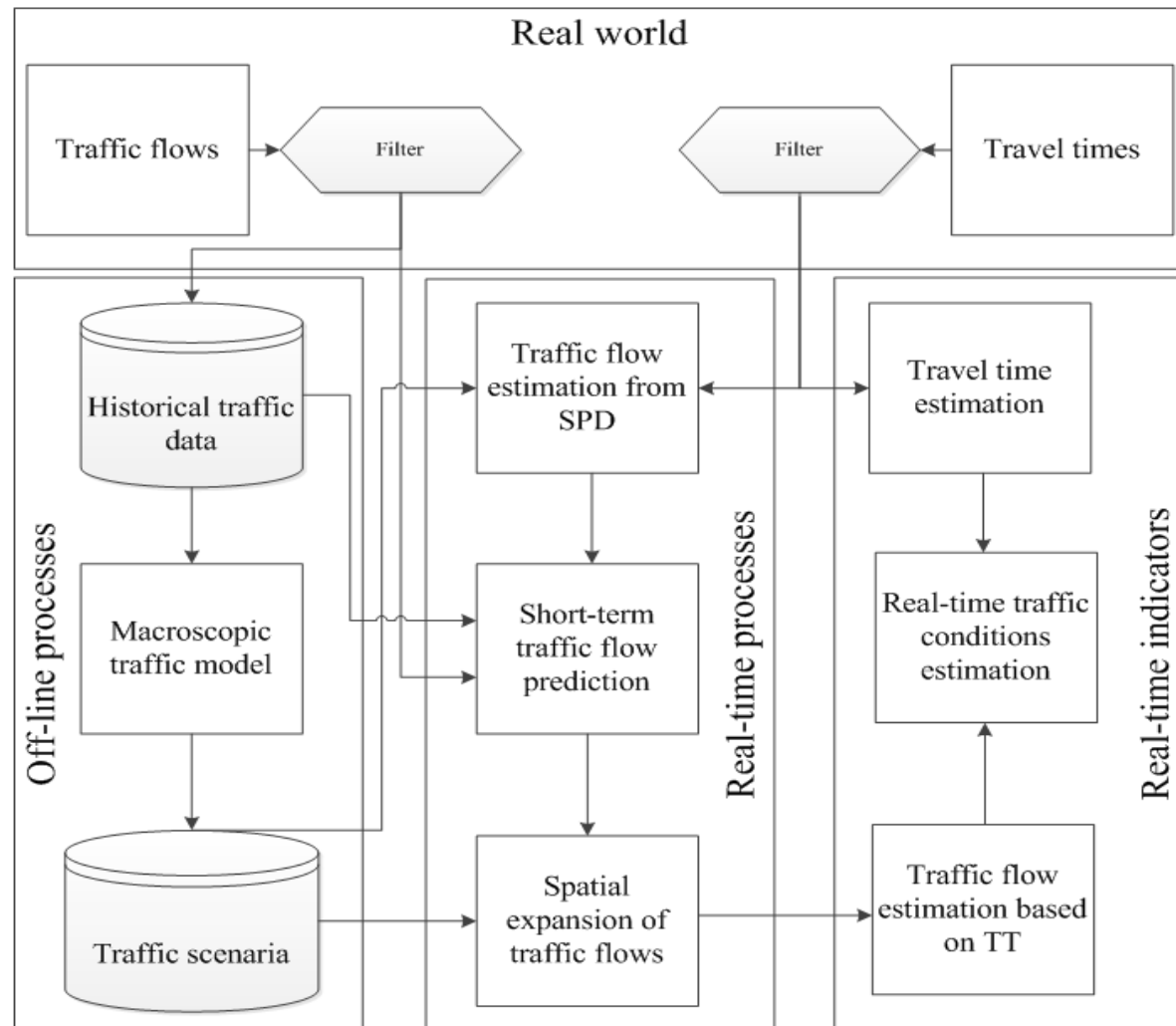


SESSION 2: TECHNICAL REQUIREMENTS AND ADDITIONAL TRANSPORT USE CASES

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Real-time traffic conditions estimation



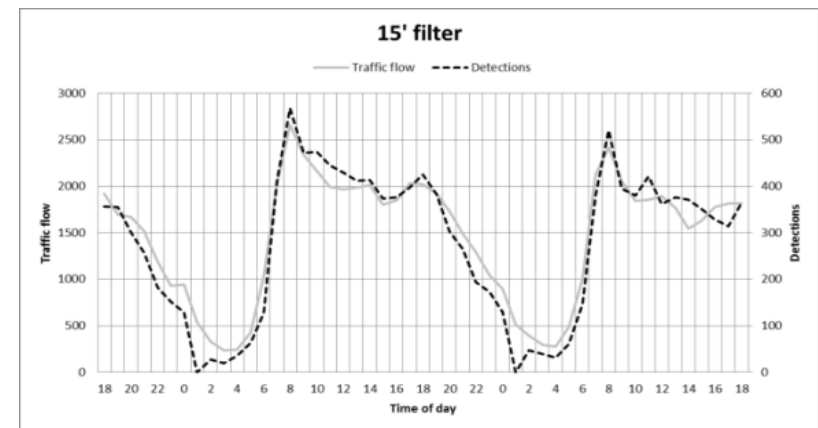
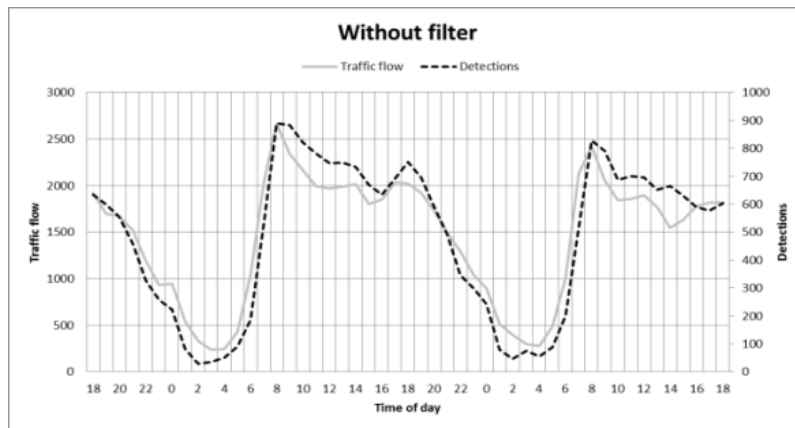
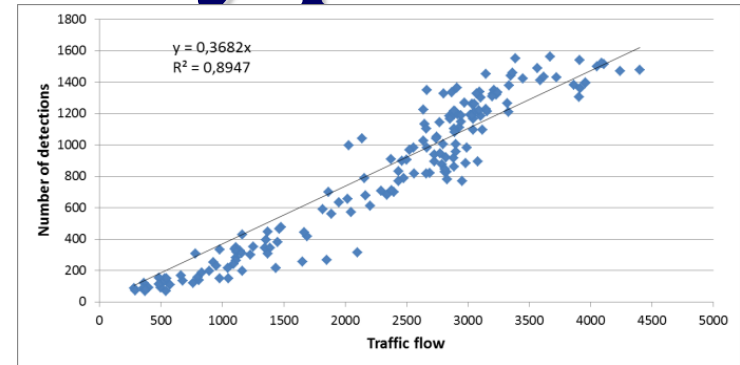
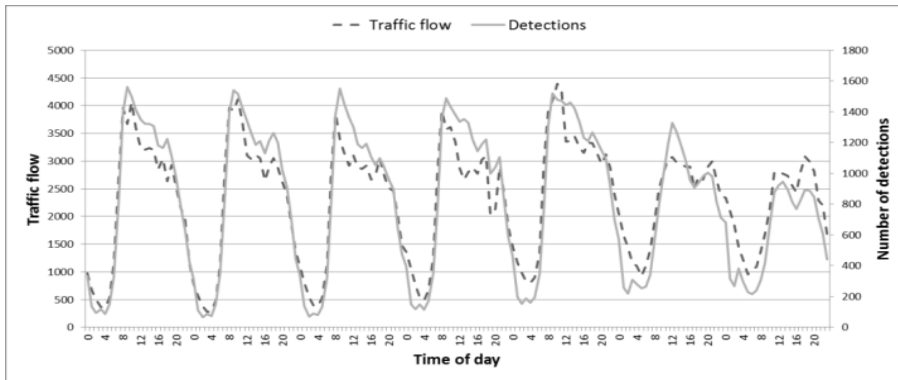


Processing of big data in Thessaloniki

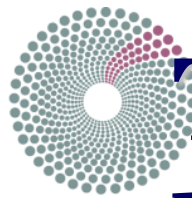
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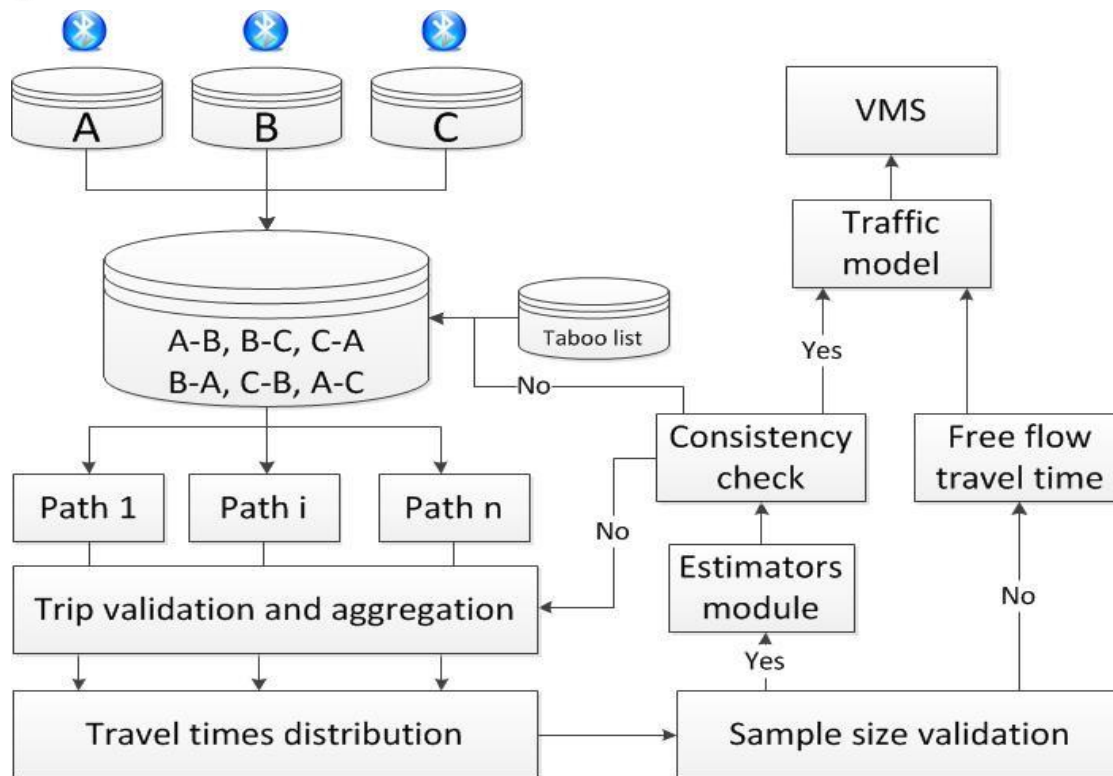
Traffic flow estimation based on stationary probe

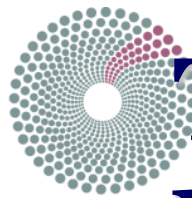


Time interval used for data filtering	Without filtering	5min filter	15min filter	60min filter
Correlation coefficient	0.3412	0.2179	0.1972	0.0442
R ²	0.9166	0.9193	0.9337	0.8594
Largest differences (absolute value and percentage ranges)	-401 / 623 -26% / 75%	-410 / 437 -23% / 61%	-336 / 389 -22% / 57%	-536 / 767 -35% / 79%

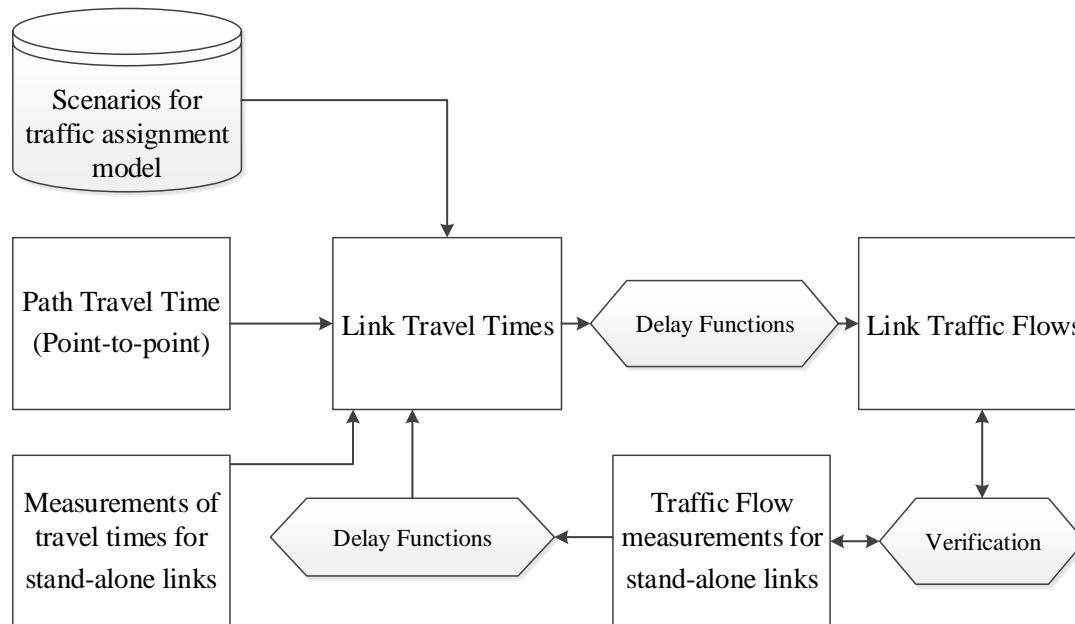


Travel time estimation based on stationary probe data





Traffic flow estimation based on travel time

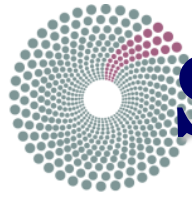


Conversion from route travel time to link travel time

$$\min \delta_1 * \sum(A * x - b) + \delta_2 * \sum((x - v_0)/v_0) \quad (1)$$

$$\text{s.t. } x_i > t_{0i} \forall i \in I \quad (2)$$

$$x_j = t_j \forall j \in J \quad (3)$$



Short-term traffic flow prediction

Linear autoregressive (AR) model

$$\varphi_k^{i+1} = \bar{\varphi} + \sum_{j=1}^N \beta_k^j * (\varphi_k^{i+1-j} - \bar{\varphi}) \quad (4)$$

Spatial expansion of traffic flow

Data Expansion Algorithm (DEA, [Lederman and Wynter 2009])

$$\text{Optimize } (D - p.* C_n) * x - (p.* C_c * c) \quad (5)$$

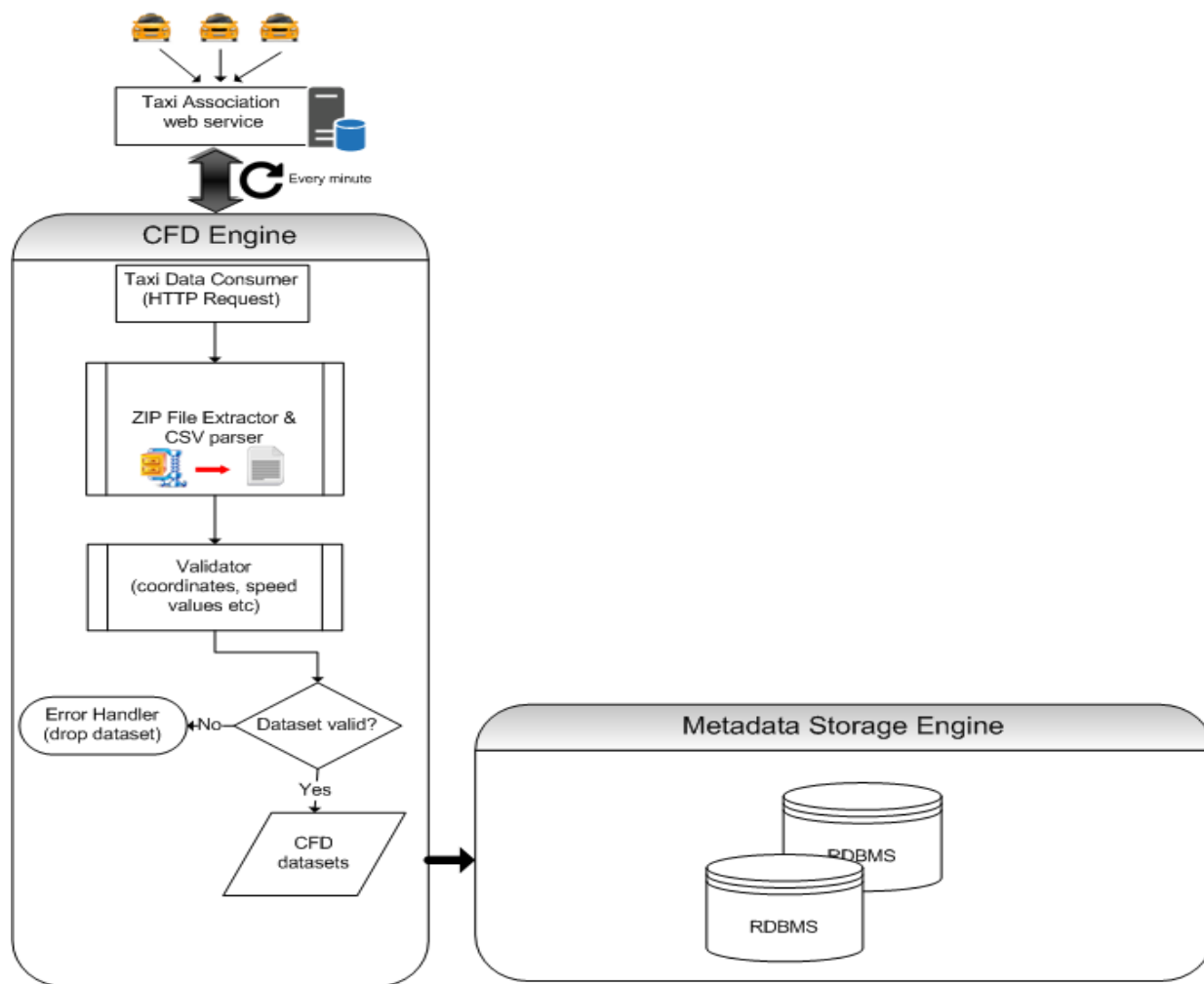
$$\text{s.t. } x_i > lb_i \forall i \in I \quad (6)$$

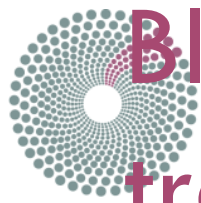
$$x_i < ub_i \forall i \in I \quad (7)$$

$$x_j = f_j \forall j \in J \quad (8)$$

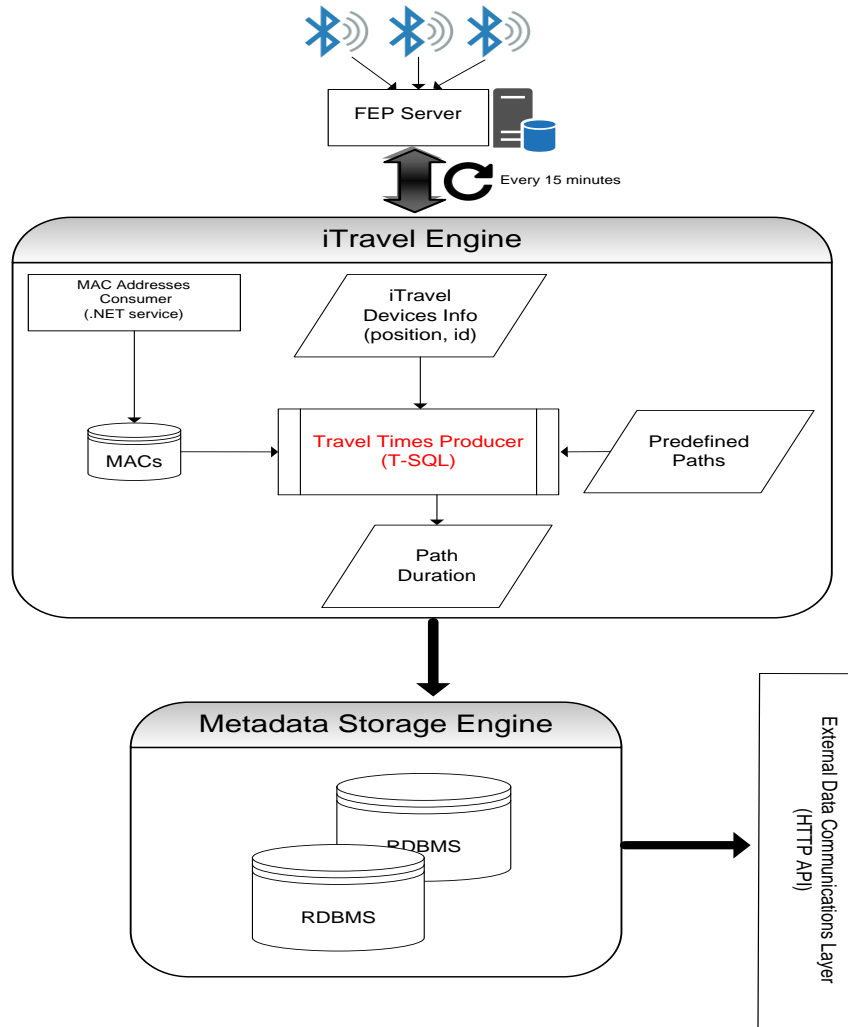


FCD data and average speeds on the road network



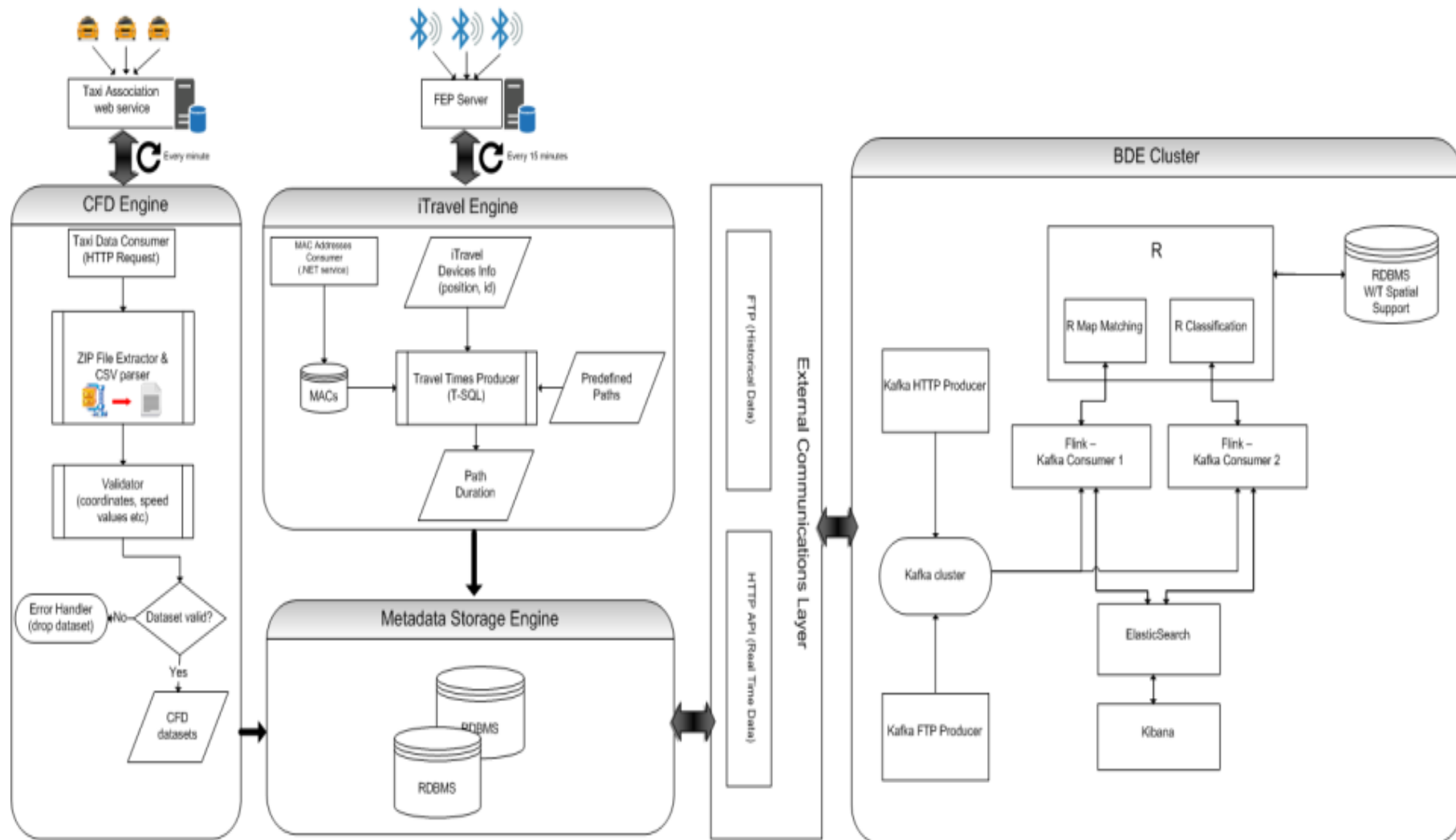


Bluetooth Sensors data and estimated travel times on the road network



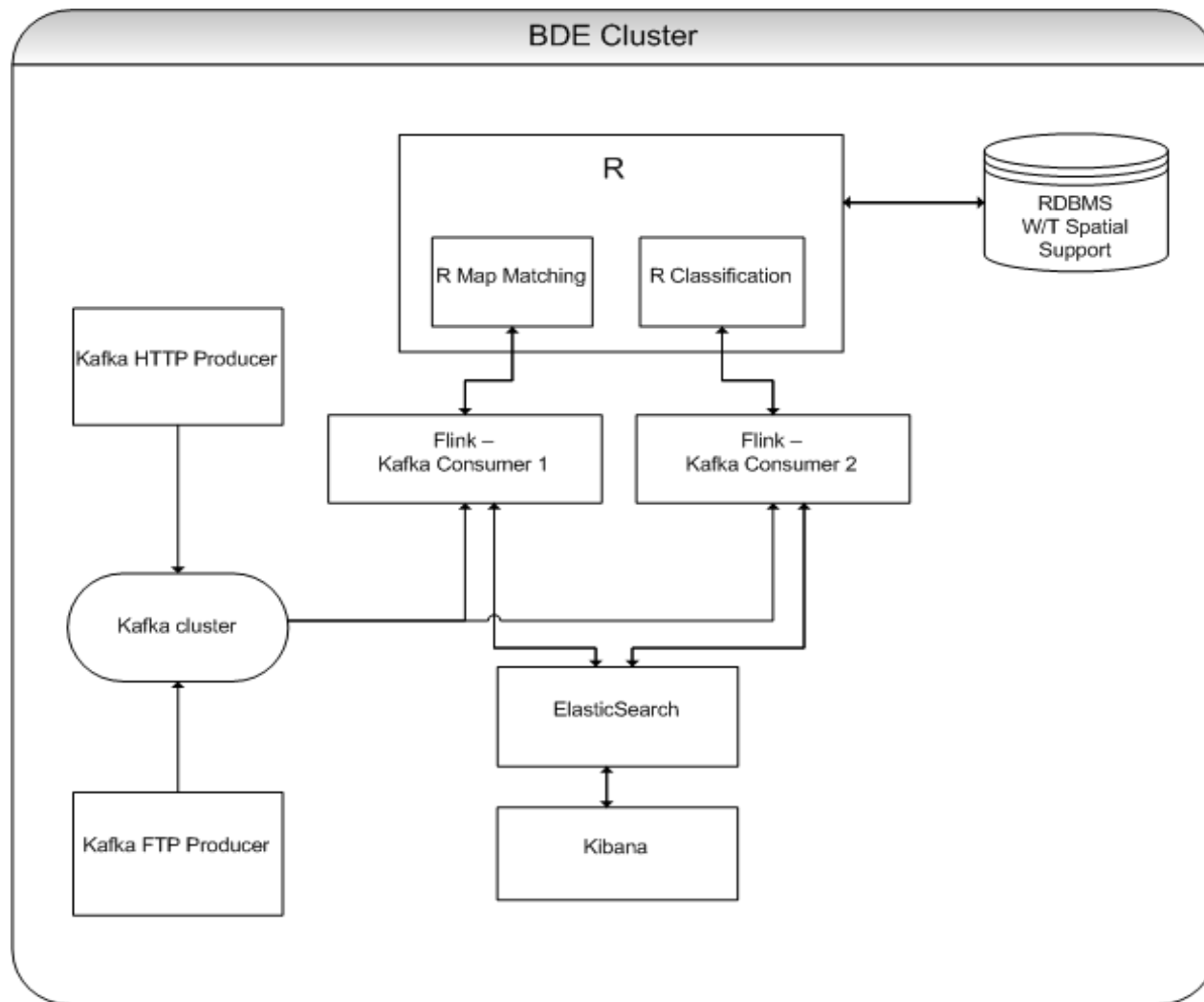


BDE Components integration with the legacy system





BDE Components





QUESTIONS

- ⊙ What are the pros and cons of the technical implementation of the platform offered by BigDataEurope?
- ⊙ How easy is to implement it to transport use case?
- ⊙ Lessons learnt from the first pilot implementation?
- ⊙ How adaptable / usable is it?



QUESTIONS

- ⊙ Any non-technical barriers to be considered? (legal, open data)
- ⊙ Does the open data flow initiative pose any threat/opportunity?
- ⊙ In which transport use case can we reproduce the pilot?
- ⊙ Which are the characteristics of the transport data that had to be considered in the design of the architecture ?



QUESTIONS

- ⊙ Any non-technical barriers to be considered? (legal, open data)
 - Privacy (the driver IDs are modified every 24 hours)
 - Data owner is a private entity (we rely on their willingness to share the data)
 - Updated maps are needed (OSM can be a solution)
 - Telecommunication costs (in our case are covered by the private company since it is crucial for their professional activity)



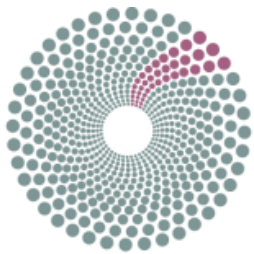
QUESTIONS

- ⊙ Does the open data flow initiative pose any threat/opportunity?
 - ++ data standardization
 - ++ data availability
 - ++ up-to-date datasets
 - -- data quality validation



QUESTIONS

- ① In which transport use case can we reproduce the pilot?
 - In any city having similar data sets
 - In other transport modes (PuT)



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<http://opendata.imet.gr/dataset>

itravel-traveltimes

Current travel times for selected paths

JSON XML CSV

fcd-compass4d

Floating car data along 2 arterials (zones)

JSON XML CSV KML MAP

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