

Université 2020

Le numérique au service de l'environnement

lab
recherche
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VINCI | PARISTECH



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Digital Traces & Mobility Computing

Studies of Floating Car Data mining

| Université 2020 – Le numérique au service de l'environnement, Novembre 2020



Agenda

Context and objectives

Study 1: Vehicle usage typology identification

Study 2: Individual significant place recognition

Study 3: Travel time estimation

Study 4: Territory functional zone discovery

Introduction

A Ph.D. thesis (2019-2021 @LVMT, ENPC, supervised by Fabien LEURENT and Xiaoyan XIE)

"FCD Mining to Feature out Mobility Patterns: Individual-Centered and Place-Based Analyses"

Context

- Understanding human movements -> critical in resolving mobility issues
- Massive digital-trajectory data -> Potentials for mobility insights
 - ❑ Fundamental knowledge of mobility entities all day long (vehicle & pedestrian)
 - ❑ A variety of sources: GSM(Mobile phone data), GPS (Floating Car Data), AFC (Smart cards, Navigo etc.)

2 key issues

- Mobility pattern analysis
 - Trajectory data mining
- ➔ Up-to-date solutions

Traditional data:

- Source: surveys (EGT), loop detectors, camera etc.
- Shortcomes: inherently limited for deploying

Trajectory data

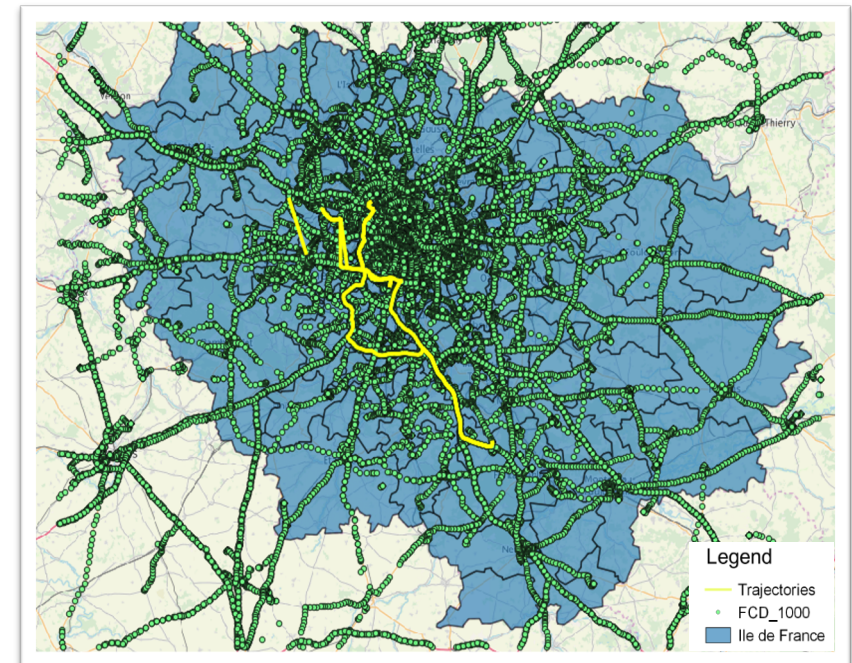
New Digital Alternative **Floating Car Data**

= GPS vehicle traces

Timestamped localizations and vehicle speeds (per 30s)
(data source: Coyote)

A cost-effective data solution

- Wide coverage (temporally 7*24 and spatially large-scale)
- Rich information along the path (intermediate points)

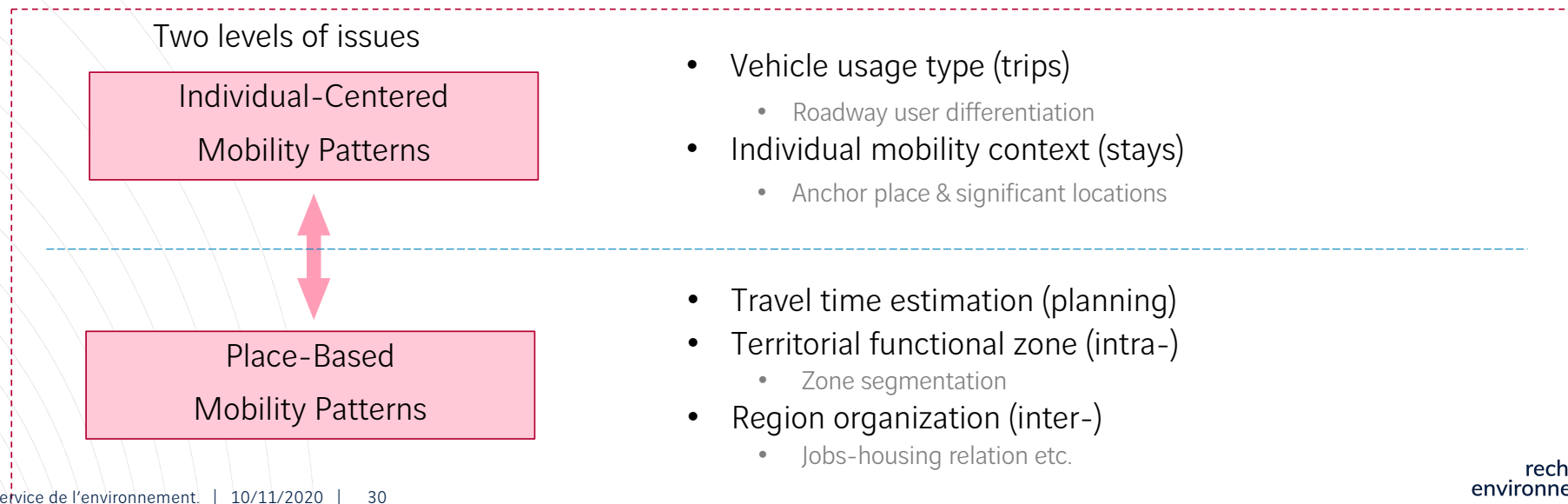




Overall objective

Study mobility patterns by FCD mining

- **What is “mobility patterns”:**
 - Recurring forms of human movements
- **What to analyze:**
 - urban dynamics : 1) travel demand of people ↔ 2) spatial configuration
- **Specific issues and aims:**





Case studies and key results



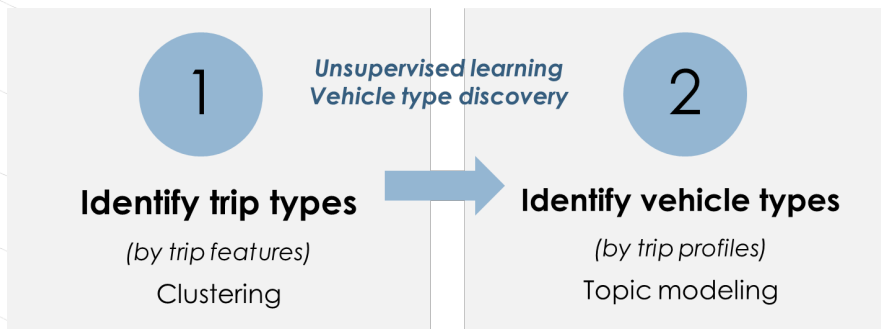
Study 1: Vehicle usage typology identification

Publication: Danyang Sun, Fabien Leurent, Xiaoyan Xie (2020). *Floating Car Data mining: Identifying vehicle types on the basis of daily usage patterns*, Transportation Research Procedia. DOI:10.1016/j.trpro.2020.03.087

Objective: • Identify vehicle usage type by mobility making • User segmentation and differentiation

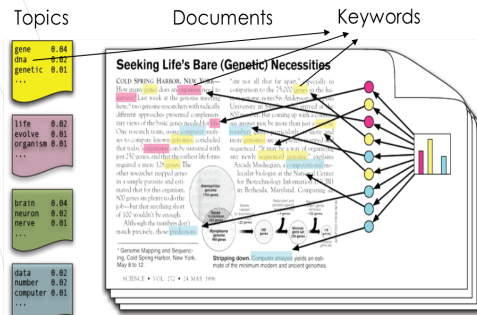
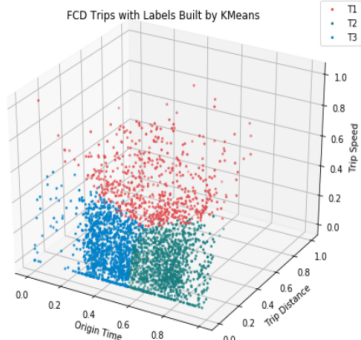
Approach: machine learning techniques

Outcome : 68,613 vehs with 196,554 trips; 2.9 trips/veh (IDF one day)

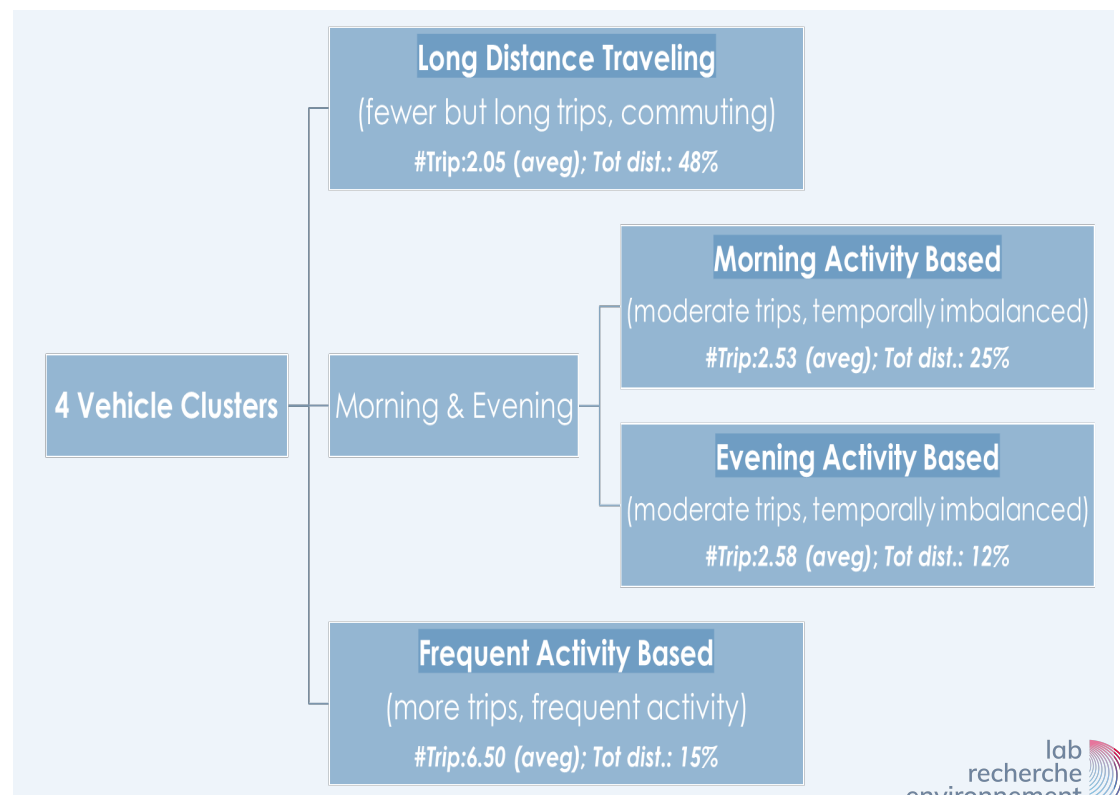


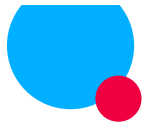
Clustering-> trip clusters

Topic Modeling-> usage "topics"



(Source: Rhody, L. M. (2012))





Study 2: Individual place recognition and significant places

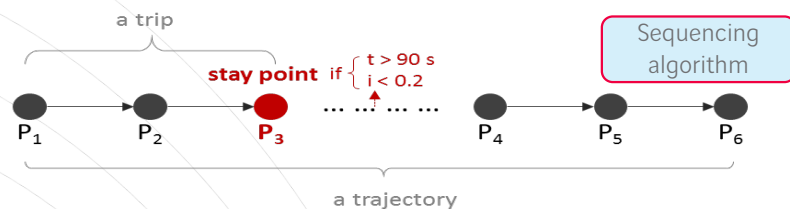
Conference communication: Danyang Sun, Fabien Leurent, Xiaoyan Xie. *Mining Vehicle Trajectories to Discover Individual Significant Places: Case Study Using Floating Car Data in the Paris Region*, Accepted for Transportation Research Board Annual Meeting, 2021, Washington DC

Objective: Mine individual mobility context and "anchor" places (homeplaces, workplaces, etc)

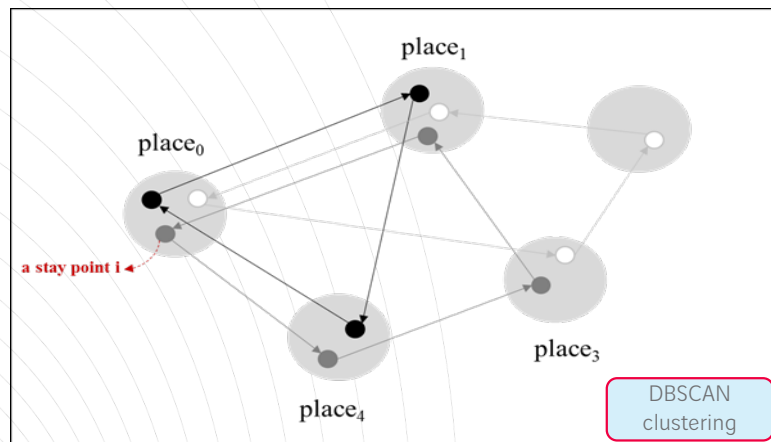
Approach: trajectory data mining

1) Extract significant-place locations

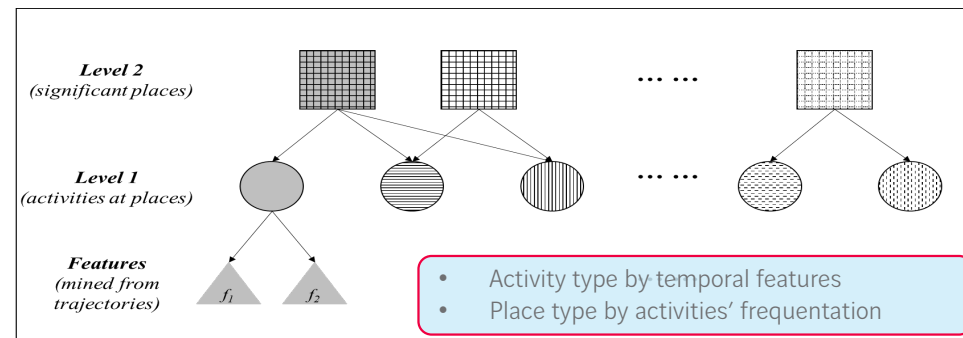
- Trajectory processing: stay points detection (stops for activities)



- Place recognition: adjacent stay points -> functional place

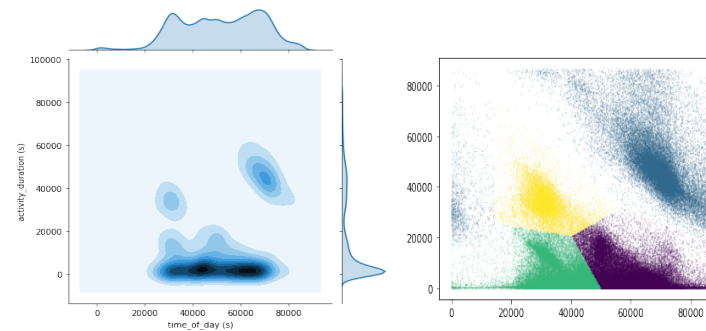


2) Identify Significant-place type by visits' frequentation and duration



Outcomes: Hierarchy of places

Homeplace & Workplace, Secondary places, Other places



c0	0.05	0.01	0.11	0.76	0.02	0.06	0.73
c1	0.07	0.01	0.08	0.1	0	0	0.05
c2	0.46	0.04	0.37	0.08	0.02	0.03	0.03
c3	0.06	0.01	0.07	0.05	0.15	0.07	0.06
c4	0.03	0	0.08	0.06	0.01	0.39	0.12
c5	0.21	0.02	0.37	0.85	0.12	0.46	0.86
c6	0.1	0.44	0.07	0.05	0.01	0.01	0.02
	wk_d_earlyday_short	wk_d_earlyday_long	wk_d_lateday_short	wk_d_lateday_long	wk_n_earlyday_short	wk_n_earlyday_long	activity_tp



Study 3: Travel time estimation

Conference communication: Fabien Leurent, Danyang Sun, Xiaoyan Xie. *Roadway Travel Times: Maximum Likelihood Estimation Based on Floating Car Data Intervals*. Accepted for 9th Symposium of the European Association for Research in Transportation, hEART 2020

- Objective:**
- Exploit massive observations from FCD to estimate local travel times
 - Compare with conventional method (point-wise average: currently widely used in the industry)

Methodology: Stochastic model

pros:

- Probabilistic specifications
- A measure of reliability
(estimation with confidence intervals rather than unique values)

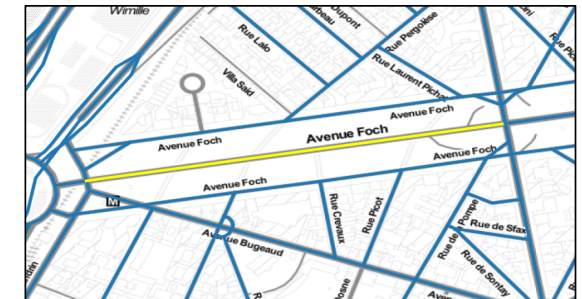
Achievements:

- obtained **more precise estimations with fewer data** by stochastic model
- resolved the underestimation of travel time problem in conventional methods

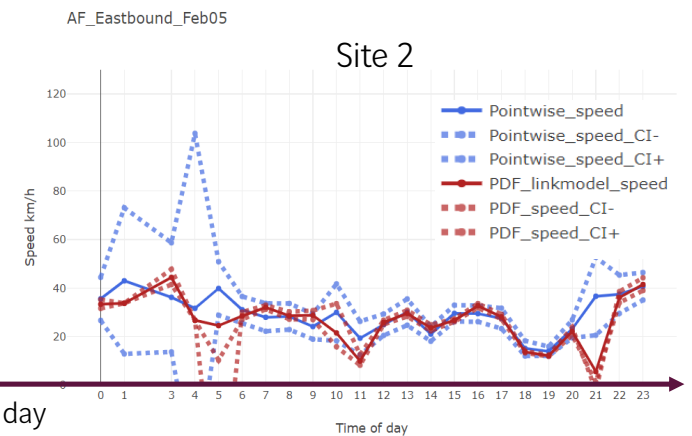
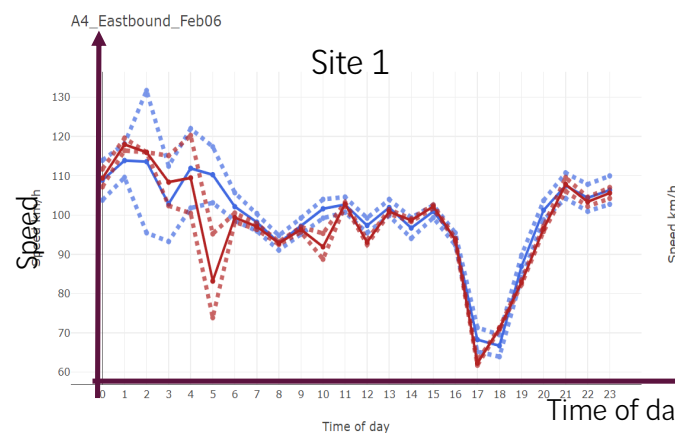
Outcomes



Site 1: A4 segment (Highway)



Site 2: Avenue Foch (Urban)





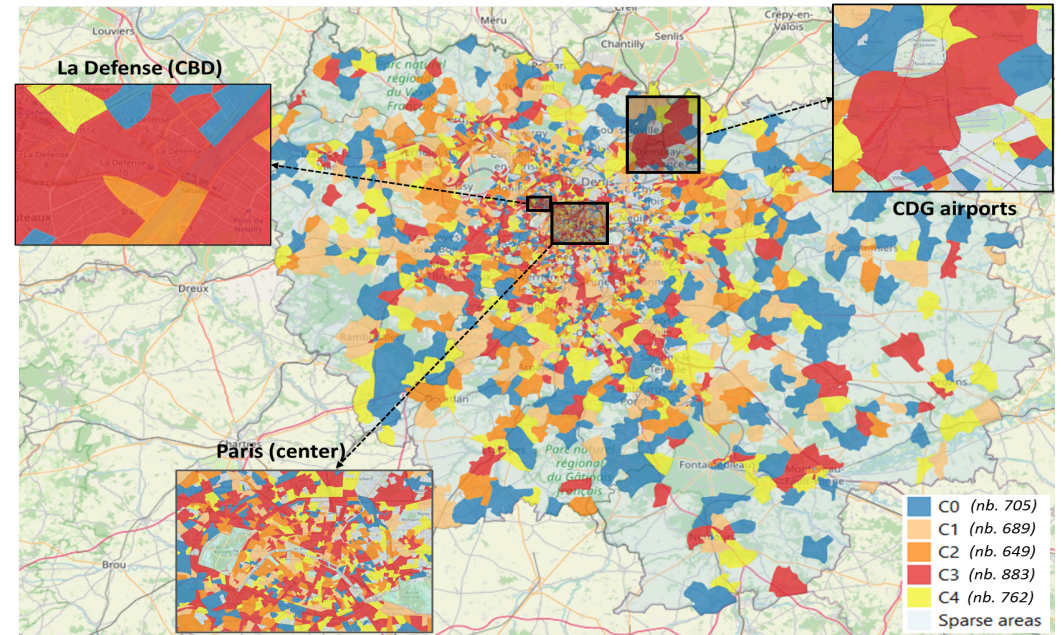
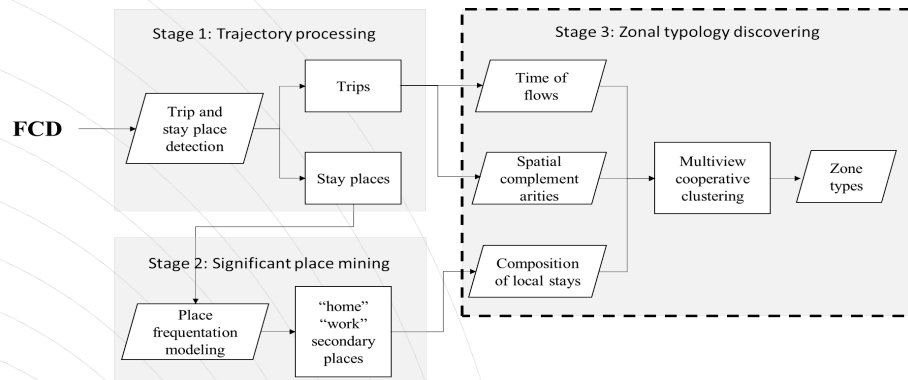
Study 4: Territory functional zones discovery

Publication: Danyang Sun, Fabien Leurent, Xiaoyan Xie. *Uncovering mobility typologies of territorial zones based on Floating Car Data mining*, Transportation Research Procedia. (in press)

Objective: Discover territory functional divisions by vehicle movements

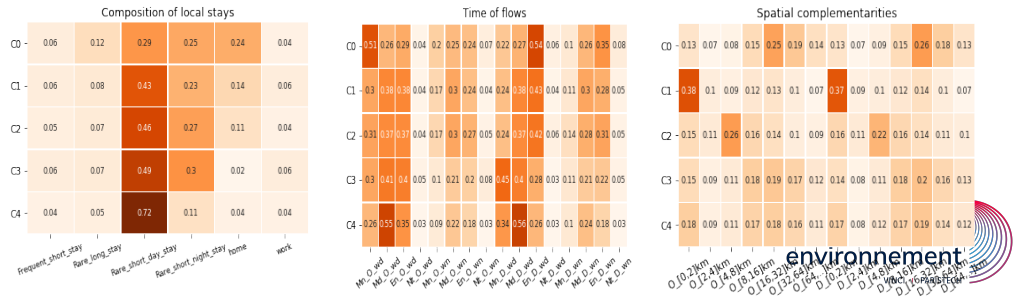
Approach: machine learning techniques

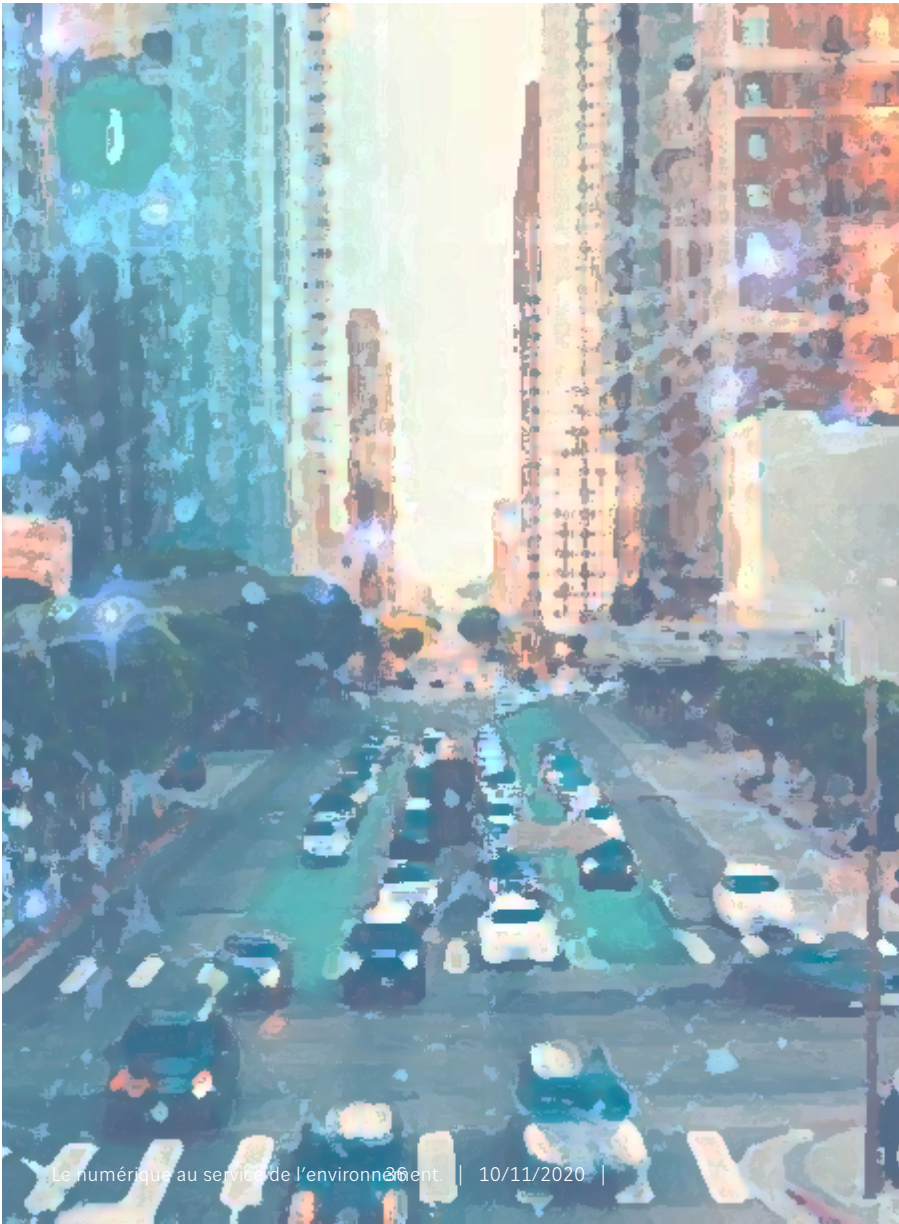
- Retrieved zonal usage attributes
 - traffic flow
 - accessibility
 - building functions
- Discover functional zone by cooperative clustering



Outcomes: 5 major types of zones in IDF (by vehicle mobility)

- C0: Residence oriented areas
- C1: Amenity-residence mixed areas (local-accessible-based)
- C2: Amenity-residence mixed areas (intermediate-range-accessible-based)
- C3: Business/employment oriented areas
- C4: Day-time mobility-oriented areas





Potential contributions

Data driven framework for automated mobility analysis

- Up-to-date and large-scale

Prevision of mobility demand

- Human mobility regularity and predictability
- Aid in proposing alternatives (mode shifting etc.)

Mobility system diagnostic

- Traffic state: congestion, road network LOS evaluation
- Planning guidance: city structure and geographical correlations etc.