

IFP ENERGIES NOUVELLES (IFPEN) IS A MAJOR FRENCH RESEARCH AND TRAINING PLAYER

Sustainable
mobility

- Located in Paris and Lyon metropolitan area
- Provides solutions to challenges in energy and climate
- Active role in the context of the EU Horizon Framework Program
- Collaborative environment with academic and industrial partners
- Promoting training in the fields of energy, motor vehicle and environment



4 strategic sectors

Climate, environment and
circular economy

Renewable energy production

Sustainable mobility

Responsible oil and gas

3 priority areas in sustainable mobility

● Electric mobility:

- increase energy efficiency via electrification
- develop devices adapted to the needs such as battery, motors and converters

● Connected mobility:

- reduce the environmental footprint and develop soft mobility
- measure and treat land transport emissions to improve air quality

● Mobility with a low environmental impact:

- maximize powertrains energy efficiency
- implement solutions for CO₂ and pollutant emissions reduction

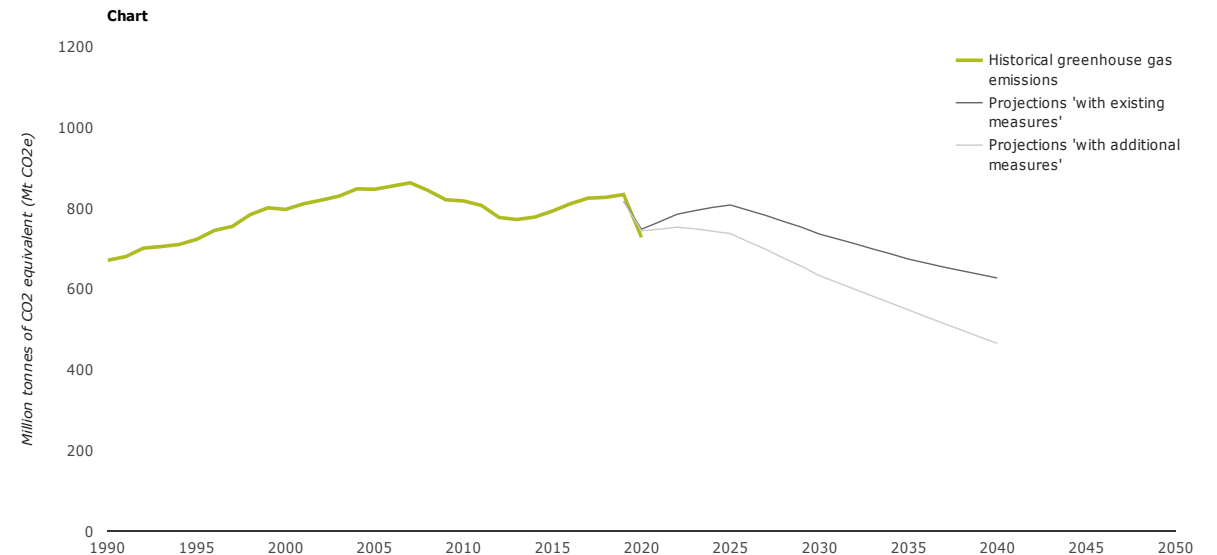
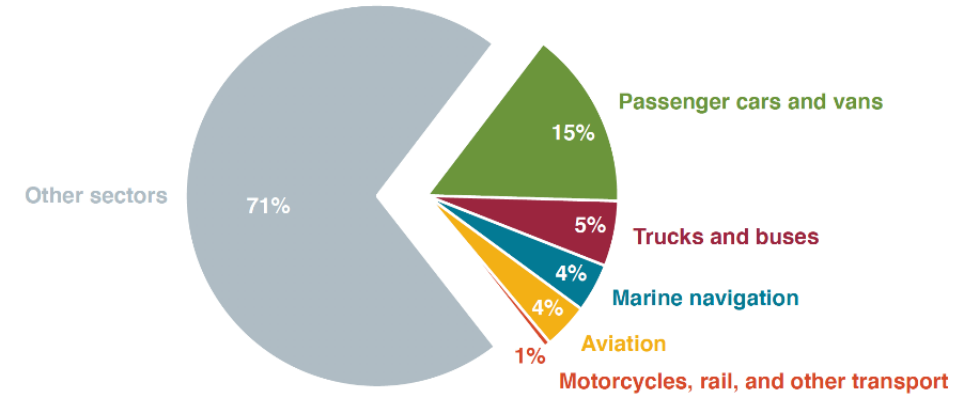
ENERGY-EFFICIENT OPTIMIZATION STRATEGIES FOR ELECTRIC AND CONNECTED MOBILITY

ANTONIO SCIARRETTA
ETH, 6 JUNE 2023

CONTEXT

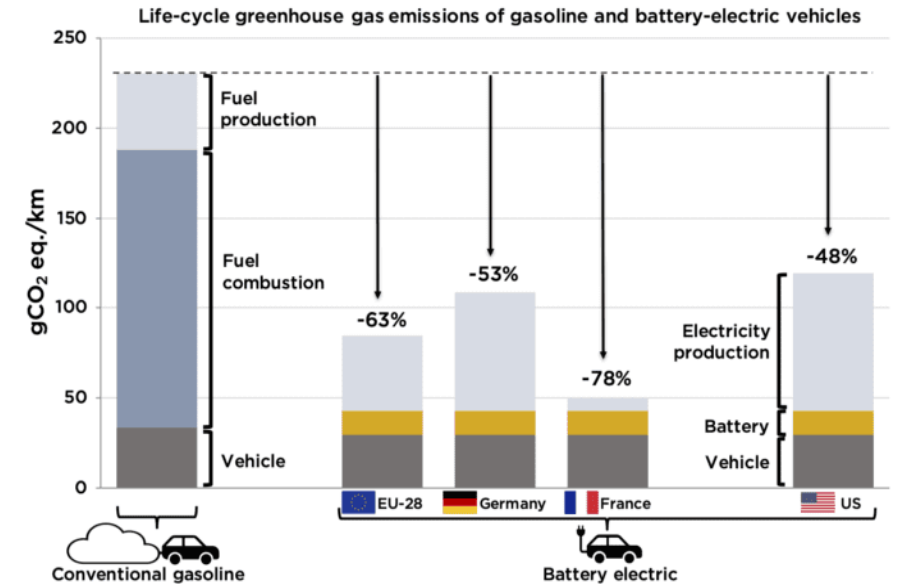
- Goal green deal: climate neutrality by 2050 and a 55% reduction of net GHG emissions by 2030, compared with 1990 levels (FitFor55 Feb 2023)
- Transport is the only sector in which emissions remain higher than 1990 → 90% reduction in transports is required
- Fit for 55: 100% ZEV (new cars and vans) by 2035 (55%-50% by 2030, compared to 1990)

Greenhouse gas emissions in the EU
2018 total: 3.8 Gt CO₂e

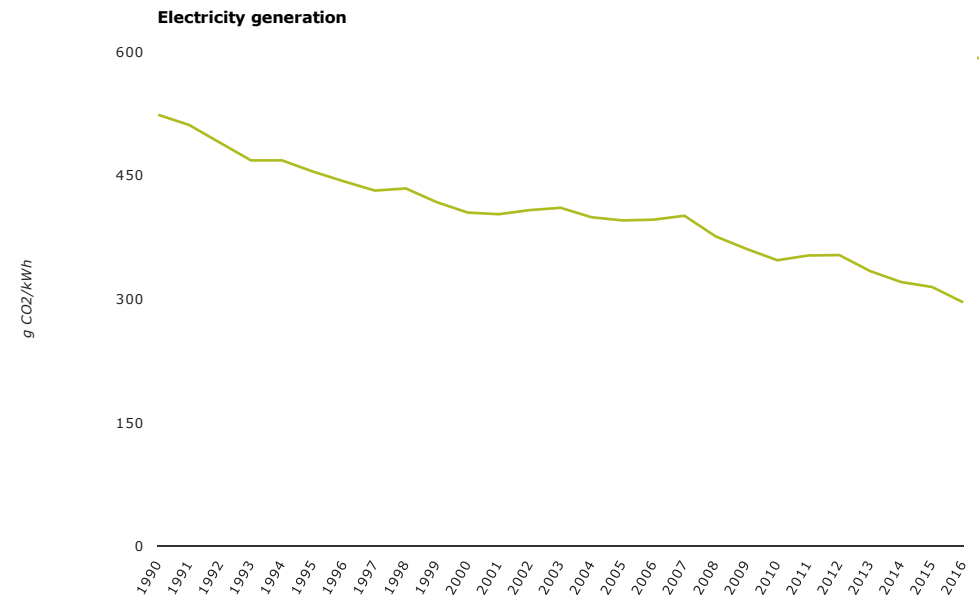


ELECTRIFICATION AND BEYOND

- Electric = 0% « taipipe » CO2 but non-zero upstream processes
- CO2 emissions from electricity generation is decreasing as EU has achieved 10,2% usage of renewables in transport (2021)
- But:
 - Strongly dependent on the country
 - Manufacturing, battery lifecycle must be considered as well
- Electric vehicles alone are not sufficient but require new mobility system and a change of habits



Source: ICCT, 2021



In addition to electrification, reducing grid-to-wheel consumption is necessary



$$E_W = \underbrace{\frac{1}{2} m_v (v_f^2 - v_i^2) + m_v g (z_f - z_i)}_{\text{« Useful » energy}} + \underbrace{m_v g C_{rr} (x_f - x_i)}_{\text{Design-dependent losses}} + \underbrace{\frac{1}{2} \rho_a A_{fr} C_D \int_0^{t_f} v^3(t) dt}_{\text{Driving-dependent losses}}$$

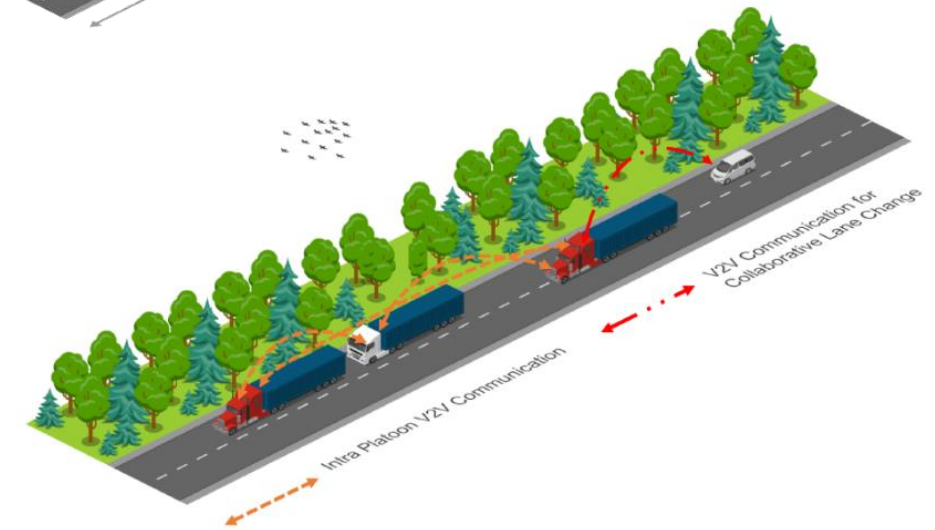
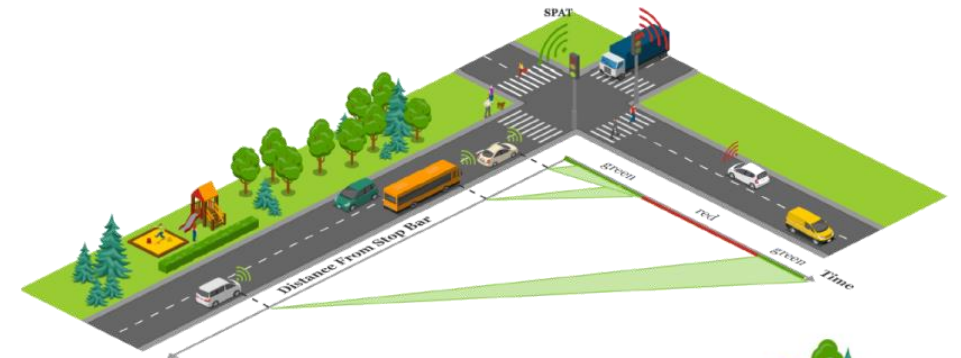
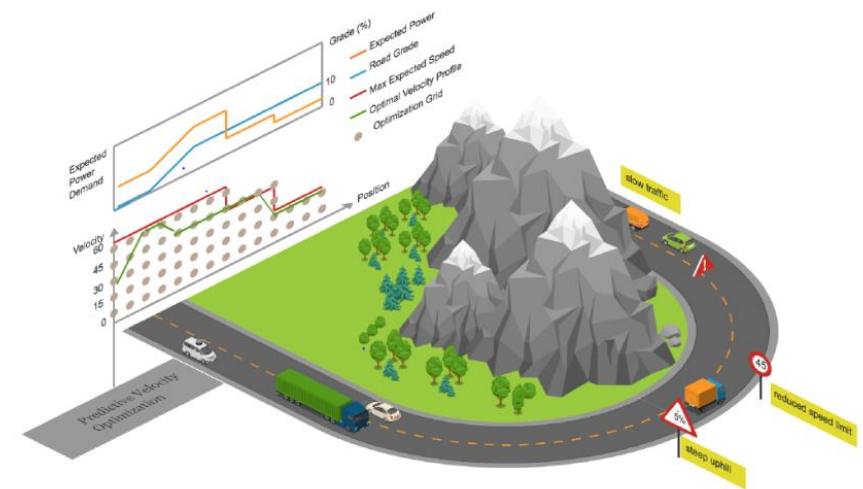
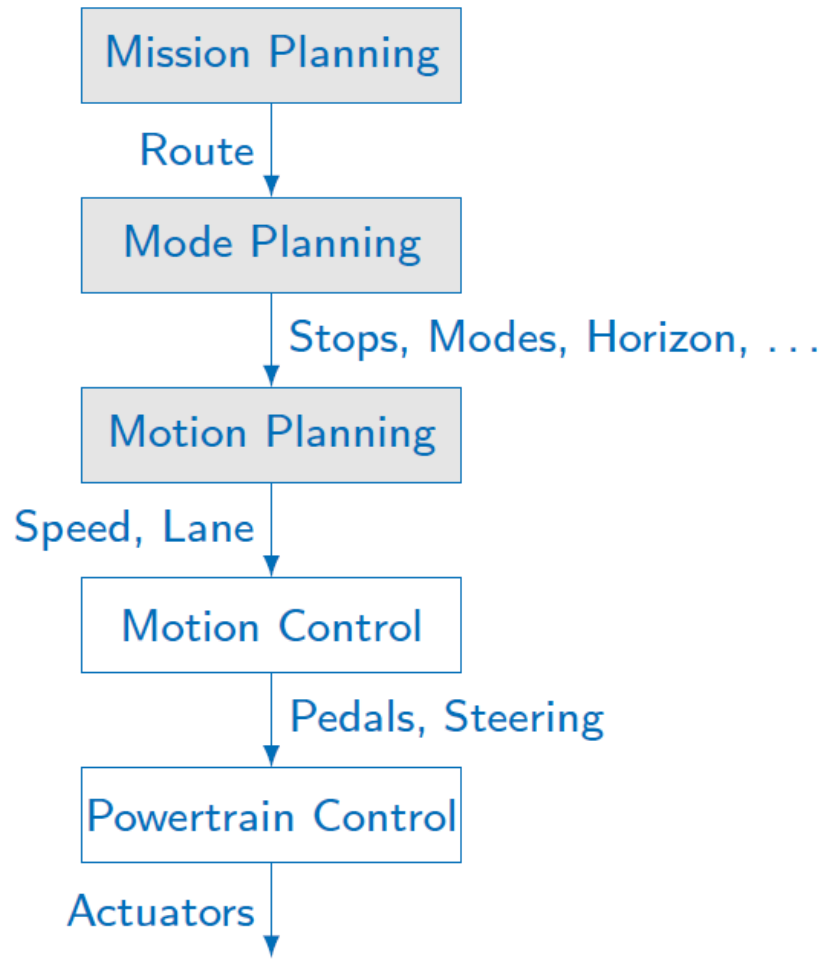
Energy at wheels =

« Useful » energy

Design-dependent losses

Driving-dependent losses

WHEEL TO WAY ENERGY MANAGEMENT AND CONNECTIVITY

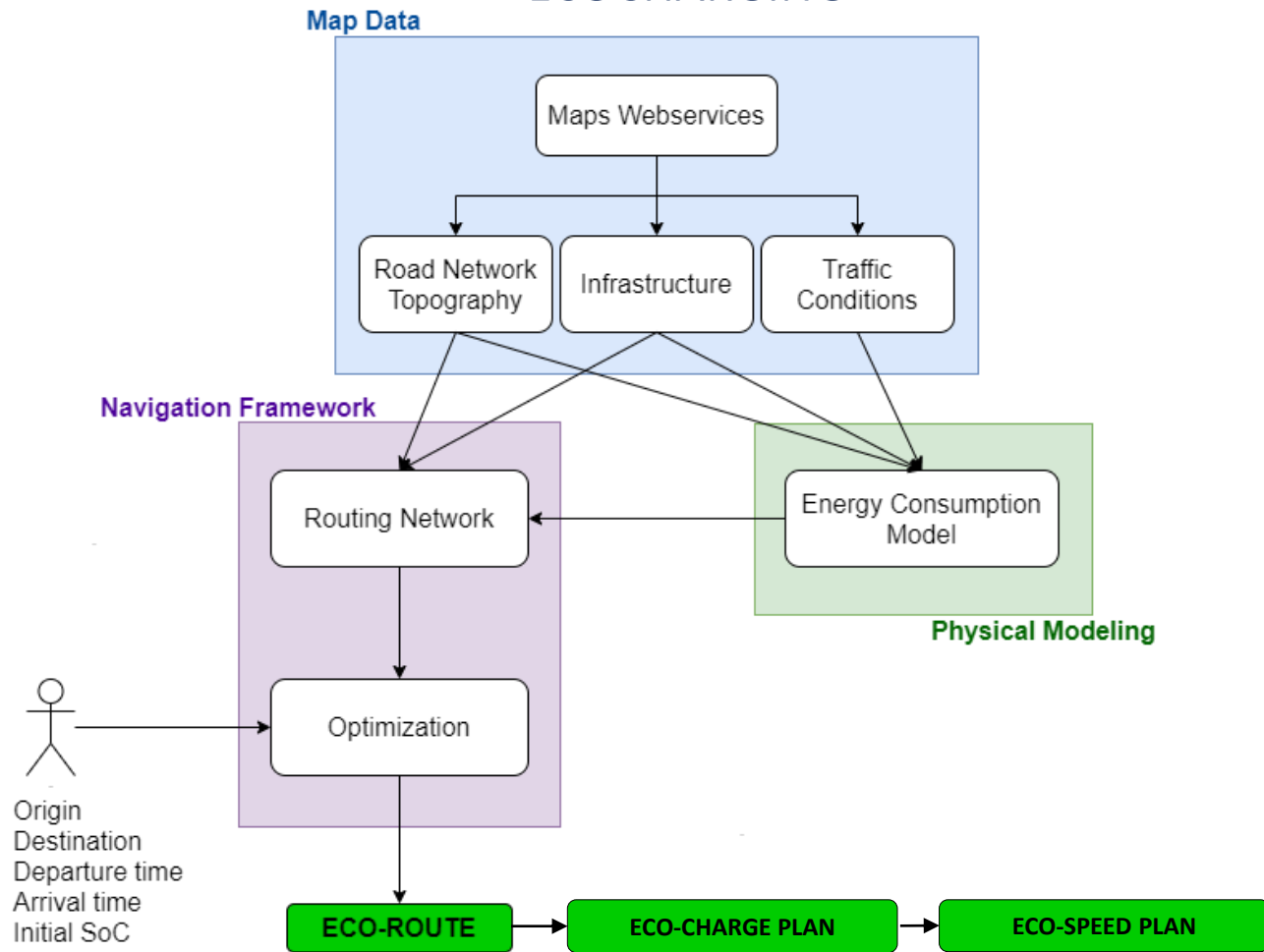


Logical scheme of planning and control layers in CAVs.

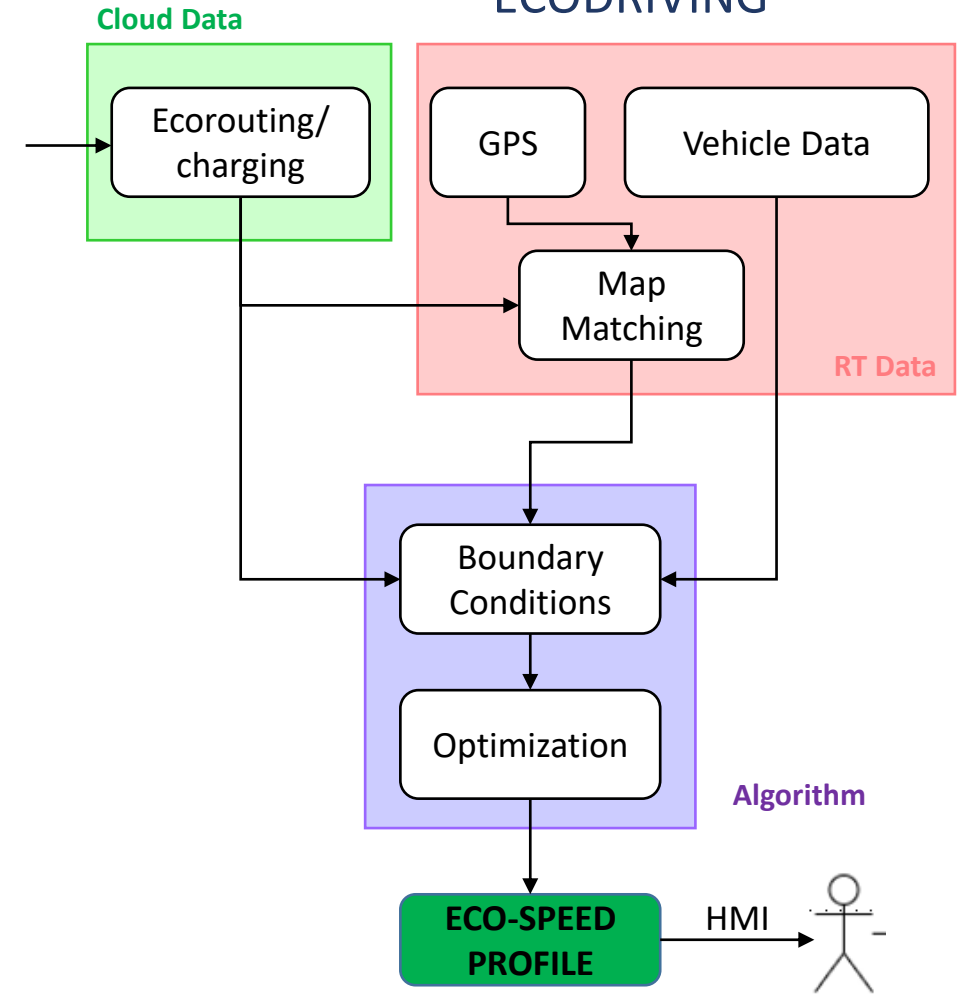
BACKGROUND

FROM ECOROUTING TO ECODRIVING

ECOCHARGING

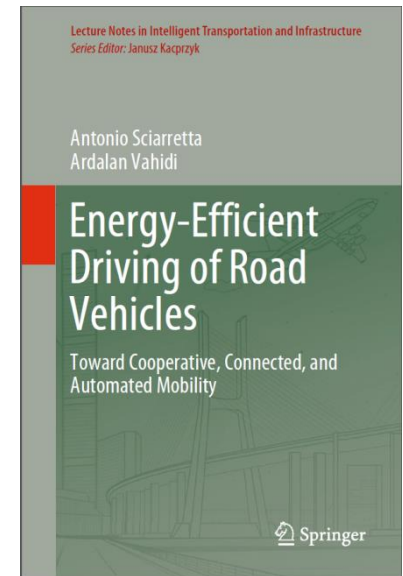


ECODRIVING



ACADEMIC COLLABORATIONS

- Chair ECAV « Chair on Electric, Connected, and Autonomous Vehicle for Smart Mobility”, IFP School, since 2019
- Monography « Energy-Efficiency Driving of Road Vehicles » (Springer) published in 2020
- Organisation of the workshop e³CAV (2 days, ~20 invited speakers) in 2019
- Collaborations with academic institutions
 - Mines Paritech, CAS (joint publication)
 - Ecole Polytechnique, CMAP (Ph.D. defended, joint publications)
 - INRIA, GIPSA-Lab (ongoing Ph.D., joint publication on traffic eco-management)
 - Clemson Univ. (scientific visitor 4 m, visiting Ph.D. 8 m, several joint publications)
 - Ohio State Univ., CAR (visiting Ph.D., joint publication)
 - Oak Ridge Nat. Lab. (joint publication)

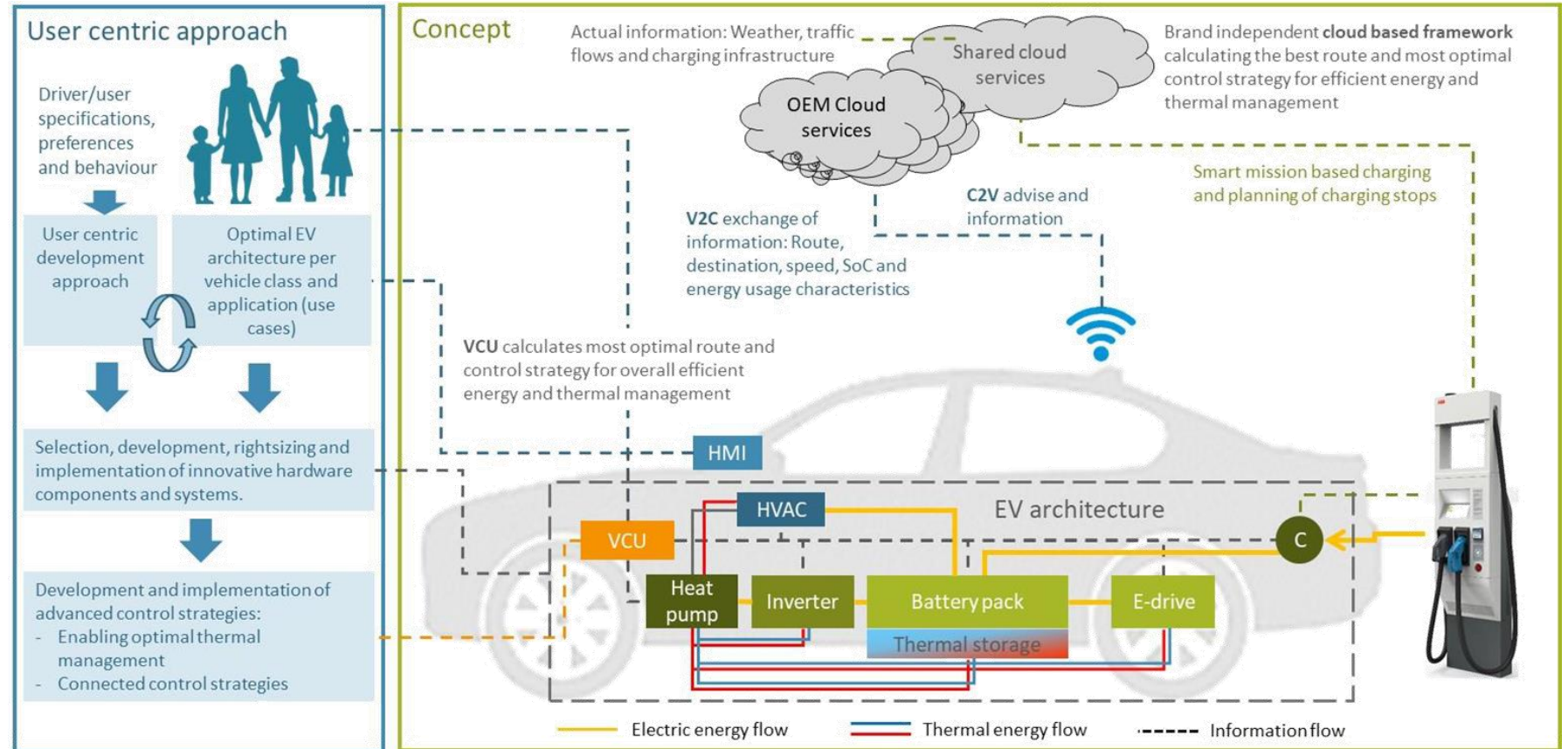


PARTICIPATION IN EU FUNDED PROJECTS

- EMPOWER (Eco-operated, Modular, highly efficient, and flexible multi-POWERtrain for long-haul heavy-duty vehicles), EU HORIZON-CL5-2022-D5-01-08 (WP leader), 2023-present.
- XL-CONNECT (Electric vehicle fleet behavior, interaction with the grid, and digital twin), EU HORIZON-CL5-2021-D5-01, 2023-present.
- LONGRUN (Eco-x strategies for electrified heavy-duty vehicles), EU H2020-LC-GV-2019, 2020-present.
- CEVOLVER (Demonstration of eco-charging and eco-driving concepts for electric vehicles), EU H2020-LC-GV-01-2018, 2018-2022.
- ADVICE (Eco-x strategies for hybrid cars), EU H2020-GV-03-2016 (WP co-leader), 2017-2020.
- OPTEMUS (Eco-routing and thermal management of an electric car), EU H2020-GV-2-2014 (also WP leader), 2014-2019.

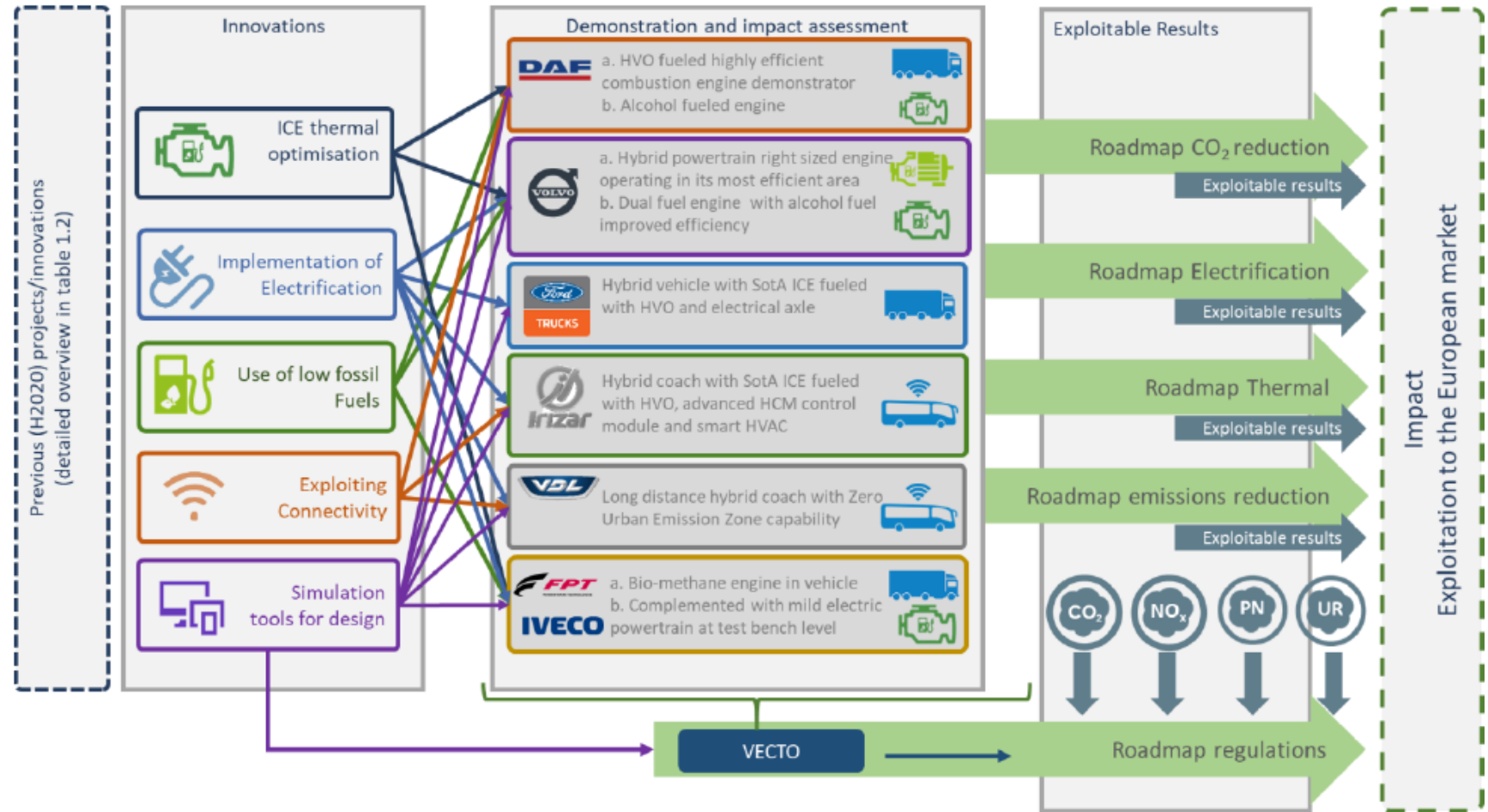
CEVOLVER PROJECT

Sustainable mobility



LONGRUN PROJECT

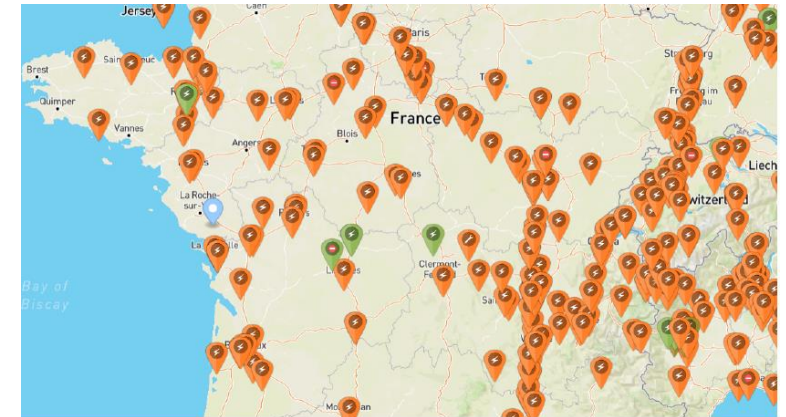
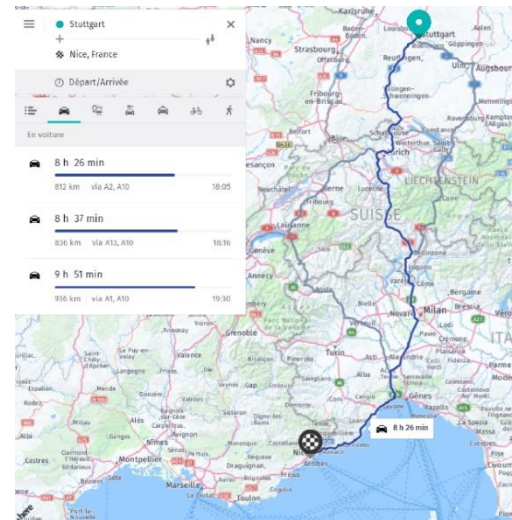
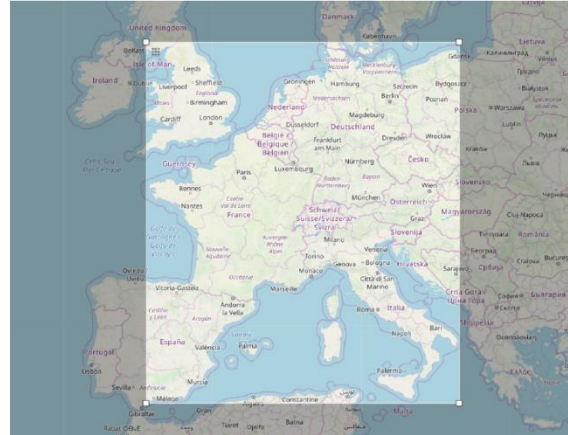
Sustainable mobility



ECOROUTING AND ECOCHARGING

ECOROUTING/CHARGING DATA

- Road topography (offline)
 - Length
 - Importance class
 - Geometry (lat, long, altitude)
 - Signalization
- Charging infrastructure (offline)
 - Location (lat, long)
 - Charger power (DC, $\geq 50\text{kW}$)
- Traffic data (online)
 - Avg. travel time
 - Avg. traffic speed
- Ambient data (online)
 - Temperature
 - Weather conditions



Source: <https://openchargemap.org/site>

OPTIMIZATION PROBLEM

Bi-objective eco-routing formulation

$$\min_{x_i} \sum_{i \in A} [\lambda e_i + (1 - \lambda)t_i] \cdot x_i$$

$$\text{s.t.} \quad \sum_{i \in u^+} x_i - \sum_{i \in u^-} x_i = \begin{cases} 1, & \text{if } u = \text{origin} \\ -1, & \text{if } u = \text{destination} \\ 0, & \text{otherwise} \end{cases} \quad \forall u \in V$$

$$x_i \in \{0, 1\} \quad \forall i \in A$$

$$\lambda \in [0, 1]$$

Objective function

Flow constraints

Binary variable constraints

Optimization weight bounds

 x_i : binary decision variable e_i : energy consumption cost on each arc u^+ : set of outgoing arcs of node u u^- : set of incoming arcs of node u t_i : travel time cost on each arc λ : optimization weight

OPTIMIZATION PROBLEM

Bi-objective formulation with speed and charging embedded in the routing graph (eco-charging)

$$\min_{x_i} \sum_{i \in A_{\mathcal{L}}} [\lambda e_i(v_i, \Delta_i) + (1 - \lambda)t_i(v_i, \Delta_i)] \cdot x_i$$

$$\text{s.t.} \quad \sum_{i \in u^+} x_i - \sum_{i \in u^-} x_i = \begin{cases} 1, & \text{if } u = \text{origin} \\ -1, & \text{if } u = \text{destination} \\ 0, & \text{otherwise} \end{cases} \quad \forall u \in V$$

$$x_i \in \{0, 1\} \quad \forall i \in A_{\mathcal{L}}$$

$$\lambda \in [0, 1]$$

Objective function

Flow constraints

Binary variable constraints

Optimization weight bounds

x_i : binary decision variable

e_i : energy consumption cost on each arc

u^+ : set of outgoing arcs of node u

u^- : set of incoming arcs of node u

t_i : travel time cost on each arc

λ : optimization weight

v_i : travel speed on each arc

Δ_i : recharged energy on each arc

OPTIMIZATION PROBLEM

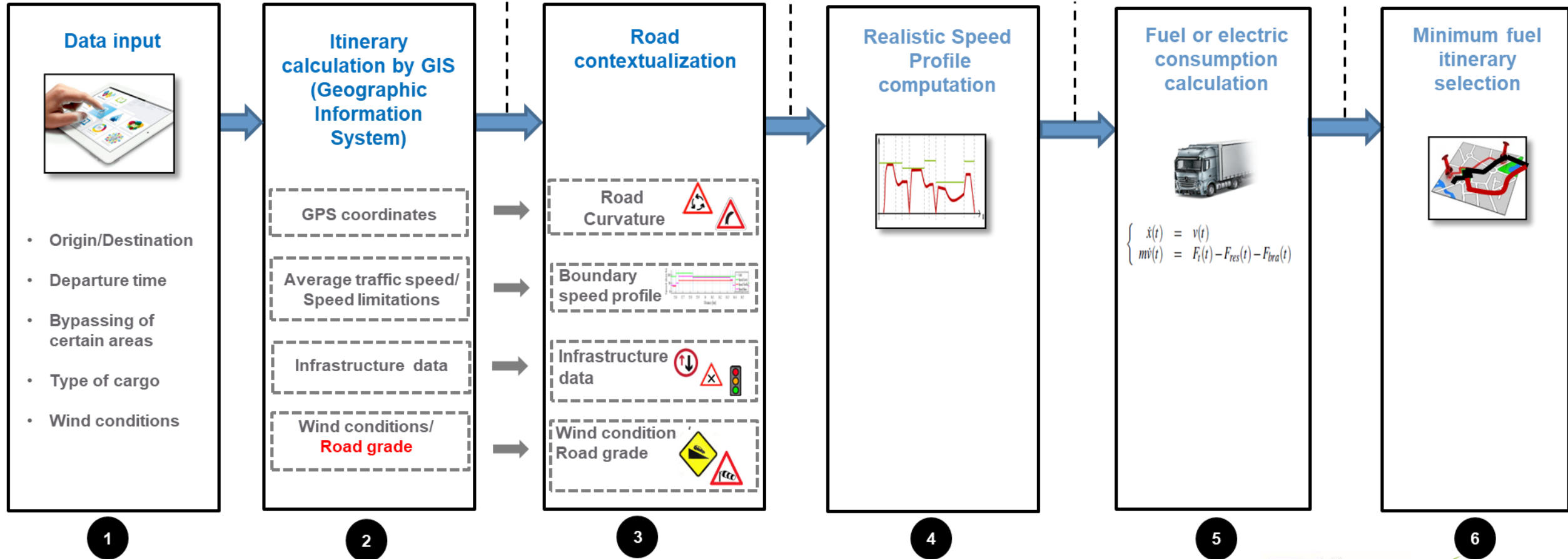
Pragmatic approach for long distance trips

Proposal of n Itineraries

Speed profile setpoint


Realistic speed profile setpoint

Fuel consumption for each itinerary




ECOROUTING/CHARGING WEBSERVICE

Sustainable mobility



Paramètres du véhicule

A  Ford

Itinéraire

Belgium, Maasmechelen, Rijksweg 477 x

67483 Edesheim, Deutschland x

+ Ajouter une destination

Partir Maintenant - 17 oct. 22, 16:17

Calculer l'itinéraire

Résultats

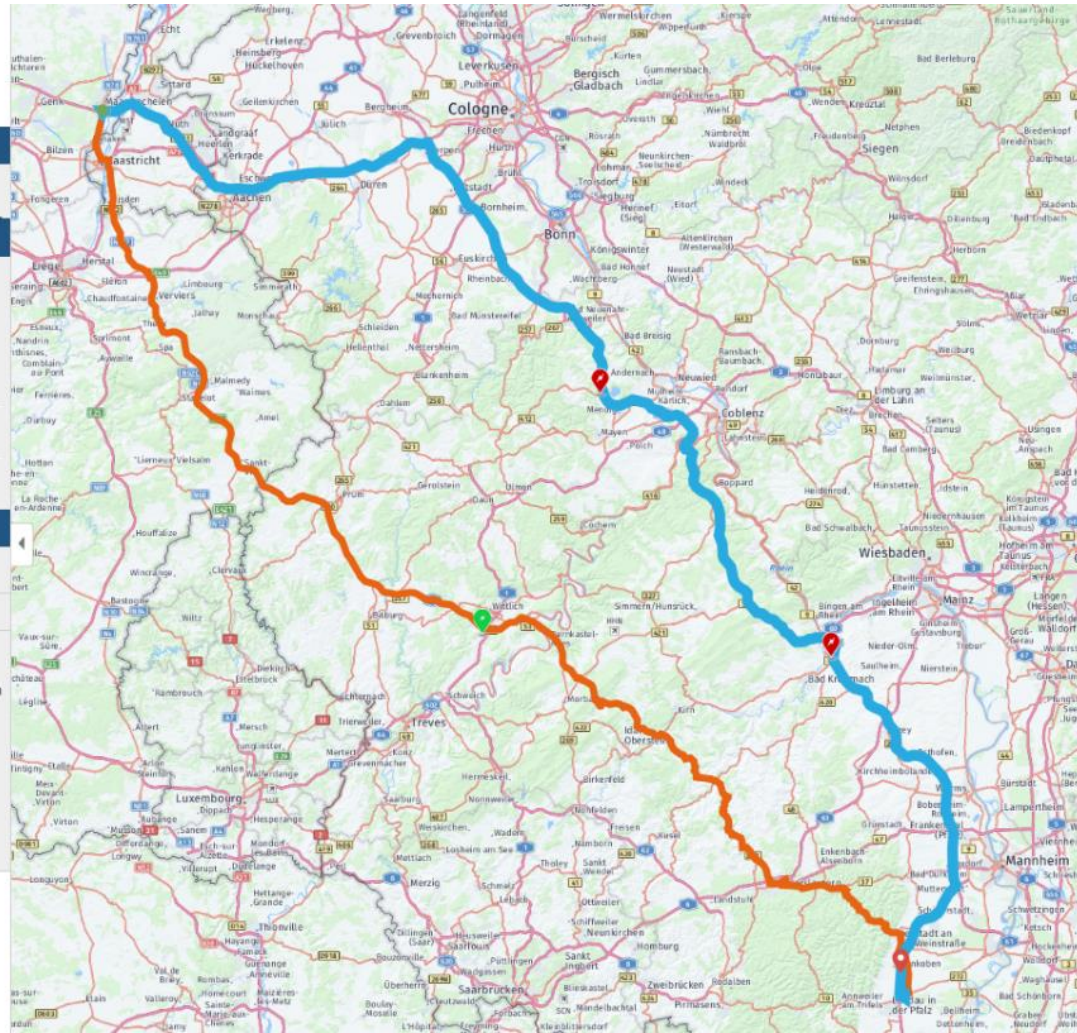
2 itinéraires trouvés Effacer les résultats

| | | | | |
|----------|------------|--------|------------|-----------|
| A | Eco Route | 338 km | 5 h 42 min | 99.0 kWh |
| A | Fast Route | 350 km | 4 h 3 min | 114.4 kWh |

Charging points

| # | Distance (km) | Heure Arrivée | Heure Départ | Durée SOC Arrivée (%) | SOC Départ (%) | Charge (kWh) |
|---|---------------|---------------|--------------|-----------------------|----------------|--------------|
| 1 | 150 | 1 h 29 min | 1 h 47 min | 18 | 22 | 71 |
| 2 | 245 | 2 h 44 min | 2 h 58 min | 14 | 28 | 67 |

Télécharger gpx




European Commission

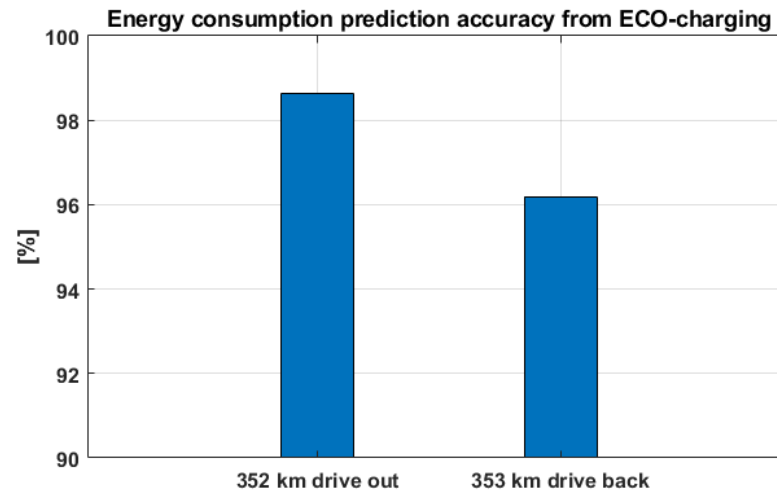
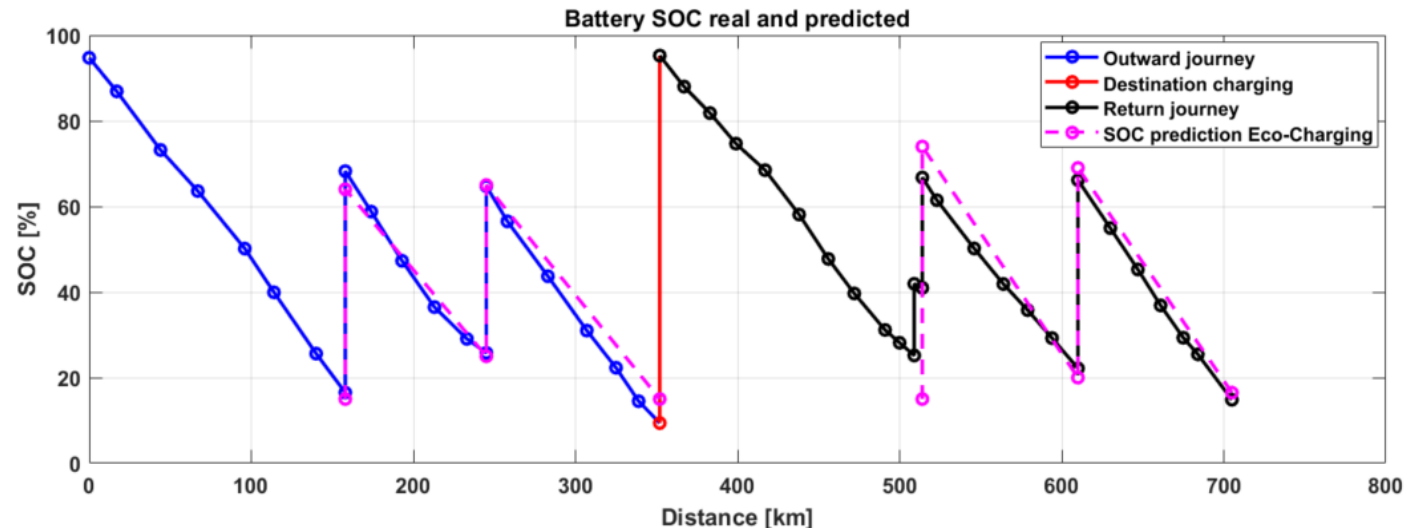
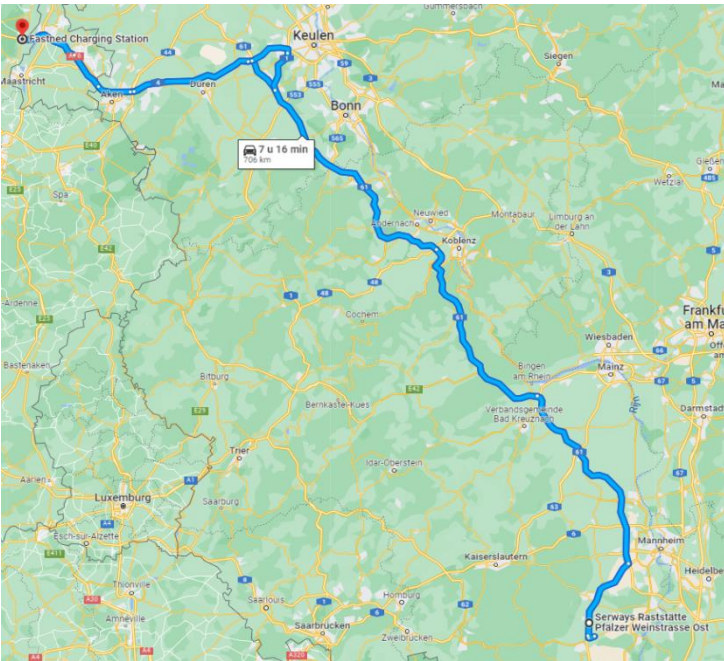


Horizon 2020
European Union funding
for Research & Innovation

ECOROUTING/CHARGING EXPERIMENTAL RESULTS FORD CAMPAIGN



Sustainable mobility



ECOROUTING/CHARGING EXPERIMENTAL RESULTS

FIAT CAMPAIGN

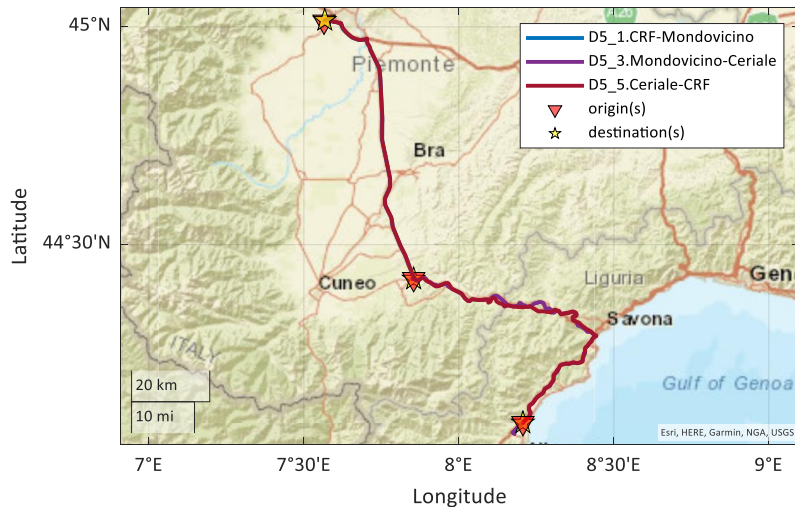


Sustainable mobility



| | η_t (Day 7) / % | η_t (Day 8) / % | η_t (average Day 7 & Day 8) / % |
|-------|----------------------|----------------------|--------------------------------------|
| Day 5 | -11.20 | -14.24 | -12.72 |
| Day 6 | -4.91 | -8.16 | -6.53 |
| Day 9 | -10.03 | -13.11 | -11.57 |

Table 7: Time gain η_t overview of trips a comparison to baseline (Day 7 and Day 8)



| | $\eta_{E,Batt}$ (Day 7) / % | $\eta_{E,Batt}$ (Day 8) / % | $\eta_{E,Batt}$ (average Day 7 & Day 8) / % |
|-------|-----------------------------|-----------------------------|---|
| Day 5 | 5.01 | 10.47 | 7.74 |
| Day 6 | 4.77 | 10.24 | 7.51 |
| Day 9 | 1.35 | 7.02 | 4.18 |

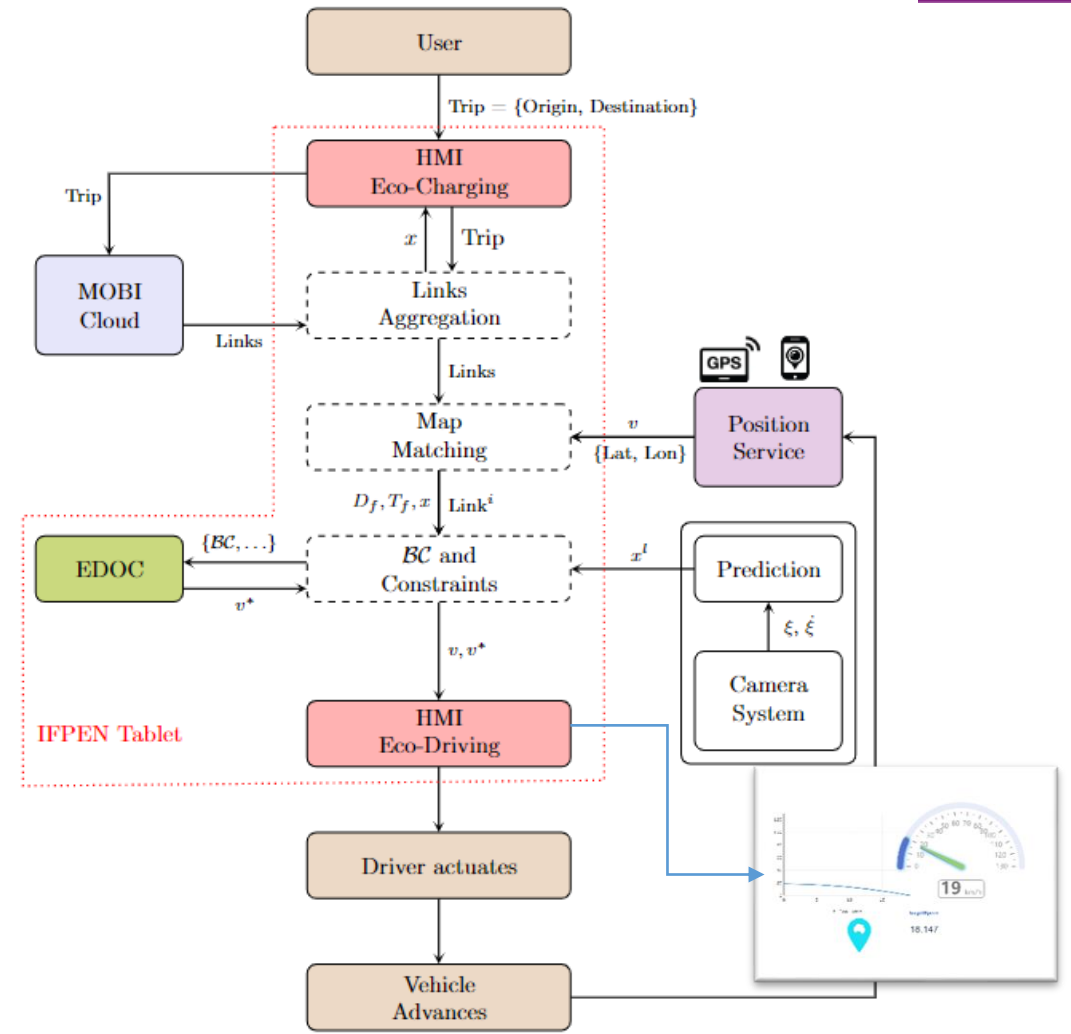
Table 8: Energy saving $\eta_{E,Batt}$ overview of trips and comparison to baseline (Day 7 and Day 8)



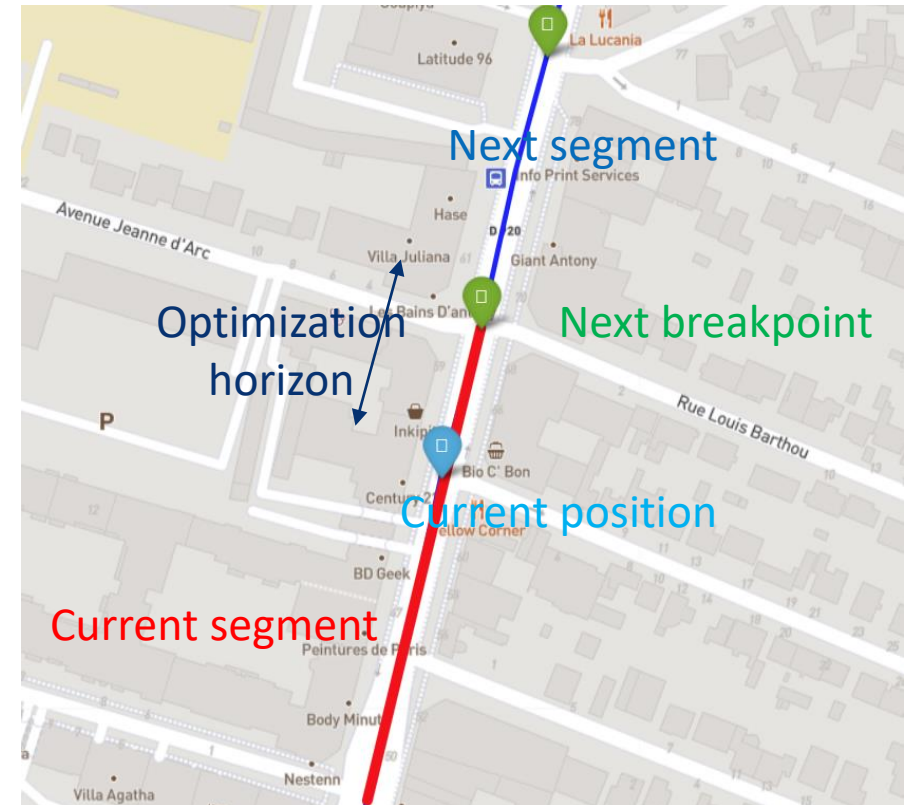
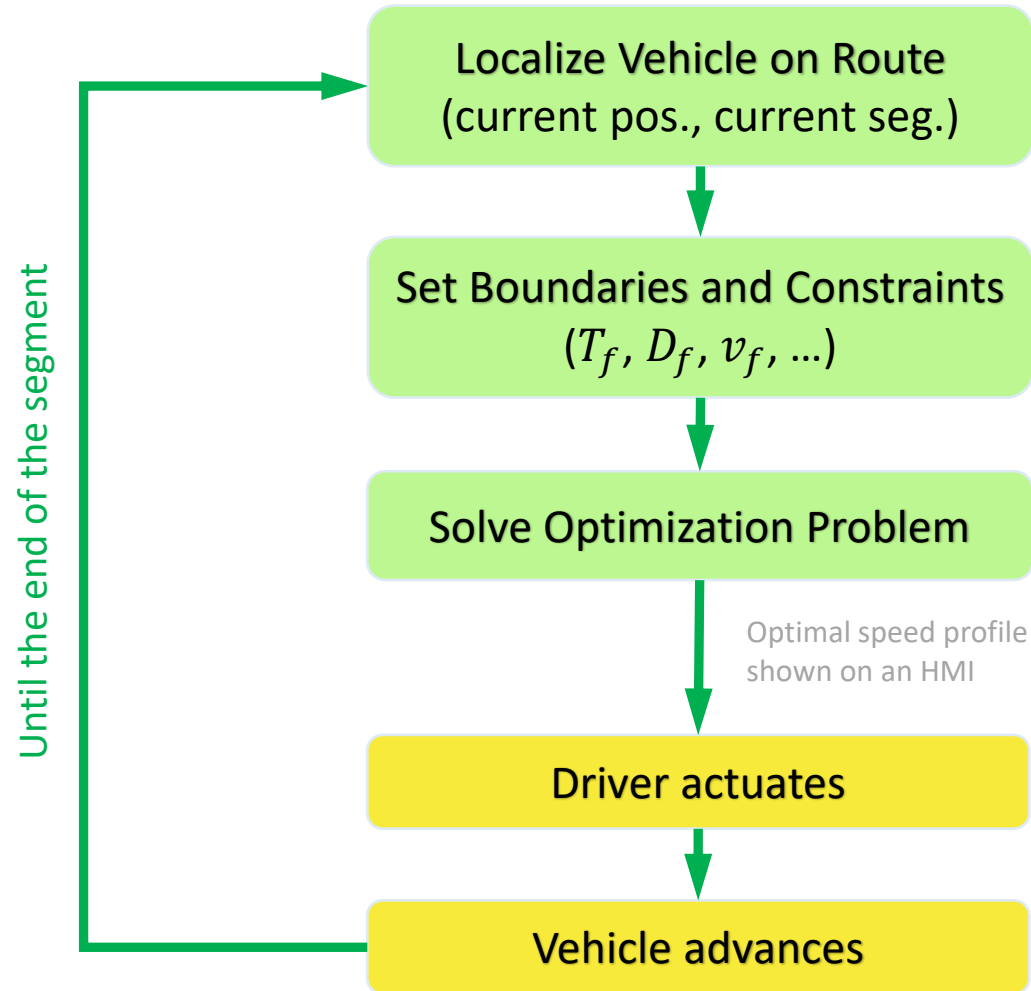
ECODRIVING

DATA AND ARCHITECTURE

- Cloud Data
 - Ecocharging webservice (ecoroute with all its attributes)
 - External webservices (traffic, weather)
- Real-Time Data
 - ADAS sensors (leading vehicle detection, traffic light) (optional)
 - In-vehicle sensors (SoC, speed, etc.)
 - GPS
- Additional Modules
 - Road links aggregation and breakpoint detection
 - Map Matching (current position, current segment on route)
 - Generation of boundary conditions and constraints for the optimization
 - HMI



ECODRIVING CONCEPT



OPTIMIZATION PROBLEM

An optimal control problem with fixed terminal time and state constraints

Control Inputs:
powertrain force,
brake force (per
unit mass)

Time horizon

minimize $u_p(t), u_b(t)$

$$J = \int_0^{t_f} P_b(u_p(t), v(t)) dt,$$

Objective Function: energy consumption
Running cost: battery power

subject to $\frac{ds(t)}{dt} = v(t),$ **State variables: position, speed**

$$\frac{dv(t)}{dt} = u_p(t) - \frac{C_1}{m}v(t) - \frac{C_2}{m}v^2(t) - h(s(t)) - u_b(t),$$

Newton's 2nd law

$v(0) = v_i, v(t_f) = v_f,$
 $s(0) = 0, s(t_f) = s_f,$ **Terminal conditions: final speed, distance horizon**

$u_{p,min}(v(t)) \leq u_p \leq u_{p,max}(v(t)),$
 $0 \leq u_b(t) \leq u_{b,max},$ **Control constraints**

$v_{min}(t, s(t)) \leq v(t) \leq v_{max}(t, s(t)),$ **Speed limits (top legal, cornering, traffic...)**

$s_{min}(t) \leq s(t) \leq s_{max}(t),$ **Position constraint induced by a preceding vehicle**

ANALYTICAL SOLUTION

Model simplification

$$P_b(t) = bu_p^2(t) + mu_p(t)v(t), \quad \dot{v}(t) = u_p(t) - h$$

Derivation of analytical solutions using PMP

Unconstrained

$$v^*(t) = v_i + \left(-\frac{4v_i}{t_f} - \frac{2v_f}{t_f} + \frac{6s_f}{t_f^2} \right) t + \left(\frac{3v_i}{t_f^2} - \frac{6s_f}{t_f^3} + \frac{3v_f}{t_f^2} \right) t^2$$

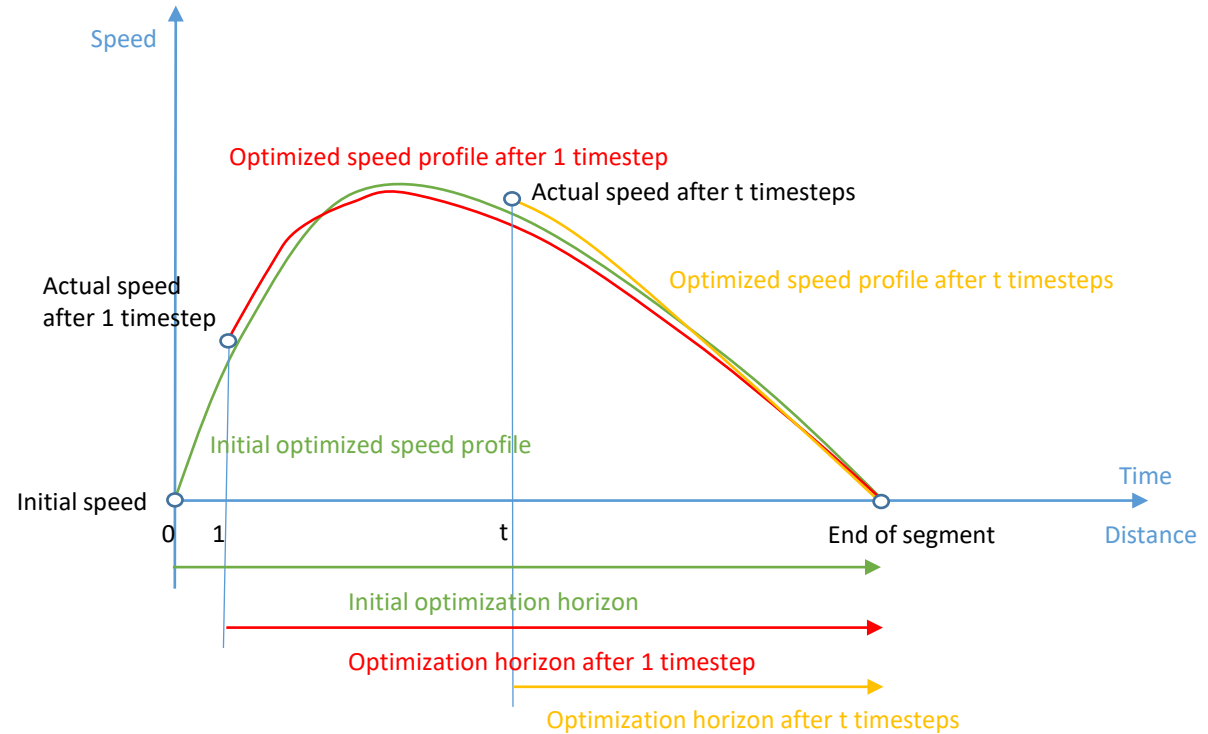
Speed-constrained

$$v^*(t) = \text{Piecewise quadratic function}(t | v_i, v_f, t_f, s_f, v_{max})$$

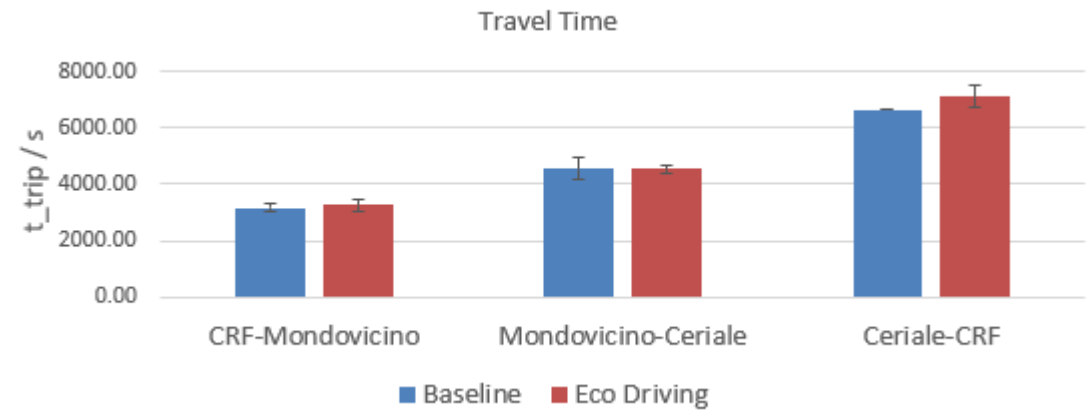
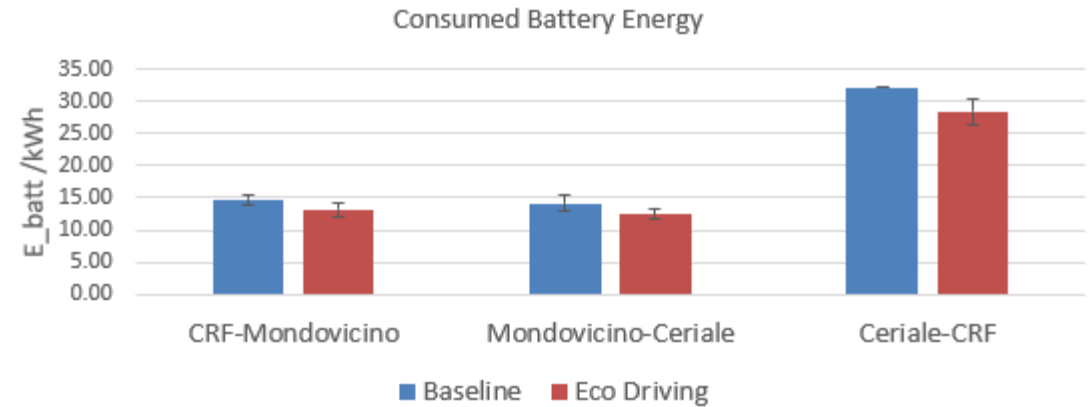
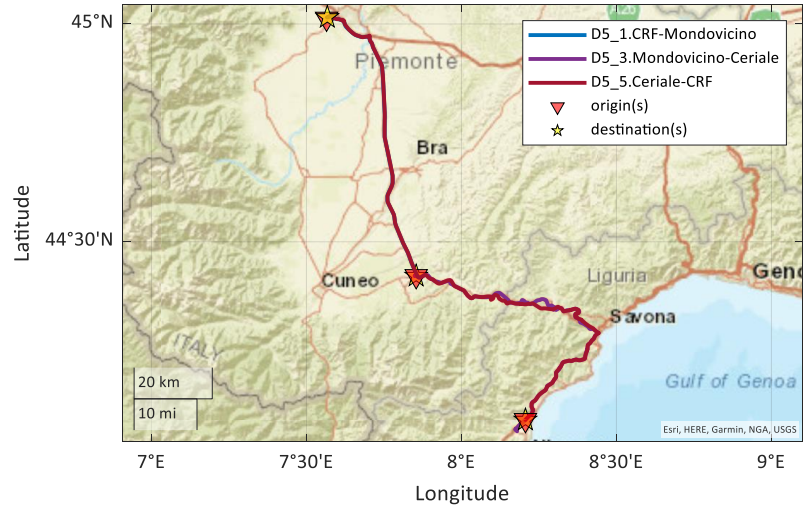
Position-constrained

$$v^*(t) = \text{Piecewise quadratic function}(t | v_i, v_f, t_f, s_f, s_{p0}, v_{p0}, a_p)$$

Application in a shrinking-horizon (MPC) framework



ECODRIVING EXPERIMENTAL RESULTS FIAT CAMPAIGN



| Trip | eta_E_Batt / % | delta_t_trip / s |
|---------------------|-----------------|------------------|
| CRF-Mondovicino | 10.8957% | 105.48 |
| Mondovicino-Ceriale | 11.9844% | -24.78 |
| Ceriale-CRF | 11.8032% | 451.13 |
| Average | 11.5611% | 177.27 |

ECODRIVING EXPERIMENTAL RESULTS

IFPEN PROPRIETARY SYSTEM

- Perception system by a French start-up company called Visual Behavior
- Stereoscopic camera with ability to detect, identify, and measure
 - Relative speed and distance of the preceding vehicle
 - Traffic light status



MOBILITY SYSTEMS

MOBILITY OBSERVATION AND ANALYSIS

Data space



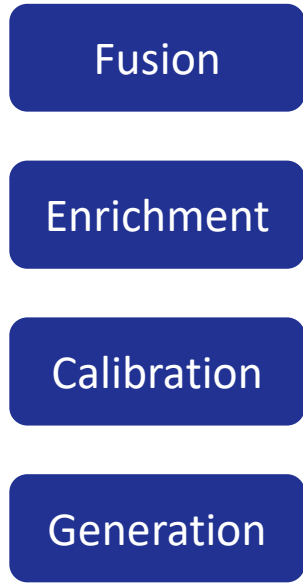
Raw data:
Counting measurement,
GPS trace,
mobility survey, ...

Data handling



Processed and exploitable data

Data enhancement



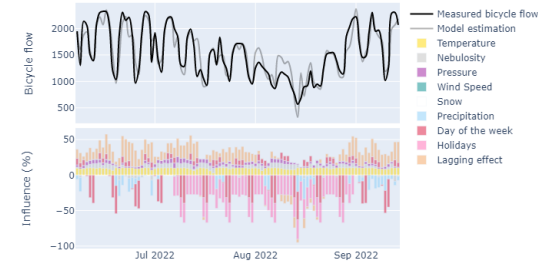
Improved, representative and calibrated data

Estimation of full flow per street by fusion

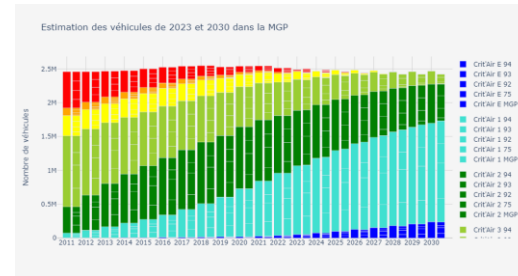
Model of counting time series for prediction and gap removing

Tuning of bike routing algorithm from user's preferences

Evaluation of prospective scenarios on territory: bicycle infrastructure, LEZ restriction



eqasim
MATSim
Multi-Agent Transport Simulation

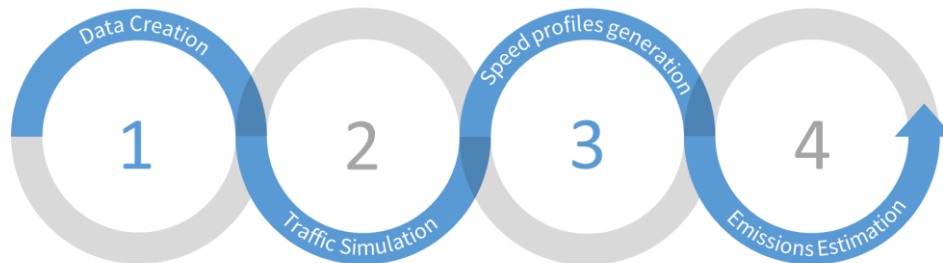
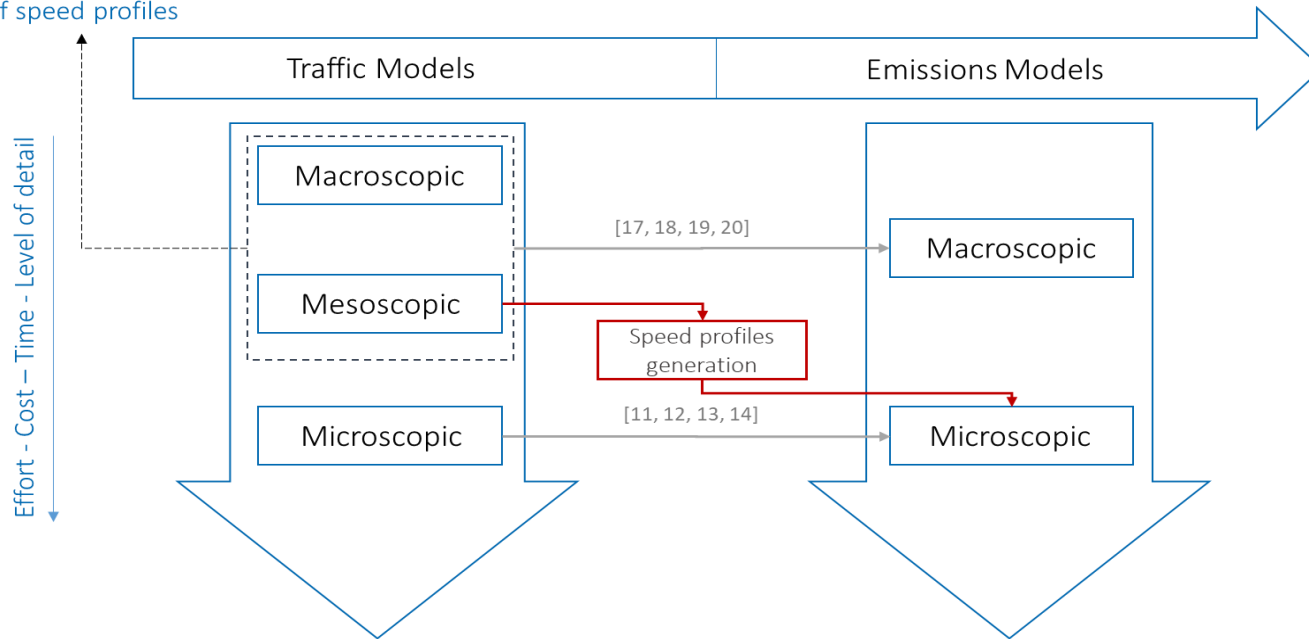


MACRO TO MICRO: SPEED PROFILES GENERATION

PHD OF ABDELKADER DIB, CO-SUPERVISED BY MILOŠ BALAČ

Goal: predict emissions from traffic and air quality

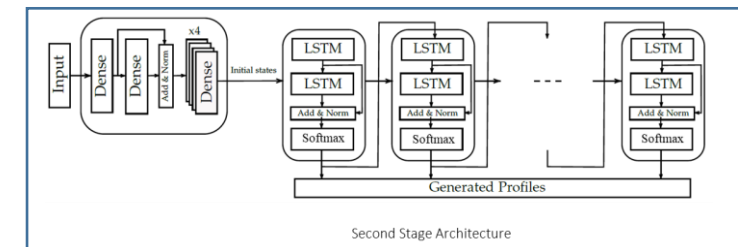
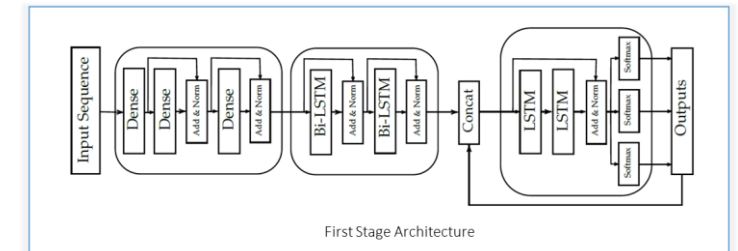
Do not give a fine description of speed profiles



EQASIM Creation of a synthetic population and organization of all needed data
MATSIM Mesoscopic simulation of the created scenario
TISPRED Speed profiles generation from vehicle average speed and road features
MOBSIM Pollutants emissions estimation

Approach: LSTM-based neural network architecture

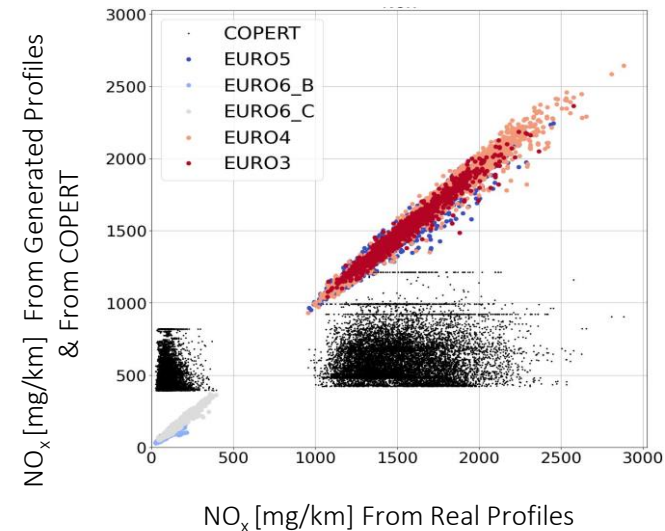
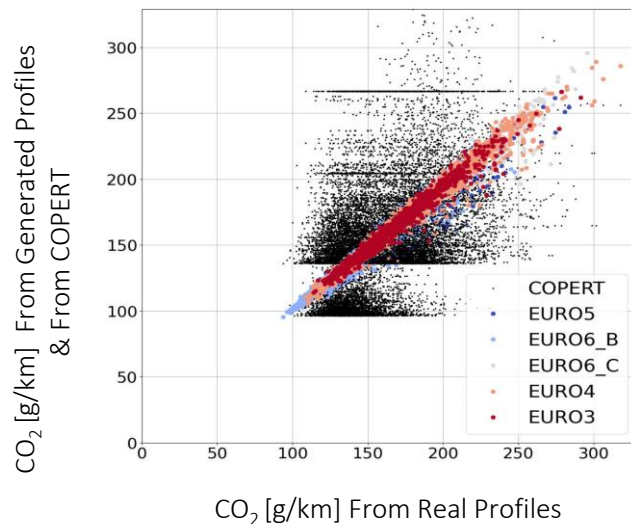
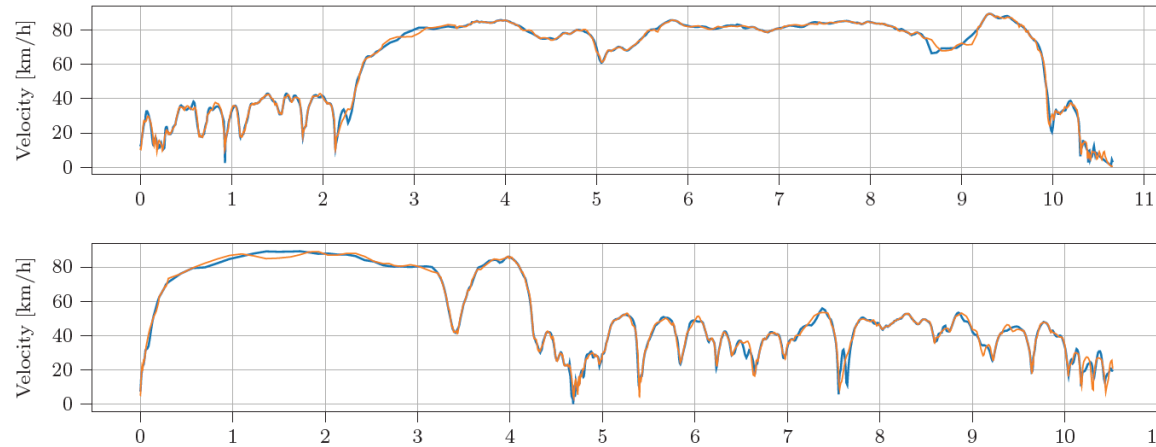
Trained through recorded trip profiles



MACRO TO MICRO: SPEED PROFILES GENERATION

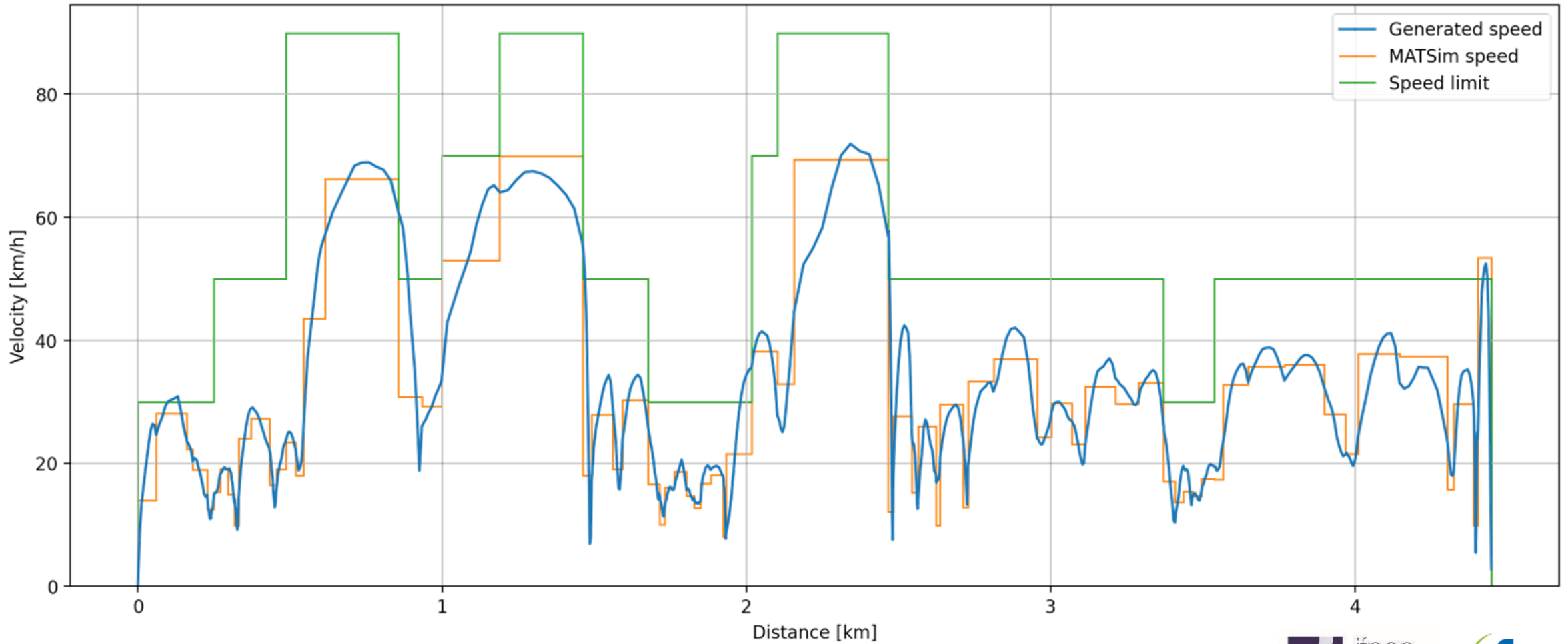
PHD OF ABDELKADER DIB, CO-SUPERVISED BY MILOŠ BALAČ

First results: reconstruction of speed profiles + emission estimation



MACRO TO MICRO: SPEED PROFILES GENERATION

PHD OF ABDELKADER DIB, CO-SUPERVISED BY MILOŠ BALAČ

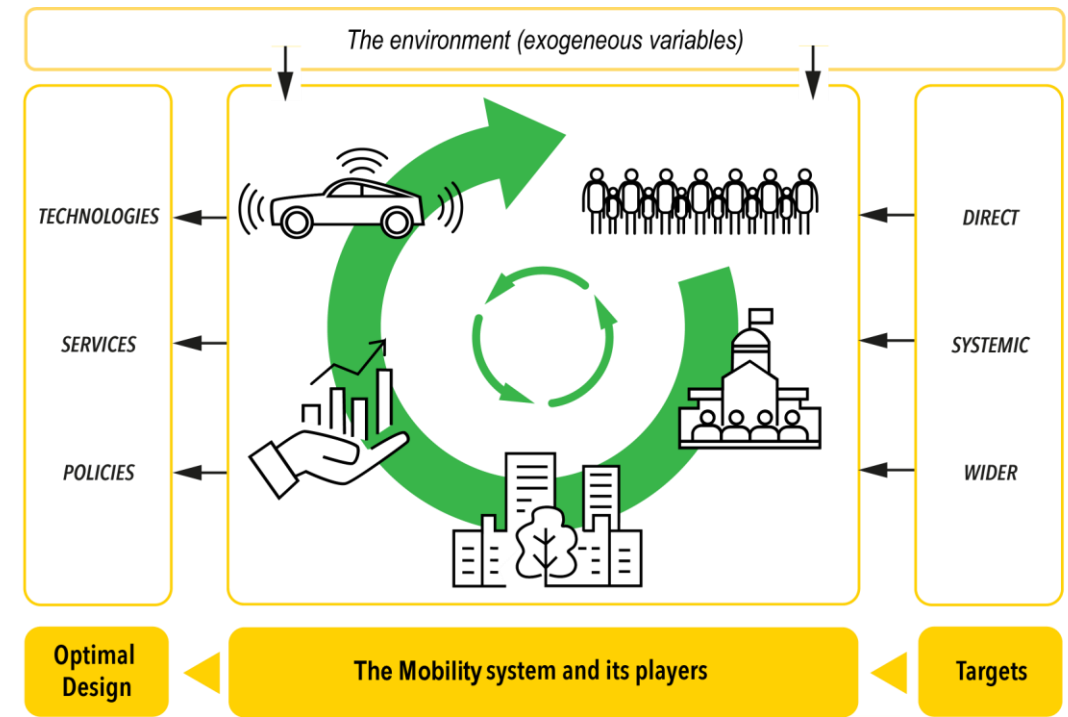
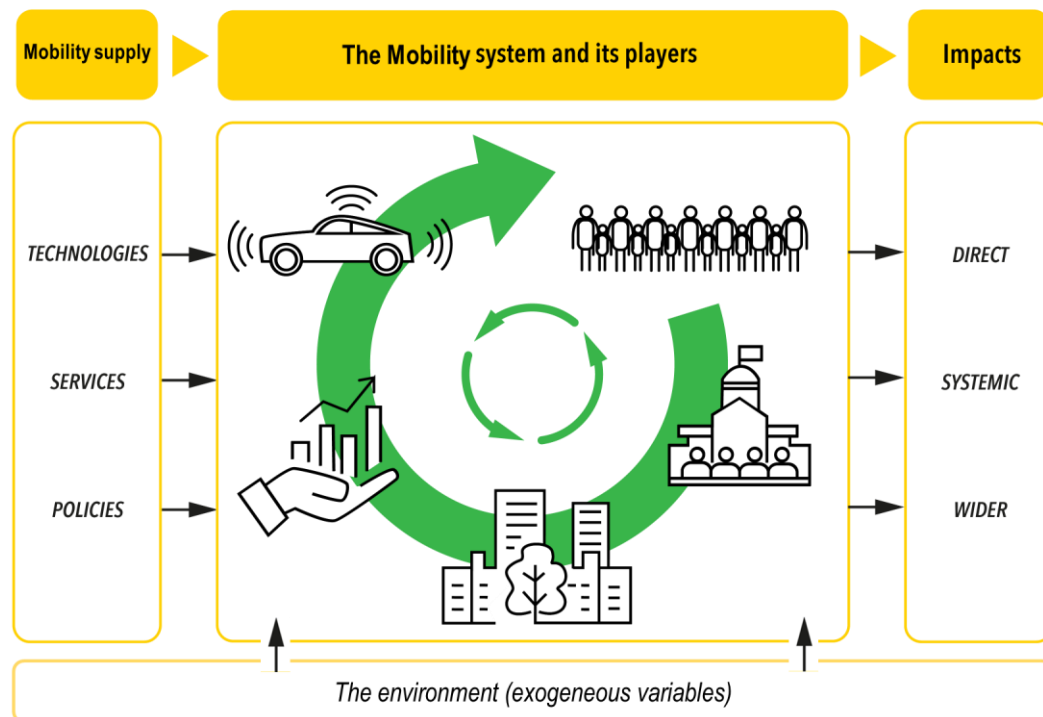


MOBILITY SYSTEMS DYNAMICS AND OPTIMISATION

UPCOMING FRENCH PROJECT « FORBAC »

Goals:

- Mobility system described by system dynamics
- From Forecasting to « backcasting » (optimal control)



Innovating for energy

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