

Ride-sharing, congestion, departure-time and mode choices: A social optimum perspective

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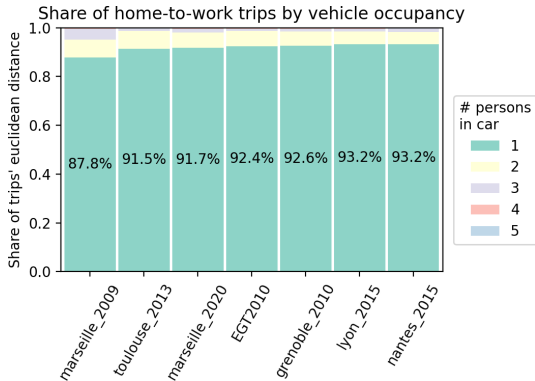
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Séminaire d'Économie des Transports – LAET
September 25, 2025

Motivation

- **Ride-sharing:** multiple passengers traveling in the same direction share a vehicle
- Expected effect: ↗ vehicle occupancy
→ ↘ car vehicle-kilometers
- High potential: In 2010, in Île-de-France, car drivers are traveling alone for 92.4% of the distance of the home-to-work trips inc. intermediate stops (60.5% for non-work purposes)



Motivation

	Benefits	Costs
Individual	Split fuel and other expenses	Temporal / spatial mismatch, inconvenience
Social	↘ congestion, ↘ CO ₂ , ↘ local pollutants, ↘ noise	Infrastructure (pick-up / drop-off zones)

- Government intervention is required (network effects, negative externalities)
- Example policies:
 - ▶ **Financial incentives** to drivers (100€ bonus in France)
 - ▶ **High-occupancy vehicle (HOV) lanes** (e.g., Boulevard Périphérique in Paris)



Contributions

Research questions:

- How to optimally match drivers and passengers?
- What is the social surplus potential?

Contributions:

- Methodology to find the socially optimal match between passengers and drivers in large-scale scenarios
- Application to Île-de-France with METROPOLIS2

	Benefits	Costs
Individual	Split fuel and other expenses	Temporal / spatial mismatch, inconvenience
Social	$\searrow$ congestion , $\searrow$ CO₂ , $\searrow$ local pollutants, $\searrow$ noise	Infrastructure (pick-up / drop-off zones)

Similar papers

	Obj. function	Time	Congestion	Network
Delle Site, de Palma, Ghosly (2022)	social costs	static	exogenous	Sioux Falls
Sun, Wu, Chen (2022)	$\sim$ individual costs	dynamic	microscopic	Chattanooga
de Palma, Stokkink, Geroliminis (2022)	individual costs	dynamic	bottleneck	single road
de Palma, Javaudin, Stokkink, Tarpin-Pitre (2024)	individual costs	dynamic	bottleneck	Île-de-France
This paper	social costs	dynamic	bottleneck	Île-de-France

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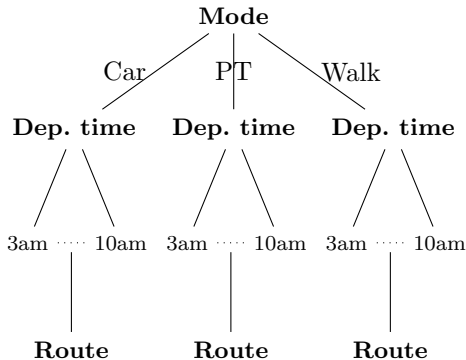
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Settings

- Home-to-work trips in the morning
- Agents select a travel mode, departure time, and route
- No ride-sharing in the baseline scenario
- Travel modes: car (as a driver), public transit (PT), walking



Preferences

Generalized cost with mode m at departure time t^d :

$$C^m(t^d) = \underbrace{\alpha_m \cdot \text{tt}^m(t^d)}_{\text{Travel cost}} + \underbrace{\beta \cdot [t^* - t^a]^+ + \gamma \cdot [t^a - t^*]^+}_{\text{Schedule-delay cost}} + \underbrace{F^m}_{\text{Fuel}},$$

- $\text{tt}^m(t^d)$: travel time with mode m at departure time t^d
- $t^a = t^d + \text{tt}^m(t^d)$: arrival time
- α_m : mode-specific value of time
- β, γ : penalties for early and late arrival
- t^* : desired arrival time
- F^m : fuel cost (only for car)

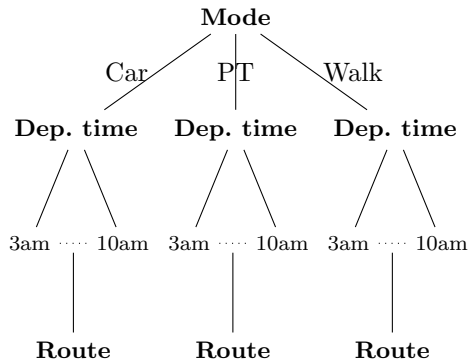
Choice models

- **Route choice:** fastest time-dependent path (car), least-cost path (PT), or shortest path (walk)
- **Departure-time choice:** Multinomial Logit

$$t^d = \arg \min_t [C^m(t) - \eta(t)]$$

- **Mode choice:** Multinomial Logit

$$m = \arg \min_m [C^m - \varepsilon^m]$$



Ride-sharing system

- System coordinated by a social planner with full information
- Participants to the system can be matched as RS driver, matched as RS passenger, or keep traveling alone
- RS drivers select their departure time and route (conditional on the detour)
- The social planner chooses the pick-up and drop-off locations of each match

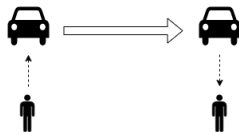
Detour scenarios



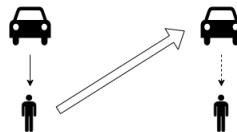
Driver (D)



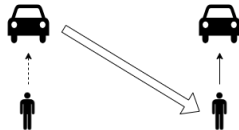
Passenger (P)



Passengers' detour at origin and destination



Drivers' detour at origin, passengers' at destination



Passengers' detour at origin, drivers' at destination



Drivers' detour at origin and destination

Ride-sharing social cost

When agent i is matched as RS driver with passenger j as RS passenger:

- Cost of driver i is:

$$C_{i,j}^{\text{Driver}} = \alpha_{\text{Driver}} \cdot \text{tt}_{i,j}^{\text{Driver}} + \beta \cdot [t_i^* - t^a]^+ + \gamma \cdot [t^a - t_i^*]^+ + F_{i,j}^{\text{Driver}} + \eta_i(t^d) + \varepsilon_i^{\text{Driver}}$$

- Cost of passenger j is

$$C_{i,j}^{\text{Pass}} = \alpha_{\text{Pass}} \cdot \text{tt}_{i,j}^{\text{Pass}} + \alpha_{\text{Walk}} \cdot \text{wt}_{i,j}^{\text{Pass}} + \beta \cdot [t_j^* - t^a]^+ + \gamma \cdot [t^a - t_j^*]^+ + \eta_j(t^d) + \varepsilon_j^{\text{Pass}}$$

- Total social cost is:

$$C_{i,j} = C_{i,j}^{\text{Driver}} + C_{i,j}^{\text{Pass}} + E_{i,j}$$

with $E_{i,j}$ cost of CO₂ emissions

Social planner optimization

Optimization program of the social planner:

$$\min_X \sum_{i,j} C_{i,j} X_{i,j}$$

subject to the constraints

$$\sum_j X_{ij} = 1, \quad \forall i$$

$$X_{ij} \in \{0, 1\}, \quad \forall i, j$$

- $X = (X_{i,j})$: matching matrix where $X_{i,j} = 1$ when i is a driver with j as passenger and $X_{i,i} = 1$ when i travels alone
- $C_{i,j}$: total social cost of matching i with j , with $C_{i,i}$ = total social cost of i traveling alone

Social planner optimization

Remarks:

- A driver can share their ride with at most one passenger
- The role of agents (driver, passenger, alone) are not pre-assigned but optimized endogenously by the social planner
- A match between i and j is feasible only if

$$C_{i,j} \leq C_{i,i} + C_{j,j}$$

i.e., Pareto-improving transfers can be implemented so that no agent is worse off

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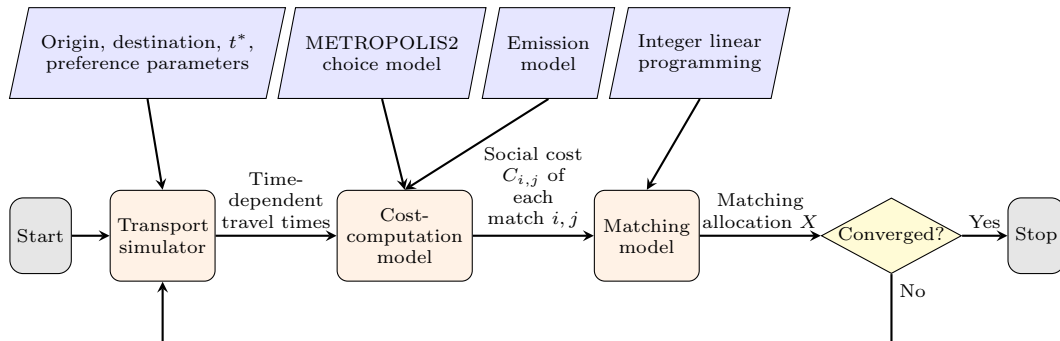
Île-de-France Application

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Endogeneity

- The optimization problem can be solved using standard integer linear programming techniques
- Endogeneity problem: social costs $C_{i,j}, \forall i, j$ are treated as fixed but they depend on congestion levels which are shaped by matching decisions X

Iterative framework



Transport simulator

METROPOLIS2 (Javaudin and de Palma, 2024):

- Population of agents with a chain of **point-to-point trips**
- Mode, departure-time and route decisions
- Endogenous congestion with dynamic link-level bottlenecks (mesoscopic simulator)
- Some agents can have fixed modes → matching decisions can be forced

Cost-computation model

For each feasible pair i, j and each detour scenario:

- driver i 's selected departure time and route are computed through METROPOLIS2
- passenger j 's walking time is computed through routing on the pedestrian network
- CO₂ emissions and fuel consumption are computed with METRO-TRACE (Le Frioux et al, 2024) based on link-level speeds and agent-level vehicle characteristics

The total social cost $C_{i,j}$ of the pair i, j is the minimum cost over all detour scenarios.

Matching model

- Integer linear program
- PuLP Python library
- Output of the model: matches $X = (X_{i,j})$

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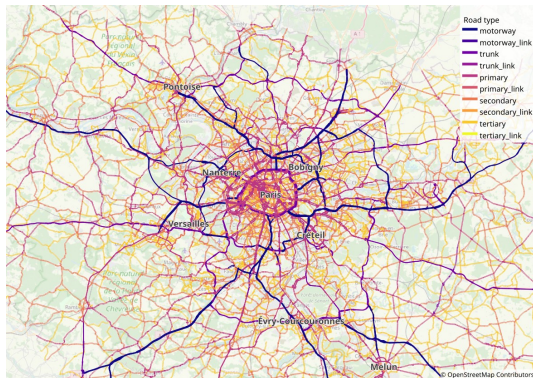
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Simulation setup

- Île-de-France region
- Road and pedestrian network from OpenStreetMap
- Public transit schedule from Île-de-France Mobilités (GTFS)
- Synthetic population with activities and trips from Hörl and Balac (2021)
- Morning period from 3 a.m. to 10 a.m.
- 10 % re-scaling: 629k agents, 819k trips
- Four-step calibration process (Javaudin, 2024)



Parameters

Parameter	Value
α_{Car}	10 €/h
α_{Walk}	10 €/h
α_{PT}	8 €/h
α_{Driver}	10 €/h
α_{Pass}	10 €/h
β	5 €/h
γ	5 €/h
Walking speed	4 km/h
Fuel cost	1.8 €/L
CO ₂ emission cost	200 €/t

Ride-sharing scheme assumptions

- Participants to the system are:
 - ▶ Car drivers in the baseline scenario
 - ▶ Direct trip from home to work
 - ▶ Additional 40 % probability
- Final share of participants: 13.2 % of simulated population
- A pair i, j is *feasible* if:
 - ▶ j 's origin is within 5 km of i 's origin
 - ▶ j 's destination is within 5 km of i 's destination
 - ▶ j 's baseline departure time is within 20 min of i 's baseline departure time

Main results

About 30 % of participants are matched either as a RS driver or RS passenger

	Baseline	Ride-sharing	Change
Car trips	2.06 M	1.95 M	−5.2 %
Vehicle-kilometers	30.70×10^6 km	30.54×10^6 km	−0.5 %
Time lost in congestion	209 526 h	200 455 h	−4.3 %
Fuel consumption	1.943×10^6 L	1.938×10^6 L	−0.2 %
CO ₂ emissions	6167 t	6155 t	−0.2 %

⇒ The potential of ride-sharing is small

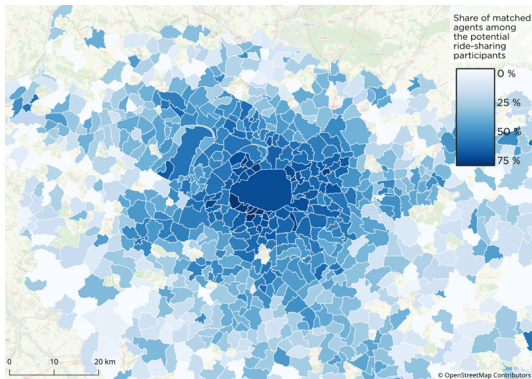
Mode switch

From \ To	Walking	Car Driver	Public transit	Car Passenger	Total
Walking	31.8	.	.	.	31.8
Car Driver	.	30.7	.	2.0	32.7
Public transit	.	0.3	35.1	.	35.4
Total	31.8	31.0	35.1	2.0	100.0

⇒ Rebound effect: 15 % of car share reduction is “absorbed” by switches from public transit (54 % of vehicle-kilometers!)

Spatial distribution

Participants are more likely to be matched when living in densely populated areas

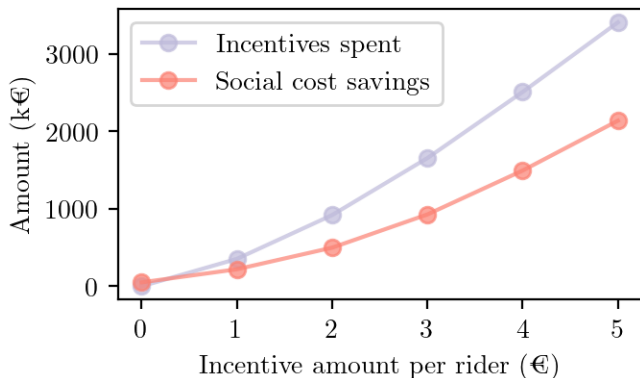


Ride-sharing incentive

Incentive amount	0€	1€	2€	3€	4€	5€
Car share	31.0 %	30.3 %	29.6 %	29.0 %	28.5 %	28.1 %
Ride-sharing share	4.0 %	5.6 %	7.3 %	8.8 %	10.0 %	10.8 %
Incentives spent (k €)	0	354	920	1,656	2,510	3,409
Social cost savings (k €)	51	220	499	928	1,495	2,139
Car VKT (10^6 km)	30.54	30.32	30.15	29.96	29.75	29.52
Time lost in congestion (10^3 hour)	200.5	195.9	191.4	187.7	184.2	182.0
CO ₂ emissions (t)	6155	6097	6084	6046	5996	5955
Mean walking distance (m)	590	610	650	680	710	730

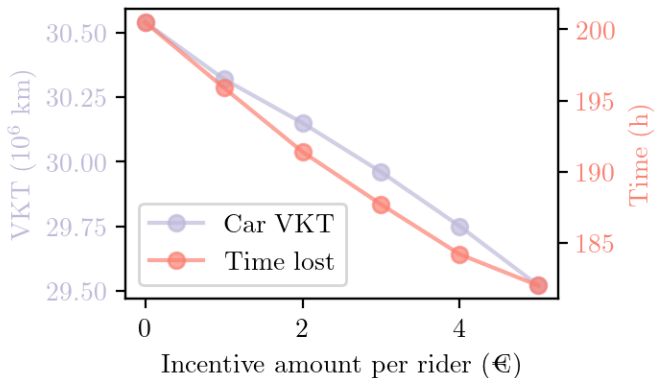
⇒ Offering incentives increase the number of matches found (up to 82 % of the theoretical maximum)

Ride-sharing incentive



⇒ The cost of incentives increases faster than the improvement in social surplus

Ride-sharing incentive



⇒ Car vehicle-kilometers and time lost in congestion decrease about linearly with the incentive amount

PT fare subsidy

	Baseline	1 € RS incentive	−0.15 € PT fare
Policy cost (k €)	0	354	343
Social cost savings (k €)	0	220	550
Car share	32.7 %	30.3 %	32.2 %
PT share	35.4 %	35.1 %	36.3 %
Car VKT (10^6 km)	30.70	30.32	30.29

⇒ A reduction of PT fare can achieve larger social benefits than ride-sharing incentives with similar cost

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Take-away

- Methodology for large-scale evaluation of a ride-sharing system with:
 - ▶ temporal matching (scheduling preferences)
 - ▶ location-based spatial matching (zones)
 - ▶ socially optimal matching (inc. CO₂)
 - ▶ endogenous congestion
- Application to Île-de-France:
 - ▶ With 13.2 % of participants: 2.0 % RS passengers; vehicle-kilometers $\searrow$ 0.5 %
 - ▶ Ride-sharing incentives are more expensive than the improvement in benefits
 - ▶ Subsidizing ride-sharing is worse than subsidizing PT

Future directions

- Local pollutants
- Arbitrary pick-up / drop-off locations
- Drivers with 2+ passengers
- Multi-hop ride-sharing
- Intermodality ride-sharing + PT
- Evening peak (round trips)
- HOV lanes

Thank you