

Deriving on-trip route choices of truck drivers by utilizing Bluetooth data, loop-detector data and variable message sign data



Data Analytics &
Traffic Simulation



Freight and Logistics



GRIP on freight TRIPS

Salil Sharma (S.Sharma-4@tudelft.nl), Maaike Snelder and Hans van Lint

Delft University of Technology, The Netherlands

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Outline

- 1 Motivation and Objectives
- 2 Route choice modeling of truck drivers
- 3 Inefficiencies in routing decisions
- 4 Conclusions and Next steps

Motivation

- On important truck-dominated motorways, a large share of traffic consists of trucks.
- Truck driver's routing decisions are different from passenger cars because of different constraints from the logistics system.
- Route choice of truck drivers is of interest to both transport planners and traffic management authorities.

¹S. Hess, M. Quddus, N. Rieser-Schüssler, and A. Daly (2015). "Developing advanced route choice models for heavy goods vehicles using GPS data". In: *Transportation Research Part E: Logistics and Transportation Review* 77, pp. 29–44

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- Truck driver's routing decisions are different from passenger cars because of different constraints from the logistics system.
- Route choice of truck drivers is of interest to both transport planners and traffic management authorities.
- A major problem for route choice modeling has always been the need to capture appropriate data¹. The strengths and weaknesses of both stated preference (SP) and revealed preference (RP) methods are widely known.
- We enrich an RP dataset with contextual information by utilizing multiple data sources to overcome the limitations of previous RP/SP studies.

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Objectives

- ① To model the route choices of truck drivers using Bluetooth data, loop detector data and variable message sign data
- ② To evaluate the efficiencies of routing decisions of truck drivers from both user's and system's perspectives

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Study area

Case study to model route choices of truck drivers between port of Rotterdam and hinterland

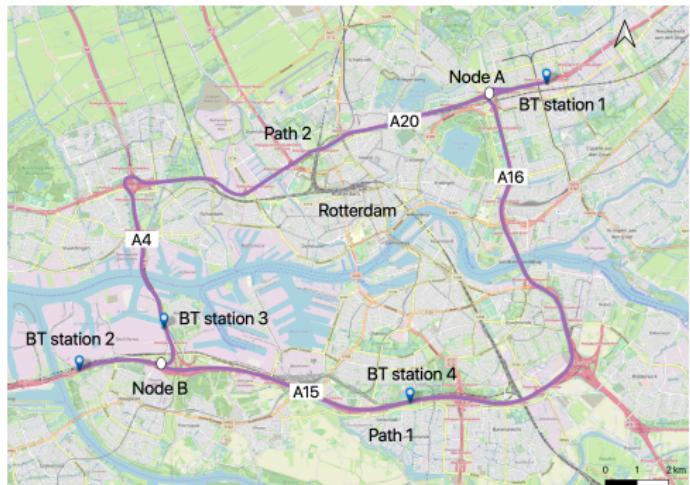


Study area: Rotterdam ring which provides a route choice for traffic destined to the port of Rotterdam

Node A as the origin and node B as the destination

Two paths: A16-A15 and A20-A4

Data collection



Origin-destination data:

Bluetooth stations located near motorway capture the time-stamps and MAC-IDs² of passing vehicles

Contextual information:

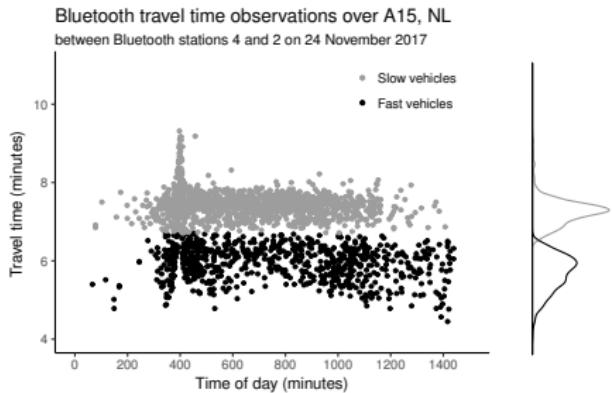
Travel time reliability and lane closures via loop-detector data and variable-message sign data

Bluetooth data do not provide mode classification!

²Media Access Control Address: unique hardware identification number

Infer trucks from Bluetooth data

Steps:



Travel time clusters are formed between short segments of motorways because of differential speed limits observed in the Netherlands

- ① For the data collection period, find all the vehicles that have passed through a path and remove outliers.
- ② Find the common vehicle Ids that belong to the **slow vehicle cluster** and to the path under consideration.
- ③ From the common vehicle Ids, select the vehicles which have traversed the path with a maximum speed of 80 km/h ³.
- ④ The vehicle Ids thus extracted can be **classified as trucks**

³80 km/h refers to the speed limit for trucks on motorways in the Netherlands.

Model specification

Utility is specified as a linear sum of the following attributes.

- Total distance of a path (TD)
- Instantaneous travel time of a path (ITT)
- Travel time unreliability of a path ($TTUR$)
- Maximum number of lanes closed along a path (LC) as a proxy for congestion

⁴ J. W. C. van Lint, H. J. van Zuylen, and H Tu (2008). "Travel time unreliability on freeways: Why measures based on variance tell only half the story". In: *Transportation Research Part A: Policy and Practice* 42.1, pp. 258–277

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$TTUR$ captures the day-to-day travel time variabilities of previous 10 working days using a skewness-based indicator⁴.

$$TTUR = \frac{T_{90} - T_{50}}{T_{50} - T_{10}}$$

$TTUR$ is time of day based: morning, afternoon, evening and night.

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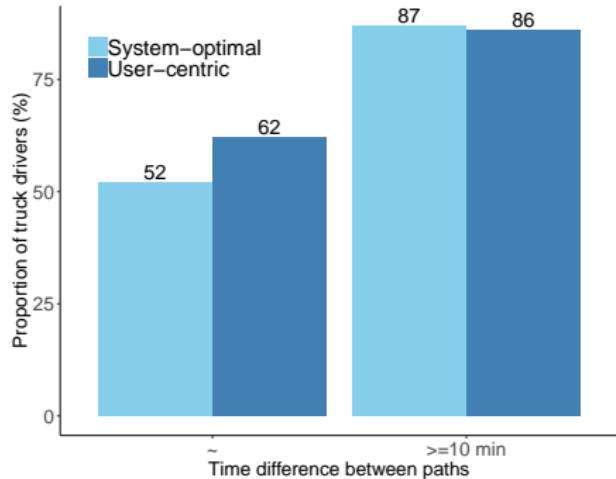
Model estimation

Parameters		Binary logit		Mixed logit	
		Value	t-test	Value	t-test
ITT (min)	Mean	-0.0866	-6.39	-0.152	-4.89
	SD			0.0197	0.21
TD (km)	Mean	-0.262	-19.54	-0.463	-6.33
	SD			0.512	4.40
TTUR	Mean	-0.00594	-1.10	-0.00899	-0.98
	SD			-0.00125	-0.53
LC	Mean	-0.229	-2.36	-0.414	-2.35
	SD			-0.493	-1.03
Number of observations			1671		1671
Number of individuals			1419		1419
$\mathcal{L}(\beta_0)$			-1158.249		-1158.249
$\mathcal{L}(\hat{\beta})$			-867.758		-848.337
$\bar{\rho}^2$			0.247		0.261

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Inefficiencies in routing decisions



User-centric: choose a path with least instantaneous travel time

System-optimal: choose a path with enough spare capacity and the instantaneous travel time on it should not be worse than that of shortest time path

Spare capacity of a path: We first compute section-specific density values. A path will have spare capacity if the maximum of all such density values is less than a nominal value (i.e., 25 veh/km/lane⁵).

⁵Y. Sugiyama, M. Fukui, M. Kikuchi, K. Hasebe, A. Nakayama, K. Nishinari, S.-i. Tadaki, and S. Yukawa (2008). "Traffic jams without bottlenecks: experimental evidence for the physical mechanism of the formation of a jam". In: *New journal of physics* 10.3, p. 033001

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Conclusions

- We model route choices of truck drivers by combining RP dataset and contextual information.
- Truck drivers significantly value time, distance and lane closures for their on-trip routing decisions.
- The mixed logit model shows that the estimate of travel distance varies significantly in the population.
- 38% of truck drivers do not take the shortest time path and 48% do not make system-optimal routing decision.
- The routing efficiencies of truck drivers can be improved by utilizing traffic management solutions.

Next steps

- To add multiple OD pairs in the present framework
- To use GPS data as a revealed preference data source
- To identify latent classes of truck drivers

Acknowledgments

