## Origin-Destination Matrix Estimation Based on Microsimulation and Optimization

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## Calibration of traffic scenarios



## Assignment and estimation



Fig. 1. Assignment and estimation process



Fig. 2. Assignment and estimation process

## Outline

- 1. User equilibrium
- 2. O/D matrix estimation
- 3. Results and discussion
- 4. Conclusions

## 1. User equilibrium

# Every driver chooses a route for which the cost (usually the travel time) is minimal



Fig. 3. Simple traffic network

## 1. User equilibrium



Fig. 4. Urban traffic network in SUMO



Fig. 5. Possible routes in the case of study

$$R_{(O_1,D_1)} = \{r_1, r_2, r_3\}$$
(3)

$$R_{(O_2,D_2)} = \{r_4, r_5, r_6\}.$$
(4)

## 1. User equilibrium



#### Fig. 6. Assignment using the shortest path





Fig. 8. Dynamic User Equilibrium for a simple urban traffic network

# $\dot{\boldsymbol{q}} = f(\boldsymbol{u})$ $\boldsymbol{u} = [V_{(O_1,D_1)}, V_{(O_1,D_2)}, V_{(O_2,D_1)}, V_{(O_2,D_2)}]^{\mathsf{T}}$ $\min_{\boldsymbol{u}} J(\boldsymbol{u}, \boldsymbol{q}, \hat{\boldsymbol{q}}) = \|\boldsymbol{q} - \hat{\boldsymbol{q}}\|_2$ subject to:

2. O/D matrix estimation

$$\dot{\boldsymbol{q}} = f(\boldsymbol{u}),$$
  
 $\boldsymbol{u}_{\min} \leq \boldsymbol{u} \leq \boldsymbol{u}_{\max},$  (6)

measured flow  $\boldsymbol{q} = [q_{r_1}, q_{r_2}, q_{r_3} + q_{r_4}, q_{r_5}, q_{r_6}]^{\mathsf{T}}$ estimated flow  $\hat{\boldsymbol{q}} = [\hat{q}_{r_1}, \hat{q}_{r_2}, \hat{q}_{r_3} + \hat{q}_{r_4}, \hat{q}_{r_5}, \hat{q}_{r_6}]^{\mathsf{T}}$ 



Fig. 9. Flow chart of the O/D estimation process

## 2. O/D matrix estimation



Fig. 10. Cost function obtained by varying  $V_{(O_1,D_1)}$  and  $V_{(O_2,D_2)}$ , and keeping the other values of the O/D matrix as  $V_{(O_1,D_2)} = V_{(O_2,D_1)} = 1000$ 

## 3. Results and discussion

#### TABLE I

#### REAL AND MEASURED TRIPS AND FLOWS FOR THE TEST CASE SCENARIO.



Variable	Real value	Measured value	Error (%)
$V_{(O_1,D_1)}$	1 200	1 158,9	3,42
$V_{(O_1,D_2)}$	1 000	1 045,7	4,57
$V_{(O_2,D_1)}$	1 000	998,5	-0,15
$V_{(O_2,D_2)}$	1 200	1 256,5	4,7
$q_{r_1}$	612	612	0
$q_{r_2}$	580	580	0
$q_{r_3} + q_{r_4}$	2 000	2 000	0
$q_{r_5}$	609	609	0
$q_{r_6}$	591	591	0

## 3. Results and discussion

## $\begin{tabular}{l} TABLE \ II \\ Initial flows obtained with the measured O/D matrix \\ \end{tabular}$



Variable	Real value	Initial conditions	Error (%)
$V_{(O_1,D_1)}$	1 200	1 158,9	3,42
$V_{(O_1,D_2)}$	1 000	1 045,7	4,57
$V_{(O_2,D_1)}$	1 000	998,5	-0,15
$V_{(O_2,D_2)}$	1 200	1 256,5	4,7
$q_{r_1}$	612	567	-7,4
$q_{r_2}$	580	582	0,3
$q_{r_3} + q_{r_4}$	2 000	2041	2,1
$q_{r_5}$	609	609	0
$q_{r_6}$	591	648	9,6

## 3. Results and discussion

### TABLE IIISolution obtained with the estimation method



Variable	Real value	Estimated value	Error (%)
$V_{(O_1,D_1)}$	1 200	1 198,3	0,1
$V_{(O_1, D_2)}$	1 000	1 042,0	4,2
$V_{(O_2,D_1)}$	1 000	958,6	4,2
$V_{(O_2, D_2)}$	1 200	1 208,9	0,7
$q_{r_1}$	612	608	0,6
$q_{r_2}$	580	583	0,5
$q_{r_3} + q_{r_4}$	2 000	2 000	0
$q_{r_5}$	609	609	0
$q_{r_6}$	591	600	1,5

## 4. Conclusions

- In this presentation, a method for estimating an O/D matrix based on microscopic simulation was described. This method consists in an optimization problem where the decision variables correspond to the trips comprising the O/D matrix, which is assumed to have a measurement error.
- This optimization iteratively runs a dynamic user equilibrium using the DUAITERATE tool found in the SUMO simulator, which implements the Gawron's Dynamic Traffic Assignment Model.
- The objective function tries to minimize the error between the flows obtained with the measured O/D matrix, and those obtained with the corrected O/D matrix.
- Simulation results showed the validity of the proposed method. Due to the dependence between the O/D matrix and the resulting traffic flows, it was difficult to consider measurement errors in the latter. Future improvements will evaluate these errors directly in the SUMO simulator.

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## Thank you.

## REFERENCES

[1] M. Jha, G. Gopalan, A. Garms, B. Mahanti, T. Toledo, and M. BenAkiva, "Development and Calibration of a Large-Scale Microscopic Traffic Simulation Model," *Transportation Research Record: Journal of the Transportation Research Board*, vol. 1876, pp. 121–131, jan 2004.

[2] C. Gawron, "An Iterative Algorithm To Determine the Dynamic User Equilibrium in a Traffic Simulation Model," *International Journal of Modern Physics C*, vol. 9, no. 3, pp. 393–407, 1998.

[3] S. Peeta and A. K. Ziliaskopoulos, "Foundations of Dynamic Traffic Assignment: The Past, the Present and the Future," *Networks and Spatial Economics*, vol. 1, no. 3, pp. 233–265, 2001.

[4] A. Acosta, J. Espinosa, and J. Espinosa, "Developing Tools for Building Simulation Scenarios for SUMO Based on the SCRUM Methodology," in *Proceedings of the 3rd SUMO User Conference*. Berlin: Deutsches Zentrum f"ur Luft- und Raumfahrt e.V., 2015, pp. 23–35.

[5] T. Van den Boom and B. De Schutter, *Optimization in Systems and Control*. TU Delft, 2007.