

Disaggregation of static OD-matrices for dynamic MATSim simulations

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Introduction and motivation

■ What do we do?

- Synthesizing coherent, distinguishable round trips from static OD matrices.
- Considering five sociodemographic groups.
- Reproducing data on the spatial distribution of home and work locations.
- Adding correct temporal structure to the round trips by reproducing activity start times and durations.

■ Why do we do it?

- Activity-based demand models are well-suited for demand synthesis of downstream agent-based models.
- However, they are data-intensive, especially up-to-date survey data is costly or sometimes unavailable.
- We bridge demand data from widespread four-step models with agent-based simulation to enable a fully disaggregated analysis.

Terminology and scenario

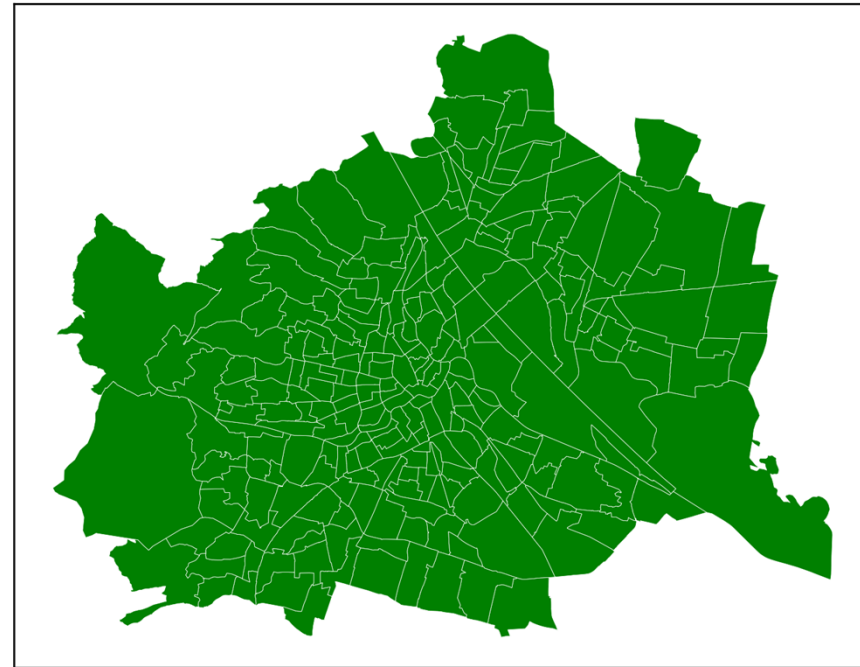
■ Round trip

- Round trip = home-based sequence of trips
- Defined by a list of departure locations and the according departure times

■ Vienna Scenario

- Population of 10,000 car drivers (2%)
- Spatial resolution: 250 zones
- Temporal resolution: 24 hours

Vienna scenario: Traffic analysis zones



Method

■ Measurement model

■ Error function: $R(x) = \sum_n |t_i - s_i(x)|$

- x ... list of n round trips
- t ... target value
- $s(x)$... sample value

■ Likelihood of a sample: $b(x) \sim e^{-\mu \cdot R(x)}$

- $R(x)$... error function
- μ ... weighting of the error function

■ Sampling from the probability distribution

■ Probability function: $\pi(x) = \frac{b(x)}{B}$

- $b(x)$... sampling weight of x
- B ... normalizing constant
... $B = \sum_x b(x)$

→ Metropolis-Hastings: $\pi(x) \sim b(x)$

- Flötteröd (2025) provides a detailed specification of the approach

Model specification

■ Likelihood function:

$$■ b(x) \sim \underbrace{e^{-E_{prior}(x)}} \cdot e^{-[\underbrace{\mu_{OD} \cdot E_{OD}(x) + \mu_{Location} \cdot E_{Location}(x) + \mu_{Time} \cdot E_{Time}(x)}]}$$

■ Maximum entropy prior

- The mean number of visited locations = 3.
- Includes plausibility constraints
 - Round trips must be completed within 24 hours.
 - Arrival time must be before departure time.

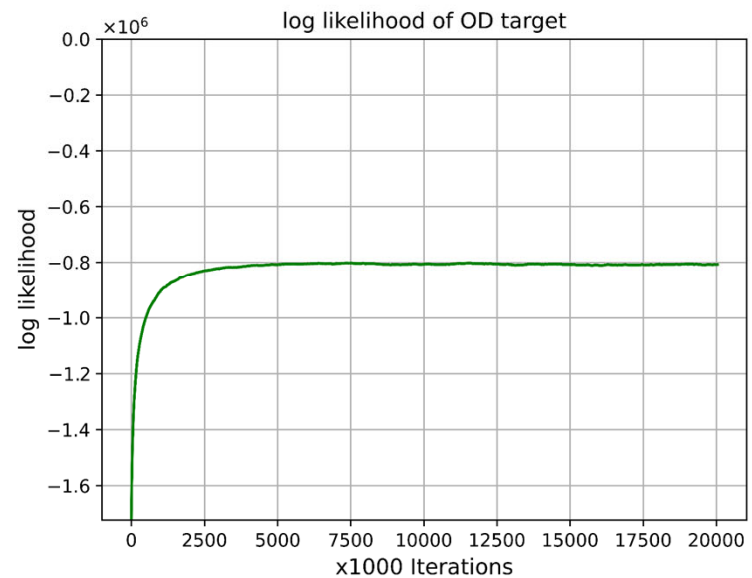
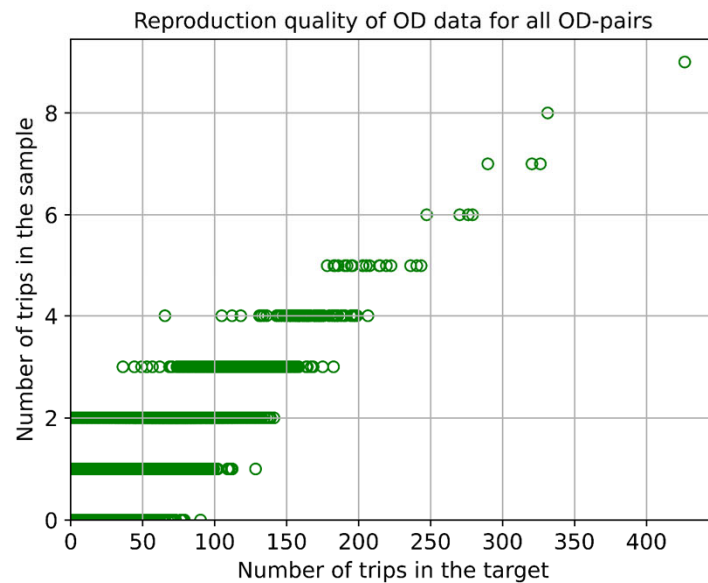
■ Reproduced data sets

- Static 24h car OD matrices of all population groups with car access (PTV VISUM Modell Verkehrsverbund Ostregion).
- Spatial distribution of home and work locations (PTV VISUM Modell Verkehrsverbund Ostregion).
- Activity durations and end times of home and work activities (Österreich Unterwegs 2014).

OD data reproduction quality

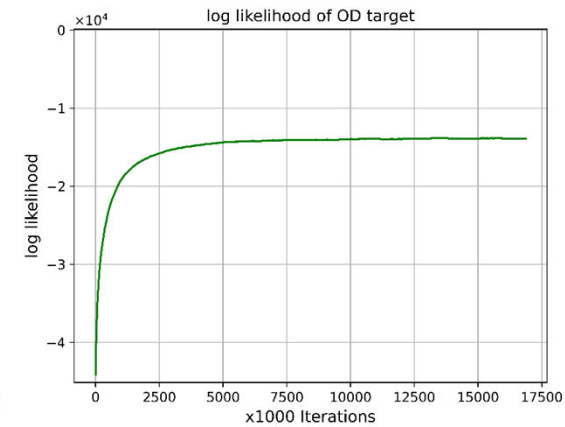
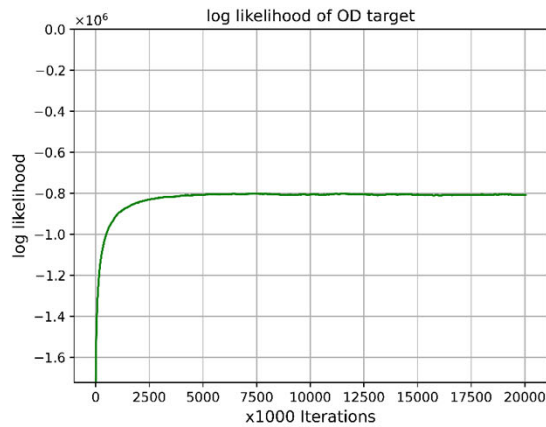
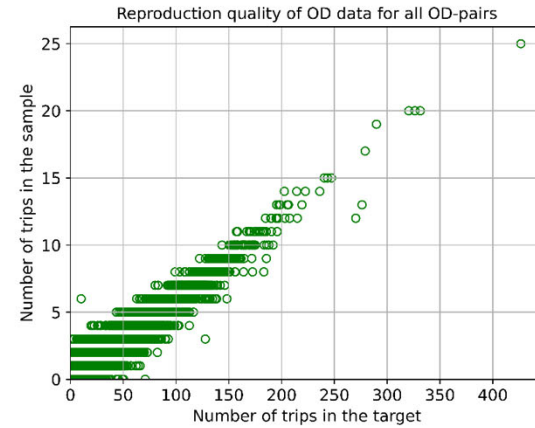
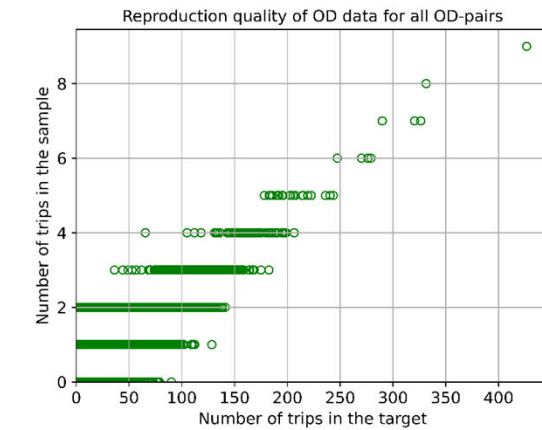
■ Likelihood function:

$$■ b(x) \sim e^{-[E_{prior}(x) + \mu_{OD} \cdot E_{OD}(x) + \mu_{Location} \cdot E_{Location}(x) + \mu_{Time} \cdot E_{Time}(x)]}$$



Increased OD data reproduction quality due to a larger population

10% population (50k plans) versus 2% (10k plans).

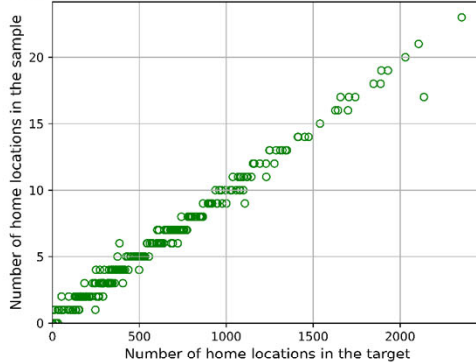


Home and work location data reproduction quality

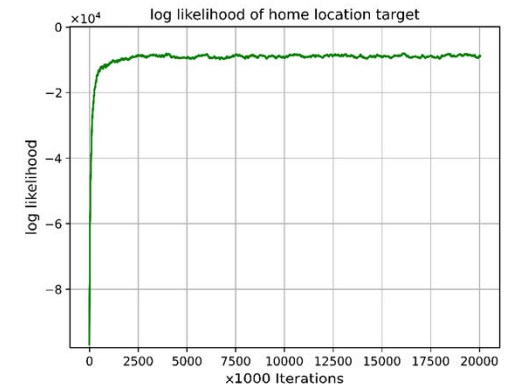
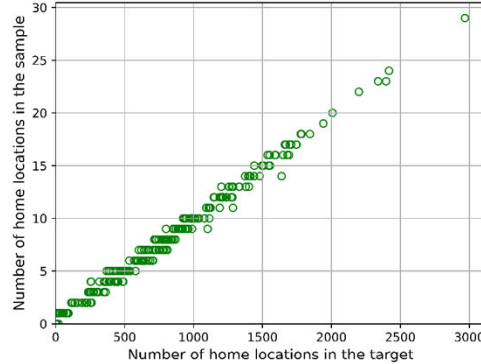
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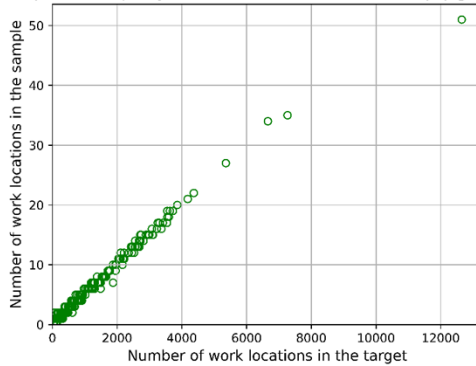
Reproduction quality of home location data for all zones (pop group 9)



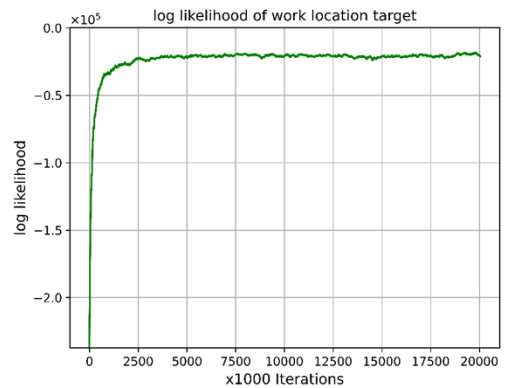
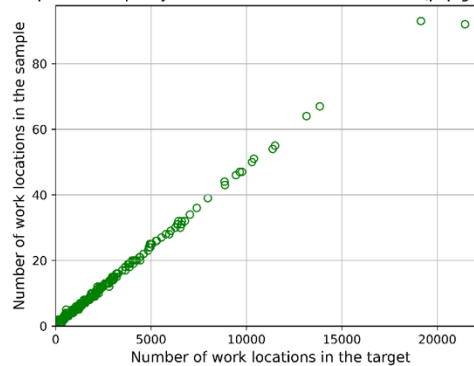
Reproduction quality of home location data for all zones (pop group 14)



Reproduction quality of work location data for all zones (pop group 6)



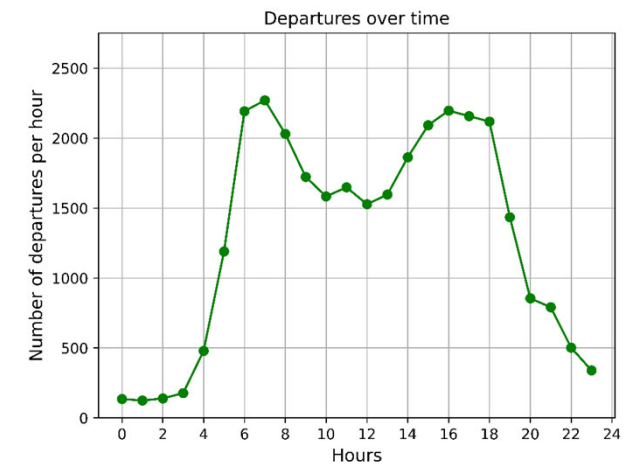
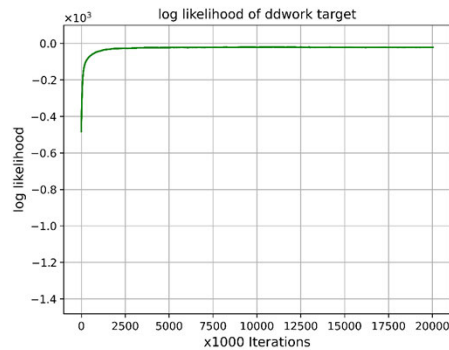
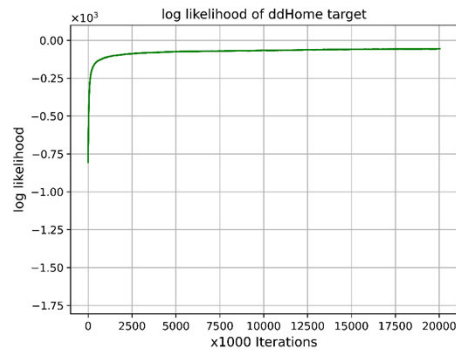
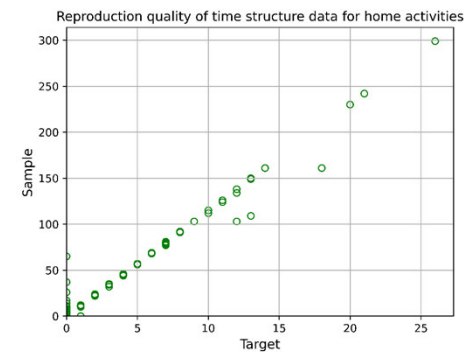
Reproduction quality of work location data for all zones (pop group 8)



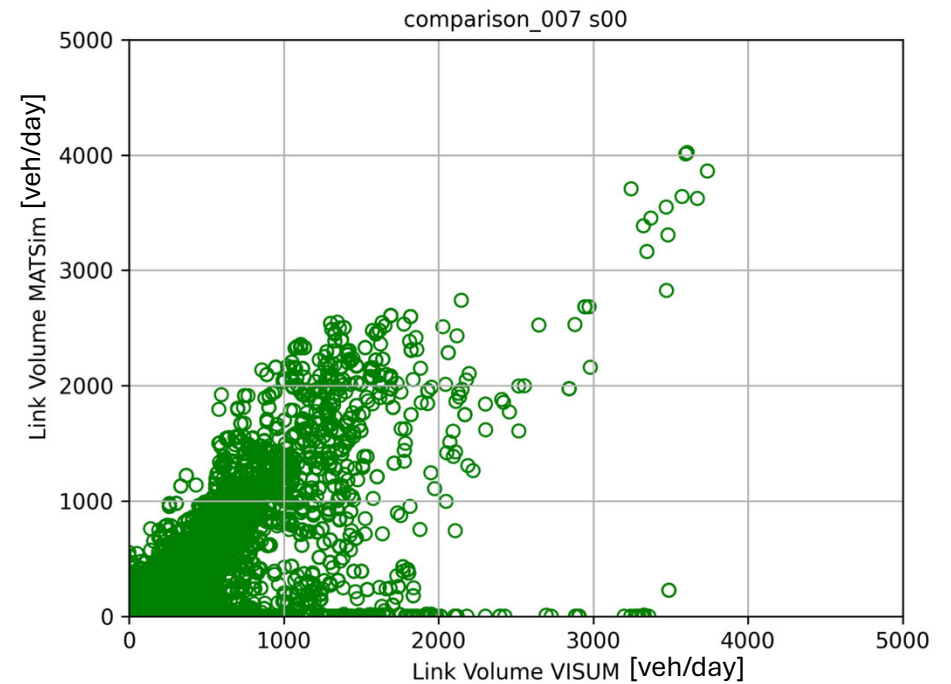
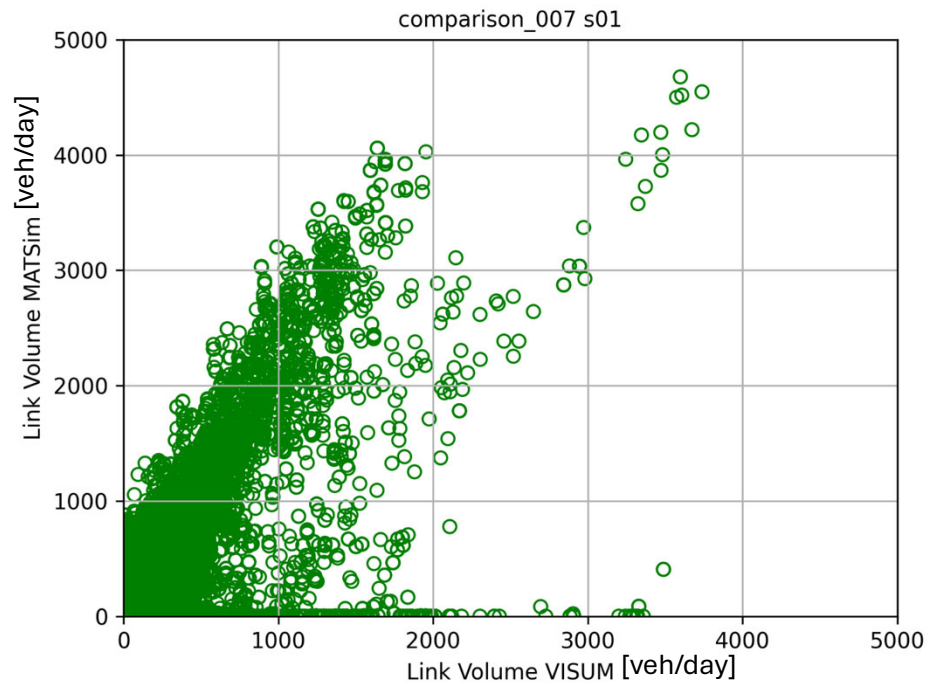
Time structure reproduction quality

■ Likelihood function:

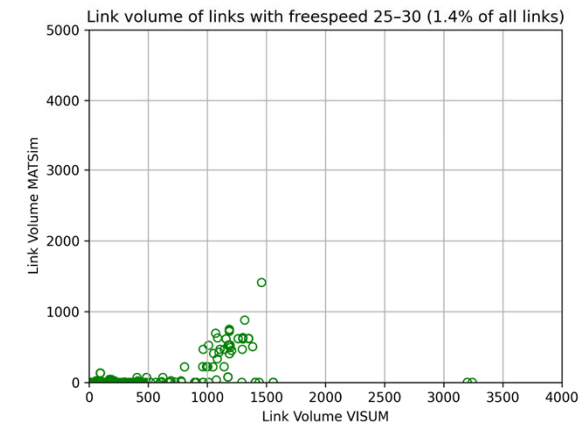
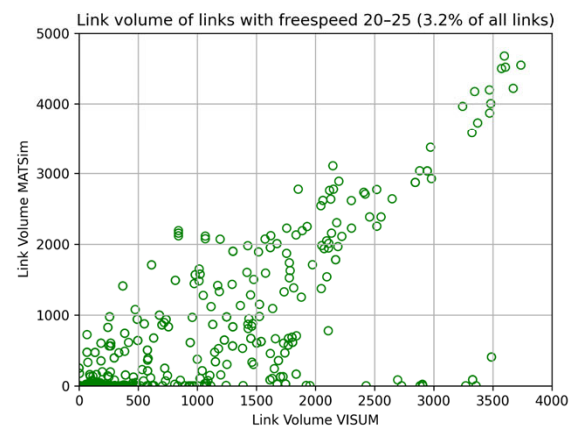
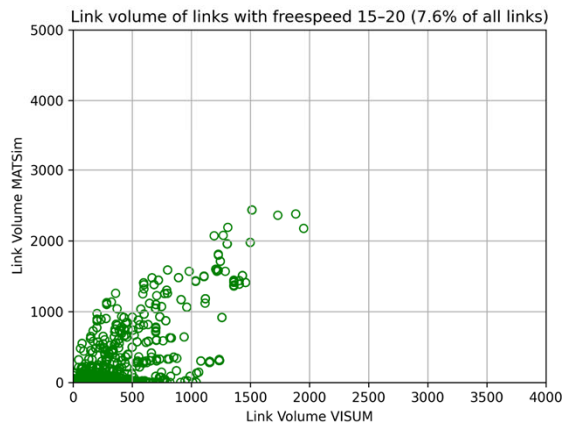
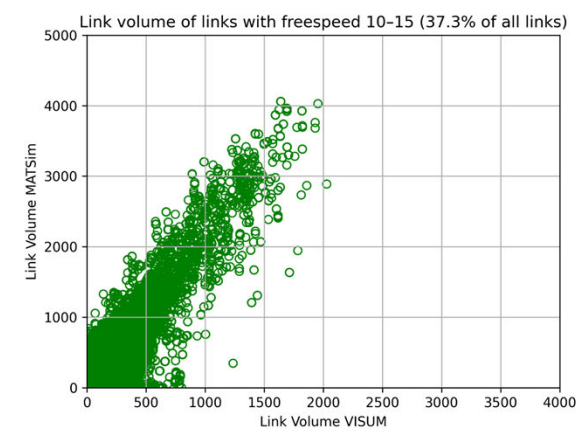
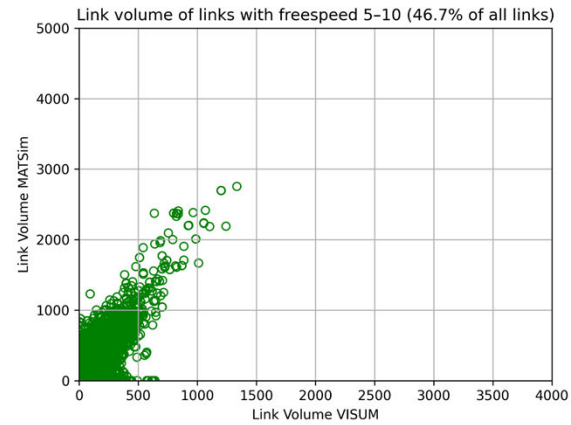
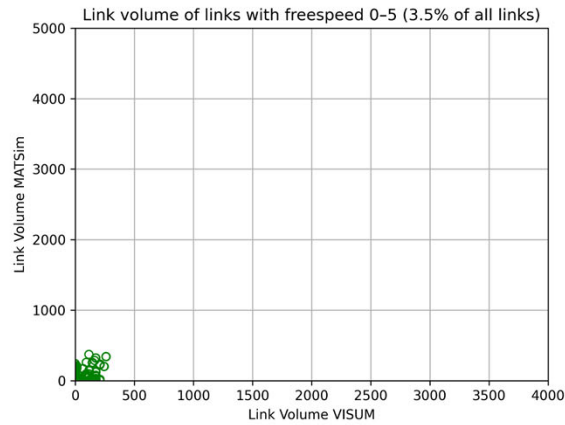
$$■ b(x) \sim e^{-[E_{prior}(x) + \mu_{OD} \cdot E_{OD}(x) + \mu_{Location} \cdot E_{Location}(x) + \mu_{Time} \cdot E_{Time}(x)]}$$



Comparing link volumes in MATSim to link volumes in VISUM



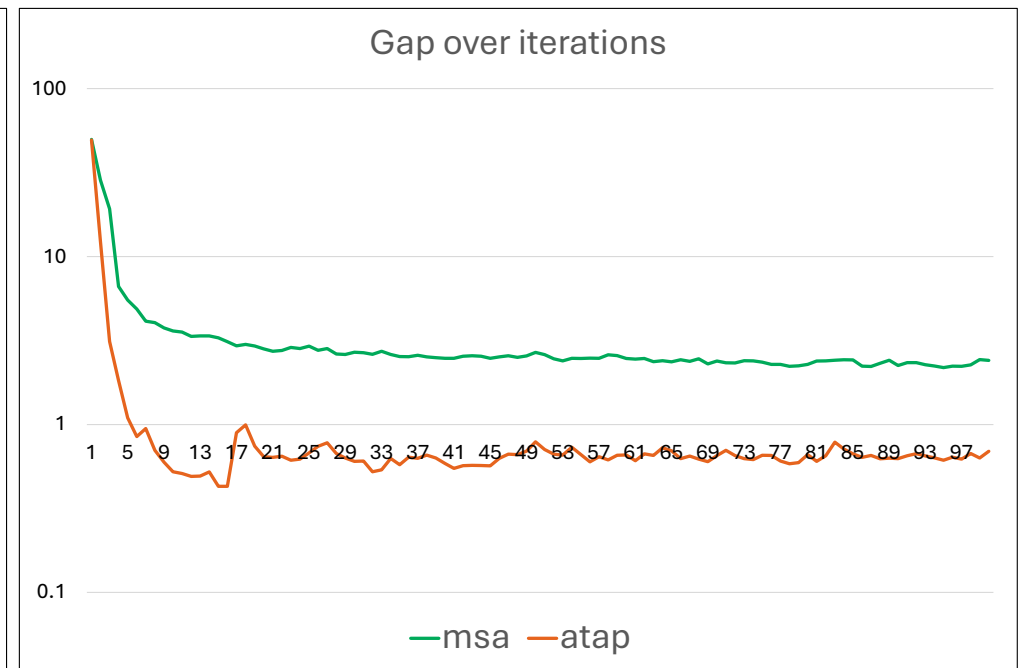
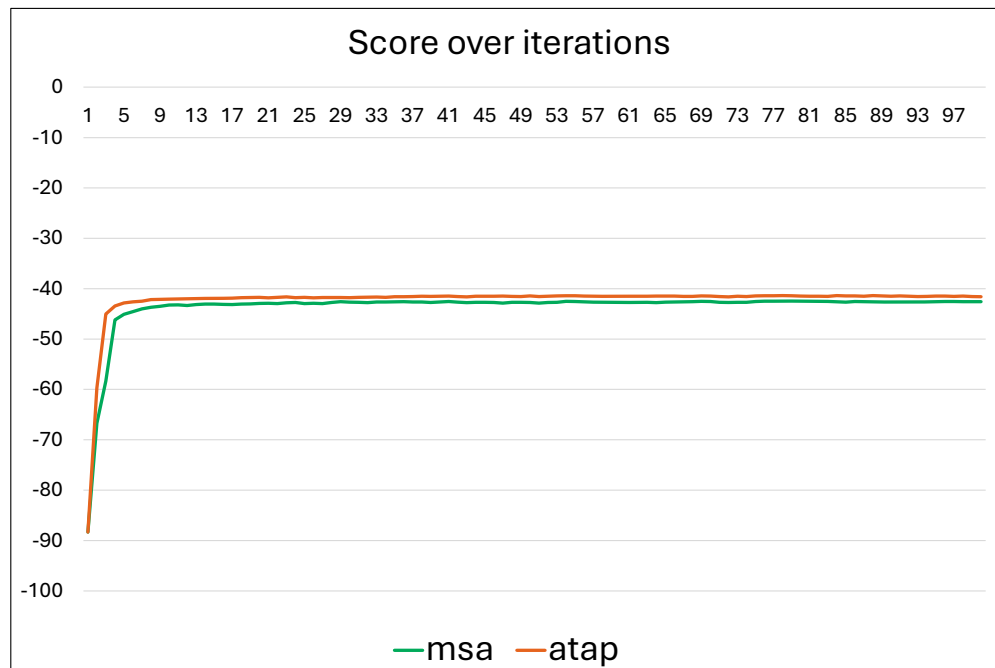
Comparing link volumes in MATSim to the VISUM model



Summary

- Data reproduction is precise and adjustable
- Population yields plausible simulation results
- Runtime depends largely on spatial resolution and population size
- Successfully transferred VISUM demand to MATSim
 - Spatially disaggregated
 - Enabling dynamic simulation
 - Enabling individual-level analysis

Digression: Score as a measure for convergence



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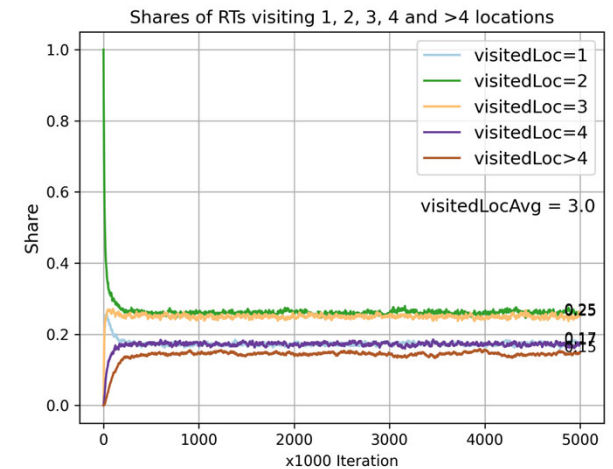
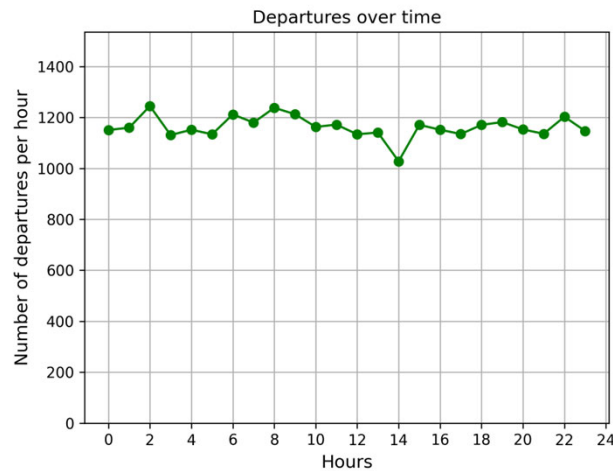
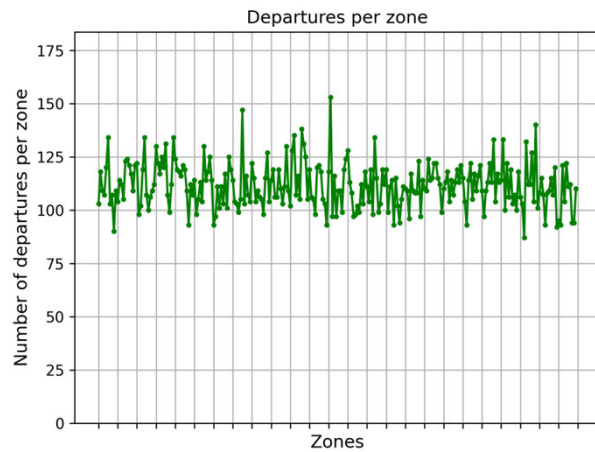
References:

Flötteröd, G. (2025). An operational alternative to origin/destination matrices. 13th Symposium of the European Association for Research in Transportation, Munich, Germany, 2025

Prior sampling (no reproduction of data sets)

■ Likelihood function:

$$■ b(x) \sim e^{-E_{prior}(x)}$$



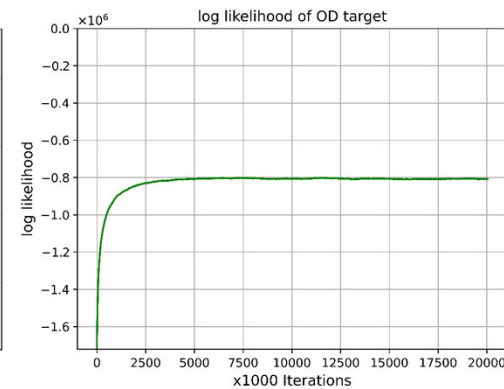
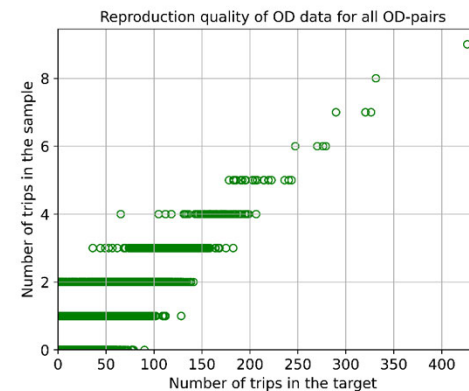
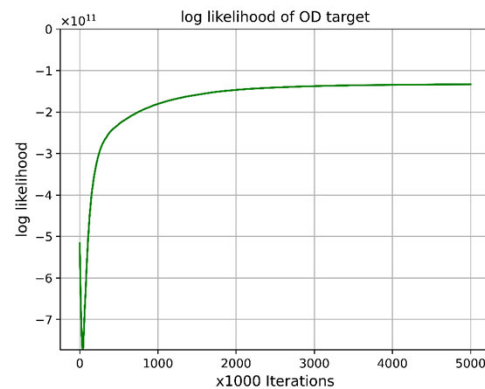
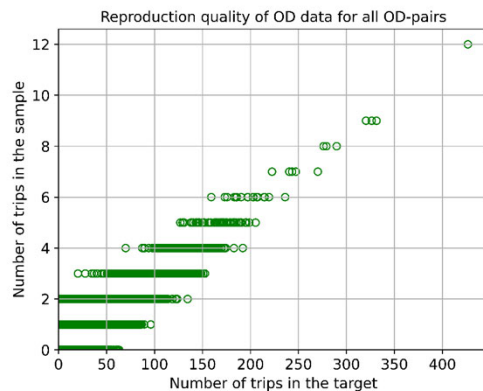
Using Gaussian distributed sampling weights

- Up to this point, all results were based on sampling weights that were two-sided exponentially distributed

- $b(x) \sim e^{-[\mu \cdot R(x)]}$

- Gaussian distributed sampling weight provide some advantages

- $b(x) \sim e^{-[0.5 \cdot \mu \cdot R(x) \cdot R(x)]}$



Backup slide

■ Error function of OD repro:

- $E_{OD}(x) = \sum_{rs} |t_{rs} - h \cdot s_{rs}|$
 - t_{rs} ... trips from zone r to s in the target OD matrix
 - s_{rs} ... trips from zone r to s in the sample OD matrix
 - h ... scaling factor; $h = \frac{\sum_{rs} t_{rs}}{\sum_{rs} s_{rs}}$

■ Population groups:

pop group 06	19-35 years	working	Car access
pop group 07	19-35 years	other	Car access
pop group 08	35-65 years	working	Car access
pop group 09	35-65 years	other	Car access
pop group 14	>65years	Not specified	Car access