

Including joint trips in a Multi-Agent transport simulation

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Agenda

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Inclusion of joint trips in MATSim

Results

Next steps

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- ▶ joint trip: several individuals traveling in the same private vehicle
- ▶ joint traveling behaviour is an important behaviour
 - ▶ occurs frequently in households
 - ▶ some policies aim at encouraging such a behaviour
 - ▶ HOV lanes
 - ▶ car-pooling services
- ▶ currently, few means of predicting such a behaviour exist
- ▶ traffic simulation is an important tool for policy evaluation
- ▶ micro-simulation, by simulating individuals explicitly, allows to simulate a wide range of behaviours

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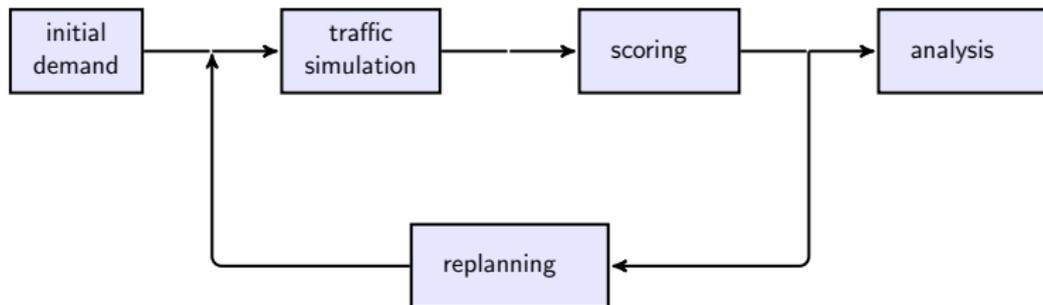
The MATSim software

- ▶ MATSim: Multi-Agent Transport Simulation
- ▶ open source software (GNU GPL)
- ▶ written in Java
- ▶ Mainly developed at ETHZ, TU Berlin, Senozon

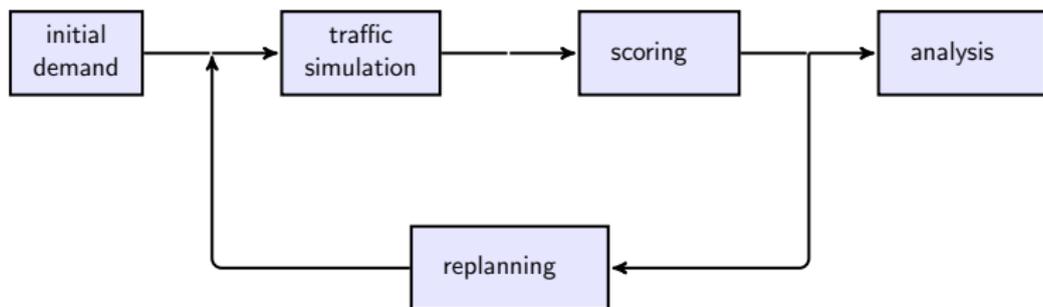
The MATSim process in a nutshell

- ▶ state of traffic in an average day: (stochastic) user equilibrium
- ▶ a strategy (daily plan) can be modified by changing dimensions easy to change in the short-term (day-to-day)
- ▶ dimensions corresponding to long-term changes (eg. home and work places) are exogenously determined (boundary conditions)
- ▶ search process: “co-evolutionary” algorithm
 - ▶ works with a population of heterogeneous agents
 - ▶ each agents i tries to solve $\max_{p_i \in P_i} U(p_i | p_{-i})$
 - ▶ influence of p_{-i} : via congestion

The MATSim process steps



The MATSim process steps



- ▶ replanning:
 - ▶ creation of new plan
 - ▶ random mutation
 - ▶ optimisation given the travel times in the previous iteration
 - ▶ selection of a past plan based on experienced score
 - ▶ probabilistic (RUM)
 - ▶ deterministic (best past plan)

MATSim and joint trips (1): MATSim

- ▶ remember the agent's problem?
 - ▶ $\max_{p_i \in P_i} U(p_i | p_{-i})$
 - ▶ $|p_{-i}$ estimated via “mobility simulation”
 - ▶ $|p_{-i}$ actually differs between iterations
- ▶ remember MATSim's process?
 - ▶ agents actually “knows” $U(p_i | p_{-i}) \approx U_I(p_i)$
 - ▶ $|p_{-i}$: effect of experienced congestion in the last execution (iteration I): “empirical” knowledge
 - ▶ this is usually valid enough:
 - ▶ changing plans of few agents only has a minor influence on the state of traffic
 - ▶ actually reproduces human learning

MATSim and joint trips (2): joint trips

- ▶ what about joint travel?
 - ▶ $p_{-i} = \{p_j\}_{j \in S_i} \cup \{p_k\}_{k \notin S_i}$ with S_i the set of co-travelers
 - ▶ S_i typically very small
 - ▶ each $\{p_j\}_{j \in S_i}$ has a lot of influence
 - ▶ participation in joint travel
 - ▶ departure time for the joint trip
 - ▶ “utility transfers” (altruistic behaviour, monetary compensation)
 - ▶ individuals typically aware of (relevant part of) $\{p_j\}_{j \in S_i}$ (agreement): “theoretical” knowledge
- ▶ necessary to find a way to actually correlate plan selection based on $U(p_i | \{p_j\}_{j \in S_i})$

MATSim and joint trips (2+1): joint trips in MATSim

To solve those problems, the equilibrium is defined over groups of agents:

- ▶ new “aggregated” data structures are defined
 - ▶ Person \rightarrow Clique
 - ▶ groups Persons which (can) travel together ($i \in \mathcal{C} \Rightarrow S_i \subset \mathcal{C}$)
 - ▶ maintains a set of JointPlans
 - ▶ Plan \rightarrow JointPlan
 - ▶ groups individual plans, always selected together
 - ▶ is affected a score (currently, the sum of the scores of individual plans: full utility transfers)
- ▶ replanning modules work at the aggregated level (competing cliques)
- ▶ joint trip: *access leg* \rightarrow *pick-up* \rightarrow *shared leg* \rightarrow *drop-off* \rightarrow *egress leg*
- ▶ mobility simulation works with individuals

Remarks on joint trip generation

- ▶ most of the joint-trip generation approaches in the literature are specific to households
- ▶ in the context of MATSim, three approaches are possible:
 - ▶ generation *a priori* (exogeneous)
 - ▶ allows to adapt to different contexts (household, car-pool. . .)
 - ▶ joint trips not part of the equilibrium
 - ▶ generation during the iterations (endogeneous)
 - ▶ joint trips truly part of the equilibrium
 - ▶ increases the search space size
 - ▶ “hybrid”
 - ▶ a limited set of possible joint trips is identified beforehand
 - ▶ joint trips from this set can be selected/unselected during the optimisation

the replanning step in more details

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the replanning step in more details

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- ▶ optimisation of activity durations and mode
 - ▶ uses Tabu Search
 - ▶ estimates travel times based on the events of the previous simulation run
 - ▶ mode is optimised at the subtour level
 - ▶ plans are synchronised by penalising unsynchronized plans

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- ▶ (joint trips selection)
- ▶ re-routing

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- ▶ re-routing
- ▶ best plan selection

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- ▶ ...

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The Scenario

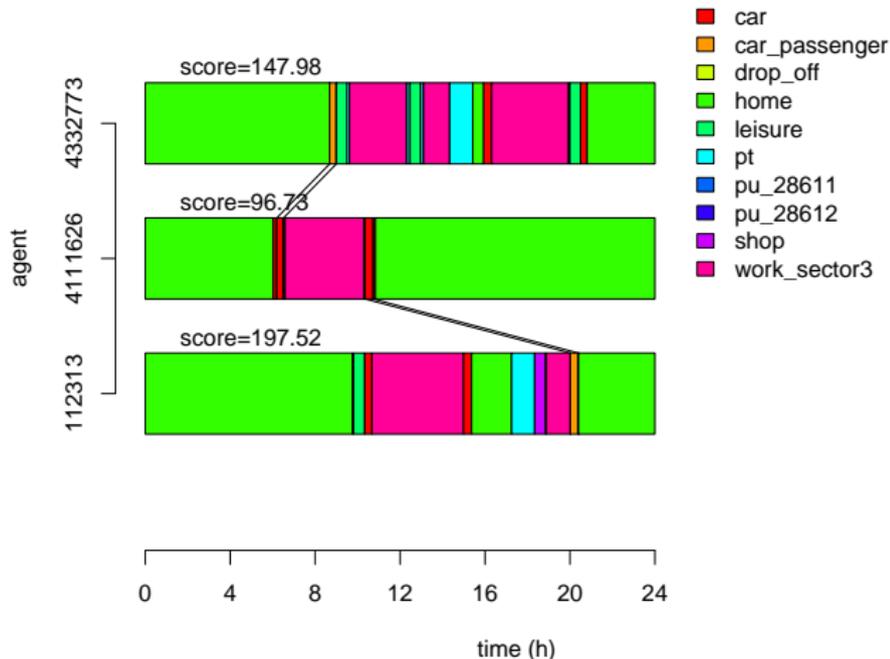
A scenario for the urban area of Zürich:

- ▶ 10% sample
- ▶ car-pooling matings computed by a partner
 - ▶ maximum detour time with time windows
- ▶ “default” (*i.e.* uncalibrated) utility parameters

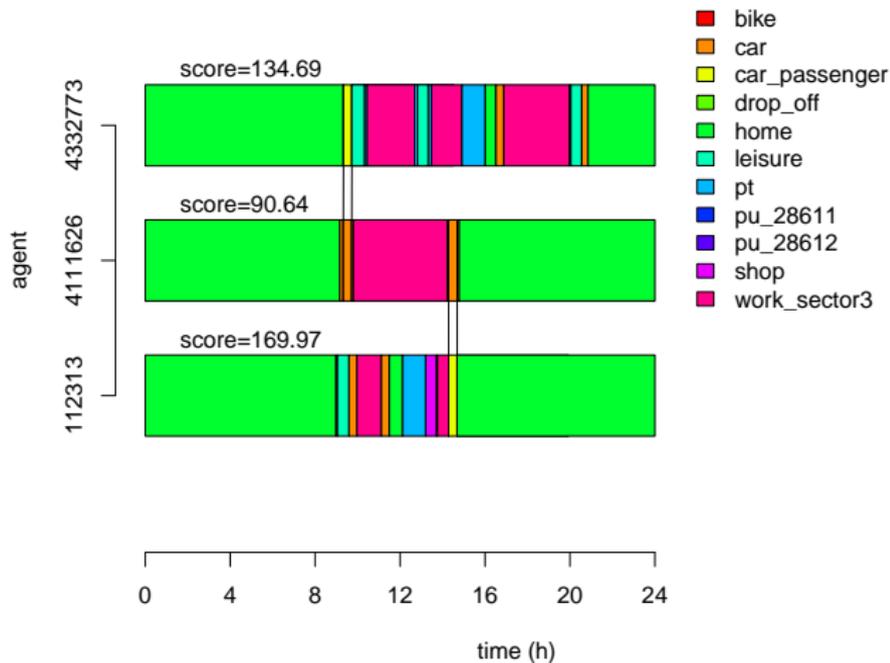
Influence of constraints

- ▶ two major constraints implied by a joint trip:
 - ▶ synchronisation
 - ▶ mode chaining
- ▶ what influence do they have on the outcome?
- ▶ 3 runs:
 - ▶ no synchronisation, no mode chaining constraints
 - ▶ no synchronisation, mode chaining constraints
 - ▶ synchronisation, mode chaining constraints

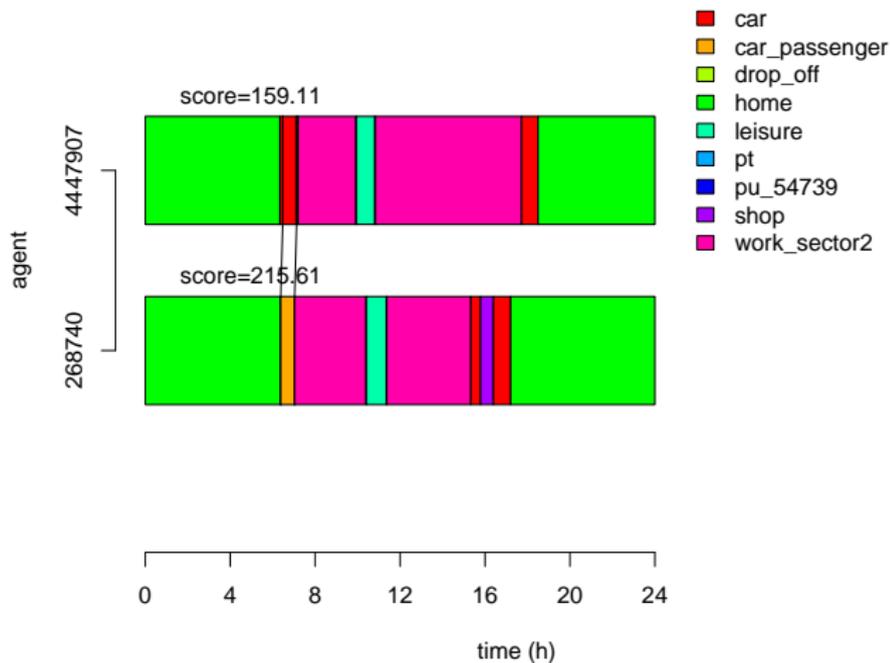
Influence of constraints: synchronisation



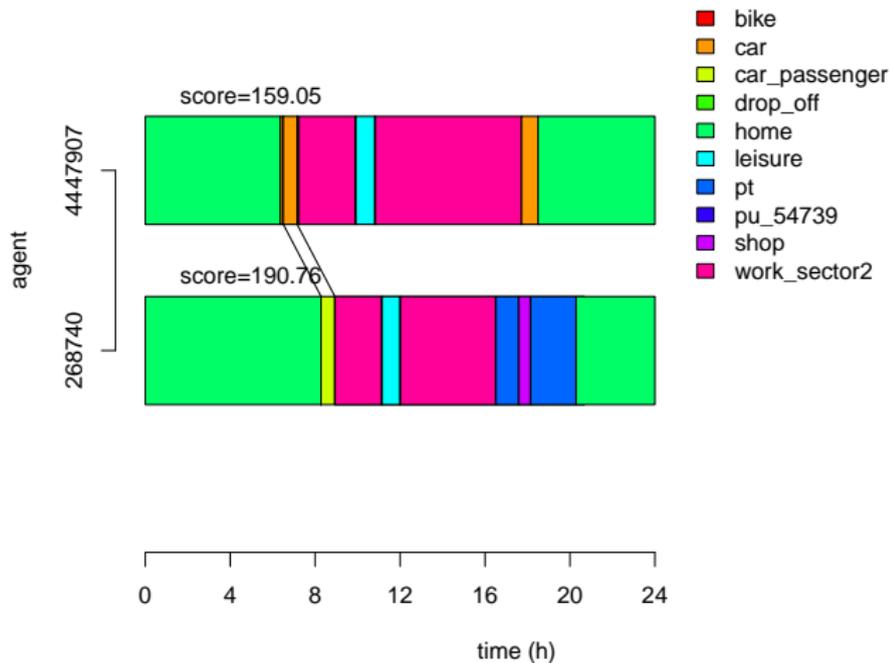
Influence of constraints: synchronisation



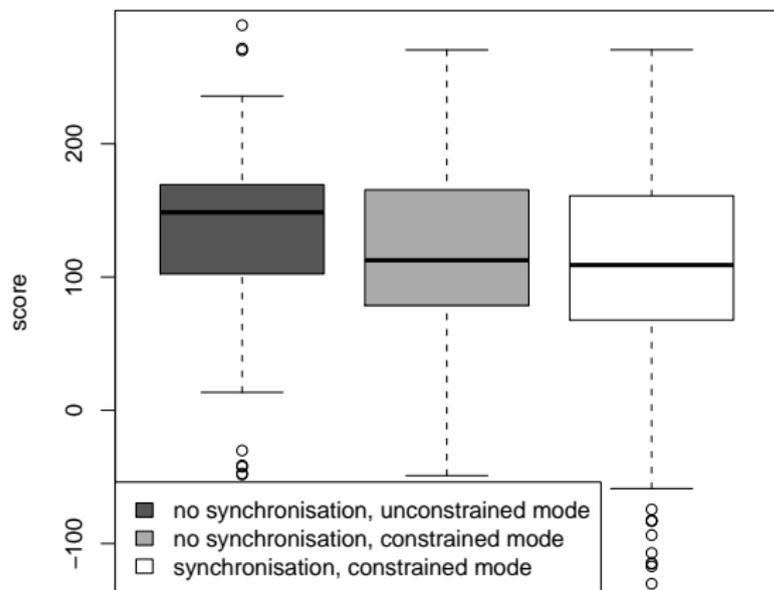
Influence of constraints: mode chaining



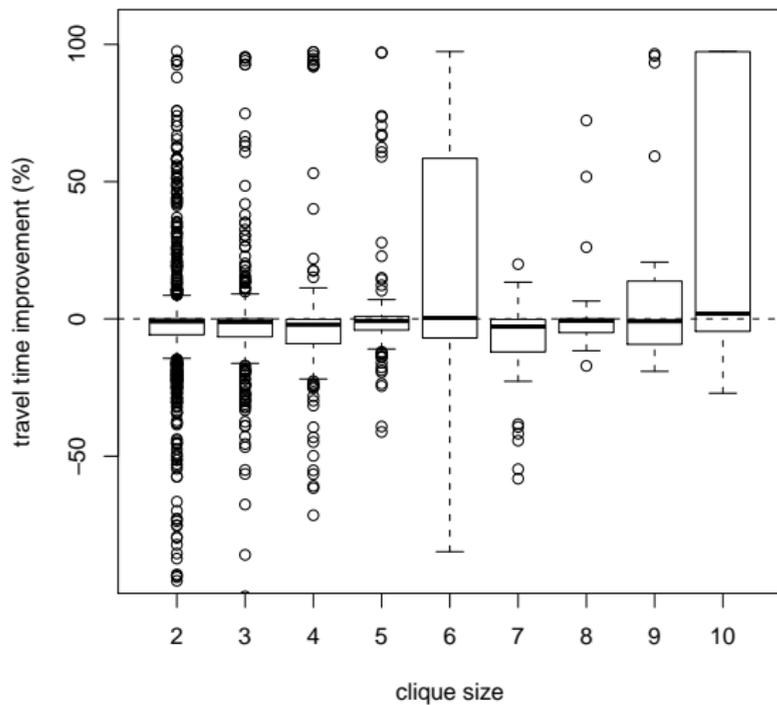
Influence of constraints: mode chaining



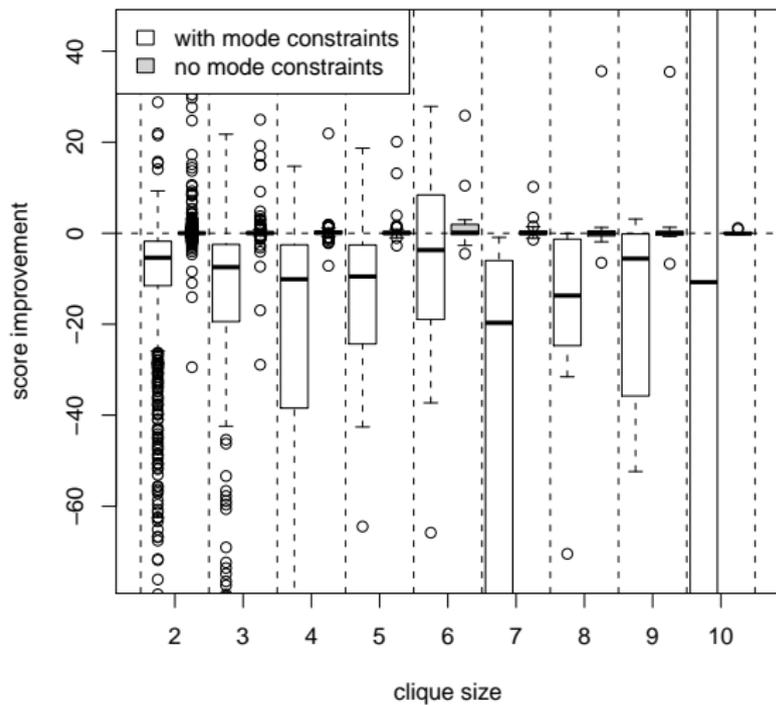
Influence of constraints: scores



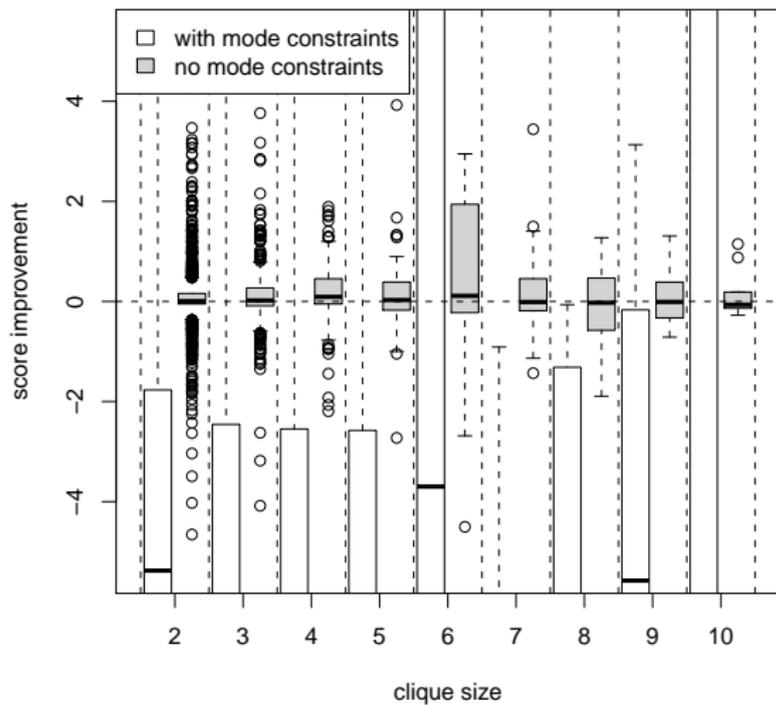
Travel time improvements



Score improvements



Score improvements



What can we get from those results?

- ▶ major influence of mode chaining constraints on the attractiveness of joint trips
- ▶ need to consider other dimensions than travel time in attractiveness of joint trips vs other modes
 - ▶ monetary costs (fuel, tolls...)
 - ▶ car availability (household)
 - ▶ willingness to share time with social contacts

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- ▶ joint trip generation/selection
 - ▶ initial demand model
 - ▶ replanning-level (for small cliques, eg. households, or social-network-based)
- ▶ include monetary cost in utility function
- ▶ relaxation of the “utility transfers” hypothesis
 - ▶ actually use $U(p_i | \{p_j\}_{j \in S_i})$ to correlate plan choice
 - ▶ deterministic: iterative removal of dominated strategies
 - ▶ stochastic: joint choice probability
 - ▶ main issue: estimate efficiently conditional utility for all possible combinations
 - ▶ finer modeling of social contacts and willingness to help
 - ▶ allows more complex networks than isolated cliques
- ▶ extend the Clique concept to represent households
 - ▶ car availability
 - ▶ joint activities
- ▶ validation against aggregate data

Thank you for your attention

Any question?