

Explicitly Correlating Agent's Daily Plans in a Multiagent Transport Simulation: Towards the Consideration of Social Relationships

Thibaut Dubernet

PhD Defense
Institute for Transport Planning and Systems (IVT)
ETH Zurich

3 Mai 2017, Zurich, Switzerland

Introduction

Equilibrium With Joint Plans

Generation of a Synthetic Social Network

Location Choice with Preference for Group Activities

Conclusions

Motivation

- ▶ Leisure activities represent a growing share of travel in developed countries
- ▶ Location of leisure activities is difficult to predict
 - ▶ Depends on variety of unobserved factors
 - ▶ Best one can do in microsimulation: add random noise (Horni 2013)
- ▶ Most important motivation for leisure: social contact
- ▶ Other important behaviors rely on *joint decisions*, particularly inside the household
- ▶ Classical equilibrium formulation of transport systems do not allow to represent such behaviors

Aims

- ▶ Design a model to represent joint decisions in microsimulations
- ▶ Design a methodology to generate a *synthetic social network*
- ▶ Test the resulting model for travel to leisure locations
- ▶ Additionally in thesis:
 - ▶ Compare two *solution concepts* in the context of households
 - ▶ Estimate *car pooling potential*

Introduction

Equilibrium With Joint Plans

Generation of a Synthetic Social Network

Location Choice with Preference for Group Activities

Conclusions

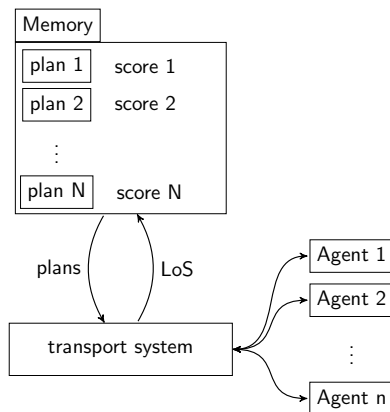
Equilibrium for Transport System

- ▶ Classical modeling assumption: rational agents
 - ▶ Preferences over routes/modes/daily plans, represented by utility
- ▶ Agents influence each other's utility (congestion, crowding)
 - ▶ Game theoretic view
 - ▶ need *Solution Concept*
- ▶ Classical way to model transport systems: some variant of Nash equilibrium
 - ▶ “no agent can *unilaterally* improve its utility”
 - ▶ Usual in traffic assignment: UE, SUE...
- ▶ MATSim: on the level of daily plans
 - ▶ Routes
 - ▶ Modes
 - ▶ Departure Time
 - ▶ (Sequence, Secondary Location...)

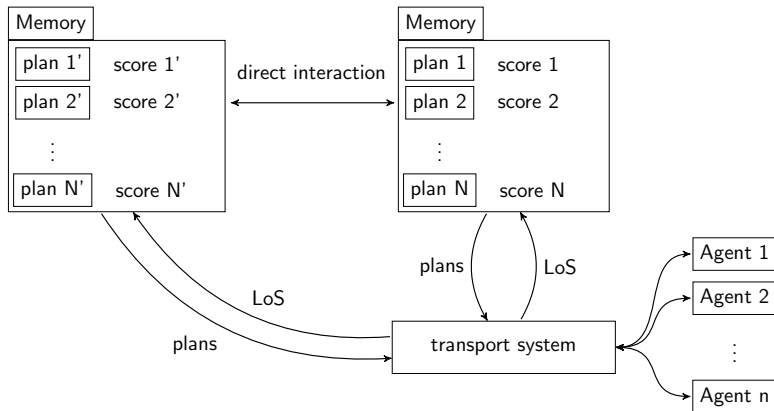
Joint Decisions

- ▶ Two kinds of co-dependence of plans' utilities
 - ▶ Field effect (congestion, crowding)
 - ▶ Direct effects (co-travelers)
- ▶ Two kinds of processes to take them into account (equilibrium):
 - ▶ Iterative learning
 - ▶ Binding agreement
- ▶ Intuitively, a decision that relies on *binding agreements*
 - ▶ for instance: going together to the cinema, sharing a ride...
- ▶ New concept in MATSim: Joint Plan
 - ▶ enforces binding agreements

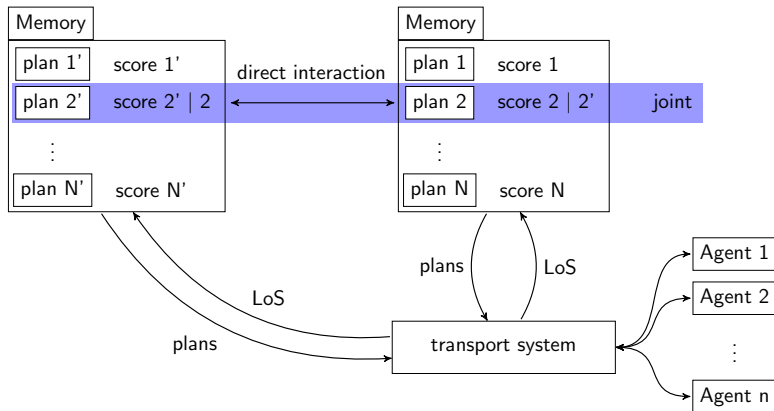
Joint MATSim: World of an Agent



Joint MATSim: World of an Agent



Joint MATSim: World of an Agent



Solution Concept for Joint Decisions

- ▶ Variants of Nash unrealistic
- ▶ Given an allocation of daily plans to agents, a group of agents represents a *blocking coalition* if:
 - ▶ they form a *clique*
 - ▶ they all can improve their (expected) utility by changing their daily plan *simultaneously*
- ▶ Includes Nash equilibrium as a special case (empty social network)

MATSim Process Without Blocking Coalitions

- ▶ Field effects taken into account as usual
- ▶ Direct effects are solved *at each iteration*
 - ▶ Scores randomized
 - ▶ Joint plans selected such that there exists no blocking coalition given the randomized scores

Introduction

Equilibrium With Joint Plans

Generation of a Synthetic Social Network

Location Choice with Preference for Group Activities

Conclusions

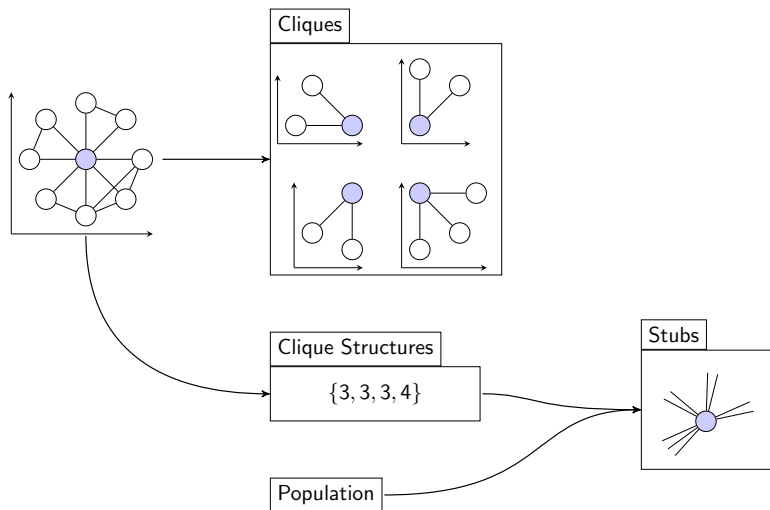
Generation of a Synthetic Social Network

- ▶ Simulating joint mobility behaviors requires *realistic synthetic social networks*
- ▶ Important characteristics:
 - ▶ Homophily (socio-demographic similarity of contacts)
 - ▶ Geography
 - ▶ Clustering (friends of friends tend to be friend)
 - ▶ in particular *clique size*
 - ▶ Large scale (several Mio agents)
- ▶ Lots of work on random networks, but no approach fulfills those requirements
 - ▶ Graph building procedures: no control
 - ▶ Statistical models: complex, poor and slow at generation

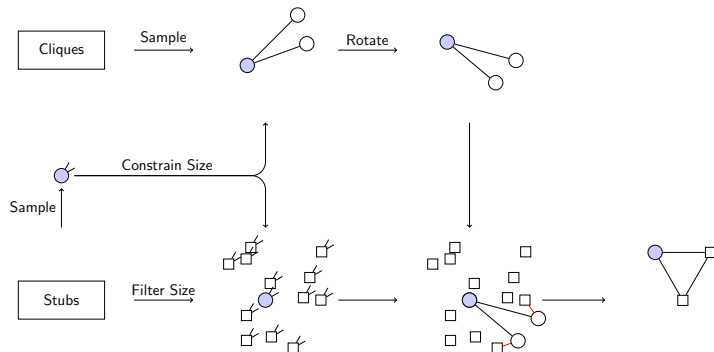
Cliques Sampling Algorithm

- ▶ Algorithm works on *ego-centric networks*
- ▶ The idea is to *sample from the observed distribution*
- ▶ Works on *positions* in a metric space, including space and socio-demographics
- ▶ Idea:
 - ▶ Sample *sets of positions* in this space
 - ▶ Link agents that are closest to those positions
- ▶ Exploits large scale of the population: there should be agents “relatively close” to most points of the space

Preparation



Cliques Sampling



Results

- ▶ Social network data from Swiss snowball sample (Kowald 2013)
- ▶ Algorithm applied to Swiss synthetic population (8,230,971 agents)
- ▶ 100 samples from 1% to 100% to test scalability
- ▶ Consider age in 5 years classes
- ▶ Consider gender
- ▶ Distance such that difference in socio-demographics more important than spatial distance

Results

- ▶ Controlled parameters (distance, age difference, gender difference, clique size) well reproduced
- ▶ Degree well reproduced (not too much overlap in data)
- ▶ Global network statistics stable with sample size and realistic:
 - ▶ One giant connected component with more than 99% of agents
 - ▶ Average social distance close to 6
- ▶ Heterogeneity of ego-centric networks pretty well reproduced, except:
 - ▶ Under-estimate number of age-homogeneous networks
 - ▶ Under-estimate number of network with high share of friends 100 and 130km away
- ▶ Spatial concentration of social contacts well reproduced

Introduction

Equilibrium With Joint Plans

Generation of a Synthetic Social Network

Location Choice with Preference for Group Activities

Conclusions

Location Choice with Preference for Group Activities

- ▶ Can social networks help reduce the amount of noise needed to get realistic travel distance to leisure?
- ▶ Use the elements presented until now
- ▶ “Proof-of-concept” simulation for joint leisure in Switzerland
- ▶ “Simplified MATSim framework”: no iteration, no externalities, only direct influence

Leisure Travel in Switzerland

Most important leisure activity types (National Travel Survey 2010, in % of leisure activities):

Activity Type	Day of the Week			
	Mon. – Fri.	Sat.	Sun.	All
Restaurants	24.8	20.7	13.5	22.2
Outdoor (non sport)	19.2	17.2	26.0	20.0
Visits	17.7	21.5	22.4	19.2
Sports	13.0	8.6	8.8	11.5
Total	74.7	68.0	70.7	72.9

Setting

- ▶ 1% Swiss sample, with social network
- ▶ Each agent “knows” 30 random locations in a radius of 30km around home, plus home
- ▶ Agents choose one of the following plans:
 - ▶ go alone at one known location (including home)
 - ▶ go as a group of 2 to one of the locations known to the participants (including home)
 - ▶ go as a group of 3 to one of the locations known to the participants (including home)
- ▶ Search for a state without blocking coalition

Preferences

- ▶ Preferences of agents represented by utility including:
 - ▶ cost of distance
 - ▶ personal preference for location $\varepsilon_{e,l} \sim \mathcal{N}(0, \sigma)$
 - ▶ preference for social contact α
 - ▶ personal preference for particular contacts
 $\eta_{e,a} = \eta_{a,e} \sim \mathcal{N}(0, \vartheta)$

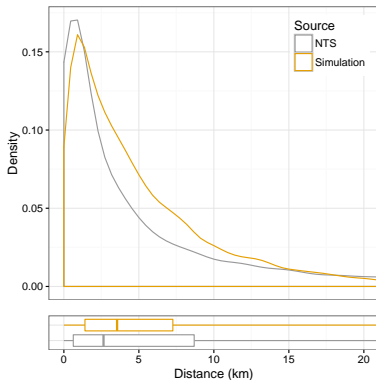
$$U(e, l, \mathcal{A}) = -d_{e,l} + \varepsilon_{e,l} + \sum_{a \in \mathcal{A}} (\alpha + \eta_{e,a})$$

- ▶ Use the following values:
 - ▶ $\alpha \in \{0 \text{ km}, 5 \text{ km}, 10 \text{ km}\}$
 - ▶ $\vartheta \in \{0 \text{ km}, 1 \text{ km}, 5 \text{ km}, 10 \text{ km}\}$
 - ▶ $\sigma \in \{0 \text{ km}, 1 \text{ km}, 5 \text{ km}, 10 \text{ km}\}$

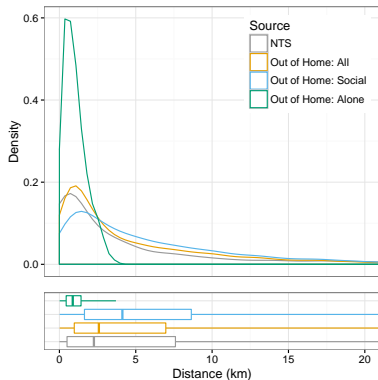
Summary Results

- ▶ Given positive α , traveled distance for visit very insensitive to parameters
- ▶ Traveled distance for Out-of-Home gets larger with σ and θ
- ▶ For higher α or θ , σ has little influence on traveled distance, but on the number of out-of-home activities
- ▶ Complex interactions between θ and σ : increasing σ makes it more difficult to find an agreement, until it is so large it is possible to find solutions acceptable for all

Traveled Distance, Visit, $\alpha = 10, \theta = 5, \sigma = 1$



(a) Visits



(b) Out of Home/Restaurant (WE)

Introduction

Equilibrium With Joint Plans

Generation of a Synthetic Social Network

Location Choice with Preference for Group Activities

Conclusions

Discussion

- ▶ Proposed algorithm able to generate realistic social network in reasonable time
 - ▶ Important non-controlled for characteristics well reproduced
 - ▶ Difficult to do better without a more “semantic” model
 - ▶ Satisfying for our purpose
- ▶ Realistic social network results in realistic traveled distances for visits
 - ▶ Not very sensitive to parameters
- ▶ Desire to meet social contact allows to reduce the level of noise needed on top of utility of facility
- ▶ Additional work to do to include it in practical simulation
 - ▶ Who performs leisure (Feil, Ordoñez, Balac)
 - ▶ Calibration / Validation of full simulation
 - ▶ Combine with household activity model

Conclusions

- ▶ A realistic social network does help to predict leisure travel distance
- ▶ Fair amount of additional complexity
- ▶ Contributions:
 - ▶ Model for joint decisions in generic social network
 - ▶ Effective algorithm to generate realistic social network
 - ▶ Demonstration of viability of using social contacts to steer leisure location choice