

# Multi-level evaluation of autonomous vehicles using agent-based transport simulation for the case of Singapore

Sebastian Hörl

March 13, 2016

# Contents

<b>1</b>	<b>Introduction</b>	<b>2</b>
<b>2</b>	<b>Perspectives on autonomous traffic</b>	<b>3</b>
2.1	The policy makers . . . . .	3
2.2	The taxi company . . . . .	3
2.3	The customer . . . . .	3
<b>3</b>	<b>Modeling framework</b>	<b>4</b>
3.1	MATSim and DVRP . . . . .	4
3.2	Corridor scenario . . . . .	4
<b>4</b>	<b>An agent-based model of autonomous taxis</b>	<b>5</b>
4.1	Marginal utility of traveling . . . . .	5
4.2	Vehicle supply . . . . .	5
4.3	Price structure . . . . .	5
<b>5</b>	<b>Simulation Results</b>	<b>6</b>
5.1	Sioux Falls . . . . .	6
5.2	Singapore . . . . .	6
5.3	Usage scenarios . . . . .	6
<b>6</b>	<b>Conclusion</b>	<b>7</b>

# Chapter 1

## Introduction

Increasing research and development of autonomous vehicles will have great impact on the future traffic situations in cities. While there are numerous advantages such as expected shorter travel times, less space usage for parking vehicles and more safety, there are still a lot of challenges. This includes policy decisions, infrastructural changes as well as economic considerations. In any case, being able to predict how the introduction of autonomous vehicle in certain usage scenarios and business models would impact the traffic in a city, could give guidelines and suggestions on the implementation process.

In this thesis a hypothetical autonomous taxi service will be implemented in the large-scale agent-based traffic simulation MATSim. The central approach of this simulation framework is the per-person generation of daily plans, the execution in a common day cycle and reconsideration of travel decisions for the next day until every person has reached an acceptable plan. This allows for a couple of interesting investigations, such as to which degree users from different travel modes like private cars or public transport will change towards the taxi service. Also, it can be found out how AV usage is distributed over travel distances or travel times.

To setup such a simulation, careful definition of the AV model has to be undertaken. It is clear that the simulation can only depict reality to a certain degree, this is even more true when one tries to implement means of traffic that are not existant today and therefore lack sensible reference measurements to calibrate. In that sense it needs to be clear, which assumptions have been made and how the choice of different model parameters affects the results in reasonable or plainly unrealistic ways.

## Chapter 2

# Perspectives on autonomous traffic

### 2.1 The policy makers

- policy side: taxes on vehicles, subsidies for real taxi services, etc... - complex situation - last-mile problem, ... - problems like cyber crime, etc. (not covered here)

### 2.2 The taxi company

- interesting economic topic subject to agent-based simulation - probably regulation needed, because - special properties of taxi services - like subsidised traffic during nights - not a real “competitive market”

- better scheduling possibilities (already used by Uber probably) - pre-scheduled trips vs. on-demand - simpler cost structure

### 2.3 The customer

- studies on reception - how likely are people to use it? - different phases - any preferences coming from different transport modes?

## Chapter 3

# Modeling framework

### 3.1 MATSim and DVRP

- explain MATSim, how the simulation, scoring and selection works - explain DVRP contribution and its advantages and disadvantages

TODO: References to problems of within-day-replanning, since this is similar... has impact on the nature of the Nash equilibrium

TODO: Mention Only one person per taxi! Replanning otherwise quite complex!

### 3.2 Corridor scenario

- description of the corridor scenario - made assumptions - baseline results (distribution of modes, etc.)

## Chapter 4

# An agent-based model of autonomous taxis

An agent-based model of an autonomous taxi service can be divided in certain main parts, which are

- a distribution model, which defines how available taxis are distributed over the network links at the beginning of the simulation,
- a usage model, which defines how agents can interact with the taxi services (i.e. taxi-on-demand or prescheduled pickups) and ,
- an allocation algorithm, defining which taxis are allocated to a certain passenger request and
- an operator model defining how trips with and without passenger account for the profits of the operator

Derived from those parts there are a couple of parameters that need to be defined:

- The number of AVs which are available (i.e. the supply)
- Marginal utility of travel, marginal utility of distance (money), constant utility (per-trip)

The biggest issue here is that the utility parameters cannot be tuned as intuitively as for other traffic modes. There usually a statistical distribution of trips and their properties is known. Then parameters are adjusted such that the relaxed state of the simulation is resembling them on the upper level.

For AVs there are at maximum preference studies, describing which percentage of people would be tempted to use a autonomous vehicle.

- number of taxis - dispatching model - marginal utility - monetary factor (pricing) - per trip, per duration, per distance, ... - per car

### 4.1 Marginal utility of traveling

- how to find a reasonable marginal utility of travel - show process for corridor scenario - dependent on share of autonomous vehicles -  $\lambda$  probably a good measure

### 4.2 Vehicle supply

- how to infer

### 4.3 Price structure

- how to infer

## Chapter 5

# Simulation Results

### 5.1 Sioux Falls

### 5.2 Singapore

### 5.3 Usage scenarios

- last mile problem etc.

## Chapter 6

## Conclusion