A holistic decision making framework for vehicle sharing systems (...and evaluation of demand forecasting)

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Outline





Framework

- The value of demand forecasting
- 6 Conclusion and future work

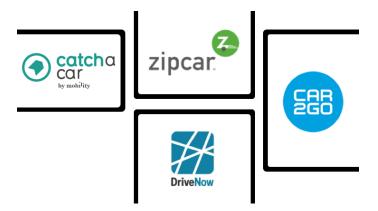
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What is a Vehicle Sharing System (VSS)?

A VSS enables users to use the available vehicles generally for short period of time by an RFID card or smart phone application identification.

- Various system configurations
 - One-way or return trip
 - Station-based or free-floating
 - Dynamic or fixed pricing
 - ...

Car-sharing companies



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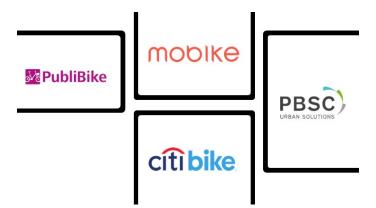
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Bike-sharing companies



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Imbalance in the network

- Bicycle-sharing systems (BSSs)
 - Capacitated traveling salesman problem (TSP) (*Pal and Zhang et al., 2017*)
 - Vehicle routing problem (VRP) with commodity flow conservation constraints (*Ghosh et al., 2016*)
 - VRP with previously clustered stations (Liu et al., 2016)
- Car-sharing systems (CSSs)
 - Multi-TSP (Nourinejad et al., 2015)
 - Mixed Integer Linear Programming (MILP) models (*Boyaci et al., 2017*)
 - Importance of the relation between demand forecasting and rebalancing (*Jorge and Correia, 2013*)
 - Denial of the requests in the case of high demand (Boyaci et al., 2017)

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Demand estimation

BSSs

- Meteorology Similarity Weighed K-Nearest-Neighbor model for station bike pick-up demand prediction (*Liu et al., 2016*)
- Simulating the demand with a Poisson process (Ghosh et al., 2016)
- Worst-case demand, which maximizes demand loss (Ghosh et al., 2016)
- CSSs
 - Forecasting OD pairs between zones with two methods (*Müller and Bogenberger, 2015*):
 - AutoRegressive Integrated Moving Average (ARIMA)
 - Holt-Winter's method

Pricing

BSSs

- Prices are assigned dynamically depending on the occupancy at the destination station. (*Chemla et al., 2013, Waserhole, 2013*)
- Incentives are linearly dependent on the additional travel time. (*Pfrommer et al., 2014*)
- Dynamic pricing improved the level of service for the weekends. (*Pfrommer et al., 2014*)
- CSSs
 - Incentives on pricing which encourages users to do trips which reduces the imbalance of the network. (*Jorge and Correia, 2013*)
 - Balance of the system is improved, but less demand is served. (*Jorge and Correia, 2013*)

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Big picture

• Shared mobility systems: an updated survey by Laporte et al., (2018)

- Two dimensional classification
 - Type of the problem
 - Decision level
- Lack of research in some specific areas
 - Pricing incentives and rebalancing at strategic level
 - Locating stations in tactical level
- This work aims to provide a holistic solution approach for the VSSs.
 - From decision maker point of view
 - Three dimensional classification
 - Decision levels: Strategic, Tactical, and Operational
 - Actors: Supply and Demand
 - Layers: Data, Models, and Actions
 - Relations between the components

Strategic level

- Corresponds to long-term decisions
 - What kind of system are we dealing with?
 - How is the scope defined?
- Planning horizon
 - More than a year

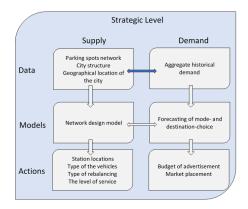


Figure: General framework - strategic level

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Tactical level

- Corresponds to mid-term decisions
 - How do we utilize the strategic level decisions?
 - Which decisions should we pass to the operational level?
- Planning horizon
 - 4-6 months

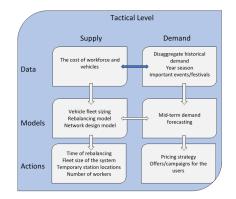


Figure: General framework - tactical level

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Operational level

- Corresponds to short-term decisions
 - What is the current situation of the system?
 - What do we do next time step?
- Planning horizon
 - Daily/hourly

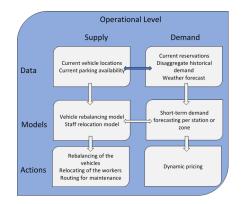


Figure: General framework - operational level

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Big picture - revisited

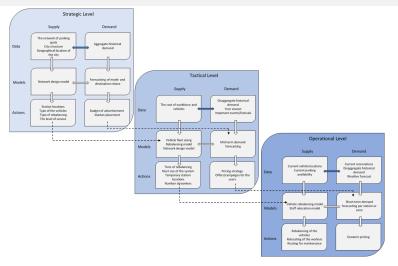


Figure: General framework and inter-relations

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Derived questions

- Demand forecasting is given a large attention,...
- ...but what about the added value from constructing a demand model?

- The literature consists of works on BSSs and CSSs.
- New types of vehicles are being introduced in VSSs.
- However, some of the approaches are inapplicable for the new types of vehicles.

The idea

- A discrete event simulation models the system demand throughout the day.
- The vehicle distribution at the end of the day is obtained from the simulation and passed to the optimization model.
- The mathematical model solves the rebalancing problem given a desired initial state for the next day.
- Two cases are investigated:
 - Known demand: the model knows a perfect demand forecast for the next day. The rebalancing is done according to this information.
 - Unknown demand: the system is rebalanced to the same initial state every day.
- The main idea is to compare the trade-off between the lost demand and rebalancing costs, between the two cases.

Setting the scene

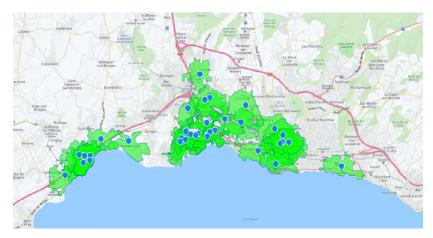


Figure: PubliBike stations and corresponding isoline polygons

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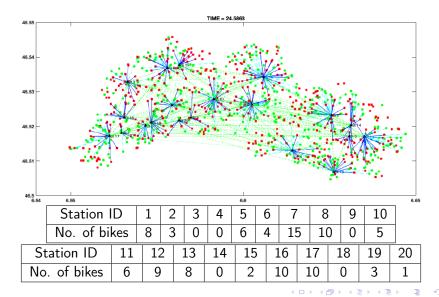
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Trips simulation



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Preliminary findings

Dav	Unknown demand		Forecasted demand		Total
Day	Lost dem.	Reb. cost	Lost dem.	Reb. cost	requests
1	172	18681	181	16538	469
2	186	16236	171	16206	477
3	174	16206	165	16264	457
4	172	15938	153	15938	482
5	173	15614	164	15484	482
6	178	15932	169	15932	494
7	172	15484	162	15484	490
8	172	16888	165	16187	499
9	173	15932	162	15932	474
10	163	15614	163	15614	465

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Conclusion and future work

- A general framework for VSSs is presented.
- Inter- and intra-relations between framework components are discussed.
- We focused on the evaluation of value of demand modeling and presented preliminary results.

- Different scenarios such as in the case of events will be evaluated.
- Different configurations of VSSs will be analyzed.
- An application will be done on newly introduced LEVs.

An application to Light Electric Vehicles (LEVs)

• A new type of vehicles



- You don't need a car driving license
- You can ride on bicycle lane
- You are protected from bad weather
- There's a room for luggage
- Free-floating parking

- The system is available to a higher portion of the population.
- Conventional rebalancing ideas should be adapted.
- Free-floating structure is not widely studied.



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