From Data to Models and Proposed Solutions

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Motivation



The problem



- Traffic congestion responsible for 20% of fuel consumption and 90% of CO in large urban areas.
- Cost of traffic congestion will reach \$2.8T in the US by 2030 (\approx annual tax revenue).
- On a per-driver basis, cost of traffic congestion is \$1740 annually in US/Europe.
- Boston recently made news being declared #1 in hours lost in rush-hour traffic per driver in 2018.





Congestion Maps for the Boston Area: 2012 \rightarrow 2015



https://salomonw.github.io/congestionmaps/DynamicPage/PM/index.html

(Salo Wollenstein)

Yannis Paschalidis, Boston University

Beyond Congestion: Maps





Transportation Network Models

• Transportation network modeled as a graph.



- Dynamics: Drivers have a congestion function function of flow for each arc and pick the cheapest arcs to traverse. Collective decisions lead to a Nash (Wardrop) equilibrium.
- To control/design we need to build accurate predictive models.
- Data: Traffic flows.
- Can we learn (the congestion function) from data?



Price of Anarchy¹

- Having the congestion function allows us to answer many "what-if-questions".
- We can also formulate a problem to obtain a socially optimal equilibrium.
- Price-of-Anarchy:

 $\mathsf{PoA} = \frac{\mathsf{Congestion \ under \ Selfish \ Behavior}}{\mathsf{Congestion \ under \ Socially \ Optimal \ Behavior}}$



• Useful to assess how good/bad things are, but also to design interventions.

¹Zhang, Pourazarm, Cassandras, Paschalidis, CDC 2016, IFAC 2017, Proceedings IEEE 2018.



PoA Boston Data



Boston Area Data²

Eastern Massachusetts (EMA) Network



- Spatial average speeds for 13,000 road segments for each minute of 2012 (50 GB) and 2015 (130 GB).
- Capacity data in different times-of-day: lanes, peak vehicles counts, etc.

²https://github.com/jingzbu/InverseVIsTraffic, https://www.kaggle.com/jingzbu/ematransportation





Price-of-Anarchy (2012)







Road Congestion: Socially Optimal vs. User Optimal

"Spreading the traffic" results in:







Control and Interventions

- Sensitivities: Where to intervene?
- Socially optimal route recommendations: Can be shown that we can achieve the Socially Optimal solution through User Optimal actions if users use a properly modified congestion function!
 - Easier to incorporate in apps, even enforce with autonomous vehicles.
 - Take the driver "out of the picture."
- S Change demand! Congestion pricing and incentives!











Final remarks

- We have developed a new general framework for modeling driver behavior using data.
- Policy space: How to address traffic allocation issues and prevent NIMBY reactions?