

Measuring and Optimizing the Network Disequilibrium Levels through Ride-sourcing Vehicle Data

Wei Ma, Sean (Zhen) Qian

Carnegie Mellon University

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Real-world traffic networks

- ▶ The public agencies and transportation researchers know little about real-world real-time traffic conditions.
 - ▶ Traffic speed, traffic volume are known
 - ▶ Route choices, departure time and traffic demand are still challenging to obtain
- ▶ Does Dynamic User Equilibrium hold in day-to-day traffic?
- ▶ How to manage the traffic with real-time data?

Ride-sourcing services

- ▶ More than \$600B questions can be addressed by ride-sourcing services
 - ▶ Car as a service (\$110B)
 - ▶ Core taxi-cab market (\$100B)
 - ▶ Public transportation service (\$175B)
- ▶ Uber has operations in 785 metropolitan areas worldwide
- ▶ Didi Chuxing processes 4,875TB data and archives 106TB trajectory data per day

Is there a way to collaborate?



VS



- ▶ Is there a way to collaborate?
 - ▶ *For TNCs:* Provide data without damaging their interests
 - ▶ *For public agencies:* Use the data to manage the traffic and reduce congestion

Contributions

A holistic traffic management framework:

- ▶ Measure of network disequilibrium level (NDL)
- ▶ NDL can be estimated using ride-sourcing vehicle (RV) data
- ▶ A data disclosure scheme for TNCs to release ride-sourcing vehicle (RV) data
- ▶ Traffic management method using NDL

Dynamic User Equilibrium

- ▶ No traveler can improve their travel time by unilaterally changing their route and departure time
- ▶ Complementarity problem

$$F_{rs}^k(t) \left(C_{rs}^k(t) - \Pi_{rs}(t) \right) = 0, \forall k \in K_{rs}, rs \in K_q, t$$

where

$$\Pi_{rs}(t) = \min_{k \in K_{rs}} C_{rs}^k(t), \forall rs \in K_q, t$$

Network Disequilibrium Level (NDL)

$$D_{rs}^{\mathcal{M}}(t) = \sum_{k \in K_{rs}} F_{rs}^k(t) \left(C_{rs}^k(t) - \Pi_{rs}(t) \right)$$
$$D_{rs}^{\mathcal{F}}(t) = \sum_{k \in K_{rs}} p_{rs}^k(t) \left(C_{rs}^k(t) - \Pi_{rs}(t) \right)$$

- ▶ $D_{rs}^{\mathcal{F}}(t)$ is free of path flow
- ▶ Merit function in Dynamic User Equilibrium
- ▶ OD based, time-dependent
- ▶ Challenging to evaluate

Estimate the NDL using RV trajectory

Facts:

- ▶ RV penetration rate 1% \sim 10% in total traffic flow, growing
- ▶ Representative: route choice behaviors of RVs are similar to the conventional vehicles (from picking up to dropping off)
- ▶ Travel time of a RV in path k for OD rs is an unbiased estimator of C_{rs}^k

We can construct an estimator of NDL using RV trajectory data.

Estimate the NDL using RV trajectory

The duration of trajectory i is denoted by γ_i

$$\gamma_i = t_i^{T_i-1} - t_i^0$$

If trajectory i departs from r and arrives at t at time t , then

$$\mathbb{E}(\gamma_i) = \sum_{k \in K_{rs}} p_{rs}^k(t) C_{rs}^k(t)$$

Then we define

$$\begin{aligned} D_{rs}^{\mathcal{D}}(t) &= \frac{1}{|N_{rs}(t)|} \sum_{i \in N_{rs}(t)} \gamma_i - \min_{i \in N_{rs}(t)} \gamma_i \\ &\rightarrow C_{rs}(t) - \Pi_{rs}(t) \end{aligned}$$

$N_{rs}(t)$ represents the set of trajectories departing at t from r to s . Then

$$D_{rs}^{\mathcal{D}}(t) \xrightarrow{P} D_{rs}^{\mathcal{F}}(t)$$

Data privacy and zone-to-zone travel time

To compute the NDL estimator, we need pick-up and drop-off location and time.

- ▶ The data might reveal the drivers and riders' personal information
- ▶ The data can estimate the revenue of TNCs

To ensure data privacy:

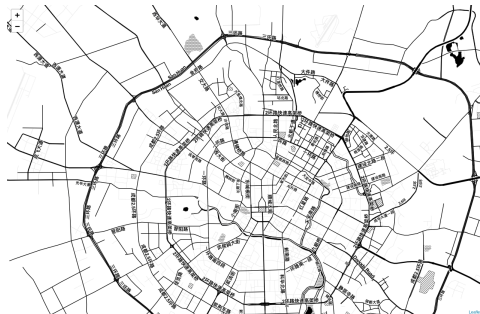
- ▶ We develop a data sharing scheme to convert the trajectory data to zone-to-zone travel time
- ▶ The disequilibrium level can be estimated from zone-to-zone travel time data

Traffic Management using NDL

- ▶ NDL represents the gap between current traffic conditions and DUE
- ▶ Minimizing NDL \rightarrow User optimal routing
- ▶ Always control the vehicle with largest NDL \rightarrow Optimal control effects

Didi Chuxing in Chengdu

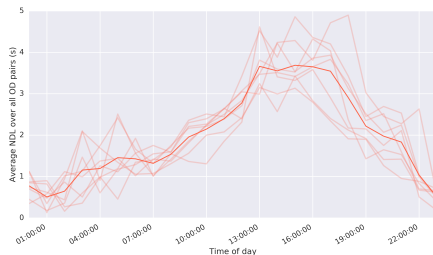
- ▶ Trajectories of all RVs operated by DiDi chuxing from Nov. 1st, 2016 to Nov. 30th, 2016



NDL pattern



(a) Weekdays

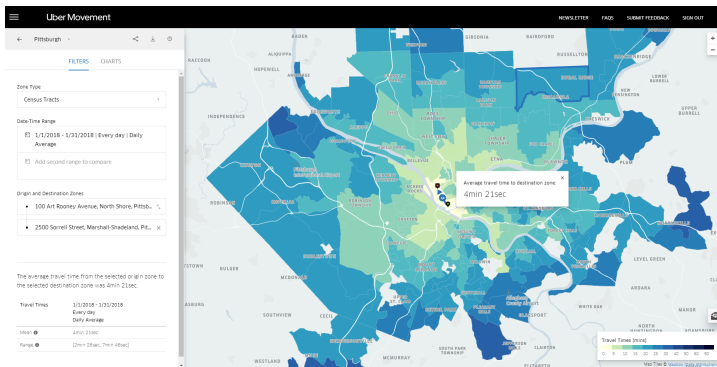


(b) Weekends

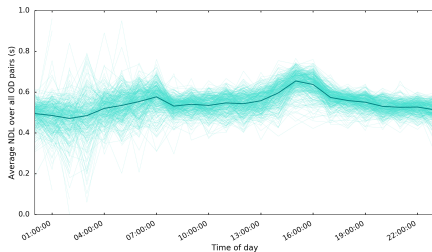
- ▶ General trend, day-to-day variation
- ▶ NDL patterns are similar to demand pattern (which is not always the case)

Uber in Pittsburgh

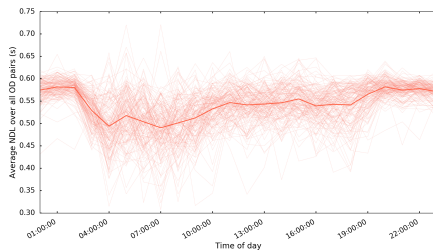
- ▶ Hourly zone-to-zone travel time within Allegheny county from Jan. 1st, 2016 to Jun. 30th, 2017



NDL patterns



(a) Weekdays



(b) Weekends

- ▶ NDL pattern differs from demand pattern (Pittsburgh is not a tourist city)
- ▶ NDL relatively stable on weekdays
- ▶ NDL is unstable on weekends

Traffic management

- ▶ We calibrate the traffic flows, route choice model and OD demand in Pittsburgh metropolitan area (traffic volume, INRIX speed)
- ▶ Controlling 1% of the vehicle will reduce the total network congestion by around 7%.

Thanks! Questions and comments?

Wei Ma, weima@cmu.edu
Civil and Environmental Engineering
Carnegie Mellon University

Website: www.weima171.com