

## Street-level Travel-time Estimation using sparse and coarse data

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- Two major areas of traffic forecasting:
  - Demand modeling (number of trips between O-D pairs)
  - Travel-time estimation (travel time between O-D pairs)
- Due to data availability, there is more research on freeways than on **arterials**. Many sources but sparse.
- **Goal:** Develop a method to leverage open (sparse) data sets to estimate arterial street-level travel times on a metropolitan road network.





### **Uber Movement**

- Summary statistics of Uber trips between traffic analysis zones (TAZ)
- movement.uber.com

### **OpenStreetMap**

 Graph representation of road network with intersections and road segments represented as vertices and edges

Road network for 27 TAZs in downtown Los Angeles

openstreetmap.org 





### **Edge Travel-time Estimation**

	Α	В	С	D	Е	F
Α				25		
В						
С						
D						6
Е						13
F						

Χ  $b_{c} = 1$ Q

**Solutions:** x = 12, y = 1, z = 12 x = 1, y = 23, z = 1y = [1, 23]



### + y + z = 25- p+z = 6 | + y + p = 13



### **Edge Travel-time Estimation**

	Α	В	С	D	Е	F
Α				25		
В						
С						
D						6
Е						13
F						



**Solutions:**  $\begin{array}{l} x = 12, y = 1, z = 12 \\ q = 1, y = 11, p = 1 \\ y = [1, 11] \end{array} \qquad \begin{array}{l} x + y + z = 25 \\ p + z = 6 \\ q + y + p = 13 \end{array}$ 





Α

Β

С

D

Ξ

F

### **Edge Travel-time Estimation**





The betweenness of an edge is the number of shortest paths that go through it for all possible vertex O-D pairs.





Types of roads based on betweenness centrality (b<sub>c</sub>) and degree (k<sub>road</sub>). (Source: Wang et al. Sci. Reports, 2012, 2, 1001)



### **Project Workflow**



Estimate edge travel times by solving constrained least-squares problem:  $\hat{t} = \operatorname{argmin}_{0.8f \le t \le 1.25t_k} \|S_k^T t - y_k\|_2^2$ 

### Update edge travel times and weight parameter: $t_{k+1} = (1 - \lambda_k)t_k + \lambda_k \hat{t}$



We estimated edge travel times for downtown LA graphs using

- Different times of day:
  - 3am = 507,449 TAZ O-D pairs
  - 6pm = 1,115,432 TAZ O-D pairs
- Different graph sizes:
  - 1-mile (M = 7,138), 2-mile (M = 22,756) and 3-mile radii (M = 40,929)
- Different number of sampled trips:
  - *N* = 1,000, 3,000, and 5,000

Error measured by 
$$\epsilon_k = \sqrt{1/N} \sum_{(i,j) \in \mathcal{U}_k} n_{ij} (\log g_{ij}(t_k) - \log G_{ij})^2$$



### **Experimental Results: Convergence**







## Initial Results on metro-scale estimation

- LA and associated Metro Areas (160k intersections / 470k links no sparsification):
  - Total coverage Approx. 1200 sq. mi
  - 64 chunks @ approx. 20 sq. mi. per chunk
  - Total running time on a single (8-core) server node : 18.5 hours per hourly window
  - With 64-node cluster ~ Approx. 25 mins (including serial overhead)
- Seattle and associated Metro areas ( / 160k links no sparsification)
  - Total coverage Approx. 500 sq. mi
  - 16 chunks @ approx. 30 sq. mi per chunk
  - Total running time on a single (8-core) server node : 5.00 hours per hourly window
  - With 16-node cluster ~ Approx. 20 mins (including serial overhead)
- Current approach cannot be used to drive-down chunk sizes as it affects accuracy
  - Imposes a limit on scaling gains
  - Projections for a 64-node cluster (12K Processors) are of about 30 mins (including overhead)
  - Potential areas of performance improvement: Python code, constrained least squares solver



## Visualization of congestion on LA metro-area map

Pasadena

- Results at 6pm on a weekday
- Known congestion areas correctly identified

Long beach

Downtown LA





## Visualization of congestion on Seattle metro-area

- 6PM
- Weekday
- OpenStreet Map

### Redmond

### Seattle Downtown

### Industrial Area





## **Time evolution of the Seattle network 4pm-6pm**

Increase in traffic congestion  $\rightarrow$  larger travel times in congested arterials







4pm

5pm

### 6pm



## **Graph Pseudo-Sparsification**

Edges with lower betweenness are more likely to have estimated travel times closer to their free-flow travel times.





 $q = \frac{|t - f|}{f}$ 



## **Graph Pseudo-Sparsification: Method**



Estimate edge travel times by solving constrained least-squares problem:  $\hat{t} = \operatorname{argmin}_{0.8f \le t \le 1.25t_k} \|S_k^T t - y_k\|_2^2$ 

Update edge travel times and weight parameter:  $t_{k+1} = (1 - \lambda_k)t_k + \lambda_k \hat{t}$ 



## **Summary and Future Work**

- Leveraged coarse-grained Uber Movement data in the form of TAZ O-D pair summary statistics to provide estimates of fine-grained, street-level travel times.
- Solved Large metropolitan areas like Los Angeles and Seattle.
- Implement HPC based software for (near)real time estimation.
- Include Multi-Modal information: bi-cycles, bus, peds, etc.
- Open source the software with Laptop/Cloud/Supercomputer support.



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# Thank you

