



Quantifying City-Scale Resilience to Extreme Congestion Events

Dan Work

Civil and Environmental Engineering and
Coordinated Science Laboratory
University of Illinois at Urbana–Champaign

Joint work with Brian Donovan (UIUC) & Dr. Jong Lee (NCSA)

[Research Sponsors: NSF, NCSA, FHWA RSI UTC, IDOT]

[Open source code and data available at [GitHub](#) and publish.illinois.edu/dbwork]

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Goal: Empirically quantify resilience to extreme congestion events



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- Extreme congestion events characterized by:
 - Changes in travel demand
 - Loss of cyber components
 - Damage to physical components
- Assess transportation network performance quantitatively
 - Recovery time
 - Deviation deviations from typical conditions



[A. Savulich, New York Daily News, 2012] 2

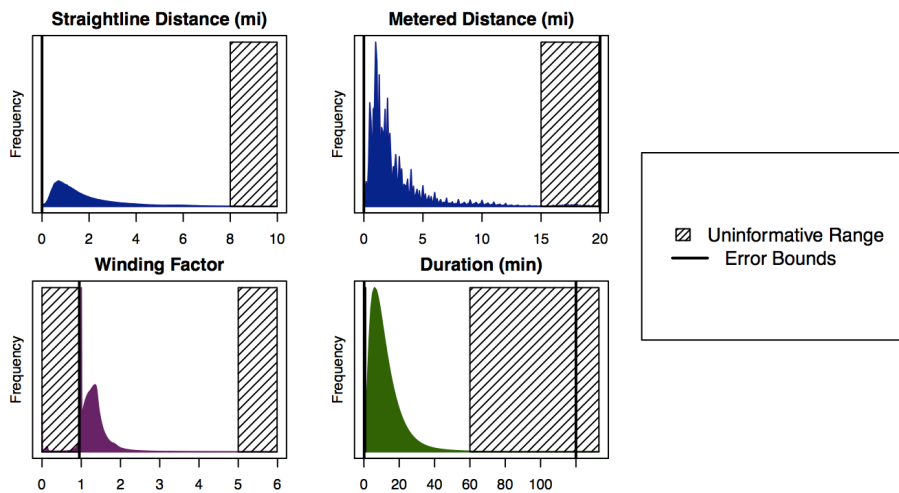
Go to “war” with the sensors you have, not the ones you want – D. Estrin

- 697,000,000 taxi trips (2010-2013)
- FOIL data: New York City Taxi and Limo Commission
- Data structure:
 - Only two GPS points
 - Start of trip
 - End of trip
 - Meter distance and travel time
 - Driver ID and Car ID
 - Fare data
 - **Download:** publish.illinois.edu/dbwork/open-data

Filtering the data



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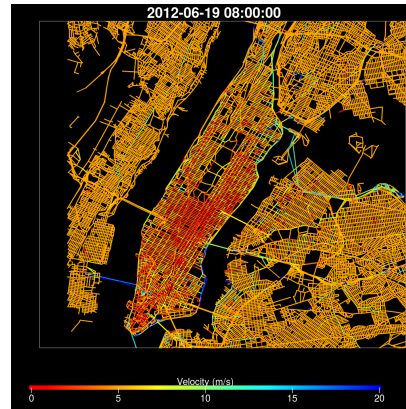
(7.5% data errors, 11% uninformative)

High resolution traffic estimation with coarse GPS data



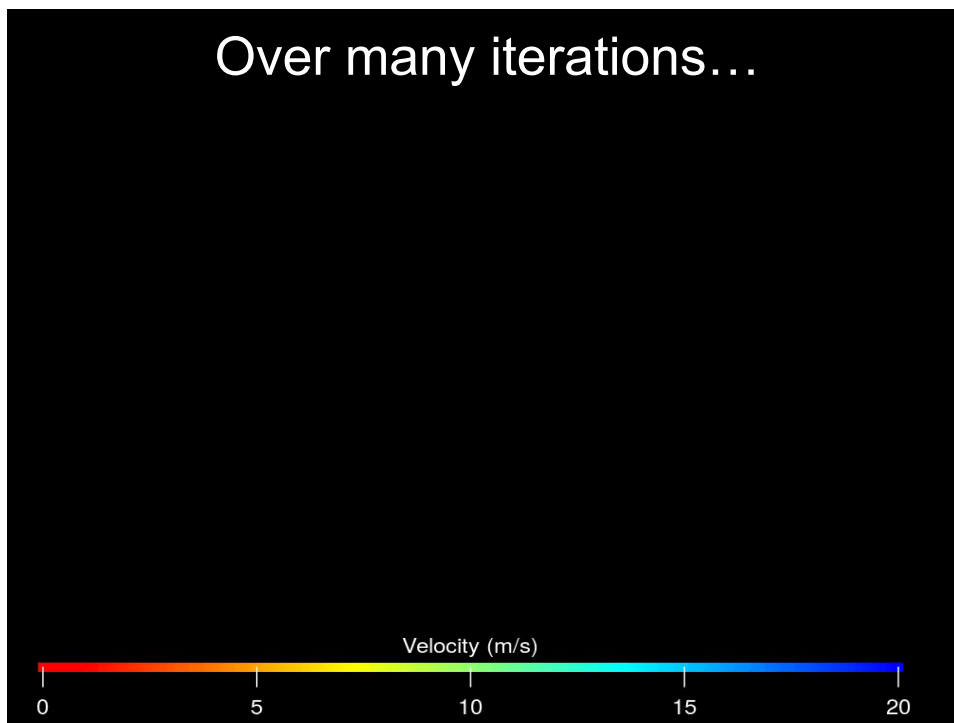
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- Compute hourly traffic paces via iterative link travel time estimator
 - **Routing:** Given link speeds, route all taxis and compute predicted travel times
 - **Link travel time correction:** Compare predicted travel times with true travel times, then adjust the link travel times to minimize the travel time prediction error



[Donovan, Lee, Work, *IEEE T-ITS*, 2016 (in review); Santi et al. *PNAS* 2014]₅

Over many iterations...

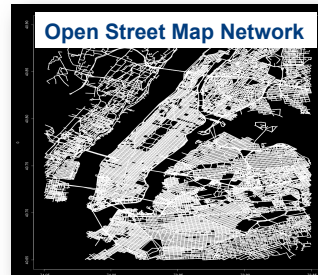


Scaling to multi-year datasets



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- NYC road graph
 - 24,000 nodes
 - 63,000 links
- Typical hour: 20,000 trips
- Computing traffic estimates
 - Bottleneck: routing step
 - Accuracy: ~2 min error per trip
 - Saved estimates:
publish.illinois.edu/dbwork/open-data



Storing traffic data in a vector



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- Build a column vector of traffic paces
- Each element corresponds to the average pace in a given hour during the four year period

$$M_1 = \begin{bmatrix} M_{11} \\ M_{21} \\ \vdots \\ M_{m1} \end{bmatrix} \begin{array}{l} \text{Pace on link 1 (Wed noon-1pm, 1/6/2010)} \\ \text{Pace on link 2 (Wed noon-1pm, 1/6/2010)} \\ \vdots \\ \text{Pace on link m (Wed noon-1pm, 1/6/2010)} \end{array}$$

Building a traffic data matrix



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- $n = 52$ (weeks/year) \times 4 (years) copies of traffic on a Wednesday from noon-1pm over a four year period

$$M = \begin{bmatrix} \begin{array}{c} M_{11} \\ M_{21} \\ \vdots \\ M_{m1} \end{array} & \begin{array}{c} M_{12} \\ M_{22} \\ \vdots \\ M_{m2} \end{array} & \cdots & \begin{array}{c} M_{1n} \\ M_{2n} \\ \vdots \\ M_{mn} \end{array} \end{bmatrix}$$

Wed noon-1pm, 1/6/2010
 Wed noon-1pm, 1/13/2010
 Wed noon-1pm, 12/25/2013

[Donovan, Lee, Work, *IEEE T-ITS* 2016 (in review)]

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Facts about traffic data matrices



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- If traffic is perfectly periodic and repeatable:

- all columns of M are identical
- M is a rank 1 matrix

$$M = \begin{bmatrix} \begin{array}{c} M_{11} \\ M_{21} \\ \vdots \\ M_{m1} \end{array} & \begin{array}{c} M_{12} \\ M_{22} \\ \vdots \\ M_{m2} \end{array} & \cdots & \begin{array}{c} M_{1n} \\ M_{2n} \\ \vdots \\ M_{mn} \end{array} \end{bmatrix}$$

Wed noon-1pm, 1/6/2010
 Wed noon-1pm, 1/13/2010
 Wed noon-1pm, 12/25/2013

- Real traffic matrices
 - Have missing data
 - Have estimation errors
 - Have some columns that are far from typical (e.g., 12/25/2013 is a federal holiday)

[Donovan, Lee, Work, *IEEE T-ITS* 2016 (in review)]

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Detecting outliers via Robust PCA



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- **Problem:** Given Data Matrix M , find a decomposition composed of:
 - M : original data matrix
 - L : low rank matrix
 - C : column sparse matrix indicating outliers

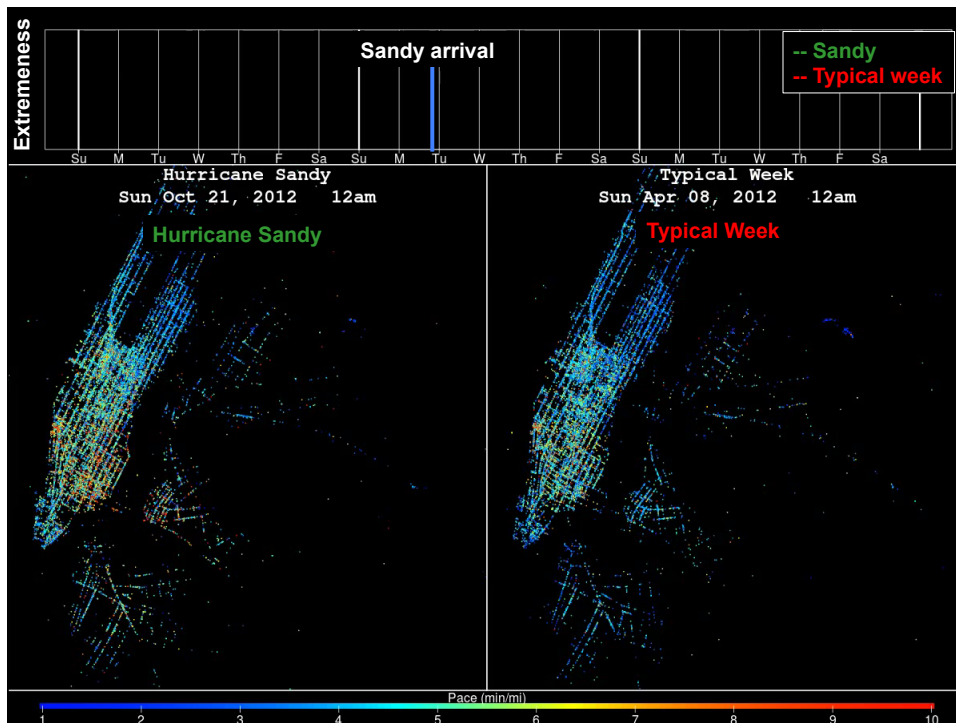
- Solved via convex optimization:

$$\begin{aligned} \text{Minimize:} & \quad \|L\|_* + \lambda \|C\|_{1,2} \\ \text{Subject to:} & \quad \|M - (L + C)\|_F \leq \varepsilon \end{aligned}$$

- **Key outcome:** Any nonzero column in C is an outlier
- Extreme points in the low dimensional subspace are also identified as outliers via Mahalanobis distance.

[Xu, Caramanis, and Sanghavi, *IEEE Trans Information Theory*, 2012; Donovan, Lee, Work, *IEEE T-ITS* 2016 (in review)]

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Worst events for New York City traffic



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| Event | Start Time | Duration (Hours) | Max Pace Dev (Min/Mi) | Min Pace Dev (Min / Mi) |
|-----------------|------------------|------------------|-----------------------|-------------------------|
| Hurricane Sandy | 2012-10-28 21:00 | 134 | 2.26 | -1.54 |
| Snowpocalypse | 2010-12-26 13:00 | 109 | 4.24 | 0.33 |
| Blizzard | 2011-01-31 10:00 | 47 | 2.01 | 0.34 |
| Hurricane Irene | 2011-08-27 13:00 | 43 | 0.65 | -1.65 |
| Blizzard | 2010-02-10 06:00 | 32 | 0.65 | -1.03 |
| Blizzard | 2013-02-08 06:00 | 27 | 1.53 | -0.59 |



Hurricane Irene



Snowpocalypse



Hurricane Sandy

Summary and future perspectives



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- Quantifying extreme events
 - Hurricane Sandy: ~5 days of recovery time
 - *Worst congestion after the hurricane hit.*
 - **Re-entry process is significantly slower than evacuation.**
- *Long term goal: Improve transportation policy*



Hurricane Irene

As Hurricane Season Strengthens, NYC Tells Residents to "Know Your Zone"

BY GRAHAM T. BECK | AUGUST 29, 2014





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