

# **Open-Source Dynamic Traffic Assignment Package DTALite: GMNS based Multi-Resolution Modeling**

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Prepared for Short Course on  
Modelling transportation supply and demand-supply interactions  
30th June — 2nd July 2022  
Organized by  
The Department of Civil Engineering  
Indian Institute of Technology (IIT), Madras, India



# Evolution of Simulation/Digital Twins

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*Computer-driven  
Simulation (1960)*



*Computer-driven  
Simulation (1985)*



*Simulation-driven System  
Design (2000)*



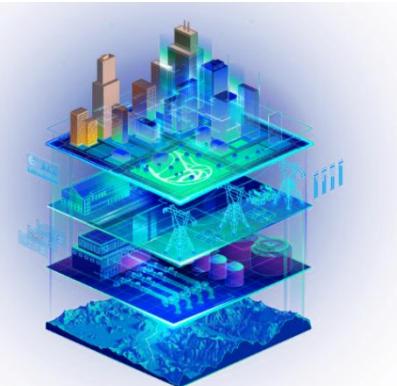
*Digital  
Twins (2015)*

AutoCAD becomes a de facto tool in nearly all engineering and design

NASA & USAF papers on digital twins (2011)



Source: chesky - stock.adobe.com



Source: www.esri.com/en-us/digital-twin/overview

## Questions to be Addressed

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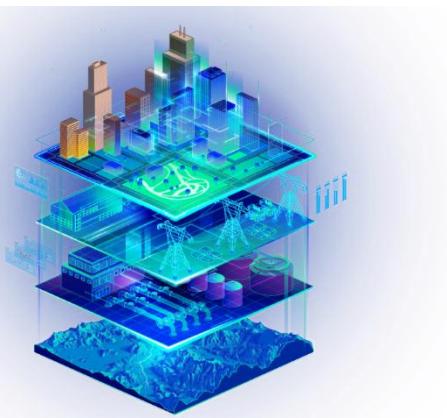
1. How to practically build a prototype of Digital Twin system for **smart cities and communities**? How to encourage **collaboration** from different software developers, researchers and planners/engineers
  
2. How to leverage the **open-source model and open data** to enable decision intelligence?
  
3. How to build a **high-fidelity multimodal simulator** that mirrors the physical world with complex interactions?
  
4. How to bridge the gap in multiresolution modeling: **From micro to macro, from macro to micro mesh network.**

# Outline/Main Research Line

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- I. Open Network Data Specification

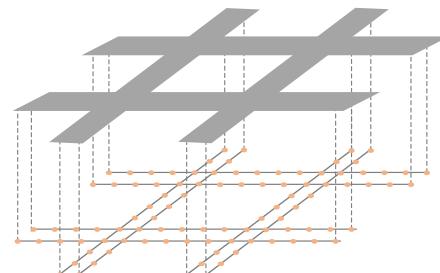
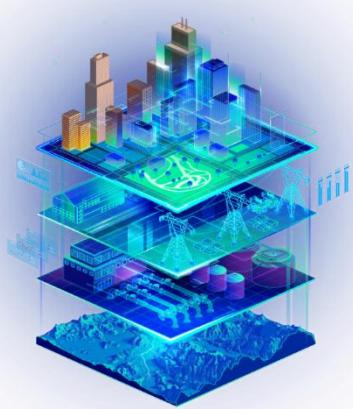


# Outline/Main Research Line

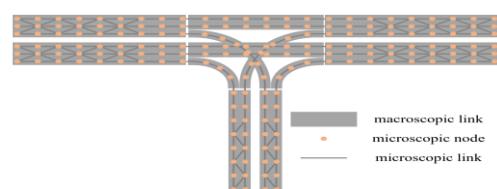
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**I. Open Network Data Specification**    **II. Open-Source Tools**



Macroscopic Layer



Microscopic Mesh Layer

macroscopic link  
microscopic node  
microscopic link

# Outline/3 Parts/DTA

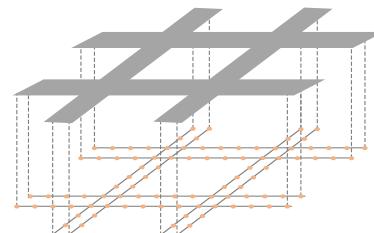
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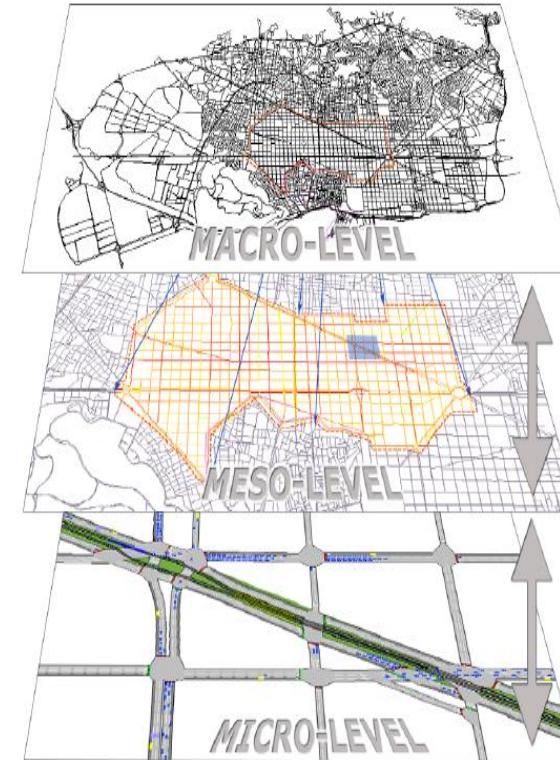
*I.* Open Data Specification



*II.* Open-Source Tools



*III.* Supply-side Analytical Models for Large Scale System

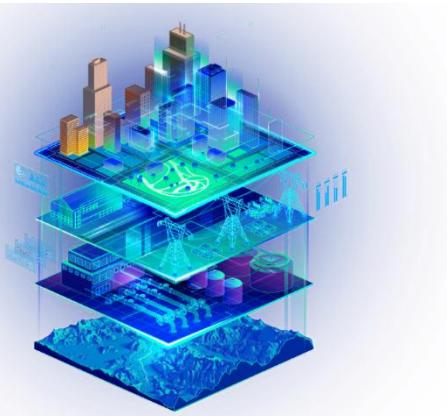


# Part I: Open Network Data Specification

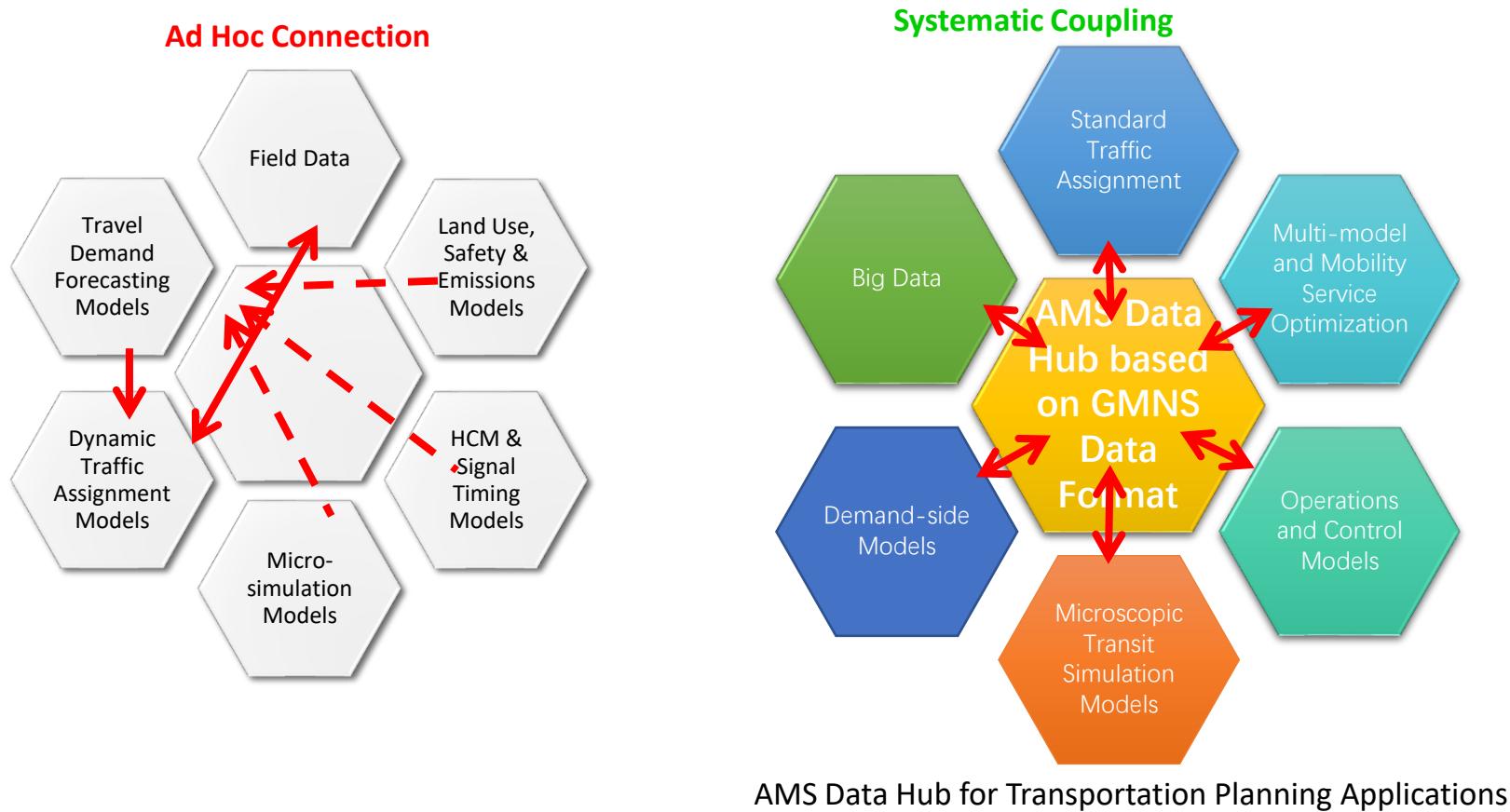
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## *I. Open Data Specification*



# Critical Challenges for Transportation Analysis Modeling and Simulation (AMS) Data Hub

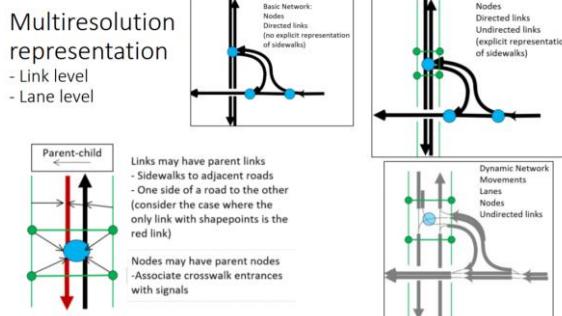


# Introducing General Modeling Network Specification (GMNS)

The objective of the GMNS is to provide a **common human and machine-readable format** for sharing routable road network files.

The project is overseen by a project management group, with MPO, city, industry, academic and US DOT participation.

[github.com/zephyr-data-specs/GMNS](https://github.com/zephyr-data-specs/GMNS)



Source: Volpe Center GMNS team  
Scott.Smith@dot.gov  
ian.Berg@dot.gov

## Movements at an intersection

### Link Level

- Link Level
- Lane Level (for the northeast approach)
- Movement attributes
  - Node
  - Inbound link and lane(s)
  - Outbound link and lane(s)
  - Type of Movement
    - left, right, thru, merge, uturn
  - Type of control
    - no\_control, yield, stop, stop\_2\_way, stop\_4\_way, signal
  - Optionally
    - Right-turn-on-red
    - Penalty
    - Capacity
    - Pct Green Time

**Permitted movements may also be time-of-day specific**

## Governance

This project is overseen by a board-approved Project Management Group (PMG) as follows:

- Joe Castiglione, SFCTA (chair, board representative)
- Michael Mahut, INRO
- Wu Sun, SANDAG
- Guy Rousseau, ARC
- Chetan Joshi, PTV
- Jeff Frkonja, Portland Metro
- Scott Smith, Volpe
- Natalia Ruiz Juri, University of Texas Center for Transportation Research
- Song Gao, UMass Amherst

# GMNS Format

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## Basic Data Elements

- node
- link
- geometry
- zone

name	node_id	osm_node	osm_highw	zone_id	ctrl_type	node_type	activity_type	is_boundar	x_coord	y_coord	main_node	poi_id	notes
	0	41459438			0				-111.928	33.4245			
	1	41520512			0				-111.944	33.42547			
	2	41520515			0				-111.944	33.42432			
	3	41520518			0				-111.944	33.42318			
	4	41520521	traffic_signals		1				-111.944	33.42189			
	5	41520523			0				-111.944	33.42064			
	6	41520525			0				-111.944	33.42035			
	7	41520528			0				-111.944	33.4194			
	8	41520531			0				-111.944	33.41892			
	9	41520533			0				-111.944	33.41778			

node.csv

name	link_id	osm_way	from_node	to_node	id	dir_flag	length	lanes	free_speed	capacity	link_type_n	link_type	geometry	allowed_us	from_biw	to_biw
South Farn	0	5590095	13	14	1	1	81.57798	1	25		residential	6	LINESTRIN	auto		1
South Farn	1	5590095	14	13	1	1	81.57798	1	25		residential	6	LINESTRIN	auto		1
South Farn	2	5590095	14	15	1	1	80.16146	1	25		residential	6	LINESTRIN	auto		1
South Farn	3	5590095	15	14	1	1	80.16146	1	25		residential	6	LINESTRIN	auto		1
South Farn	4	5590095	15	16	1	1	240.2824	1	25		residential	6	LINESTRIN	auto		1
South Farn	5	5590095	16	15	1	1	240.2824	1	25		residential	6	LINESTRIN	auto		1
South Farn	6	5590095	16	17	1	1	84.15426	1	25		residential	6	LINESTRIN	auto		1
South Farn	7	5590095	17	16	1	1	84.15426	1	25		residential	6	LINESTRIN	auto		1
South Farn	8	5590095	17	18	1	1	83.10715	1	25		residential	6	LINESTRIN	auto		1
South Farn	9	5590095	18	17	1	1	83.10715	1	25		residential	6	LINESTRIN	auto		1

link.csv

# OpenStreetMap

OpenStreetMap (OSM) is a free, open-source, editable map website that can provide free downloads. osm2gmns, as a data conversion tool, can directly convert the OSM map data to node and link network files in GMNS format. Users can convert and model drivable, walkable, railway, or aeroway networks with a single line of Python code.



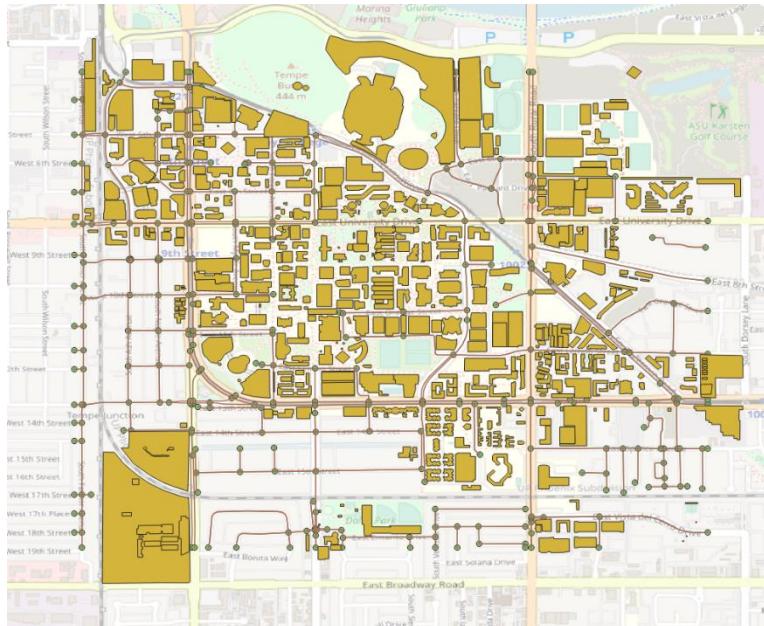
## Findings: High Map Accuracy across 30 Cities\*

	% of tags correct in OSM
Road class	98.6%
Road directionality	98.9%
Road names	99.8%
Intersection restrictions (Turn Restrictions)	94%
On/Off Ramp Signage	89%
Destination Signage	88%
Lane counts	66.8%

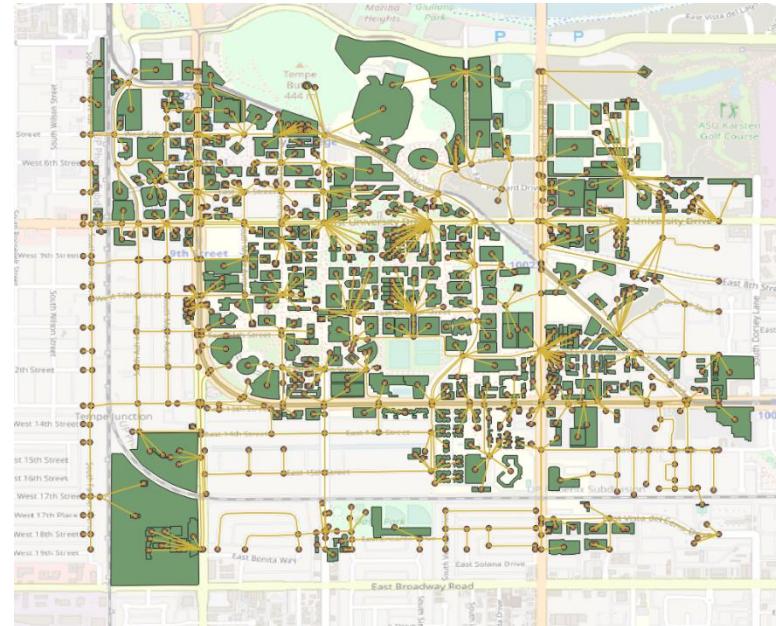
We found that core features of OpenStreetMap roads are correct more than 95% of the time relative to what exists in the real world. Data critical to safe navigation, such as left turn restrictions, are correct more than 85% of the time. Nationwide, these estimates are precise to within 5% sampling uncertainty. The regional uncertainty varies more based on region-level dynamics, visible in the figures at the end of this post.

<https://eng.lyft.com/how-lyft-discovered-openstreetmap-is-the-freshest-map-for-rideshare-a7a41bf92ec>

# POIs, Activity Locations and Network



Network with POIs



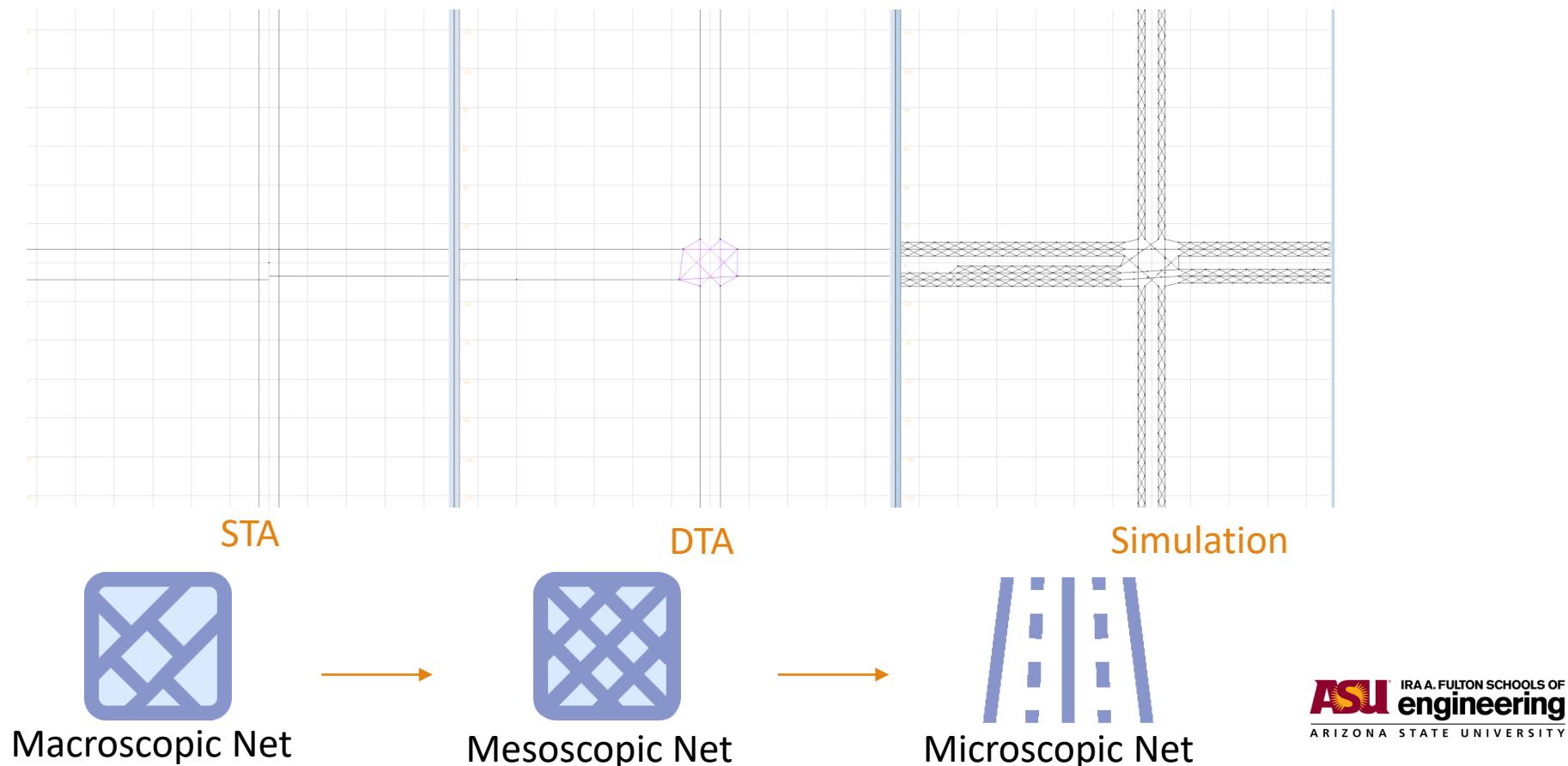
Connect POIs with network

More sample networks

<https://osm2gmns.readthedocs.io/en/latest/sample-net.html>

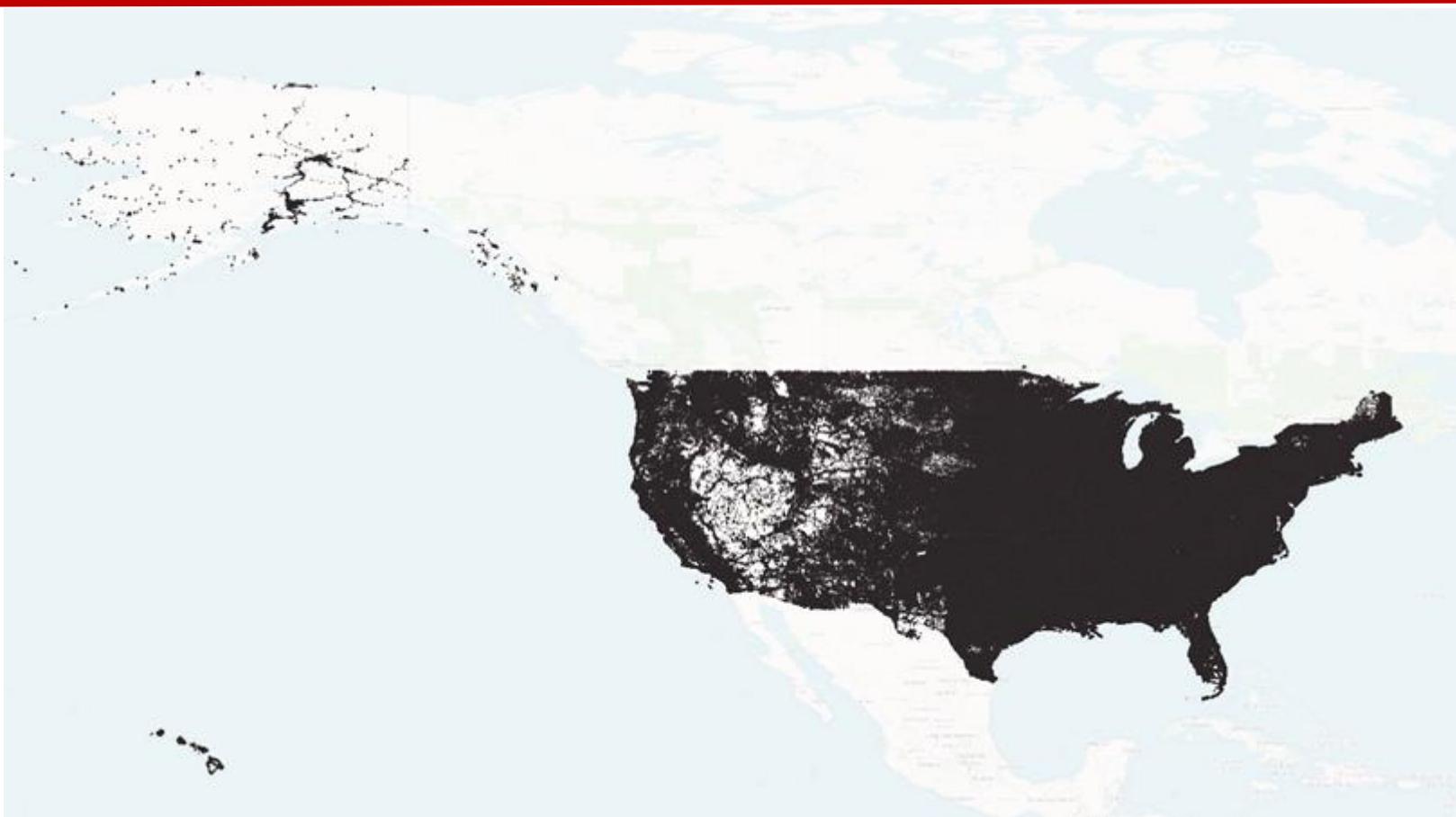
# Multi-resolution Network

For any given networks that meet the GMNS standard, net2cell helps users automatically generate hybrid (macroscopic, mesoscopic and microscopic) transportation networks to accommodate different modelling needs.



# Multi-resolution Network

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Using ASU research computing facilities, we are able to produce the entire U.S. driving network from OpenStreetMap with 20 million nodes.

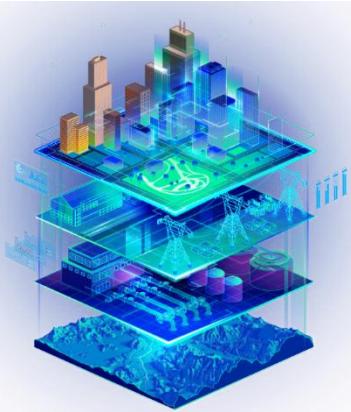
<https://github.com/asu-trans-ai-lab/asu-trans-ai-lab.github.io>

# Part II: Open-Source Tools

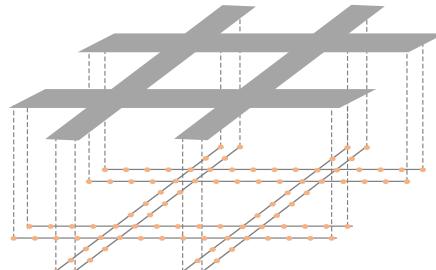


I. Open Data  
Specification

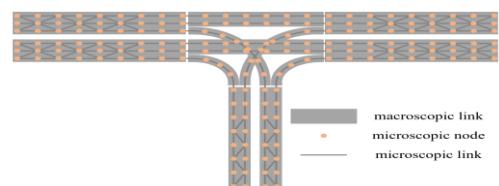
II. Open-Source Tools



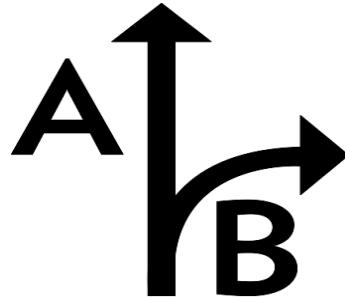
Macroscopic Layer



Microscopic Mesh Layer



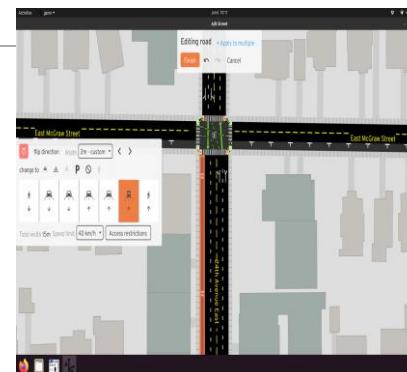
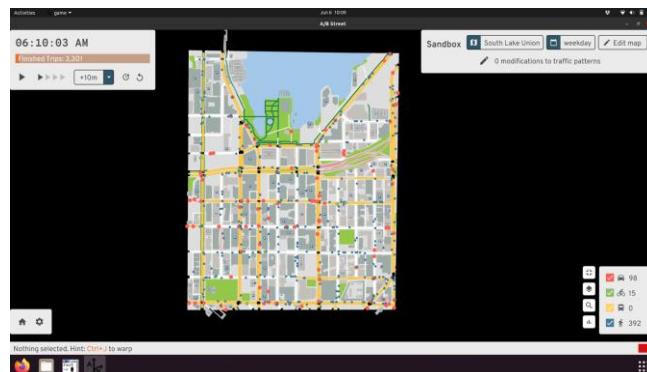
# (a) Smart Planning, Digital Twins, & Open Data



**STREET**

DUSTIN CARLINO, MICHAEL KIRK, & YUWEN LI

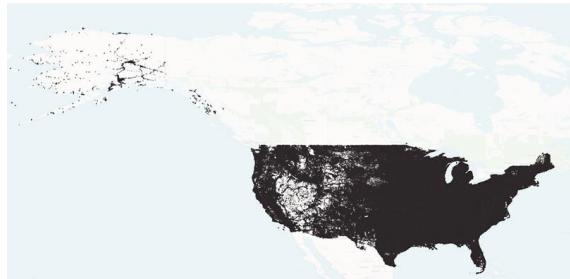
[HTTPS://DOCS.GOOGLE.COM/PRESENTATION/D/1TT6OENTUH\\_Q-WPIC8AUX67HBOCPTJ7YWNKZN-X6NY8W/EDIT?USP=SHARING](https://docs.google.com/presentation/d/1TT6OENTUH_Q-WPIC8AUX67HBOCPTJ7YWNKZN-X6NY8W/edit?usp=sharing)



## (b) Network Data Set: OSM2GMNS

OpenStreetMap (OSM) is an open-source map website providing free download.

OSM2GMNS can convert the OSM map data to multi-resolution routable network files in GMNS format. Users can convert and model drivable, walkable, railway, or aeroway networks with a few lines of Python code.



<https://pypi.org/project/osm2gmns/>

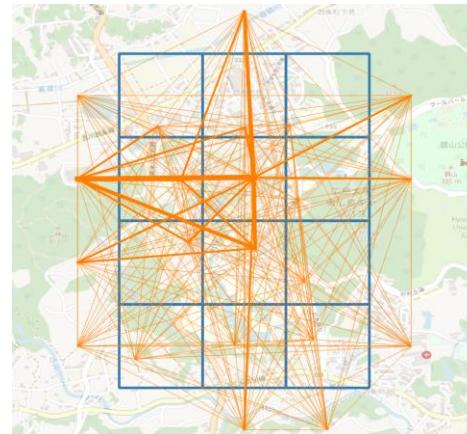
PyPI link	<a href="https://pypi.org/project/osm2gmns">https://pypi.org/project/osm2gmns</a>
Total downloads	46,148
Total downloads - 30 days	2,747
Total downloads - 7 days	538

**Jiawei (Jay) Lu**  
PhD. student at Arizona State University



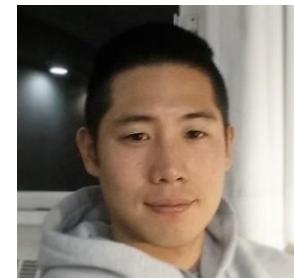
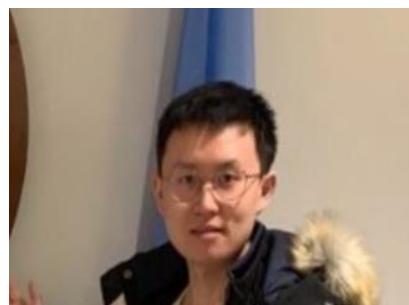
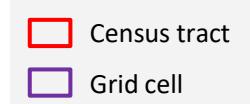
## (c) Demand Generation: Grid2Demand

Grid zones can be automatically generated by a given number of grid cells or cell's width and height in meters.



Advantages:

- ◆ Compared with traditional TAZs like census tracts, grid zones could reduce spatial biases.
- ◆ Trip generation directly comes from activity nodes, which is aggregated in grid zones.
- ◆ Cell size is flexible to satisfy multi-resolution models.
- ◆ [Open-source quick transportation demand generation tool grid2demand – YouTube](#)
- ◆ Developers: Anjun Li, Entai Wang and Taehooie Kim



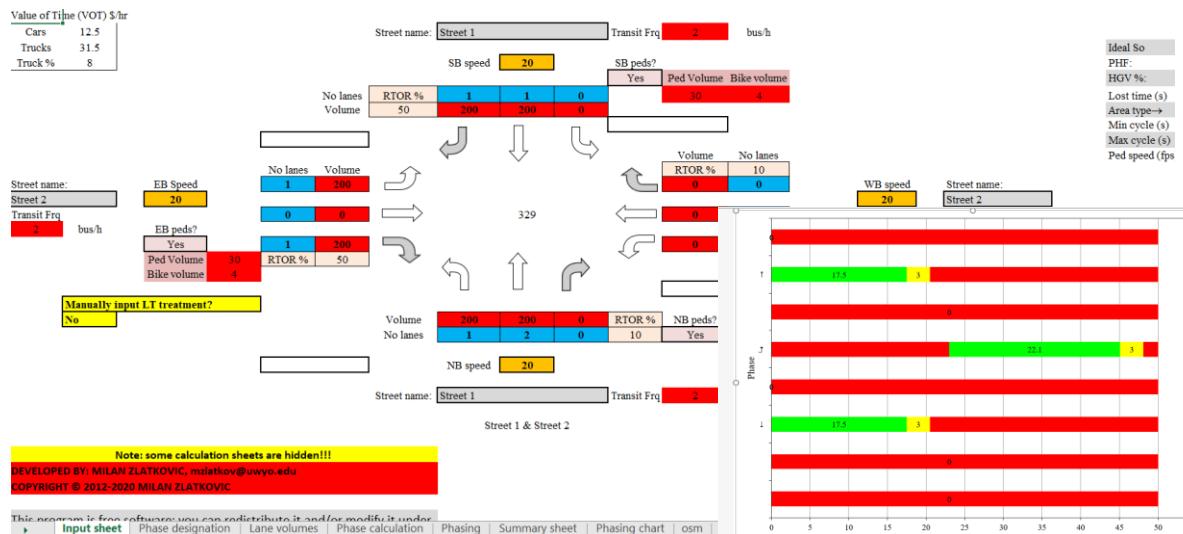
# (d) Sigma-X: Excel-based computational engine for signalized intersections,

<https://github.com/milan1981/Sigma-X>

Excel->GMNS, Synchro UTDF -> GMNS

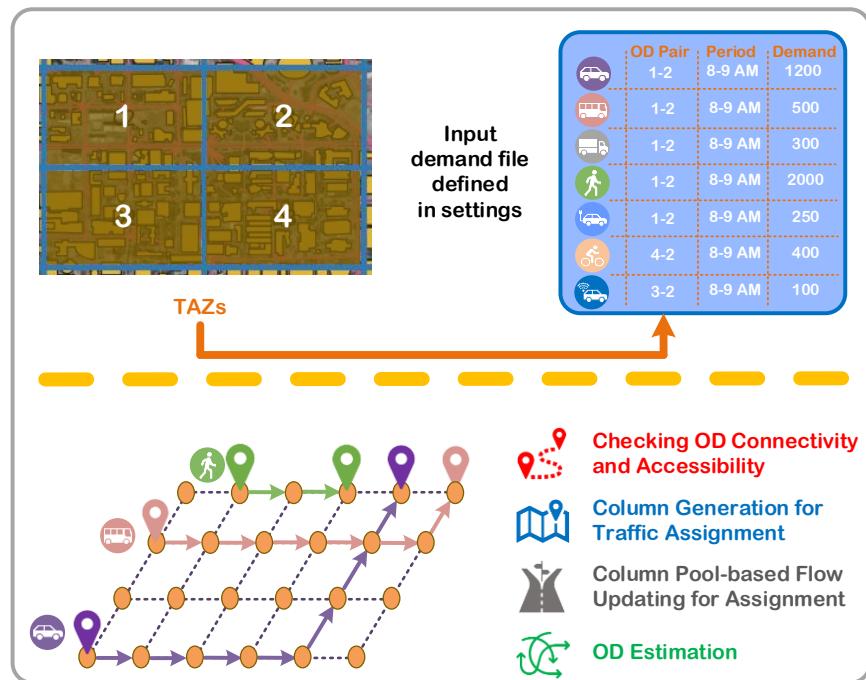
Contributor: Prof. Milan Zlatkovic

[http://www.uwyo.edu/civil/faculty\\_staff/faculty/milan-zlatkovic/index.html](http://www.uwyo.edu/civil/faculty_staff/faculty/milan-zlatkovic/index.html)



Python Package: <https://github.com/asu-trans-ai-lab/signal4gmns>

## (e) Path4GMNS



PyPI link

<https://pypi.org/project/path4gmns>

Total downloads

16,865

Total downloads - 30 days

812

Total downloads - 7 days

201

The screenshot shows the PyPI project page for "path4gmns 0.7.2". The page includes a search bar, navigation links (Help, Sponsors, Log In, Register), and a release note indicating it is the "Latest version" released on June 6, 2021. The main content area describes Path4GMNS as an open-source, cross-platform, lightweight, and fast Python path engine for networks encoded in GMNS. It highlights features such as static and time-dependent shortest path finding, column-based modeling, and activity-based demand modeling. A "Project description" section is expanded, showing the project's purpose and functionality. A "Project links" section provides a link to find static shortest paths between two nodes.

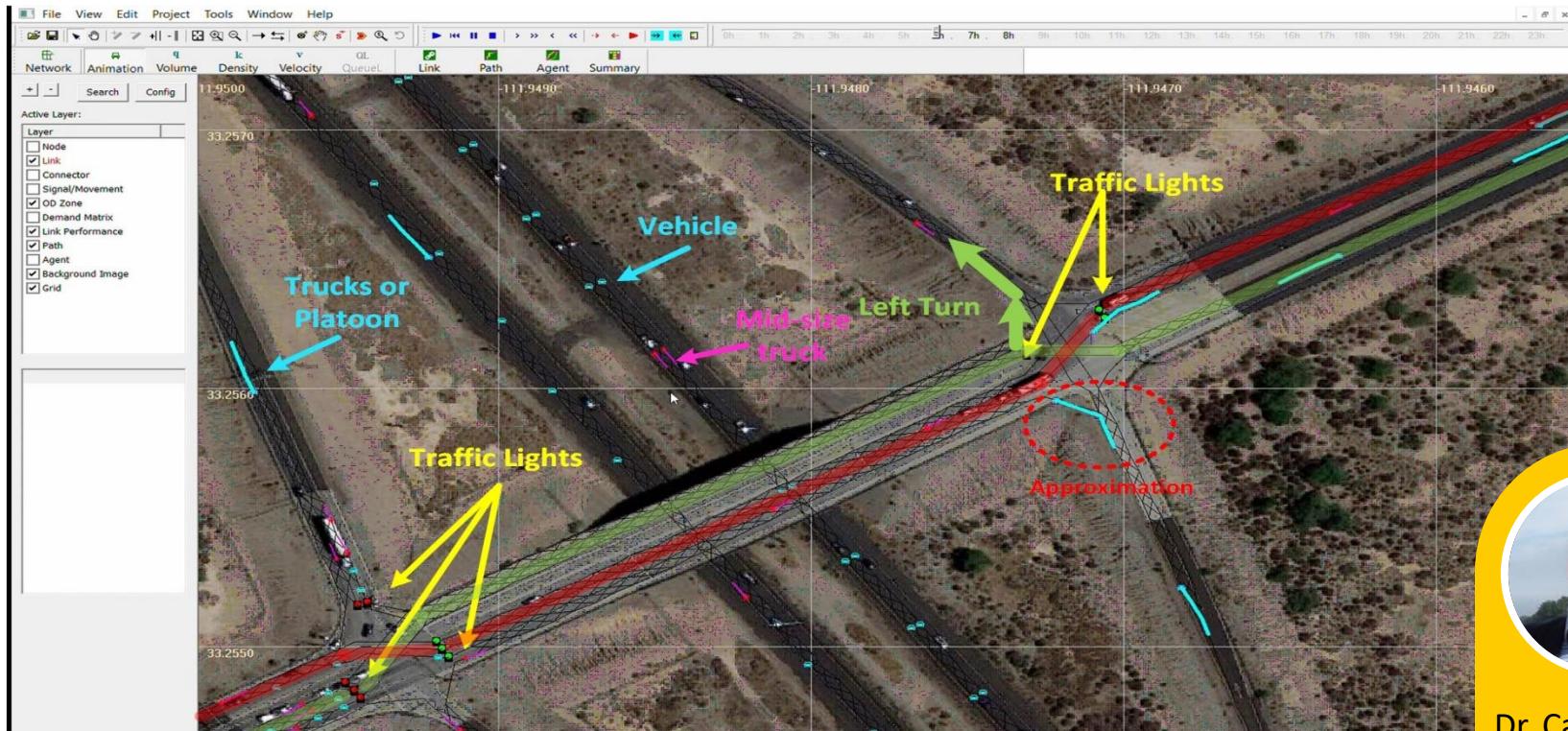


Dr. Peiheng Li

Operations Research  
Analyst at Norfolk  
Southern Corporation

ASU Graduate Ph.D.  
(2017)

## (f) DLSim: Deep Learning Based Traffic Simulation



Dr. Cafer Avci

Simulation  
Engine

# (g) GTFS2GMNS

Contributors:

Fang (Alicia) Tang  
[fangt@asu.edu](mailto:fangt@asu.edu)

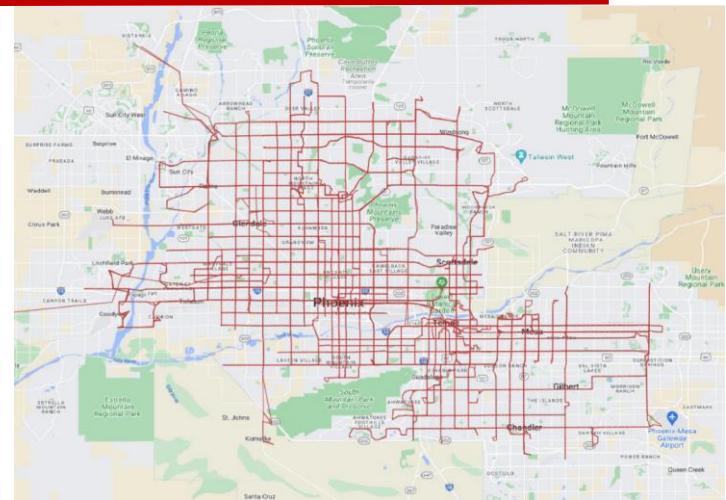
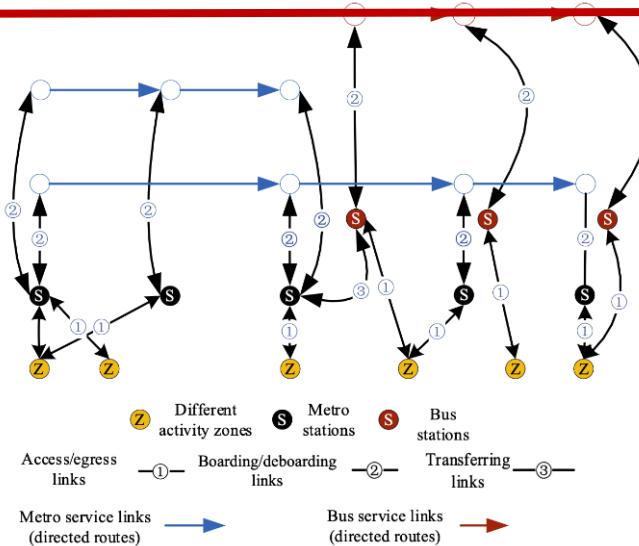
Xin (Bruce) Wu  
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Xianting (Alice) Huang  
[xantinh@andrew.cmu.edu](mailto:xantinh@andrew.cmu.edu)

Cafer Avci  
[cafer.avci@aalto.fi](mailto:cafer.avci@aalto.fi)

Taehooie Kim  
[taehooie@urbansim.com](mailto:taehooie@urbansim.com)

Xuesong (Simon) Zhou  
[xzhou74@asu.edu](mailto:xzhou74@asu.edu)



gtfs2gmns is an open-source Python code that enables users to conveniently construct any multi-modal transit network from [General Transit Feed Specification \(GTFS\)](#).



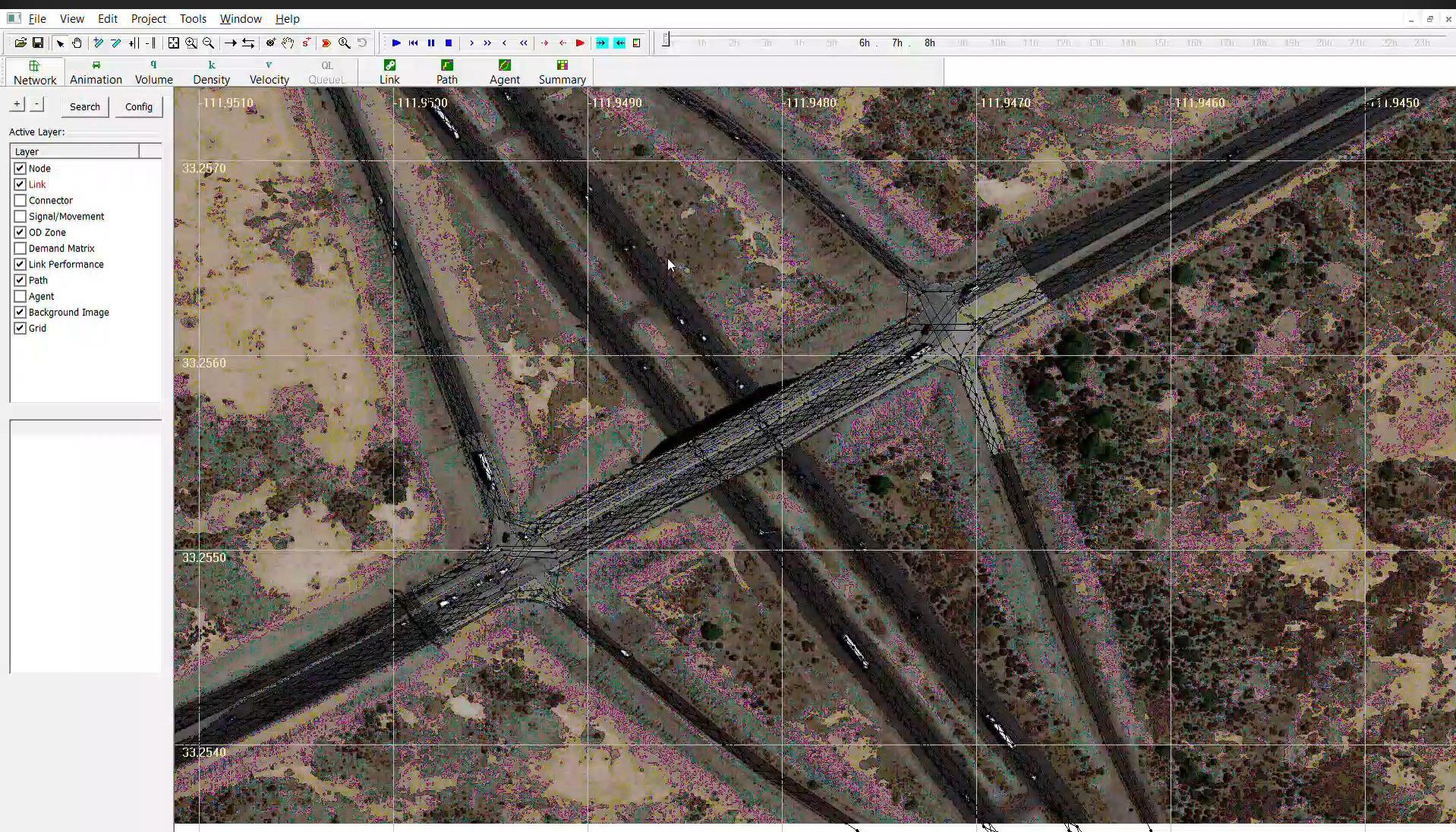
Fang (Alicia) Tang at ASU,



Xianting (Alice) Huang at CMU

**DLSIM**

# Sample Simulation of DDI



## (h) DTALite and NeXTA

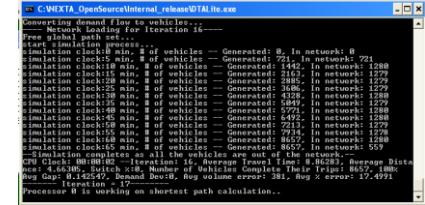
### NeXTA: front-end GUI (C++)

- Version 2: GUI for TRANSIMS and DYNASMART
- Version 3: GNU Open-source data hub

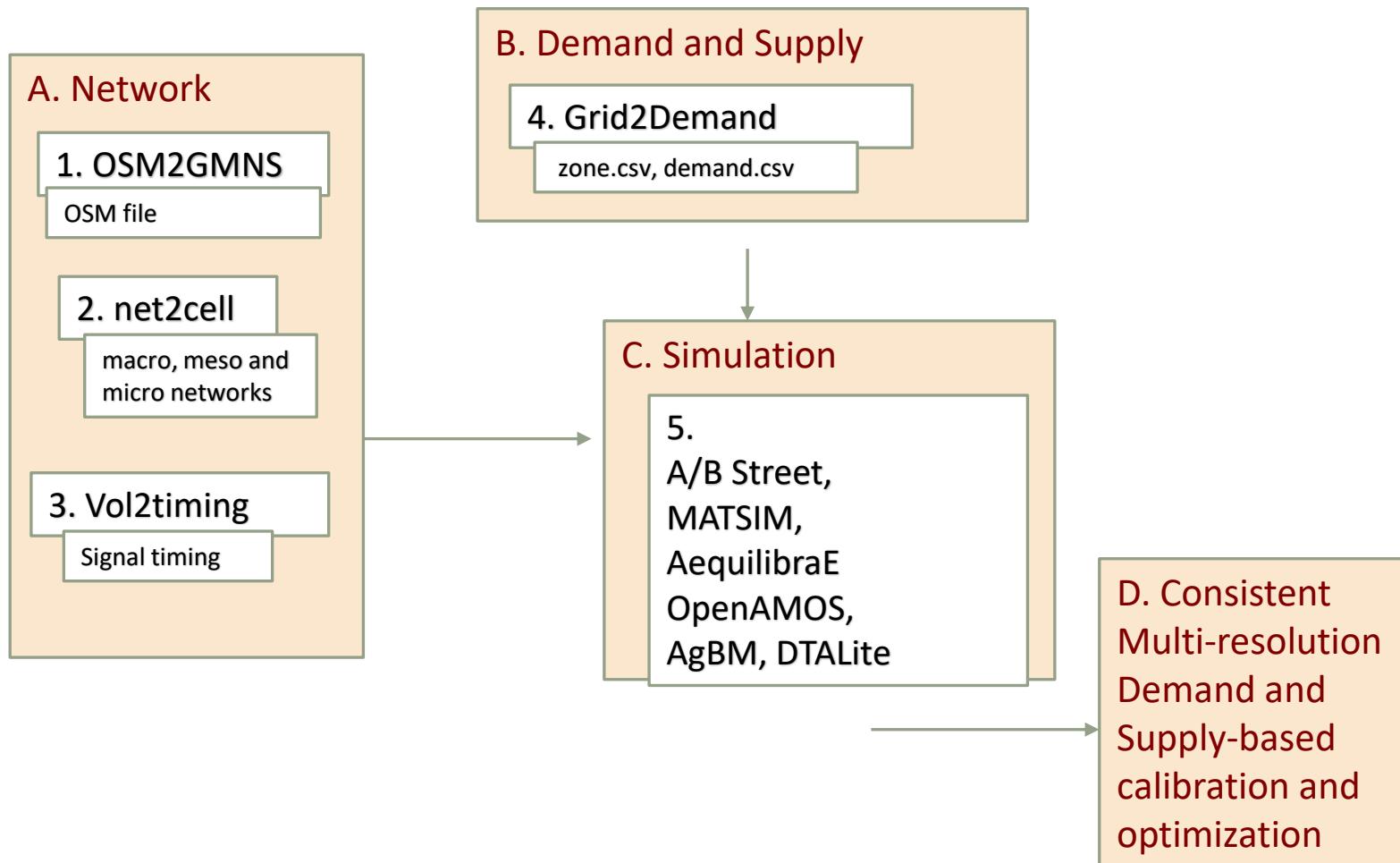


### DTALite: Open-source computational engine (C++)

- Light-weight and agent-based DTA



# Long-term Goal: Integrate Network, Demand and Supply Elements



# Case study of Indian Institute of Technology (IIT), Madras, India



[https://github.com/asu-trans-ai-lab/gmns\\_education\\_data\\_sets/tree/main/Universities/Asia/India/IIT%20Madras](https://github.com/asu-trans-ai-lab/gmns_education_data_sets/tree/main/Universities/Asia/India/IIT%20Madras)

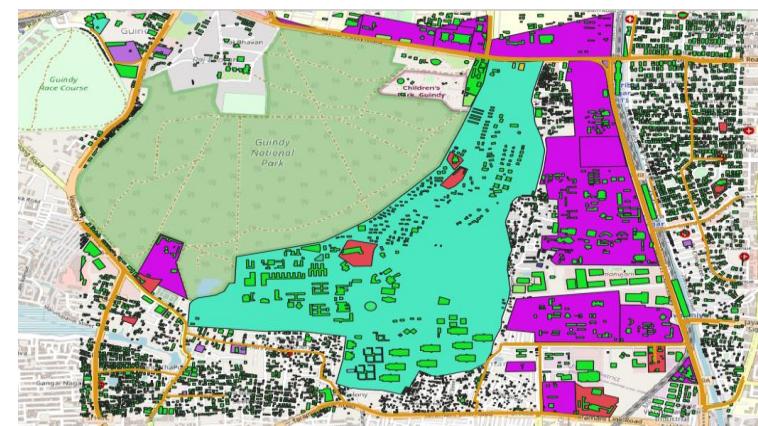
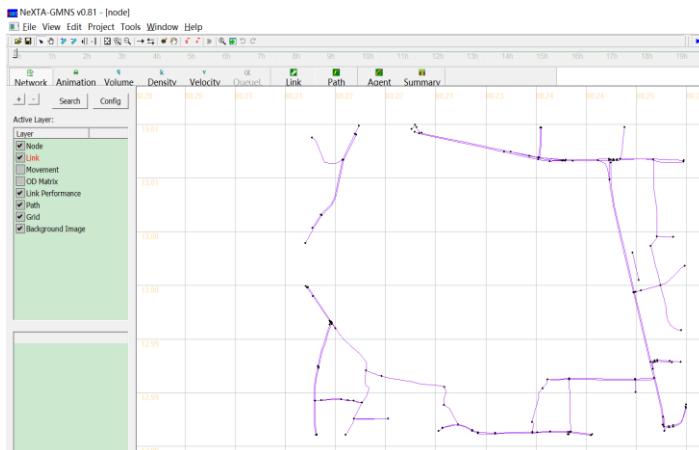
# Step 1: Open-Source Network Visualization

NEXTA: front-end Graphical User Interface GUI (C++)  NEXTA.exe

<https://github.com/xzhou99/NeXTA-GMNS>

QGIS: a free, open source, cross platform (lin/win/mac) geographical information system (GIS) 

<https://qgis.org/en/site/forusers/download.html>



More sample networks

<https://osm2gmnns.readthedocs.io/en/latest/sample-net.html>

# Step 2: Open-source MRM Tools: osm2gmns

## Some examples

Download OSM data

<https://osm2gmns.readthedocs.io/en/latest/quick-start.html>

Get a network in GMNS format

```
>>> import osm2gmns as og  
>>> net = og.getNetFromOSMFile('map.osm')  
>>> og.outputNetToCSV(net)
```

Consolidate Intersections

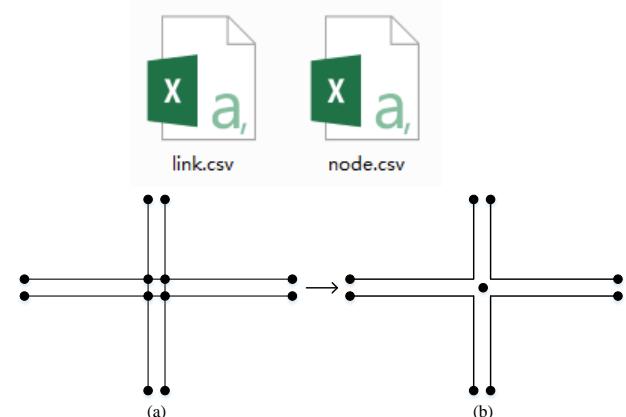
```
>>> og.consolidateComplexIntersections(net)
```

Generate movements at intersections

```
>>> og.generateMovements(net)
```

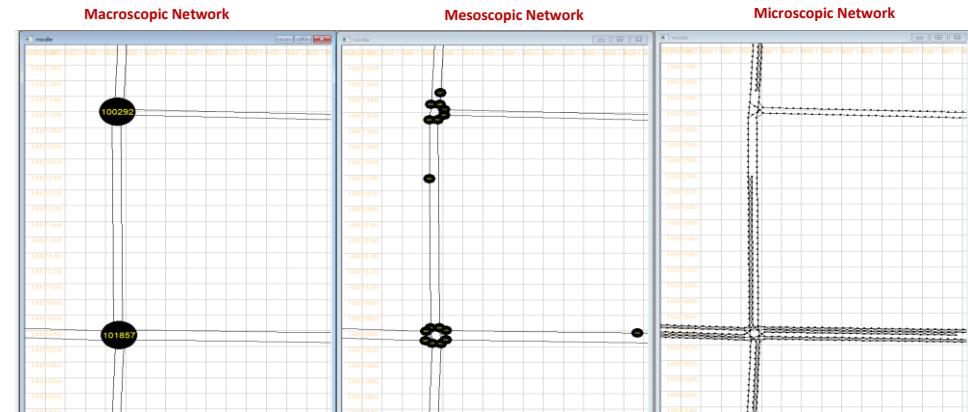
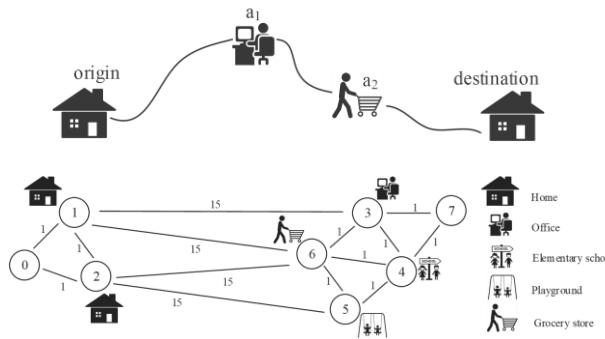
Network Types and POI

```
>>> net = og.getNetFromOSMFile('map.osm', network_type=('auto','bike','walk'))  
  
>>> net = og.getNetFromOSMFile('map.osm', POIs=True)  
  
>>> og.connectPOIWithNet(net)
```

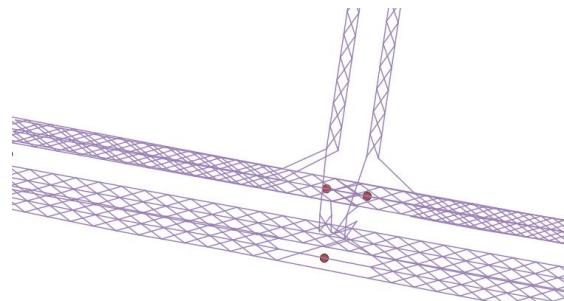


# Step 3: Multi-Resolution Network Generation

Open-source tool for creating cell-based (microscopic) and mesoscopic networks



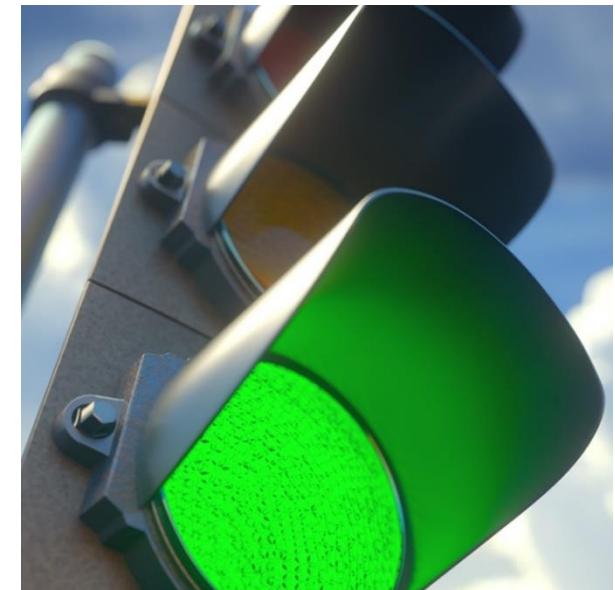
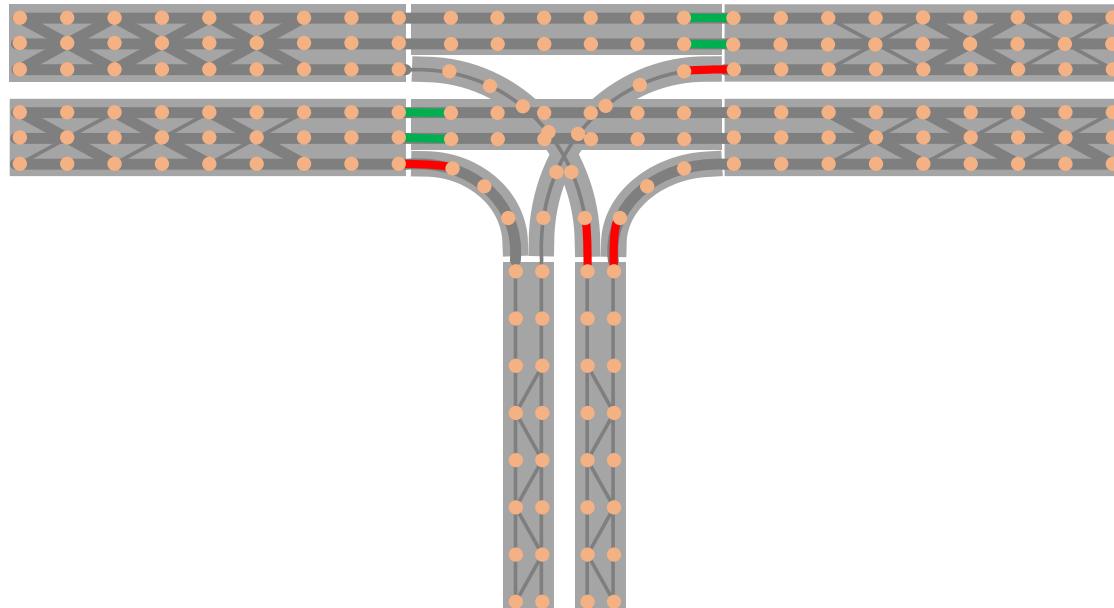
- mesonet
- micronet
- link.csv
- map.pbf
- movement.csv
- mrm.qgz
- node.csv
- poi.csv



## Step 4: Signal Timing Generation Tool

Signal4gmns download: <https://pypi.org/project/signal4gmns/>

### Signal Control Logic



- Map signal timing data on microscopic cells to simulate signal control
- Make it extremely efficient for signal optimization modeling after introducing expanded space-time network

## Step 4: Signal Timing Generation Tool

Signal4gmns download: <https://pypi.org/project/signal4gmns/>

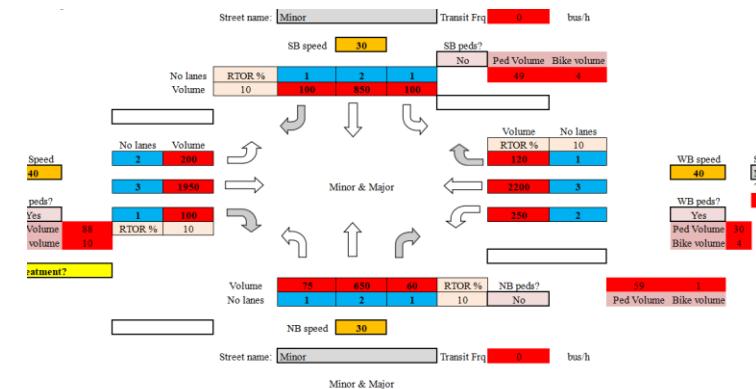
Excel-based and Python based

<https://github.com/milan1981/Sigma-X>

HCM-based computational engine for signalized intersections

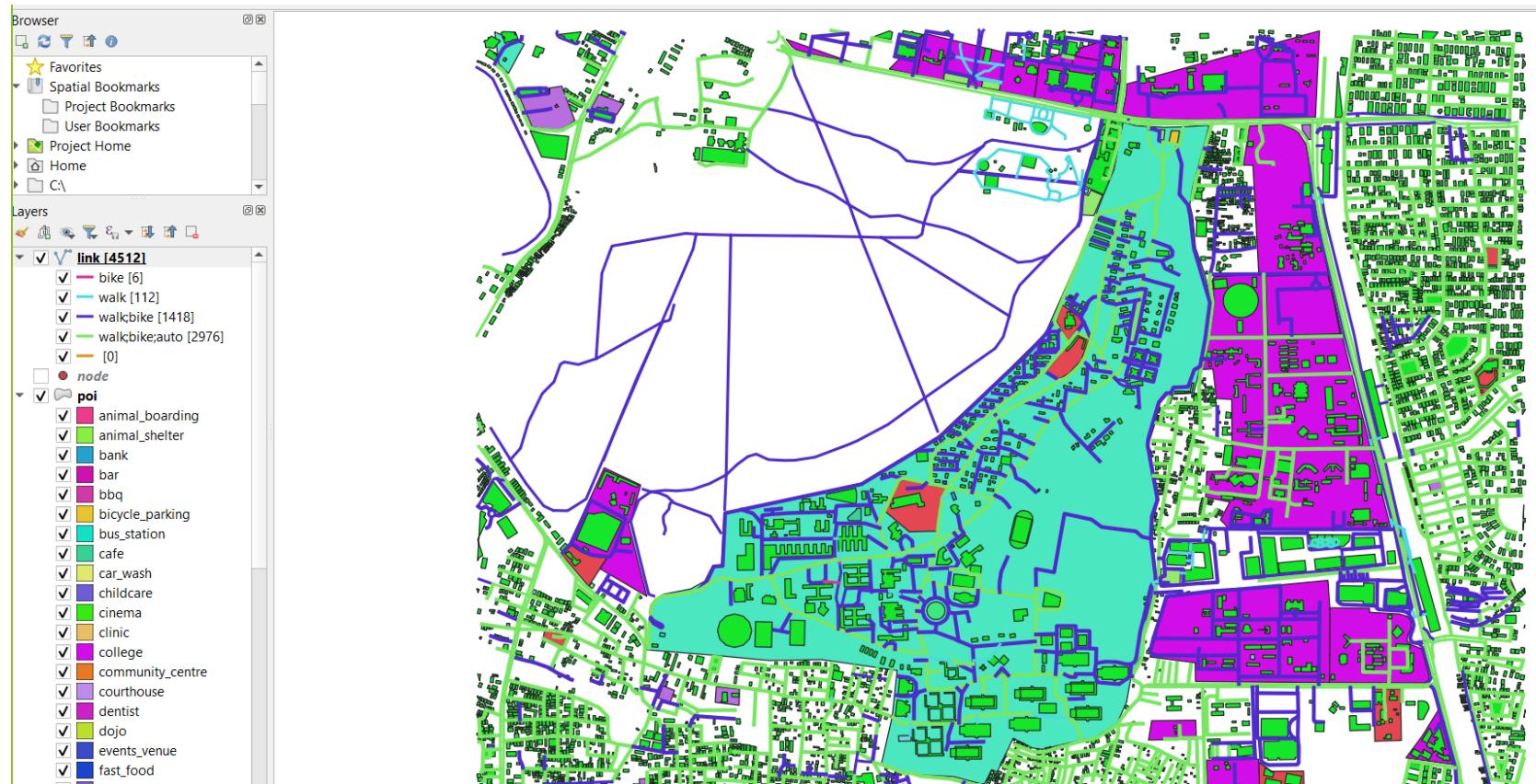


- [signal\\_movement\\_setting.csv](#)
- [signal\\_node\\_setting.csv](#)
- [signal\\_phase\\_mvmt.csv](#)
- [signal\\_timing\\_phase.csv](#)
- [timing.csv](#)



## Step 5: Demand Generation

Grid2Demand download: <https://pypi.org/project/grid2demand/>

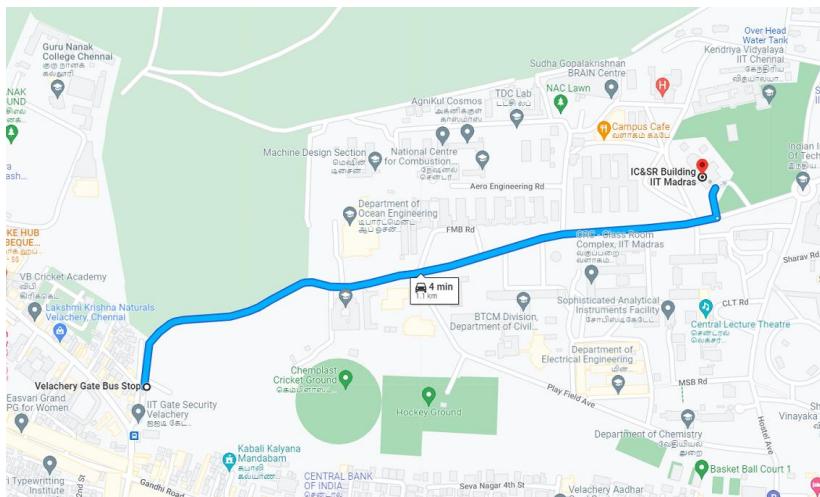


# Step 5: Demand Generation

Grid zones can be automatically generated by a given number of grid cells or cell's width and height in meters.

Advantages:

- ◆ Compared with traditional TAZs like census tracts, grid zones could reduce spatial biases.
- ◆ Trip generation directly comes from activity nodes, which is aggregated in grid zones.
- ◆ Cell size is flexible to satisfy multi-resolution models.
- ◆ [Open-source quick transportation demand generation tool grid2demand – YouTube](#)



Origin: Velachery Gate Bus Stop, Anna Garden  
Destination:

## Step 6: Shortest Path for GMNS

---

Path4GMNS download: <https://pypi.org/project/path4gmns/>

Path4GMNS is an open-source, cross-platform, lightweight, and fast Python path engine for networks encoded in GMNS. Besides finding static shortest paths for simple analyses, its main functionality is to provide an efficient and flexible framework for column-based (path-based) modeling and applications in transportation (e.g., activity-based demand modeling). Path4GMNS supports, in short,

- finding (static) shortest path between two nodes,
- constructing shortest paths for all individual agents,
- performing path-based User-Equilibrium (UE) traffic assignment,
- evaluating multimodal accessibility.



# Outline

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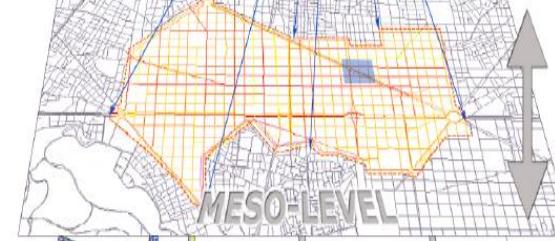
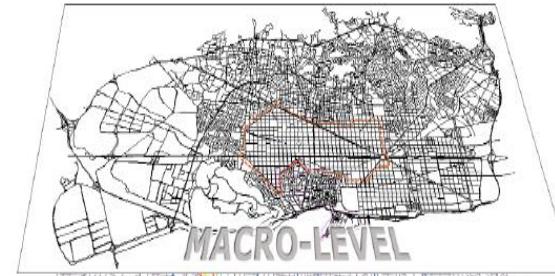
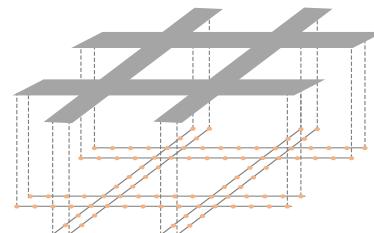
I. Open Data Specification



II. Open-Source Tools



III. Supply-side Models for Large Scale System



## Part III: Supply-side Models for Cross-resolution Analysis/Simulation

---

### Motivations

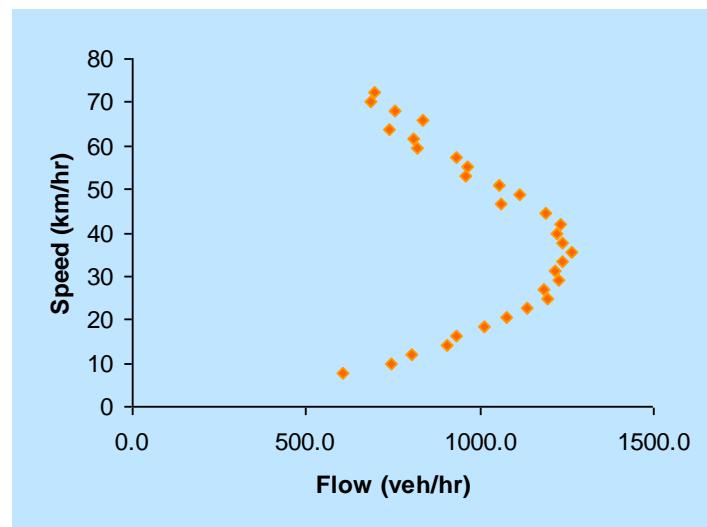
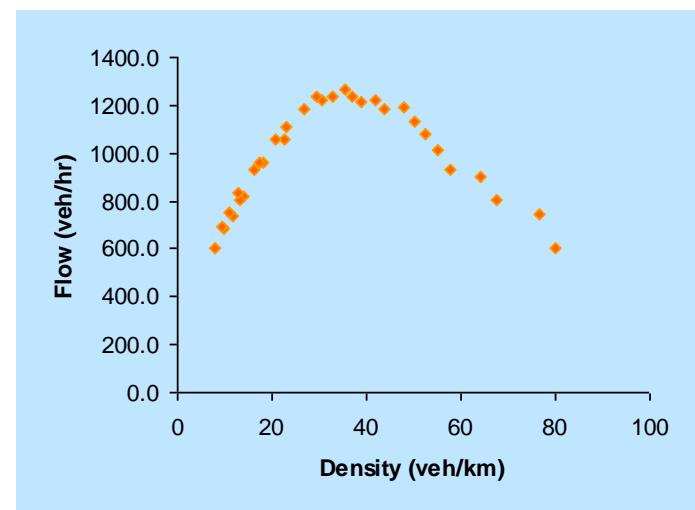
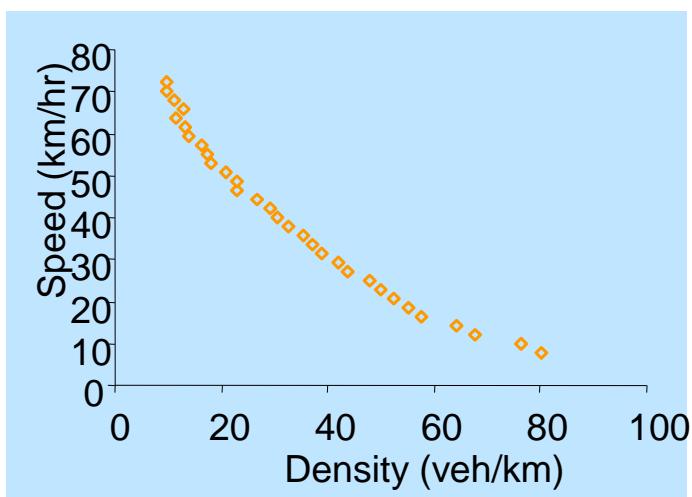
- Existing technical barriers (based on DTA user survey, TRB network modeling committee, 2009)
  - Require **too many input data**: 47%
  - Take **too long** to run: 35%
  - **Model is unclear**: 35%
- Our goals
  - **Simplified data input** from traffic assignment
  - Use **parallel computing** capability, simplified routing and simulation
  - **Open-source Visualization**: **Seeing is believing**
  - **Excel Tools**: **Start from basics**

## Part III: Keywords on Supply-side Models

---

- A. Macroscopic traffic stream models
- B. Simulation model for traffic Bottlenecks (Kinematic wave)
- C. Continuous-time fluid-based approximation for describing queue dynamics
- D. Queue-based volume-Delay function: QVDF

## A.1: Field Data (Holland Tunnel, Eddie 63)



## A.2: Single-regime models

---

- Greenberg's model →

$$u = u_m \ln(k_j / k)$$

- Underwood's model →

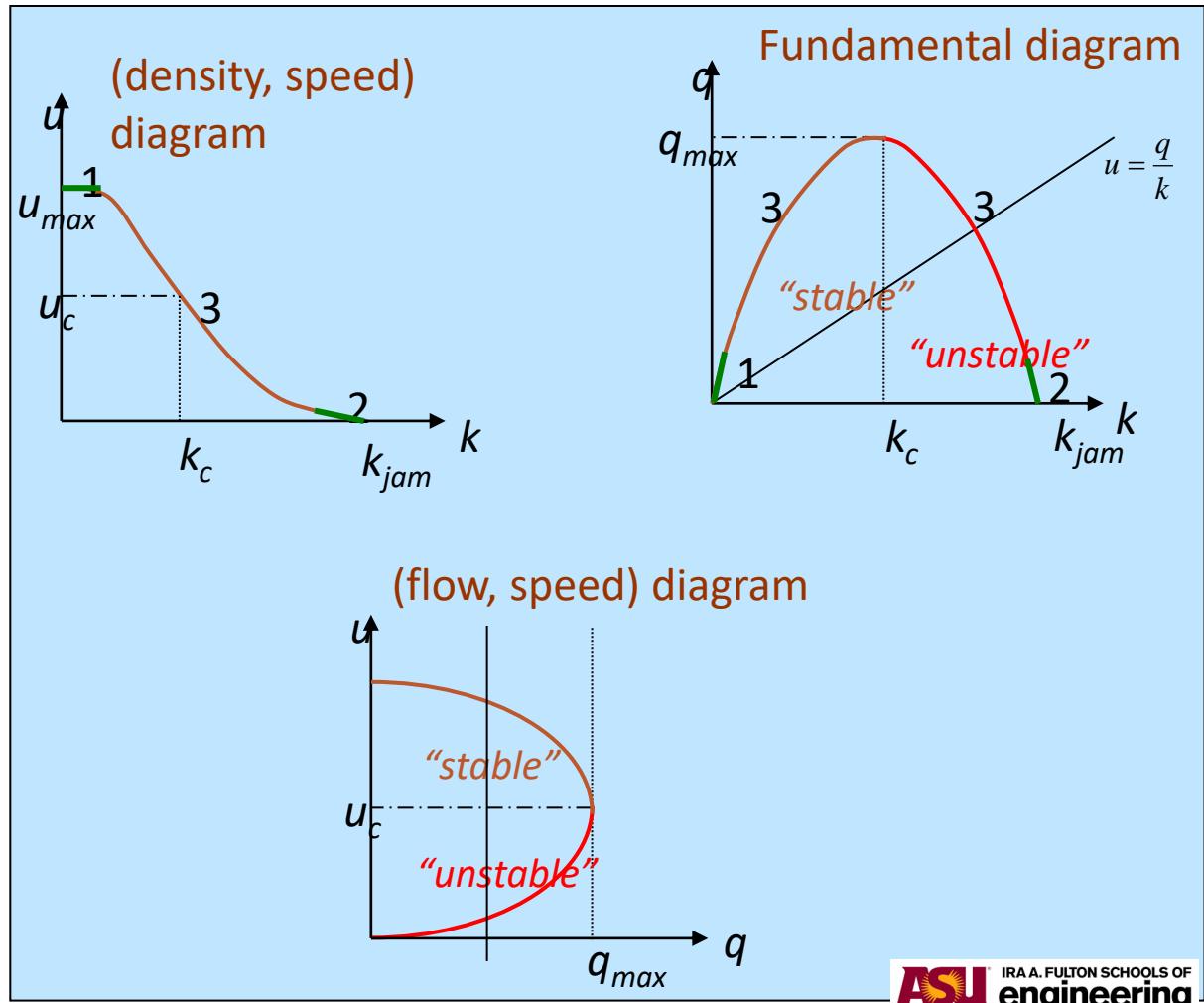
$$u = u_f e^{-k/k_m}$$

$u_f, k_j, k_m$ , parameters to be calibrated

## A.3: Multi-regime Relationships

### Structure

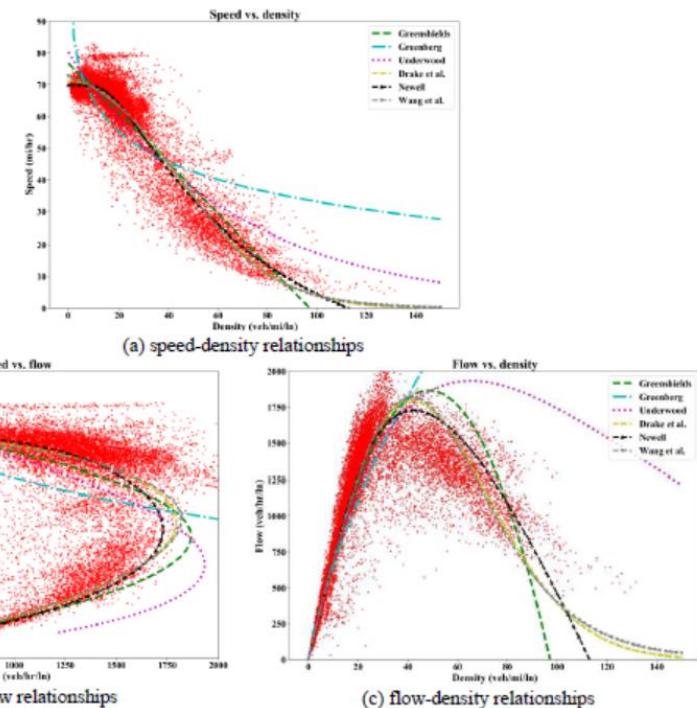
- **Single regime:** same functional form under all traffic conditions
- **Multiple regime:** different models for different traffic conditions



## A.4: Nonlinear Single-regime Traffic Flow Models

Single-regime models	Functional form	Parameters
Greenshields et al. (1935)	$v = v_f \cdot (1 - k/k_{jam})$	$v_f, k_{jam}$
Greenberg (1959)	$v = v_c \cdot \ln(k_{jam}/k)$	$v_c, k_{jam}$
Underwood (1961)	$v = v_f \cdot \exp(-k/k_c)$	$v_f, k_c$
Newell (1961)	$v = v_f \cdot \left[ 1 - \exp\left(-\frac{\lambda}{v_f} \cdot \left(\frac{1}{k} - \frac{1}{k_{jam}}\right)\right) \right]$	$v_f, k_{jam}, \lambda$
Drake et al. (1967)	$v = v_f \cdot \exp[-(k/k_c)^2/2]$	$v_f, k_c$
Pipes (1967)	$v = v_f \cdot (1 - (k/k_{jam})^m)$	$v_f, k_{jam}, m$
Drew (1968)	$v = v_f \cdot \left[ 1 - (k/k_{jam})^{m+\frac{1}{2}} \right]$	$v_f, k_{jam}, m$
Ardekani and Herman (1987)	$v = v_f \cdot (1 - f_{z,min})^{m+1} \cdot \left[ 1 - (k/k_{jam})^l \right]^{m+1}$	$v_f, k_{jam}, f_{z,min}, m, l$
Kerner and Konhäuser (1994)	$v = v_f \cdot \left[ \frac{1}{1 + \exp((k/k_c - 0.25)/0.06)} - 3.72 \times 10^{-6} \right]$	$v_f, k_c$
Castillo and Benítez (1995a)	$v = v_f \cdot \left[ 1 - \exp\left(\frac{ C_j }{v_f} \cdot \left(1 - \frac{k_{jam}}{k}\right)\right) \right]$	$v_f, k_{jam},  C_j $
Jayakrishnan et al.	$v = v_f \cdot \frac{(1 - k/k_{jam})(1 - k/k_c)^m}{k} = \frac{1}{c_1 + \frac{c_2}{v_f - v} + c_3 v}$	$v_f, v_c, k_{jam}, m$
Van Aerde (1995)		$v_f, c_1, c_2, c_3$
MacNicholas (2011)	$v = v_f \cdot \left( \frac{k_{jam}^m - k^m}{k_{jam}^m - c \cdot k^m} \right)$	$v_f, k_{jam}, m, c$
3PL model in Wang et al. (2011)	$v = \frac{v_f}{1 + \exp[(k - k_c)/\theta]}$	$v_f, k_c, \theta$
Ni et al. (2016)	$k = \frac{1}{(\gamma v^2 + \tau v + l) \left[ 1 - \ln(1 - v/v_f) \right]}$	$v_f, \gamma, \tau, l$
This paper	$v = \frac{v_f}{\left[ 1 + (k/k_c)^m \right]^{2/m}}$	$v_f, k_c, m$

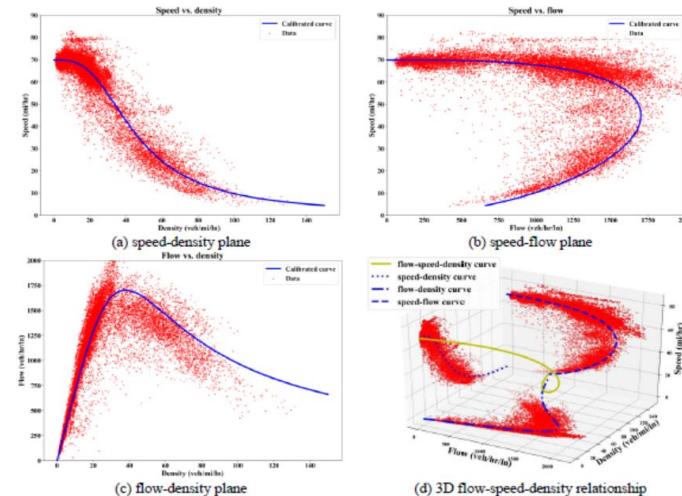
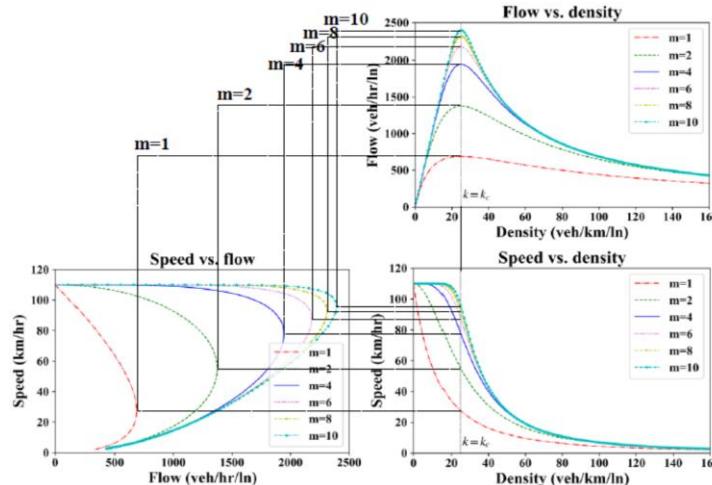
Cheng, Q., Liu, Z., Lin, Y., & Zhou, X. S. (2021). An s-shaped three-parameter (S3) traffic stream model with consistent car following relationship. *Transportation Research Part B: Methodological*, 153, 246-271.



## A.5: S3 Model

3 parameter: Vf, Kcritical, m: flow maximizing inertia factor

Parsimony and interpretability



**Advantages:**

1. With only three parameters to cover **both light-traffic and jam conditions**.
2. Propose a three-dimensional (3D) traffic flow model with the **3D surface connecting the flow, speed, and density simultaneously**.

An s-shaped three-parameter (S3) traffic stream model with consistent car following relationship – ScienceDirect

Cheng, Q., Liu, Z., Lin, Y., & Zhou, X. S. (2021). An s-shaped three-parameter (S3) traffic stream model with consistent car following relationship. *Transportation Research Part B: Methodological*, 153, 246-271.

## A.6: S-shaped Three-parameter (S3) Fundamental Diagram

---

$$v = \frac{v_f}{\left[1 + (k/k_c)^m\right]^l} \quad q = \frac{k \cdot v_f}{\left[1 + (k/k_c)^m\right]^l}$$

where  $v_f$  is the free flow speed,  $k_c$  is the critical density which can result in a maximal traffic flow rate, and  $m$  and  $l$  represent the flow maximization ratio (**FMR**) to be determined.

$q(k)$  is continuously differentiable. When  $k=k_c$ , we have  $q=q_{max}$ .

$$\left. \begin{aligned} \frac{dq}{dk} &= \frac{v_f}{\left[1 + (k/k_c)^m\right]^l} - \frac{k \cdot v_f}{k_c} \cdot \frac{l \cdot m \cdot (k/k_c)^{m-1}}{\left[1 + (k/k_c)^m\right]^{l+1}} \\ \frac{dq}{dk} \Big|_{k=k_c} &= 0 \Rightarrow \frac{v_f}{2^l} - v_f \cdot \frac{l \cdot m}{2^{l+1}} = 0 \Rightarrow l = \frac{2}{m} \end{aligned} \right\} \Rightarrow \left\{ \begin{aligned} v &= \frac{v_f}{\left[1 + (k/k_c)^m\right]^{2/m}} \\ q &= \frac{k \cdot v_f}{\left[1 + (k/k_c)^m\right]^{2/m}} \end{aligned} \right.$$

## A.7: S-shaped Three-parameter (S3) Fundamental Diagram

---

Averaged maximal flow rate (or capacity):

$$k = k_c \Rightarrow q_{\max} = \frac{k_c \cdot v_f}{2^{2/m}}$$

Explanation of the **flow-maximization** parameter  $m$ :

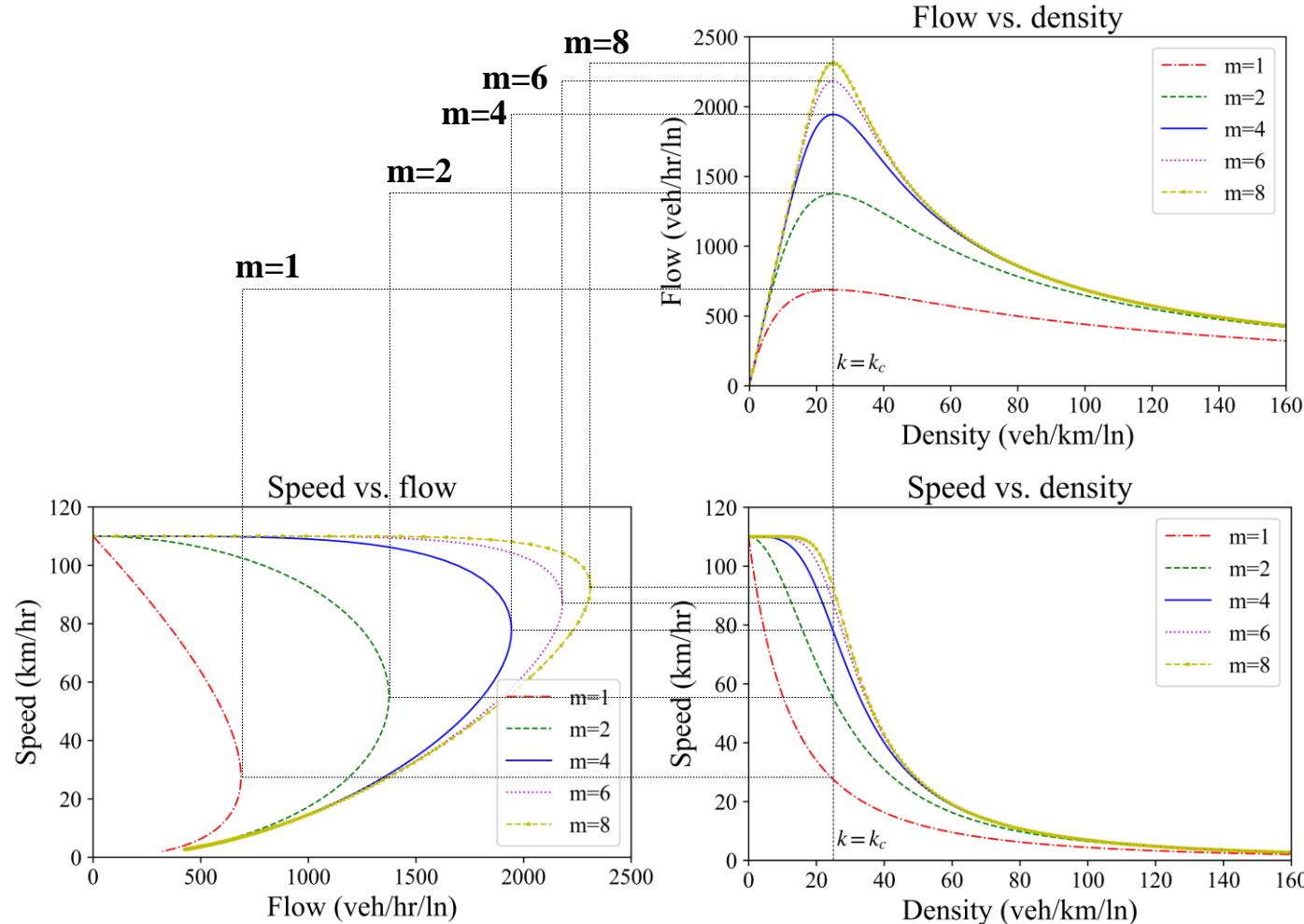
$$v_c = \frac{v_f}{\left[1 + \left(k_c/k_c\right)^m\right]^{2/m}} \Rightarrow v_c = \frac{v_f}{2^{2/m}} \Rightarrow m = \frac{2 \ln 2}{\ln(v_f/v_c)}$$

where  $v_c$  is the speed at the critical density.

## A.8: S-shaped Three-parameter (S3) Fundamental Diagram

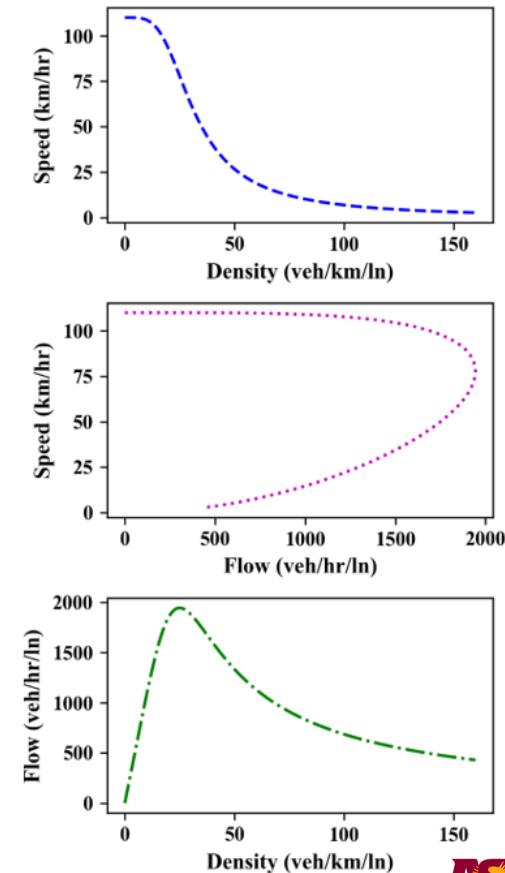
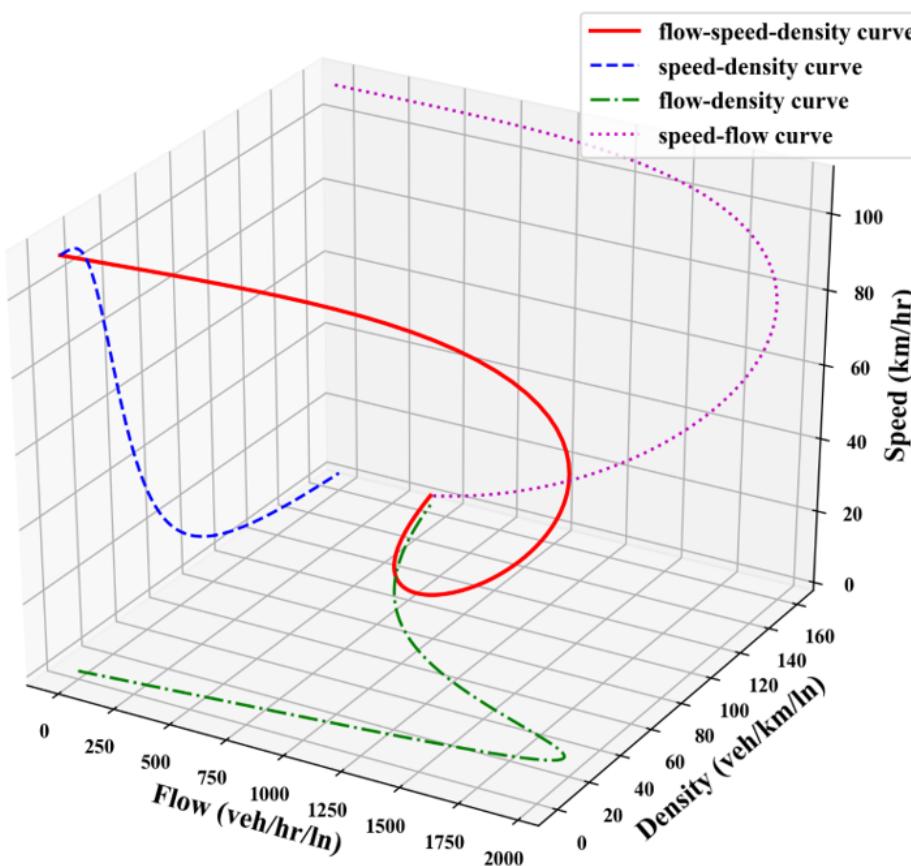
Projections to the 2D planes:

Discuss



## A.9: S-shaped Three-parameter (S3) Fundamental Diagram

3D Space when  $m=4$



## A.10: Deriving Macro-to-Micro Car-following Models

---

Macroscopic speed-density

$$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^m\right]^{2/m}}$$

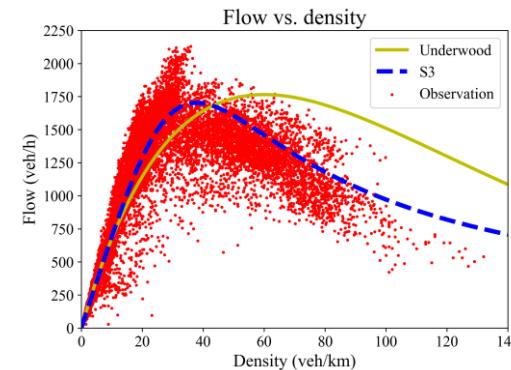
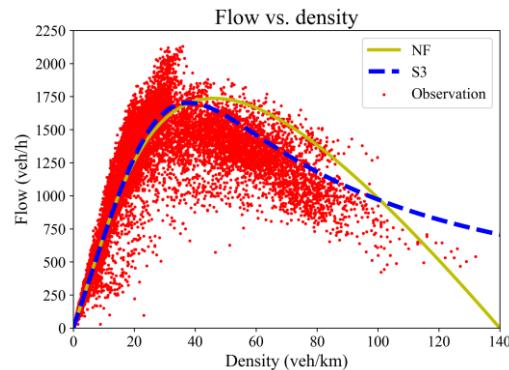
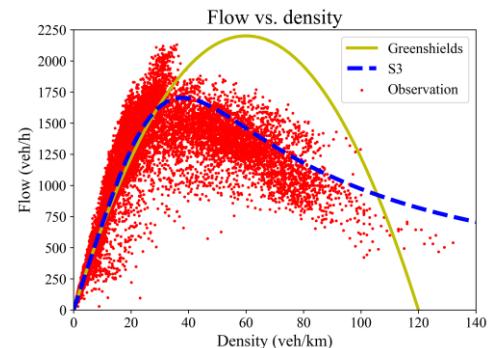
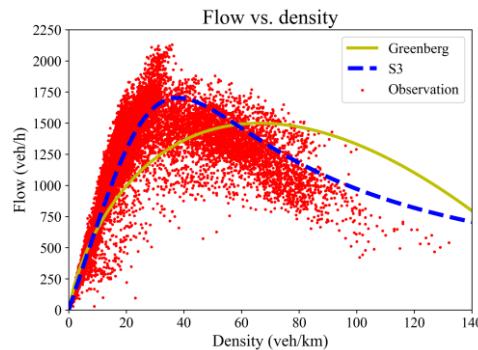
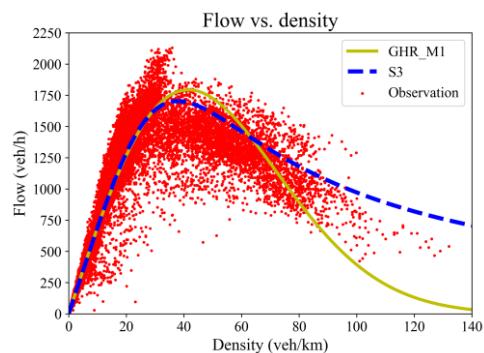
Microscopic car-following

$$a_n(t + \tau_n) = \frac{\alpha \cdot \Delta x_n(t) \cdot \Delta v_n(t)}{\left[1 + \left(\beta \cdot \Delta x_n(t)\right)^m\right]^{1+\frac{2}{m}}}$$

FoC	Speed-density function	Car-following model	Connection
$m=1$	$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^2\right]^2}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^3\right]^3} \cdot \Delta v$	
$m=2$	$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^2\right]}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^2\right]^2} \cdot \Delta v$	
$m=3$	$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^3\right]^{2/3}}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^3\right]^{5/3}} \cdot \Delta v$	$\alpha = 2v_f\beta^2$
$m=4$	$v = \frac{v_f}{\sqrt{1 + \left(k/k_c\right)^4}}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^4\right]^{3/2}} \cdot \Delta v$	$\beta = k_c$
$m=5$	$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^5\right]^{2/5}}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^5\right]^{7/5}} \cdot \Delta v$	
$m=6$	$v = \frac{v_f}{\left[1 + \left(k/k_c\right)^6\right]^{1/3}}$	$a = \frac{\alpha \cdot \Delta x}{\left[1 + \left(\beta \cdot \Delta x\right)^6\right]^{4/3}} \cdot \Delta v$	

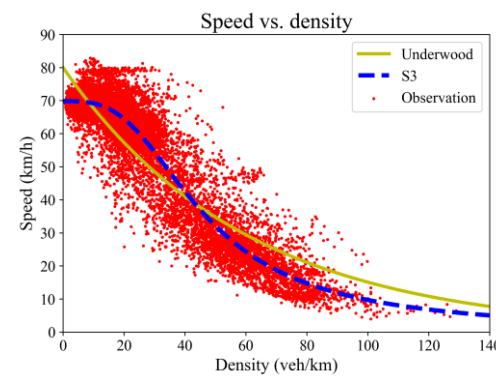
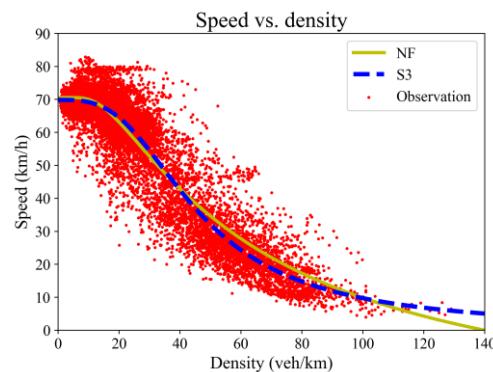
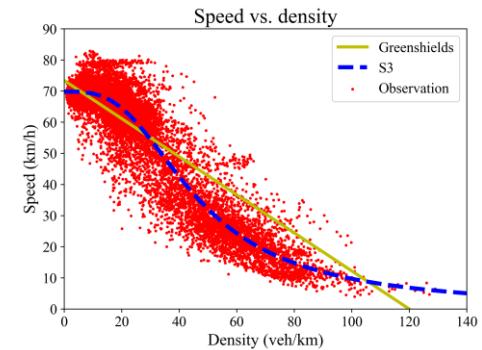
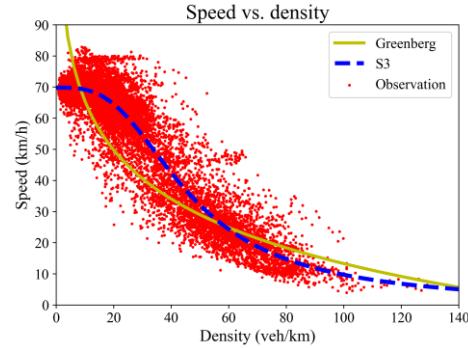
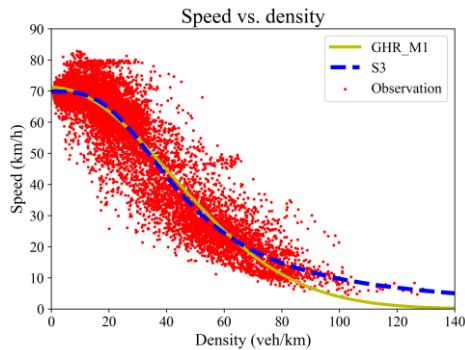
## A.11: Model calibration

### Flow vs. Density



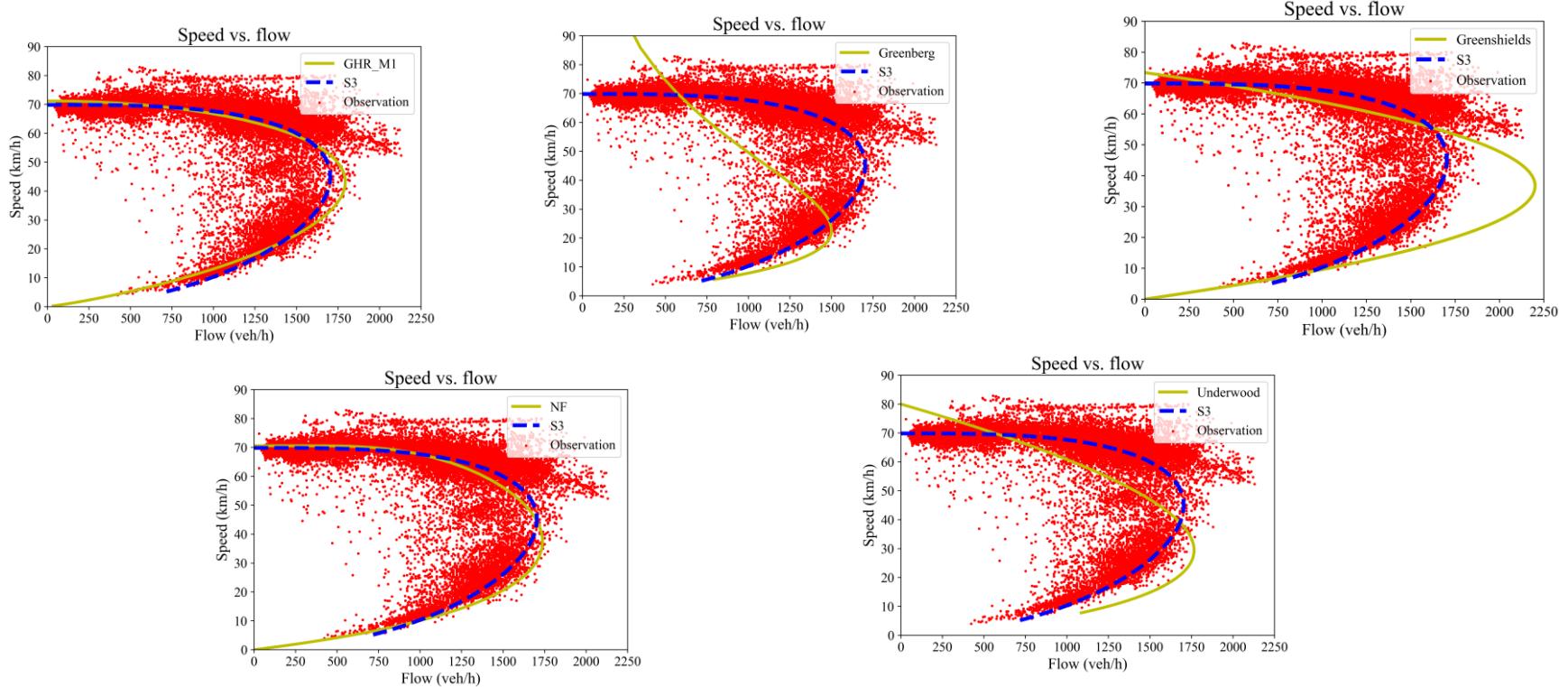
## A.11: Model calibration

### Speed vs. Density



## A.11: Model calibration

### Flow vs. Speed

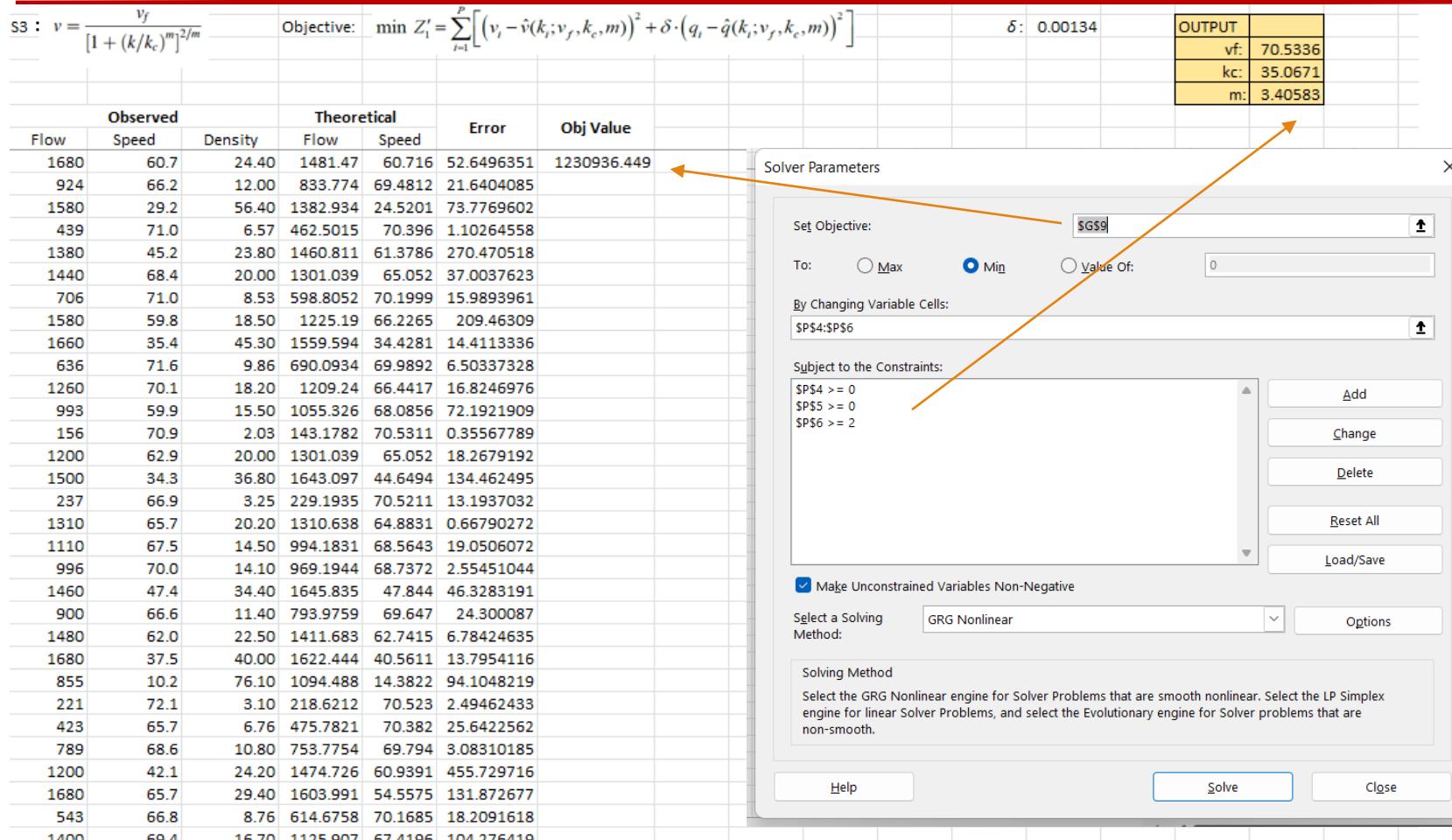


# Comparison of MRE on the estimated speed with only speed information in the objective function

Literature	Density range (veh/mi)												
	0~10	10~20	20~30	30~40	40~50	50~60	60~70	70~80	80~90	90~100	>100	Avg	Std
GS	3.59%	6.69%	9.43%	22.30%	40.92%	53.27%	60.08%	78.51%	76.80%	55.59%	69.70%	43.35%	28.46%
GB	22.34%	17.09%	25.32%	23.93%	19.32%	18.07%	23.32%	36.90%	43.60%	48.46%	62.35%	30.97%	14.82%
UW	7.22%	9.02%	14.51%	18.99%	23.94%	28.24%	34.01%	52.19%	60.34%	65.41%	86.86%	36.43%	26.12%
NW	2.96%	5.10%	9.00%	18.82%	25.00%	25.68%	26.92%	35.73%	34.36%	25.27%	33.74%	22.05%	11.68%
DK	3.15%	5.20%	8.41%	20.02%	26.53%	22.62%	21.29%	24.31%	29.83%	49.23%	65.00%	25.05%	18.44%
PP: M1	6.15%	7.37%	10.96%	20.33%	35.52%	46.74%	54.96%	76.71%	81.98%	73.35%	68.49%	43.87%	29.32%
PP: M2	6.76%	6.43%	10.76%	19.26%	28.93%	32.63%	33.58%	41.93%	36.43%	22.31%	41.67%	25.52%	13.27%
DR	6.15%	7.37%	10.96%	20.33%	35.52%	46.74%	54.96%	76.71%	81.98%	73.35%	68.49%	43.87%	29.32%
KK	3.65%	5.34%	8.46%	20.42%	27.41%	22.73%	21.23%	24.79%	31.82%	49.92%	62.19%	25.27%	17.92%
JA	6.76%	6.43%	10.76%	19.26%	28.93%	32.63%	33.58%	41.93%	36.43%	22.31%	41.67%	25.52%	13.27%
VA	220.45 %	22.06%	26.10%	24.76%	17.60%	13.02%	13.17%	12.22%	10.32%	11.20%	10.97%	34.72%	61.87%
MN	2.81%	5.11%	8.28%	19.33%	23.61%	20.44%	21.18%	24.80%	22.87%	24.95%	28.02%	18.31%	8.71%
3PL	3.65%	5.34%	8.46%	20.42%	27.41%	22.73%	21.23%	24.79%	31.82%	49.92%	62.19%	25.27%	17.92%
4PL	3.18%	5.21%	8.24%	20.01%	24.16%	19.59%	21.15%	24.64%	23.51%	24.69%	54.18%	20.78%	13.70%
5PL	2.71%	5.13%	8.20%	18.68%	21.59%	19.45%	21.51%	28.18%	29.45%	29.67%	49.13%	21.25%	13.22%
S3 model	2.71%	5.11%	8.22%	18.96%	22.26%	19.66%	21.28%	26.54%	25.31%	22.49%	26.48%	18.09%	8.64%

Discussion: Under different density ranges, S3 model has the least avg and std error in terms of MRE

# Case study: Excel-based Traffic Stream Model Calibration



[https://github.com/asu-trans-ai-lab/Traffic-Flow-Fundamental-Diagram/tree/main/Traffic\\_Model\\_Calibration%20S3](https://github.com/asu-trans-ai-lab/Traffic-Flow-Fundamental-Diagram/tree/main/Traffic_Model_Calibration%20S3)

# Case study: Python-based Traffic Stream Model Calibration

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from math import sqrt
from sklearn.metrics import mean_squared_error, r2_score
plt.rcParams.update({'figure.max_open_warning': 0})
plt.rc('font',family='Times New Roman')
plt.rcParams['mathtext.fontset']='stix'

class fundamental_diagram_model():

    def __init__(self, observed_flow, observed_density, observed_speed):
        self.observed_flow = observed_flow
        self.observed_density = observed_density
        self.observed_speed = observed_speed

    def S3(self, beta):
        vf, kc, foc = beta
        estimated_speed = vf/np.power(1 + np.power((self.observed_density/kc), foc), 2/foc)
        f_obj = np.mean(np.power(estimated_speed - self.observed_speed, 2))
        return f_obj

class first_order_derivative():

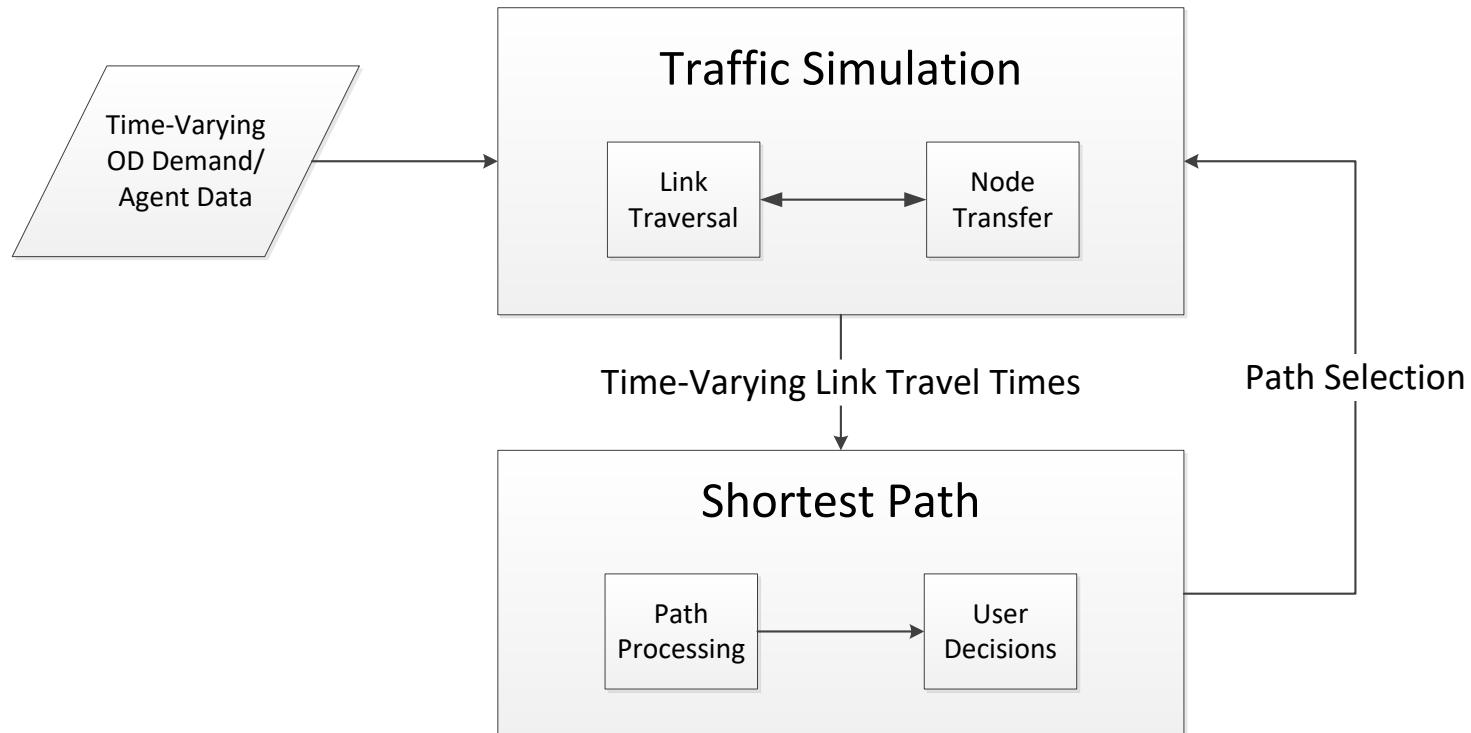
    def __init__(self, observed_flow, observed_density, observed_speed):
        self.observed_flow = observed_flow
        self.observed_density = observed_density
        self.observed_speed = observed_speed

    def S3(self, beta):
        vf, kc, foc = beta
        intermediate_variable = np.power(self.observed_density/kc, foc)
        first_order_derivative_1 = 2*np.mean((vf/np.power(1 + intermediate_variable, 2/foc) - self.observed_speed) / np
        first_order_derivative_2 = 2*np.mean((vf/np.power(1 + intermediate_variable, 2/foc) - self.observed_speed) * 2,
        first_order_derivative_3 = 2*np.mean((vf/np.power(1 + intermediate_variable, 2/foc) - self.observed_speed) * 2,
        first_order_derivative = np.asarray([first_order_derivative_1, first_order_derivative_2, first_order_derivative_3])
        return first_order_derivative

class estimated_value():

    def __init__(self, observed_flow, observed_density, observed_speed):
        self.observed_flow = observed_flow
        self.observed_density = observed_density
        self.observed_speed = observed_speed
```

## B: Bottleneck based Traffic Simulation



## B.1 Traffic Flow Theoretical Foundation: LWR Equations

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Hyperbolic system of conservation laws

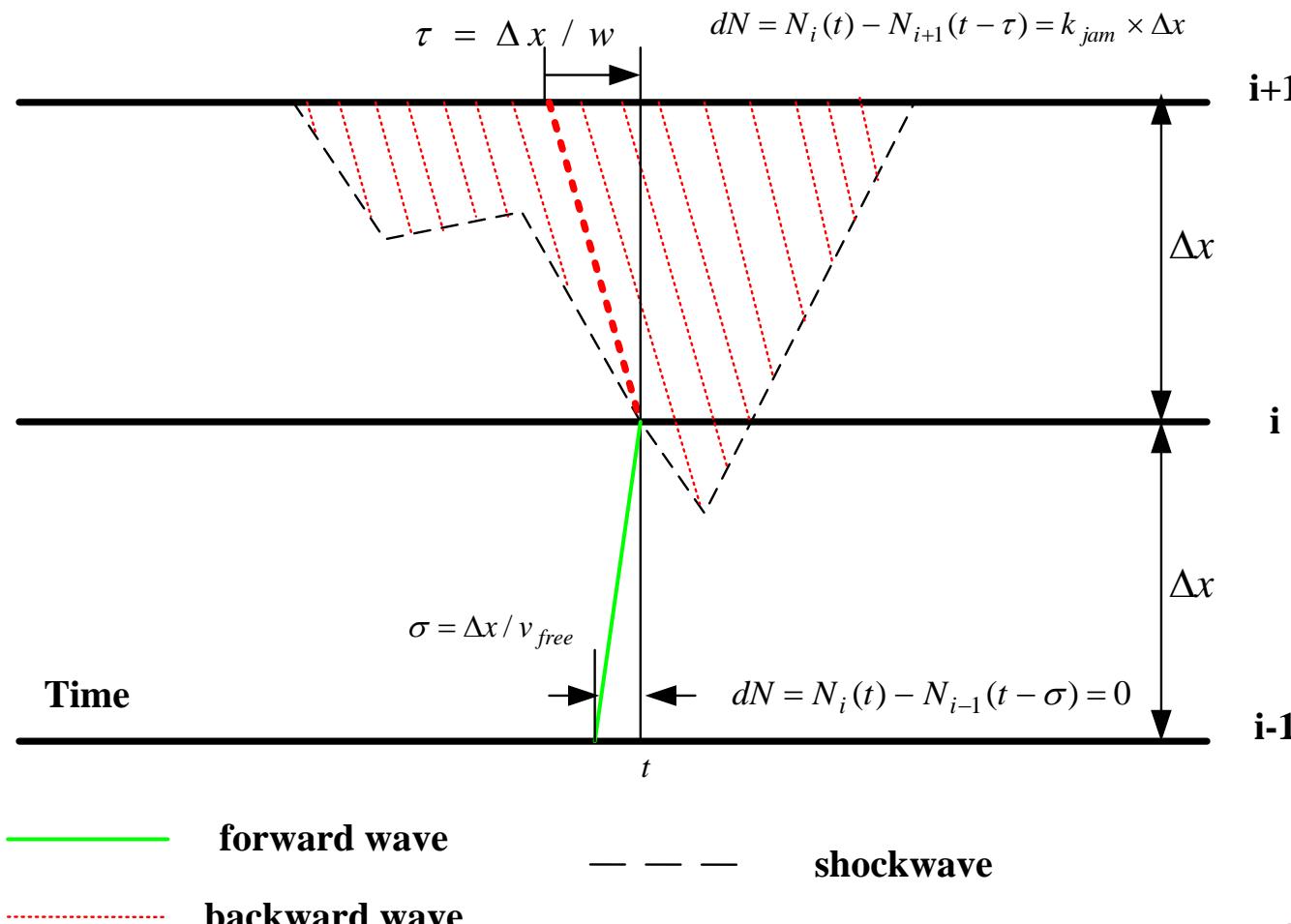
$$\frac{\partial q}{\partial x} + \frac{\partial k}{\partial t} = g(x, t)$$

where  $q$ ,  $k$  are flow and density, respectively, and  $g(\cdot)$  is the net vehicle generation rate.

Traffic flow model

- Speed (or flow) of traffic is a function of density only
- E.g. Greenshields model:  $V = V_{\text{free}} (1 - k/K_{\text{jam}})$

## B.2: Illustration of N-Curve Computation For Tracking Queue Spillback



## B.3: Numerical Calculation Scheme Based on Link Flow/Density (Used in CTM)

---

### Notation

- $\Delta t$  = length of simulation interval (e.g. 1 sec, 6 sec)
- $\Delta x$  = cell length (e.g. 0.5 mile)
- $k_{i,t}, v_{i,t}$  = the prevailing density and mean speed in cell  $i$  during the  $t^{\text{th}}$  time step
- $q_{i,t}$  = transfer flow rate from section  $i$  to cell  $i+1$  during the  $t^{\text{th}}$  time interval  $[t, t+\Delta t]$ .

### Computation Procedure

- Step 1: Calculate **prevailing speed**  $v_{i,t}$  according to a traffic flow model
- Step 2: Calculate **flow ready to move** from cell  $i-1$  to section  $i$ :  $v_{i,t} \times k_{i,t}$
- Step 3: Calculate **transfer flow** from section  $i$  to cell  $i+1$

$$q_{i,t} = \text{Min} \{ v_{i,t} \times k_{i,t}, q_{\max i+1,t} \}$$

- Step 4: Update **prevailing density** at cell  $i$

$$k_{i,t+1} = k_{i,t} + (q_{i-1,t} - q_{i,t}) \times \Delta t / \Delta x$$

Issue: transfer flow  $q_{i,t}$  is an explicit function of the occupancy at the current cell, other than the downstream capacity

- Transferred flow exceeds the spatial storage capacity at the next cell

## B.4: Link Transmission Model

along the corridor, and  $N(x, t)$  is the cumulative flow count at location  $x$  and time  $t$  of a link. The change of  $N(x, t)$  along a characteristic line (wave) is represented as follows:

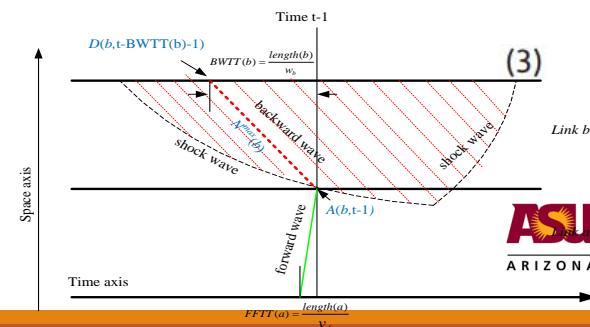
$$dN(x, t) = \frac{\partial N}{\partial x} dx + \frac{\partial N}{\partial t} dt = qdt - kdx \quad (1)$$

A wave represents the propagation of a change in flow and density along the roadway, and the wave speed is the slope of the characteristics line  $w = \frac{\partial q}{\partial k} = \frac{dx}{dt}$ . Along the movement of a wave, we substitute  $dt = \frac{dx}{w}$  into the above equation, so that we can link the difference of cumulative flow counts together through

$$dN(x, t) = qdt - kdx = \left( -k + \frac{q}{w} \right) dx \quad (2)$$

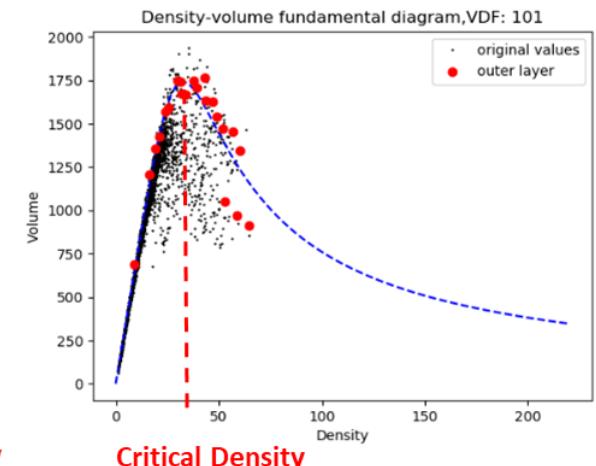
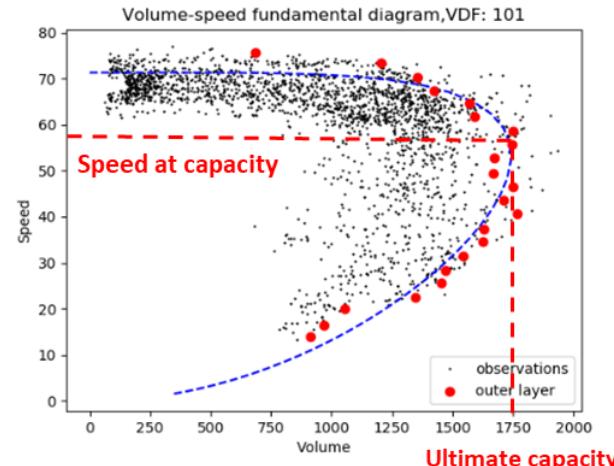
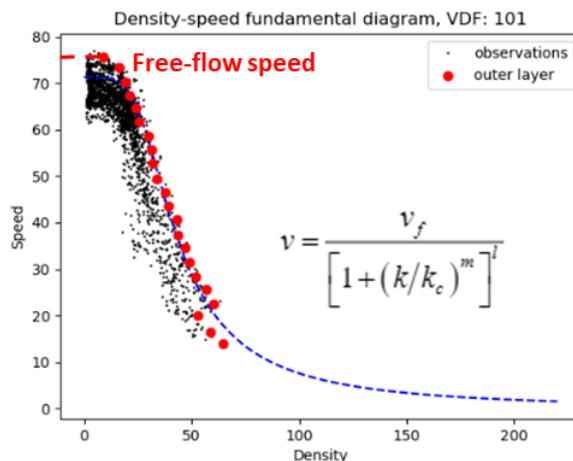
For the triangular-shaped flow-density relationship with constant forward and backward wave speeds, it is easy to verify that, when the speed of the forward wave is  $v_f$ , the general cumulative flow count updating formula reduces to  $-k + \frac{q}{v_f} = -k + k = 0$ . Under congested traffic conditions with a constant backward wave speed  $w_b$ , we have  $-k + \frac{q}{w_b} = -k_{jam}$ , and this equation can be rewritten as:

$$dN = \left( -k + \frac{q}{w_b} \right) dx = -k_{jam}(a) \times \text{length}(a) \times n\text{lanes}(a)$$



## B.5: Derivation of the Queued Demand

Supply side (traffic stream model calibration):



$$c = v_{\max} = \frac{k_c u_f}{2^{\frac{2}{m}}}$$

*Calibrated fundamental diagram , Freeway, CBD area*

[https://www.researchgate.net/publication/341104050\\_An\\_s-shaped\\_three-dimensional\\_S3\\_traffic\\_stream\\_model\\_with\\_consistent\\_car\\_following\\_relationship](https://www.researchgate.net/publication/341104050_An_s-shaped_three-dimensional_S3_traffic_stream_model_with_consistent_car_following_relationship)

## B.6: Dynamic Traffic Assignment (DTA)

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Existing simulation-based DTA models

- Integration (Vertical queue)
- DynaSMART (Link density function, vertical queue)
- DynaMIT (Link density function, moving queue)
- VISTA (Cell Transmission Model)
- Vissum (Car following)
- Transsims (Cellular Automata)
- Transmodeler (Car following)
- DTALite (Simplified kinematic wave model)

# B9: Describing Traffic Congestion Propagation using Simulation Models

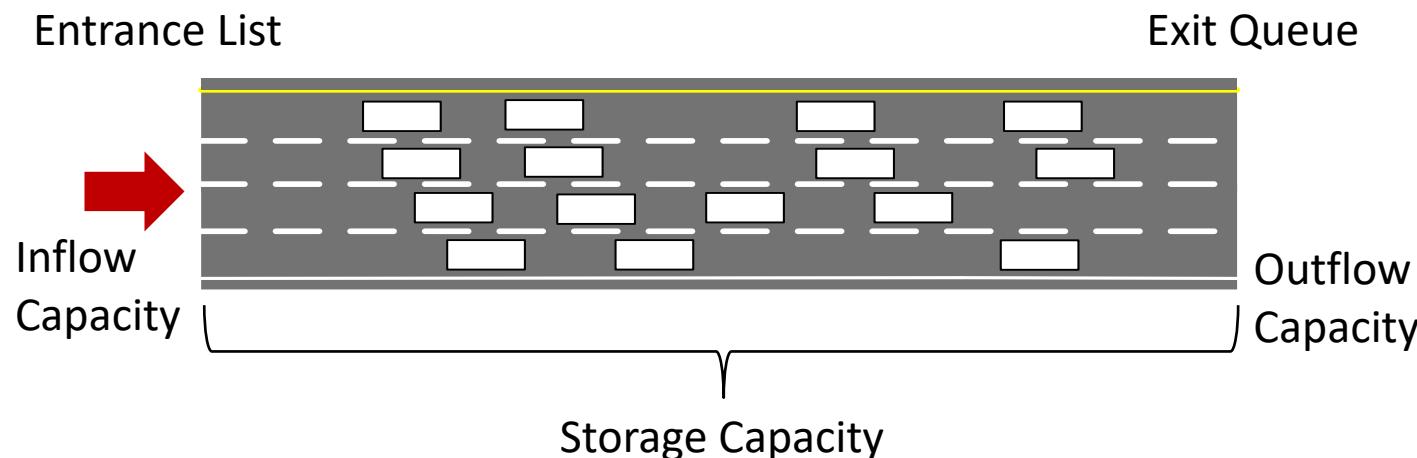
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## Newell's simplified kinematic wave model

Outflow capacity

Inflow capacity

Storage capacity

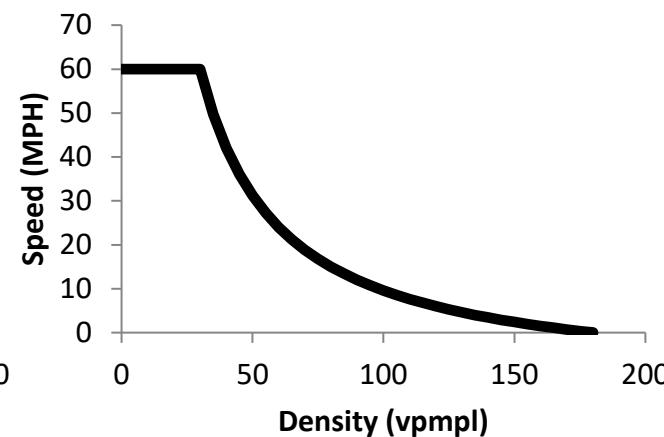
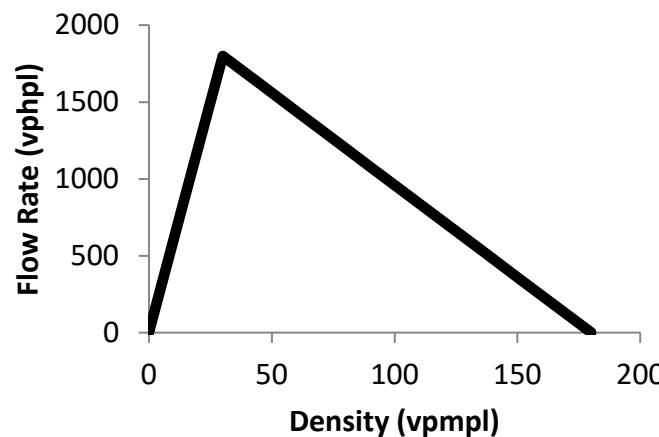


## B.10: Bottleneck-based Traffic Simulation Models

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### Newell's simplified kinematic wave model

- Triangular flow-density relationship
- Free flow speed, jam density, backward wave speed

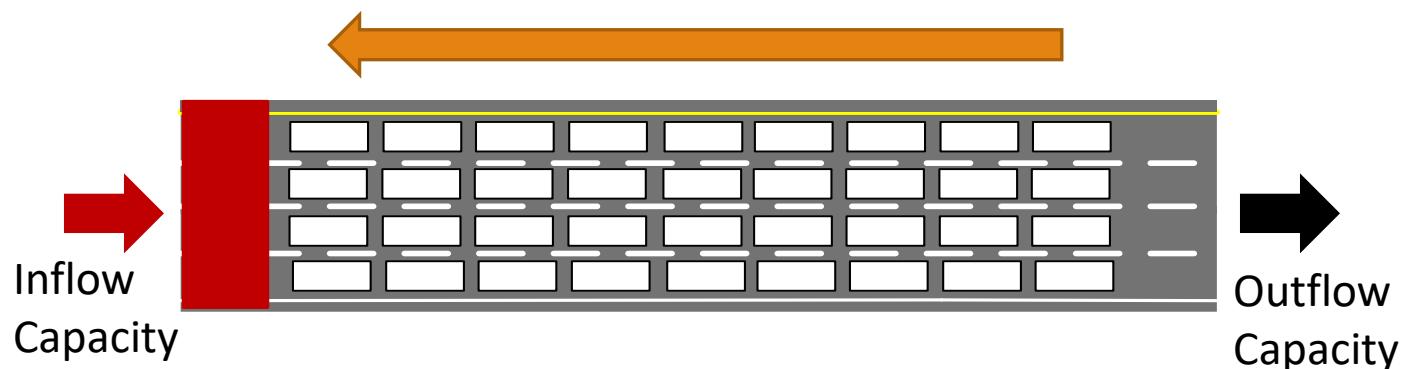


## B.11: Bottleneck-based Traffic Simulation Models

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### Queue propagation

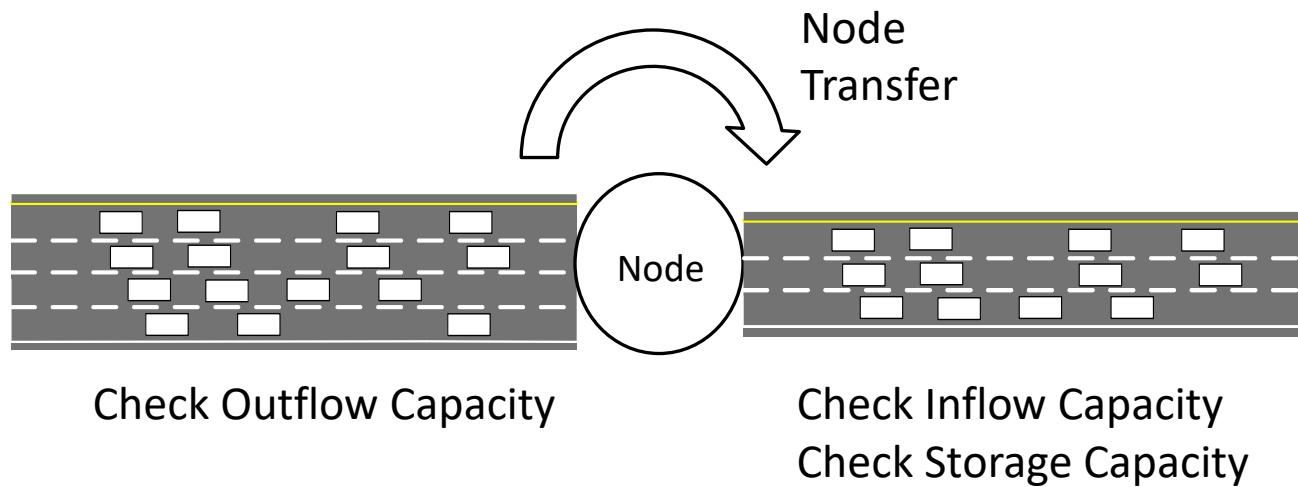
- Inflow capacity = outflow capacity



## B.12: Bottleneck-based Traffic Simulation Models

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### Node transfer

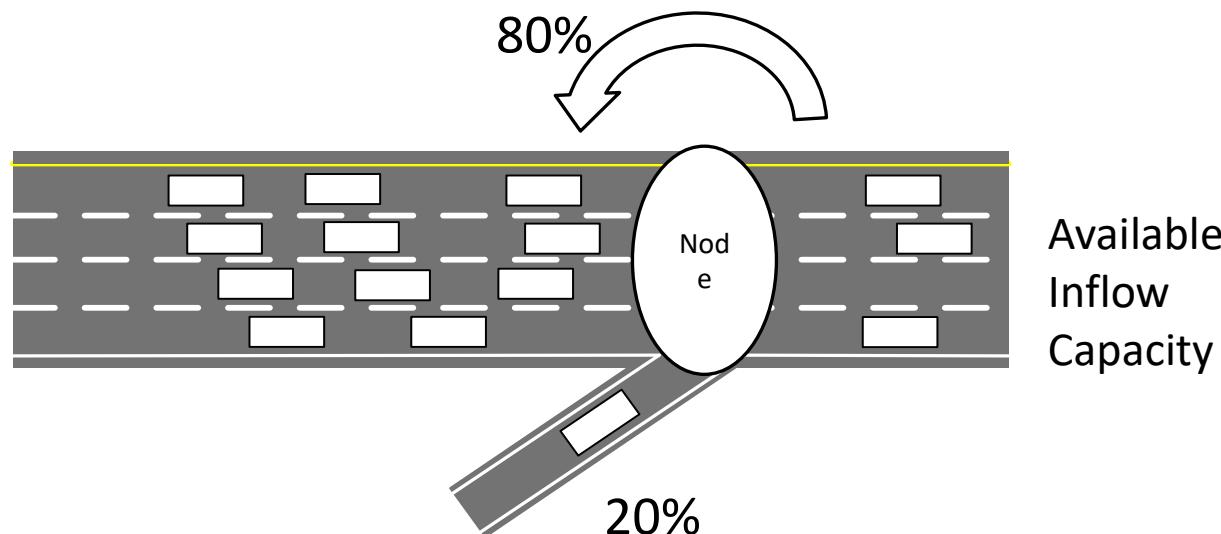


## B.13: Bottleneck-based Traffic Simulation Models

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### Merge Models

- Distribute inflow capacity to upstream links
- Lane & demand-based methods

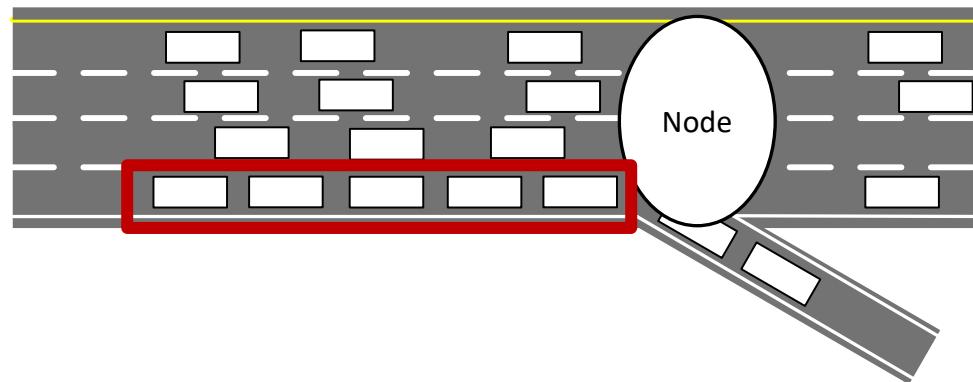


## B.14: Bottleneck-based Traffic Simulation Models

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### Diverge Models

- Different conditions by lane
- First-In-First-Out (FIFO) constraint
- Relaxation to prevent extreme bottlenecks

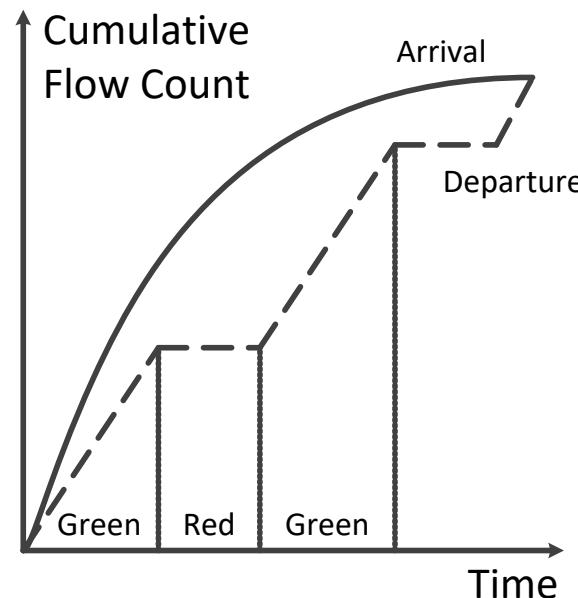


## B.15: Multiple Traffic Simulation Models

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### Signalized Intersections

- Effective green time, saturation flow rate, movement-based capacity
- Relaxed inflow constraints

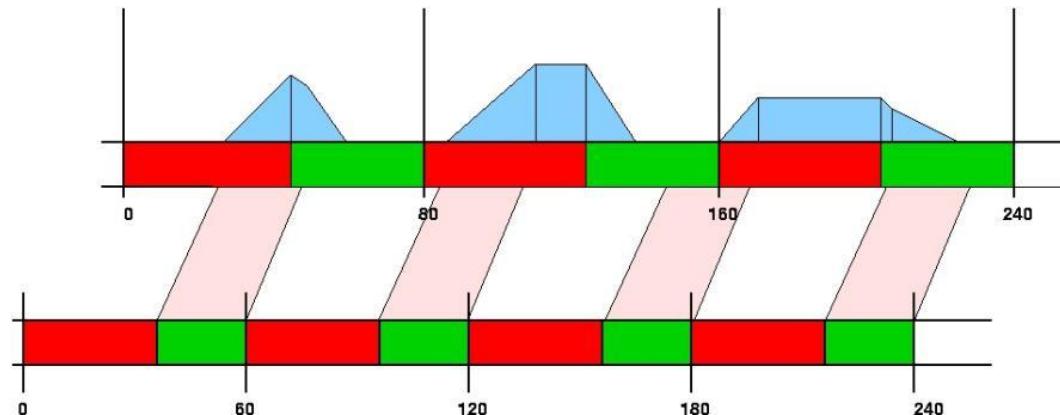


## B.16: Multiple Traffic Simulation Models

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### Signal Timing & Hourly Capacity

- Input: Average hourly capacity, cycle time, offset at node
- Output: Effective green time per cycle(capacity/saturation flow rates)



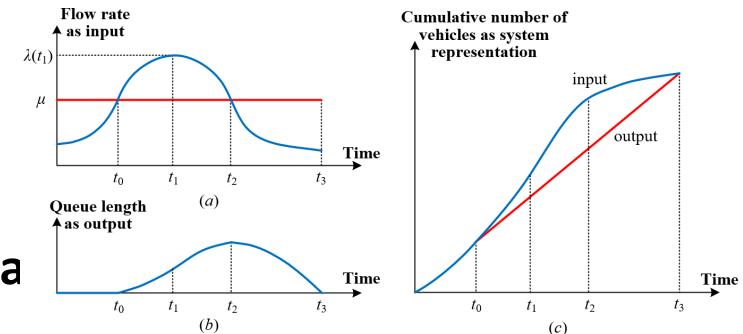
# C: Continuous-time Fluid based Queuing Model

## Typical Operations Questions about Traffic Dynamics:

How long will people have to wait to turn left from a driveway?

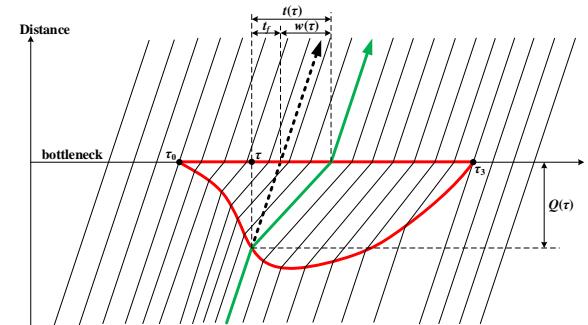
How much delay will people face at this signal?

What is the “capacity” or “queue-discharge rate” of a freeway?

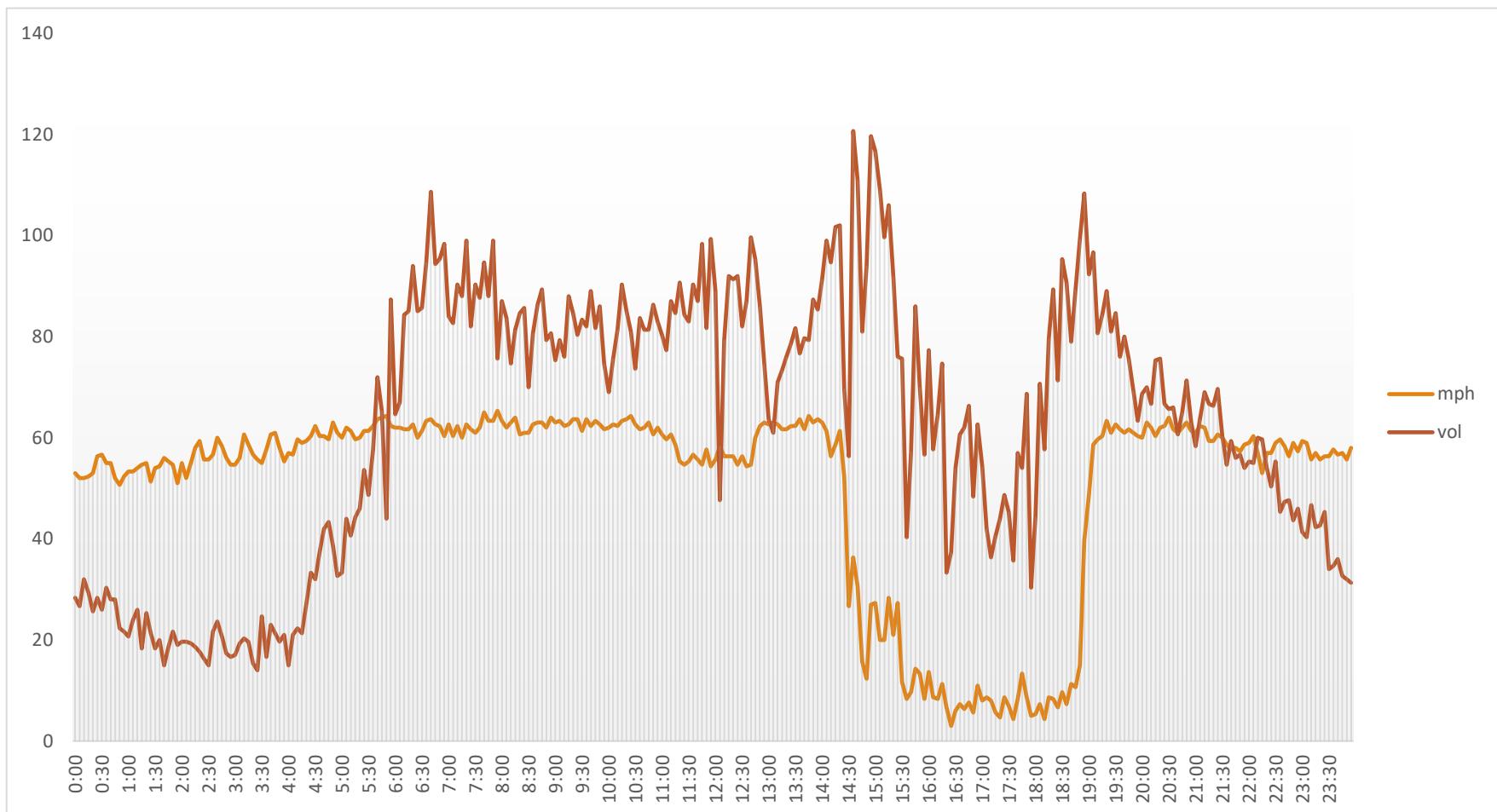


## Proposed Models: Fluid Queue /PAQ Polynomia Arrival Queue

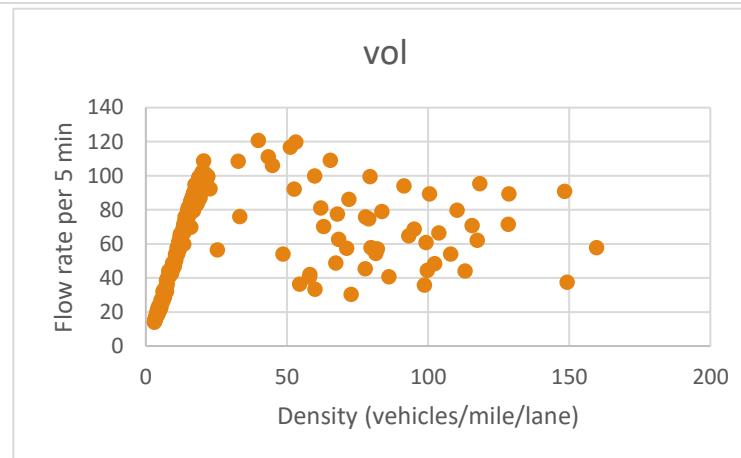
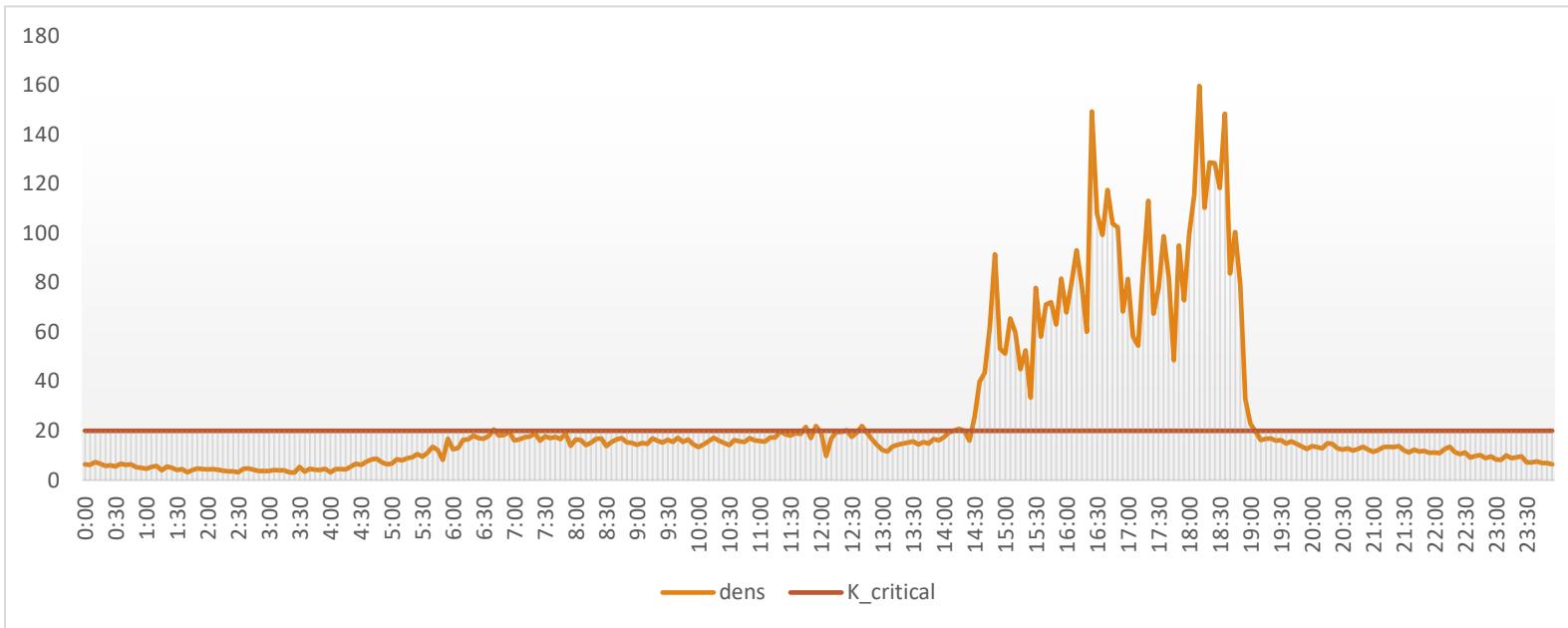
- Applications of Queueing Theory
- By G. F. Newell
- Chapter 2: Deterministic fluid approximation – single server



## C.1: Understand Volume vs. Speed Profile



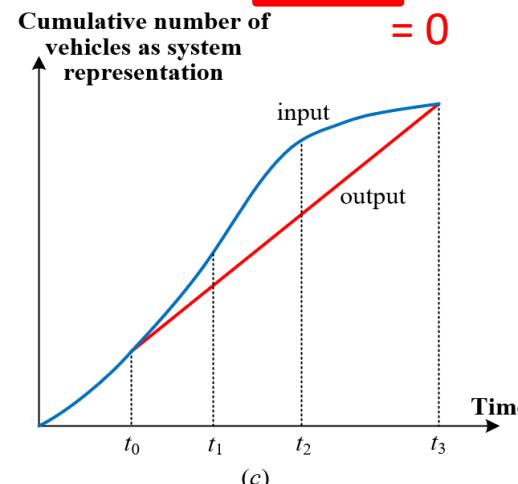
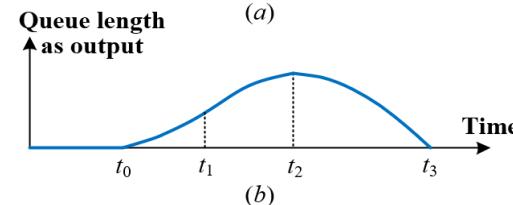
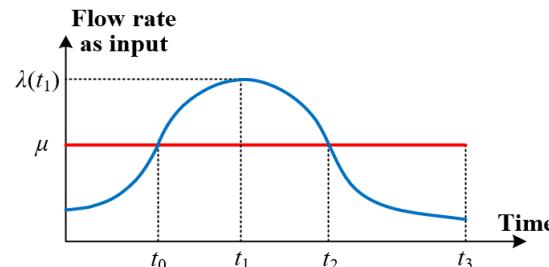
## C.2: Understand Density vs. Volume Profile



## C.3: Assumption

Newell (1982)

- Assumption: quadratic inflow rate  $\lambda(t) = \lambda(t_1) + \boxed{\lambda'(t_1)} \cdot (t - t_1) + \boxed{\frac{\lambda''(t_1)}{2}} (t - t_1)^2$



$\lambda(t)$ : inflow rate at time  $t$

$t_0$ : start time of congestion period

$t_1$ : time index with maximum inflow rate

$t_2$ : time index with maximum queue length

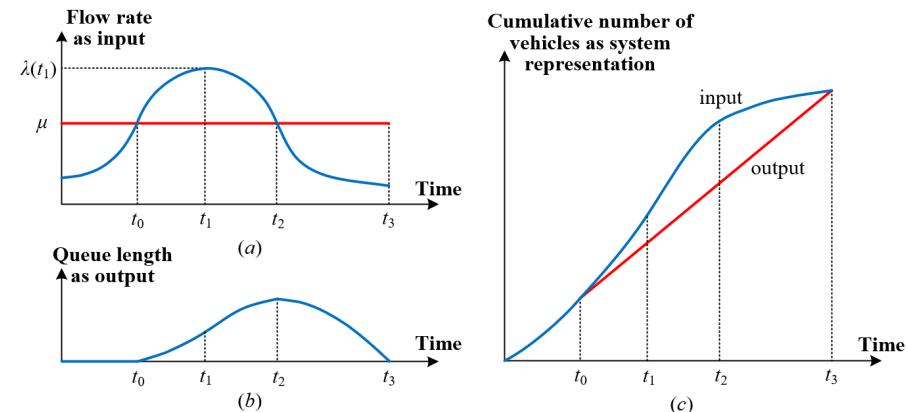
$t_3$ : end time of congestion period

$\mu$ : discharge rate (or capacity)

## C.4: Factored Form of Net Flow

Newell (1982)

- $\lambda(t) = \lambda(t_1) - \rho(t - t_1)^2$



$$\because \mu = \lambda(t_0) = \lambda(t_2) \quad \therefore \mu = \lambda(t_1) - \rho(t_0 - t_1)^2 = \lambda(t_1) - \rho(t_2 - t_1)^2$$

$$\therefore t_0 = t_1 - \left[ \frac{\lambda(t_1) - \mu}{\rho} \right]^{1/2} \quad t_2 = t_1 + \left[ \frac{\lambda(t_1) - \mu}{\rho} \right]^{1/2}$$

➤ Factored form of :  $\lambda(t) - \mu$        $\lambda(t) - \mu = \rho(t - t_0)(t_2 - t)$

## C.5: Closed Form of Queue Length(t) as boundary points of t0 and t3

Newell (1982)

- Queue length:  $A(t) - D(t)$

$$Q(t) = A(t) - D(t) = \int_{t_0}^t [\lambda(\tau) - \mu] d\tau$$

➤ Maximum queue length at time  $t_2$ :

$$= \int_{t_0}^t [\rho(\tau - t_0)(t_2 - \tau)] d\tau$$

$$= \rho(t - t_0)^2 \left[ \frac{t_2 - t_0}{2} - \frac{t - t_0}{3} \right]$$

$$Q(t) = \frac{\rho}{3}(t - t_0)^2(t_3 - t)$$

$$Q(t_2) = \frac{\rho}{6}(t_2 - t_0)^3 = \frac{4[\lambda(t_1) - \mu]^{3/2}}{3\rho^{1/2}}$$

➤ Queue dissipating time  $t_3$ :

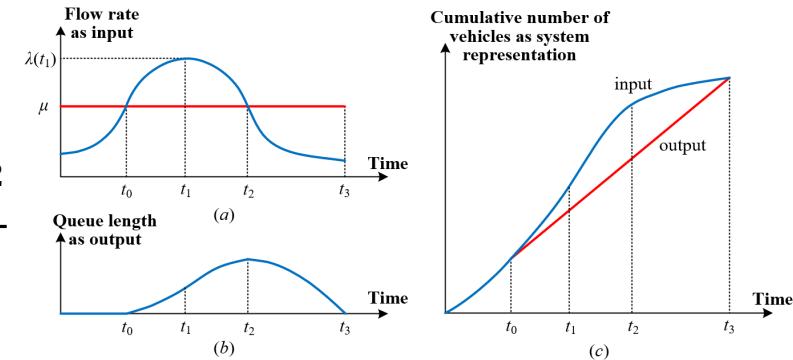
$$t_3 = t_0 + \frac{3}{2}(t_2 - t_0) = t_0 + 3(t_1 - t_0)$$

## C.6: Closed Form of Avg Waiting Time

Newell (1982)

- Total delay:  $W$

$$W = \int_{t_0}^{t_3} Q(\tau) d\tau = \frac{\rho}{36} (t_3 - t_0)^4 = \frac{9[\lambda(t_1) - \mu]^2}{4\rho}$$



➤ Average delay:

$$w = \frac{W}{d_3 - d_0} = \frac{\rho}{36} \cdot \frac{P^4}{D} = \frac{\rho}{36\mu} \cdot \left(\frac{D}{\mu}\right)^3$$

$$\mu = \frac{d_3 - d_0}{t_3 - t_0} = \frac{D}{P}$$

➤ Average travel time function:

$$t = t_f + w = t_f \left[ 1 + \frac{\rho}{36\mu \cdot t_f} \cdot \left( \frac{D}{\mu} \right)^3 \right]$$

Only two parameters:  $\rho, \mu$

## C.7: Recap of Problem Statement

Closed-form solution of delays

Key Parameters:

$\lambda(t)$ : inflow rate at time  $t$

$t_0$ : start time of congestion period

$t_1$ : time index with maximum inflow rate

$t_2$ : time index with maximum queue length

$t_3$ : end time of congestion period

$\mu$ : discharge rate (or capacity)

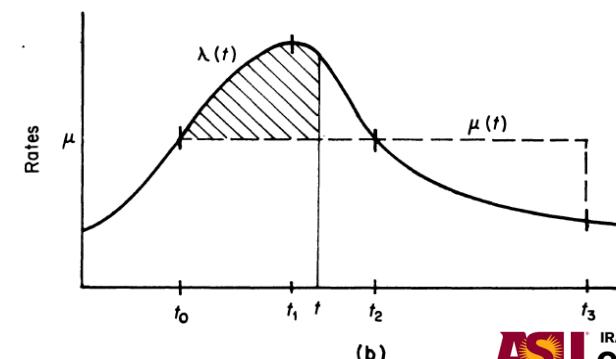
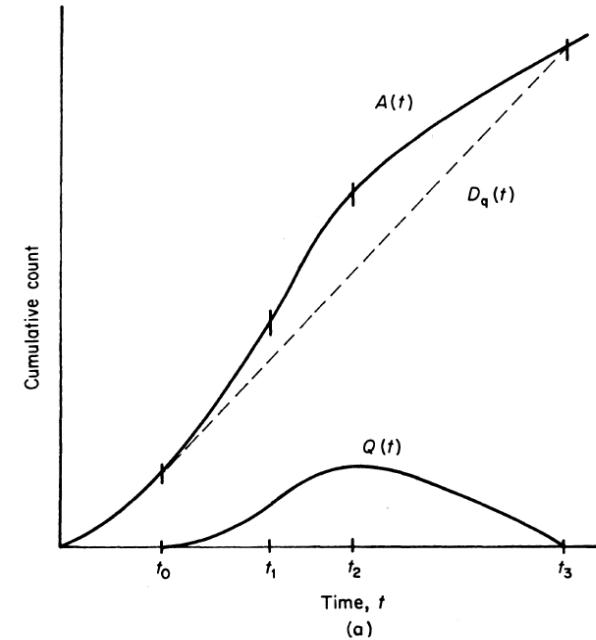


Figure 2.3 Graphical construction of queue (ARIZONA STATE UNIVERSITY)

## C.8: Understand Time-dependent Queue

The queue at any time  $t_0 < t < t_3$  is obtained by substitution of (2.4a) into (2.2),

$$Q(t) = \beta(t-t_0)^2 \left[ \frac{(t_2-t_0)}{2} - \frac{(t-t_0)}{3} \right]. \quad (2.5)$$

$$\begin{aligned}\lambda(t) - \mu &= \beta(t-t_0)(t_2-t) \\ &= \beta[-t^2 - t_0t_2 + (t_0+t_2)t]\end{aligned}$$

$$\begin{aligned}Q(t) &= \int_{t_0}^t [\lambda(\tau) - \mu] d\tau \\ &= \int_{t_0}^t [\beta(\tau-t_0)(t_2-\tau)] d\tau \\ &= -\beta \int_{t_0}^t \tau^2 d\tau + \beta(t_0+t_2) \int_{t_0}^t \tau d\tau \\ &= -\frac{\beta}{3} [t^3 - t_0^3] + \frac{1}{2} \beta(t_0+t_2)(t^2 - t_0^2)\end{aligned}$$

The length of the queue at time  $t$

$$Q(t) = A(t) - D_q(t) = \int_{t_0}^t [\lambda(\tau) - \mu(\tau)] d\tau \quad (2.2)$$


$$\begin{aligned}\mu &= \lambda(t_1) - \beta(t_0 - t_1)^2, \\ t_0 &= t_1 - \left[ \frac{\lambda(t_1) - \mu}{\beta} \right]^{1/2}, \\ t_2 &= t_1 + \left[ \frac{\lambda(t_1) - \mu}{\beta} \right]^{1/2}.\end{aligned} \quad (2.4)$$

It is convenient now to write  $\lambda(t) - \mu$  in the factored form

$$\lambda(t) - \mu = \beta(t-t_0)(t_2-t). \quad (2.4a)$$

## C.9: Average Delay $\bar{w}$

Average delay  $\bar{w}$

$$\bar{w} = \frac{W}{\mu(t_3 - t_0)} = \frac{1}{\mu} \cdot \frac{\beta}{36} (t_3 - t_0)^3 = \frac{9[\lambda(t_1) - \mu]^2}{4\beta\mu(t_3 - t_0)}$$

Average travel time calculation

$$\begin{aligned} tt &= t_f + \bar{w} \\ &= t_f + \frac{1}{\mu} \cdot \frac{\beta}{36} (t_3 - t_0)^3 \\ &= t_f + \frac{1}{\mu} \cdot \frac{\beta}{36} \left(\frac{D}{\mu}\right)^3 \\ &= t_f \left[ 1 + \frac{1}{t_f} \cdot \frac{1}{\mu} \cdot \frac{\beta}{36} \left(\frac{D}{\mu}\right)^3 \right] \end{aligned}$$

$$t_3 - t_0 = P$$

$$\mu P = D$$

$$(t_3 - t_0)^3$$

$$= \left(\frac{D}{\mu}\right)^3$$



Quadratic form ↳

$\lambda(t) = -\xi(t - t_0)(t - t_2) + \mu, \xi > 0 \Leftrightarrow$

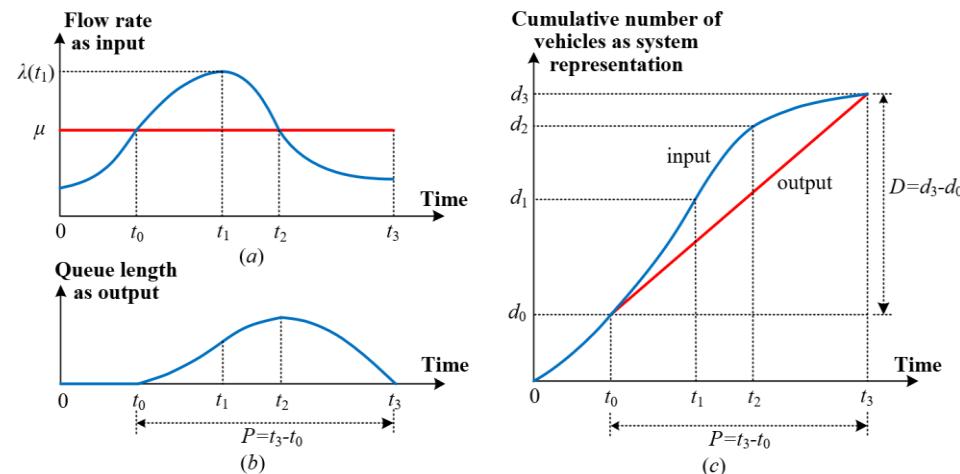
$$tt = t_f \cdot \left[ 1 + \frac{\xi}{36\mu \cdot t_f} \cdot \left(\frac{D}{\mu}\right)^3 \right] \Leftrightarrow$$

$$\begin{aligned} \lambda(t) - \mu &= -\beta(t - t_0)(t - t_2) \\ &= \beta(t - t_0)(t_2 - t) \end{aligned}$$

## C.10: Cubic Arrival Rate Function

$$\lambda(t) = \lambda(t_1) + \boxed{\lambda'(t_1)}(t - t_1) + \frac{1}{2}\boxed{\lambda''(t_1)}(t - t_1)^2 + \frac{1}{6}\boxed{\lambda'''(t_1)}(t - t_1)^3$$

$$= 0 \qquad \qquad \qquad = -\gamma_1 \qquad \qquad \qquad = -\gamma_2$$



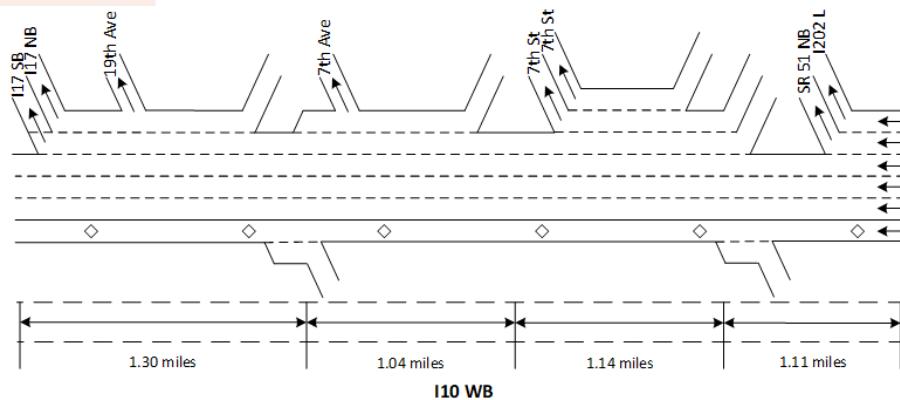
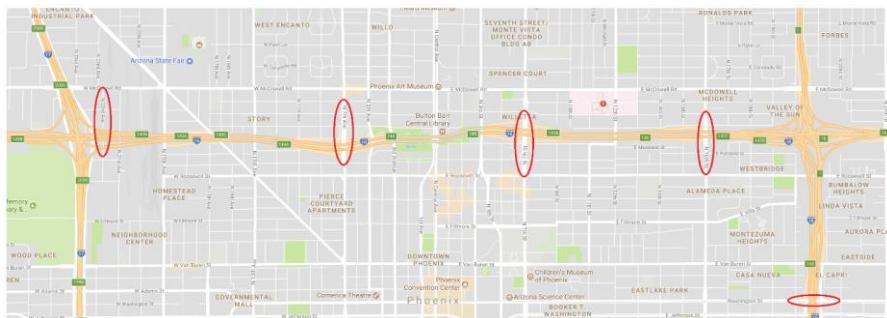
- $\lambda(t)$ : inflow rate at time  $t$
- $t_0$ : start time of congestion period
- $t_1$ : time index with maximum inflow rate
- $t_2$ : time index with maximum queue length
- $t_3$ : end time of congestion period
- $\mu$ : discharge rate (or capacity)

Cheng, Qixiu, Zhiyuan Liu, Jifu Guo, Xin Wu, Ram Pendyala, Baloka Belezamo, and Xuesong Simon Zhou  
 "Estimating key traffic state parameters through parsimonious spatial queue models." *Transportation Research Part C: Emerging Technologies* 137 (2022): 103596.

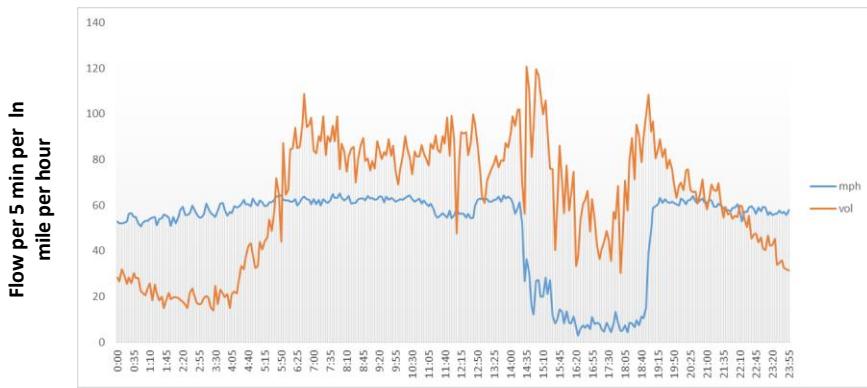
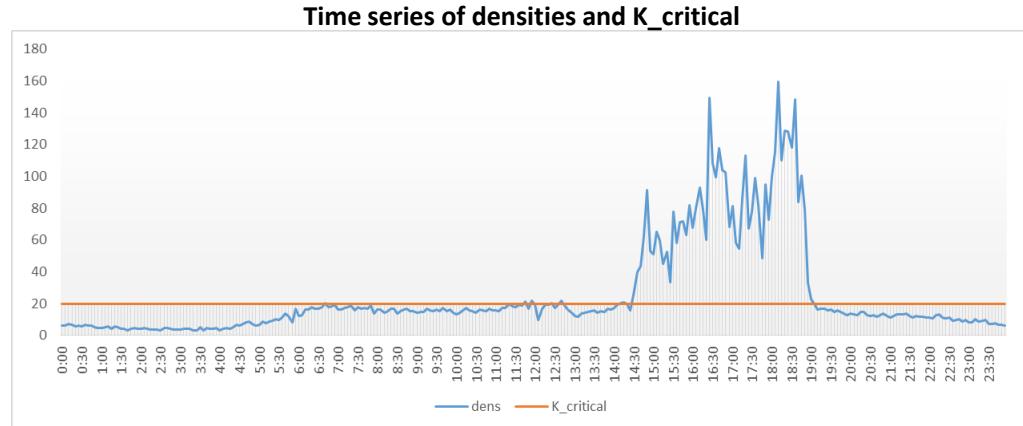
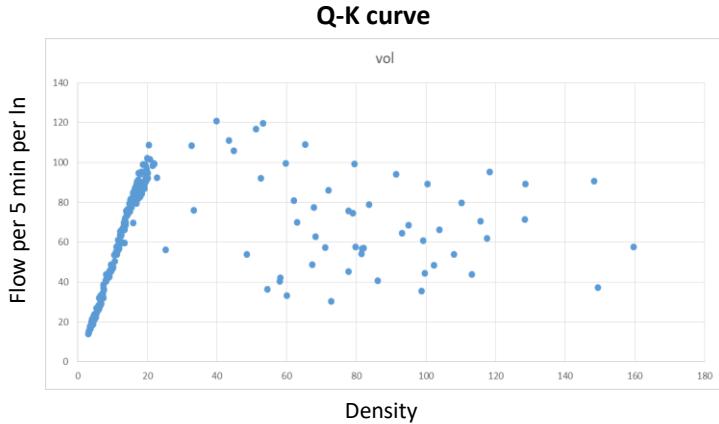
# Case study: PAQ Approximation in Phoenix

Test set: The traffic data are collected from I-10 freeway corridor (Jan 5<sup>th</sup> 2016)

Detector	Description
<b>Loop detectors</b>	<b>78, 84, 137, 139</b>
<b>Location</b>	I-10 freeway corridor, Westbound direction
<b>Traffic data collected</b>	Speed (mph) and volume (vehicle/5 minutes)
<b>Traffic data collection period</b>	Date: January 5 <sup>th</sup> , 2016 Interval: 5 minutes (from 00:00 to 23:55)



# Calibration results: sensor 137

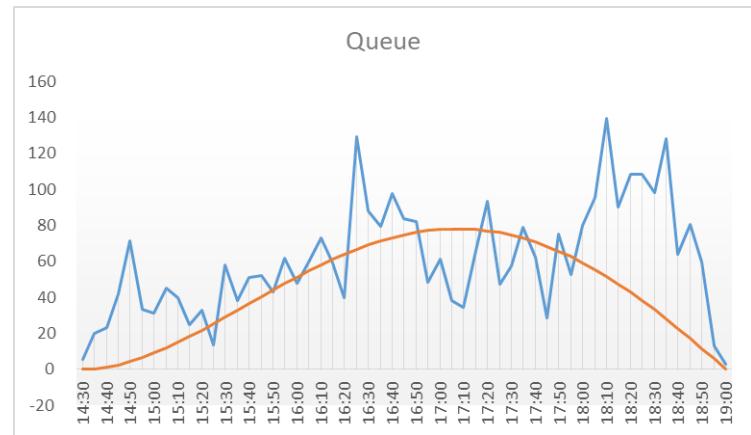
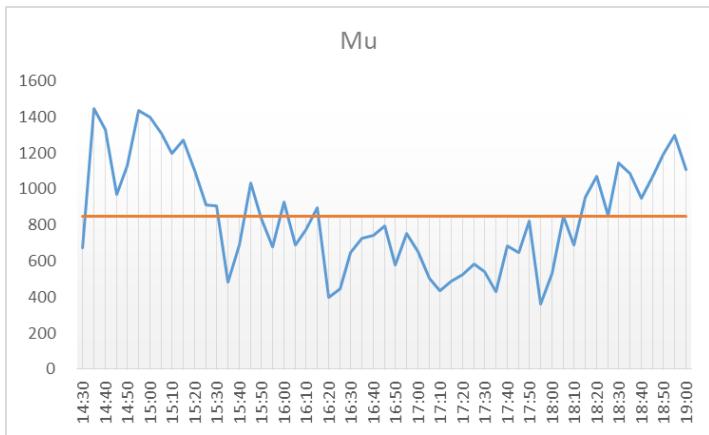


INPUT		
t0	14:30	14.50
t3	19:00	19.00
m	0.58	
t2	17:07	17.13
t_bar	20:30	20.50
n	3	
v_f	53	mile/hr
v_mu	13	mile/hr
delta_t	5	minutes

In our paper, this is a parameter to be calibrated. Here we treat it a constant which equals 7/12.  
 $t_2 = t_0 + m^*(t_3 - t_0)$   
 $t_{\bar{}} = t_0 + [3*(t_3 - t_0) - 4*(t_2 - t_0)]/(4 - 6*m)$

# of lanes immediately downstream the bottleneck

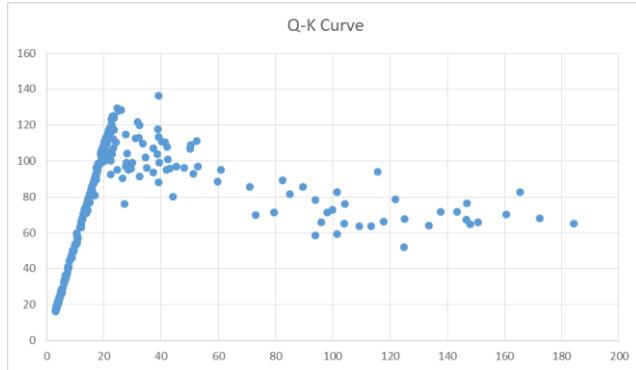
# Calibration results: sensor 137



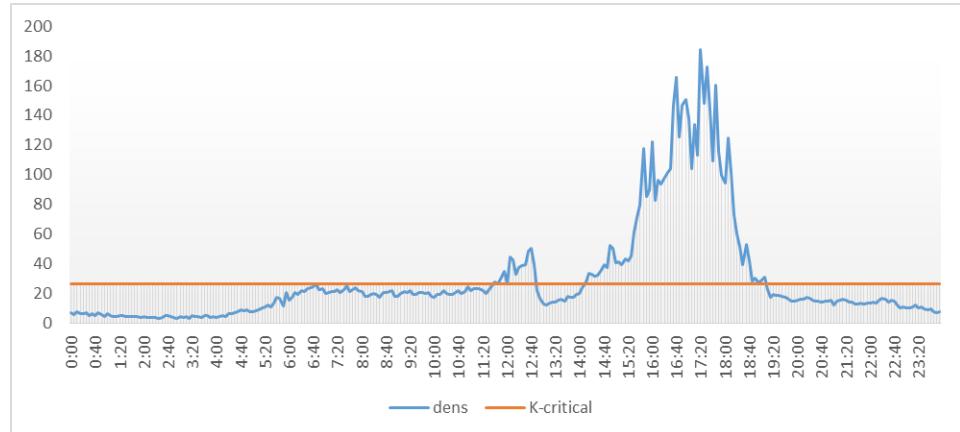
OUTPUT	
mu	849 veh/hr/ln
gamma	4.147966 veh/hr^4

# Calibration results: sensor 78

Flow per 5 min per ln

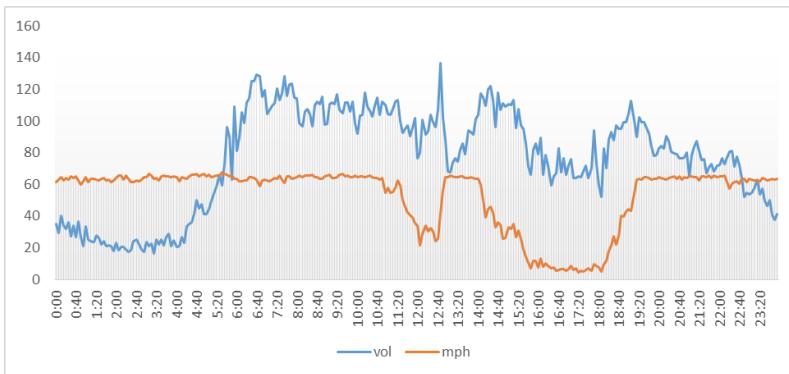


Time series of densities and K\_critical



Time series of speed and flow

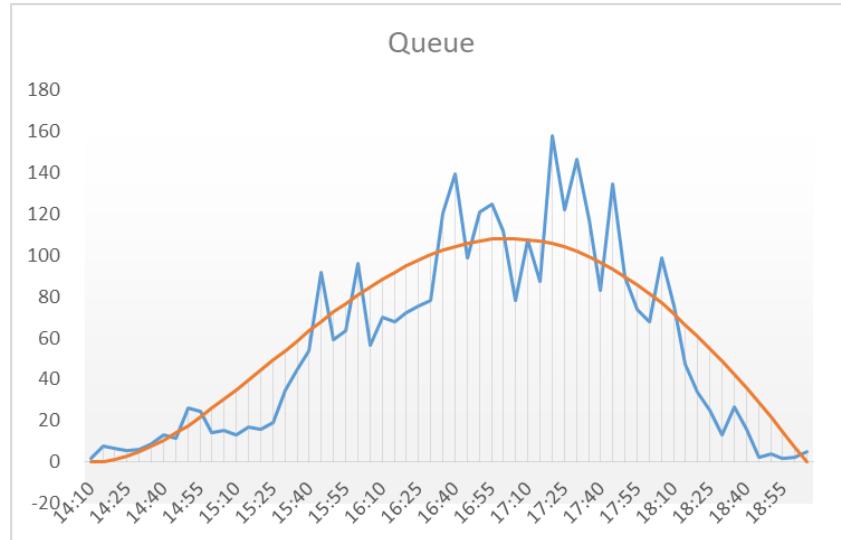
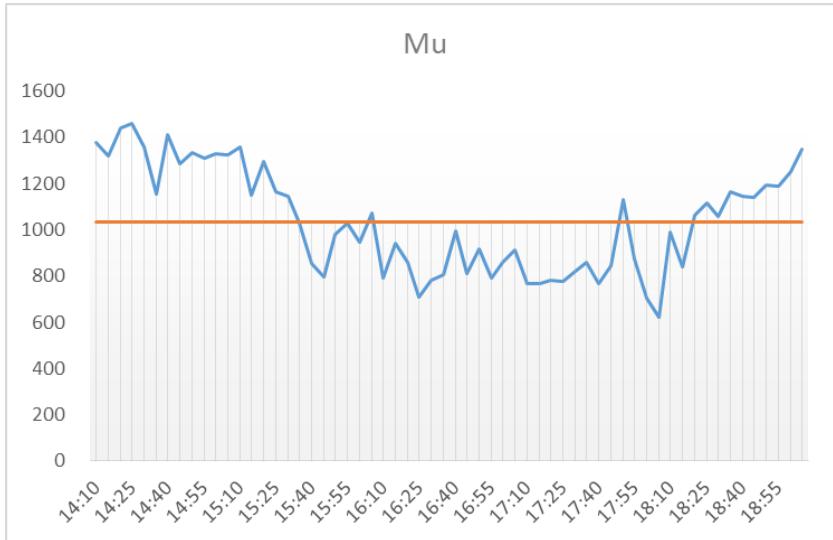
Flow per 5 min per ln  
mile per hour



INPUT		
t0	14:10	14.17
t3	19:05	19.08
m	0.58	
t2	17:02	17.03
t_bar	20:43	20.72
n	4	
v_f	53	mile/hr
v_mu	19	mile/hr
delta_t	5	minutes

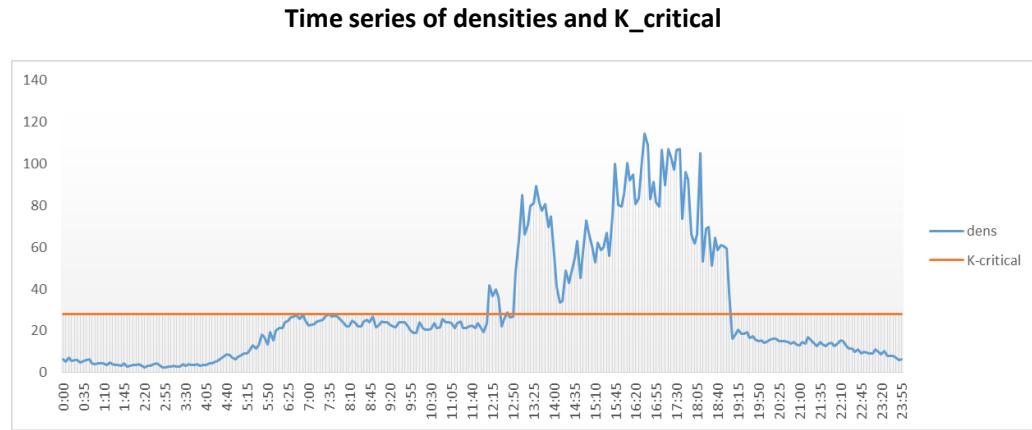
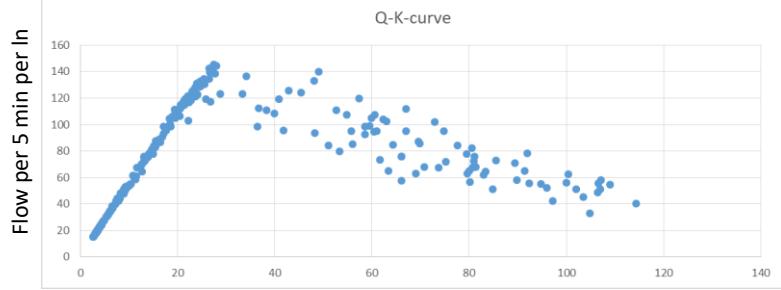
In our paper, this is a parameter to be calibrated. Here we treat it a constant which equals 7/12.  
 $t_2 = t_0 + m * (t_3 - t_0)$   
 $t_{\bar{}} = t_0 + [3 * (t_3 - t_0) - 4 * (t_2 - t_0)] / (4 - 6 * m)$   
# of lanes immediately downstream the bottleneck

# Calibration results: sensor 78

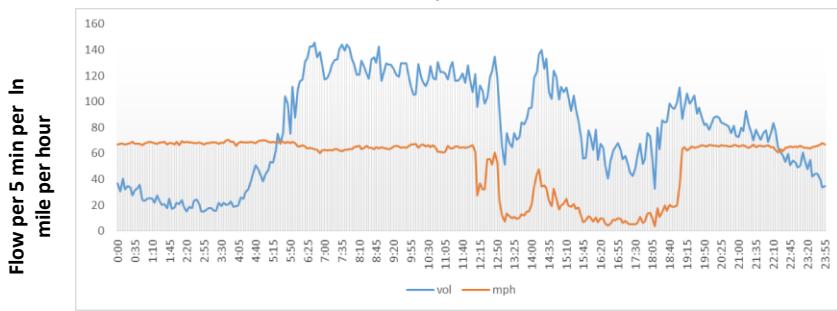


OUTPUT		
mu	1033	veh/hr/ln
gamma	3.42798	veh/hr <sup>4</sup>

# Calibration results: sensor 84



**Time series of speed and flow**

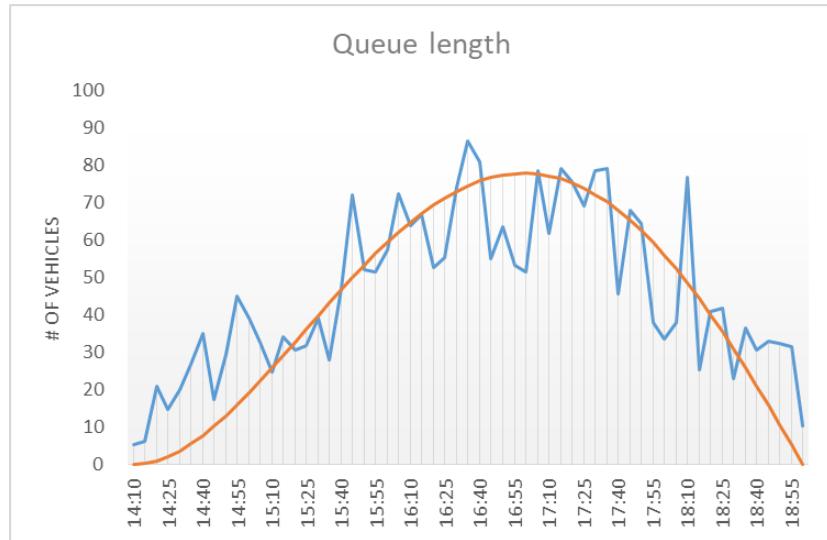
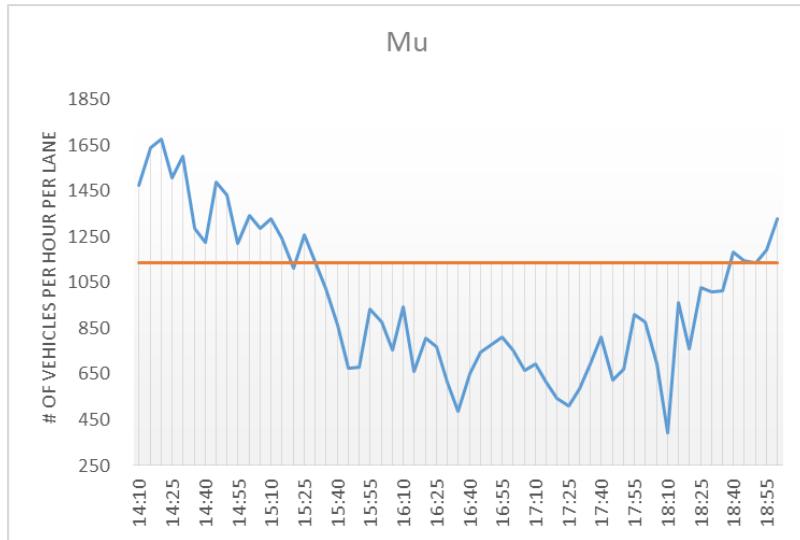


INPUT	t <sub>0</sub>	t <sub>3</sub>
t <sub>0</sub>	14:10	14.17
t <sub>3</sub>	19:00	19.00
m	0.58	
t <sub>2</sub>	16:59	16.99
t <sub>bar</sub>	20:36	20.61
n	4	
v <sub>f</sub>	53	mile/hr
v <sub>mu</sub>	16	mile/hr
delta_t	5	minutes

In our paper, this is a parameter to be calibrated. Here we treat it a constant which equals 7/12.  
 $t_2 = t_0 + m * (t_3 - t_0)$   
 $t_{bar} = t_0 + [3 * (t_3 - t_0) - 4 * (t_2 - t_0)] / (4 - 6 * m)$   
# of lanes immediately downstream the bottleneck

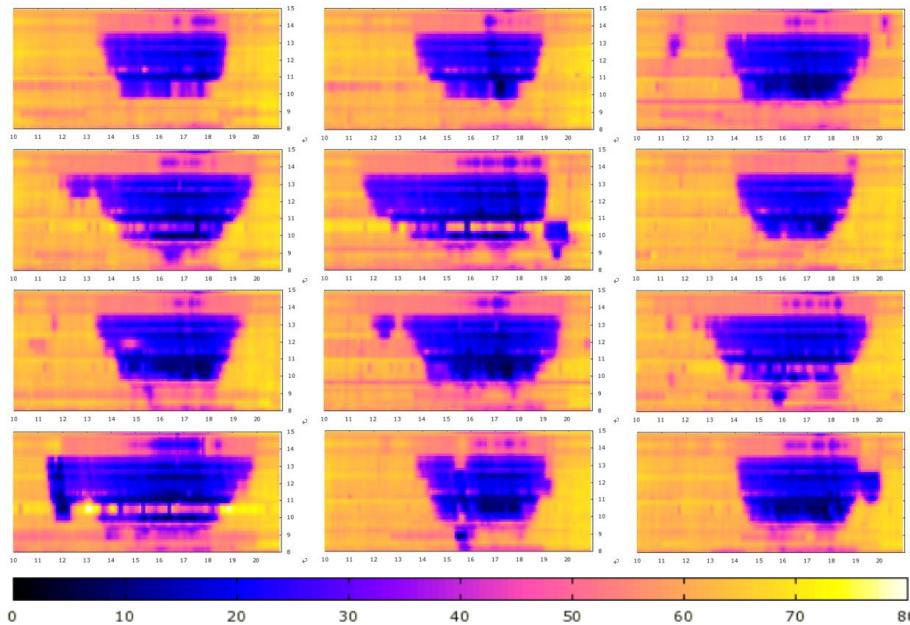


# Calibration results: sensor 84



OUTPUT	
mu	1133 veh/hr/ln
gamma	2.919345 veh/hr^4

# How to calibrate it with data from multiple days?



Speed evolution process from 10:00 am to 21:00 pm

Day $N$	$t_0$	$t_3$	$P$
Day 1	15:00	18:00	3 hour
Day 2	15:25	18:35	3.17 hour
Day 3	14:45	18:10	3.42 hour
Day 4	15:40	18:20	2.67 hour
Day 5	15:30	18:30	3 hour
Day 6	15:10	18:00	2.83 hour
...	...	...	...

# Summary: Polynomial inflow rate-based Queueing Model

A family of polynomial-function-based formulations

Arrival rate form	Arrival rate function	Average travel time function
Constant form	$\lambda(t) = \begin{cases} \pi_1 > \mu, t_0 \leq t < t_2 \\ \pi_2 < \mu, t_2 \leq t \leq t_3 \end{cases}$	$tt = t_f \cdot \left[ 1 + \frac{(\pi_1 - \mu)(\mu - \pi_2)}{2\mu(\pi_1 - \pi_2) \cdot t_f} \cdot \left( \frac{D}{\mu} \right) \right]$
Linear form	$\lambda(t) = -\kappa(t - t_2) + \mu, \kappa > 0$	$tt = t_f \cdot \left[ 1 + \frac{\kappa}{12\mu \cdot t_f} \cdot \left( \frac{D}{\mu} \right)^2 \right]$
Quadratic form	$\lambda(t) = -\rho(t - t_0)(t - t_2) + \mu, \rho > 0$	$tt = t_f \cdot \left[ 1 + \frac{\rho}{36\mu \cdot t_f} \cdot \left( \frac{D}{\mu} \right)^3 \right]$
Cubic form	$\lambda(t) = \gamma(t - t_0)(t - t_2)(t - \bar{t}) + \mu$	$tt = t_f \cdot \left[ 1 + \frac{\gamma \cdot g(m)}{\mu \cdot t_f} \cdot \left( \frac{D}{\mu} \right)^4 \right]$

Related to free flow travel time, capacity, demand, shape parameter, (m).

## D.1: Motivation for Queue-based Volume-Delay Function

---

### Functional requirements

- Capture queue building-up, propagation, dissipation
  - Key: how to detect queue spills back upstream link(s)
- Represent multiple origin-destination flow
  - Freeway merge and diverge
  - Arterial intersections (signal, stop signs, yield signs, no control)

### Evaluation Criteria

- Theoretically rigorous
- Numerically reliable
- Computationally efficient

## D.2: Volume-Delay Function (VDF)

### Typical volume-delay function

Literatures	Mathematical forms	$tt/t_0$ when $V = C$	$tt/t_0$ when $V = 0$
CATS (1960); Muranyi (1963)	$tt = t_0 \cdot 2^{(V/C)}$	2	1
Smock (1962, 1963)	$tt = t_0 \cdot \exp\left(\frac{V}{C} - 1\right)$	1	0.37
BPR (1964)	$tt = t_0 \cdot \left[1 + 0.15 \cdot \left(\frac{V}{C}\right)^4\right]$	1.15	1
Davidson (1966) <sup>1</sup>	$tt = t_0 \cdot \left(1 + \frac{J_D \cdot V}{C - V}\right), V < C$	$+\infty$	1
Akçelik (1978) <sup>2</sup>	$tt = \begin{cases} t_0 \cdot \left(1 + \frac{J_D \cdot V}{C - V}\right), & \text{if } V \leq mC, 0 < m < 1 \\ t_m + K_m(V - mC), & \text{if } V > mC, 0 < m < 1 \end{cases}$	Related to $J_D$ and $m$	1
Spiess (1990)	$tt = t_0 \cdot \left[2 - \beta - \alpha \left(1 - \frac{V}{C}\right) + \sqrt{\alpha^2 \left(1 - \frac{V}{C}\right)^2 + \beta^2}\right],$ where $\beta = \frac{2\alpha - 1}{2\alpha - 2}, \alpha > 1$	2	1
Tisato (1991) <sup>3</sup>	$tt = \begin{cases} t_0 \cdot \frac{1 - mx}{1 - x}, & x \leq x_c \\ t_0 \cdot \frac{1 - mx_c}{1 - x_c} + 30T(x - x_c), & x > x_c \end{cases}$ where $x = \frac{V}{C}, x_c = 1 - \sqrt{\frac{t_0}{30T}(1-m)}$	Related to $T$ and $m$	1
Akçelik (1991)	$tt = t_0 + 0.25T \cdot \left(\frac{V}{C} - 1 + \sqrt{\left(\frac{V}{C} - 1\right)^2 + \frac{8J_A V}{TC^2}}\right)$	Related to $J_A$ and $T$	1

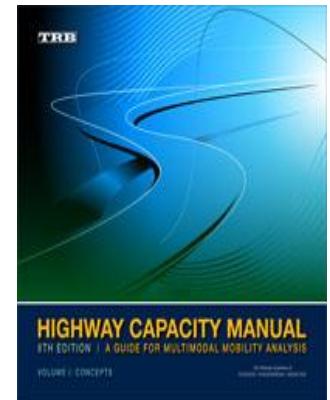
$tt$ : travel time

$t_0$ : free flow travel time

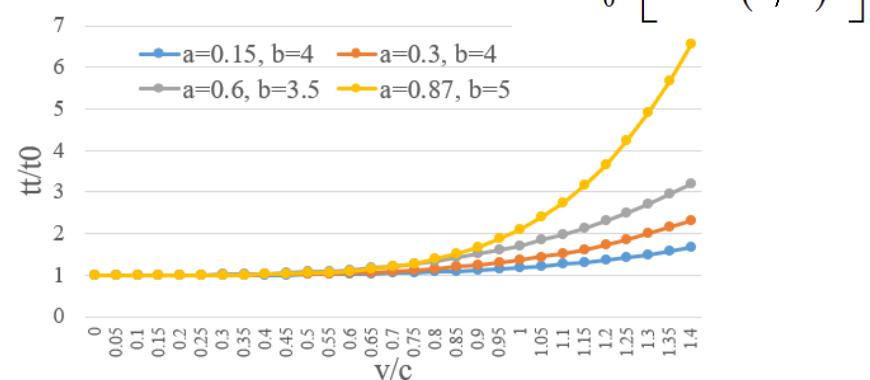
$V$ : traffic volume

$C$ : capacity

$J_A, JD, \alpha, \beta$ : parameters



BPR Function



$$\log(tt - t_0) = \log t_0 + \log a + b \log(v/c)$$

# Background and introduction

## BPR volume-delay function

Speed Limit (mph)	Practical Capacity (vehicle per hour)	Model Parameters	
		a	b
0 - 30	0 - 240	0.7312	3.6596
0 - 30	249 - 499	0.6128	3.5038
0 - 30	500 - 749	0.8774	4.4613
0 - 30	750 - 999	0.6846	5.1644
0 - 30	1000+	1.1465	4.4239
31 - 40	250 - 499	0.6190	3.6544
31 - 40	500 - 749	0.6662	4.9432
31 - 40	750 - 999	0.6222	5.1409
31 - 40	1000+	1.0300	5.5226
41 - 50	500 - 749	0.6609	5.0906
41 - 50	750 - 999	0.5423	5.7894
41 - 50	1000+	1.0091	6.5856
50+	500 - 749	0.8776	4.9287
50+	750 - 999	0.7699	5.3443
50+	1000+	1.1491	6.8677

Reference: Mannering et al. (1990)

Facility Type	Free-Flow Speed	"a"	"b"
6-Lane Freeway	70 mph	0.88	9.8
	60 mph	0.83	5.5
	50 mph	0.56	3.6
4-Lane Multilane Highway	70 mph	1.00	5.4
	60 mph	0.83	2.7
	50 mph	0.71	2.1

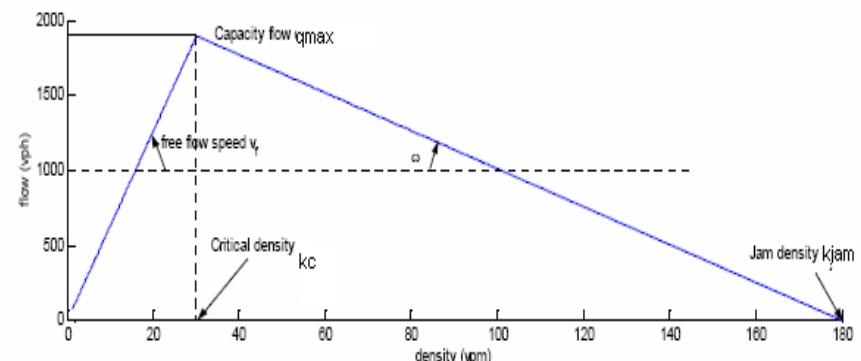
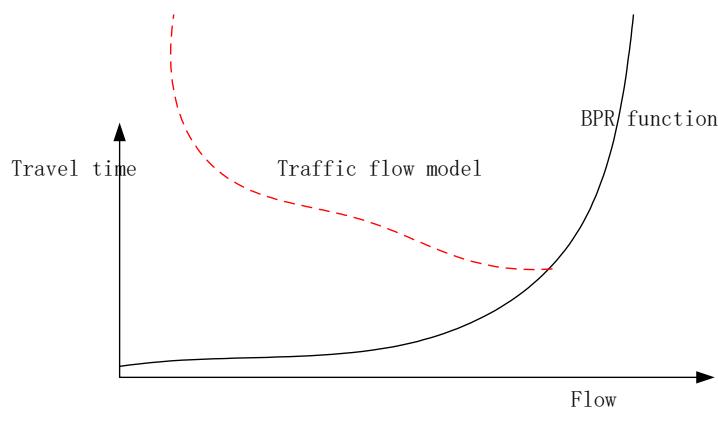


Both for highway and urban road.  
Related to speed limit and practical capacity.

## D.3: Volume-Delay Function (VDF)

BPR volume-delay function

- Widely used in static traffic assignment
- Fail to capture **queue evolution**, building-up, propagation, dissipation
- Fail to represent a **bottleneck** with low flow but high travel time



# Recap

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## Static traffic assignment

- computational efficiency
- easy for implementation
- cannot capture traffic flow dynamics
- cannot describe queue evolution
- cannot be used for oversaturated traffic conditions
- volume over capacity or demand over capacity?
- the used values of alpha and capacity are on the high side

$$t = t_0 \left[ 1 + 0.15 \cdot (V/C)^4 \right]$$

## Dynamic traffic assignment

- capture traffic flow dynamics
- describe queue evolution
- point queue or physical queue
- CTM- or LTM-based model
- high input-detail burden
- computational inefficiency

## D.4: Discussion in TMIP: STA vs DTA



Thu, 03/28/2019 - 4:58pm

HOME FORUM WE #6

John Gibb

Norman's article points out the absurdity of volume greater than capacity, but it's a little more tolerable to consider it demand rather than volume per se. Of course, we know that's not what actually flows through or downstream of a bottleneck; the rest of the problems are well known.

Three different impetuses for some form or another of DTA are:

(1) The well-known theoretical problems of static assignment,

(2) To analyze operations and design alternatives not adequately represented by link-level capacities and delay functions,

(3) The limitations of microsimulation without time-dependent traffic-aware path choice.

Planning agencies in mature, built-out areas with operations-oriented project lists rather than freeway plans (e.g. San Francisco) have responded the most to impetus 2 with fairly detailed large-area DTA models. Analysts working with microsimulation as their starting point can say what they will about impetus 3.

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- "... I argue that static assignment should not be used in **freeway capacity planning** studies - either separately or in combination with microsimulation..."
- "... it doesn't work to extract **over-capacity** subarea trip tables and then input them into **capacity-constrained DTA or microsimulation** ..."
- "**Is there a non-microsimulation DTA that's better than STA?** (Some DTAs have theoretical problems of their own, e.g. unreasonable FIFO violation.)"
- "**Is it good enough to be a replacement for STA as part of a full-region demand-equilibrium model?**"
- "**Since all models are wrong but some are useful**, what's *better* or *good enough* depends; more detail is not always better."
- "... but a **high input-detail burden** upon humans under limited budgets and schedules can make such a model less useful and more wrong than a simpler model, particularly for future scenario evaluation..."

## D.5: Discussion in TMIP: STA vs DTA



Thu, 03/28/2019 - 4:58pm

HOME FORUM WE #6

John Gibb

Norman's article points out the absurdity of volume greater than capacity, but it's a little more tolerable to consider it demand rather than volume per se. Of course, we know that's not what actually flows through or downstream of a bottleneck; the rest of the problems are well known.

Impetus 1 does not in itself require microsimulation, nor even HCM-style intersection analysis. Norman's study used DTAlite, a kinematic-wave DTA with link capacities. Two separate questions that can follow are:

(1) Is there a non-microsimulation DTA that's better than STA? (Some DTAs have theoretical problems of their own, e.g. unreasonable FIFO violation.)

(2) Is it good enough to be a replacement for STA as part of a full-region demand-equilibrium model?

Since "all models are wrong but some are useful," what's "better" or "good enough" depends; more detail is not always better. A model can be utterly faithful to the effects of every geometric and operational detail scientifically examined, but a high input-detail burden upon humans under limited budgets and schedules can make such a model less useful and more wrong than a simpler model, particularly for future scenario evaluation.

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### Our perspectives:

- Bottleneck discharge rate
- Highway capacity
- Queue evolution:  
Deterministic fluid  
approximation vs stochastic  
m/m/1
- Queue-oriented travel time  
performance function

## D.6: well-behaved VDF requirements by Spiess

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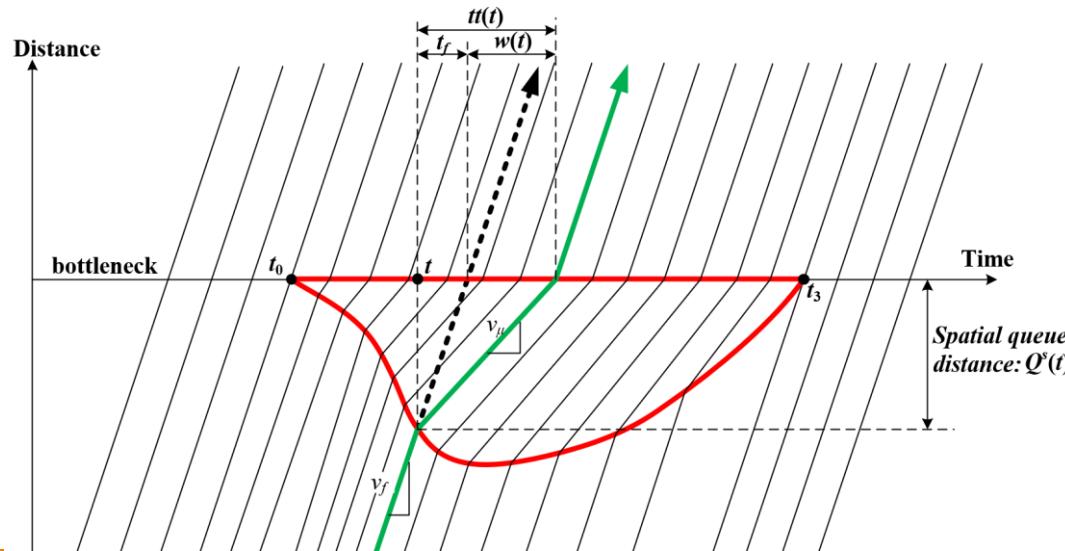
Consider a link delay function  $f(x) = \alpha x^\beta$ , and revisit the basic well-behaved VDF requirements proposed by **Spiess (1990)** as follows

1.  $f(x)$  should be strictly increasing for a feasible range of  $x$ . This is the necessary condition for the traffic assignment to converge to a unique solution.
2.  $f(0) = 0$  and  $f(1) = \text{constant } \alpha$ . The conditions ensure compatibility with the BPR form.
3.  $f'(x) = \alpha\beta x^{\beta-1}$  exists and is strictly increasing. This ensures the convexity of the congestion function, which is important when calculating marginal costs in system optimum assignment.
4.  $f'(1) = \alpha\beta$ , and the exponent  $\beta$  indicates how sudden the congestion effects change under the oversaturated condition.
5.  $f'(x) < M\beta$ , where  $M$  is a positive constant. The steepness of the congestion curve is limited so that the derived average delay is reasonable.
6.  $f'(0) \geq 0$ , which guarantees the uniqueness of the link volumes.
7. The evaluation of  $f(x)$  should be computationally efficient.

## D.7: Requirements for a well-behaved queue-based volume-delay function

Based on our definition of the inflow D/C ratio, we extend the basic requirements proposed by Spiess (1990) to develop a well-behaved QVDF. The additional requirements are listed as follows:

1. [Consistency of congestion demand] The inflow demand  $D$  should be consistent with the congestion duration  $P$  and discharge rate  $\mu$  in the queueing dynamics, that is,  $D = P\mu$ .
2. [Consistency of time-averaged delay] The average speed within the congestion duration should be consistent with the time-averaged delay.
3. [Consistency between macroscopic parameters and mesoscopic parameters]
4. [First-In-First-Out] The time-dependent travel time in the underlying mesoscopic vehicular fluid model should satisfy the FIFO property and capacity constraints.



$$\text{travel time in queue} = \frac{Q^s(t)}{v_\mu}$$

$$= t_f + w(t) = \frac{Q^s(t)}{v_f} + w(t)$$

$$w(t) = \frac{Q^v(t)}{\mu}$$

$$Q^p(t) = \frac{Q^v(t)}{1 - \frac{v_\mu}{v_f}}$$

## D.8: 2 Assumptions for linking D/C to lowest speed

- Calibration method: a recent queue-based volume delay function by Zhou et al., (2022)

Basic assumptions

Assumption 1: Link congestion duration and D/C ratio

$$P = f_d \cdot \left( \frac{D}{C} \right)^n$$

P: congestion duration

fd: congestion duration constant

D: total inflow demand

C: ultimate capacity

n: oversaturation-to-duration elasticity factor

Assumption 2: Link magnitude of speed reduction and congestion duration

$$MSR = \frac{v_{co} - v_{t_2}}{v_{t_2}} \quad MSR = f_p \cdot (P)^s$$

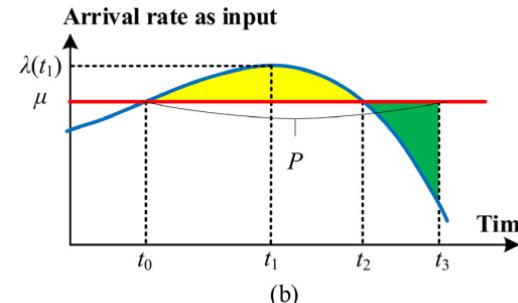
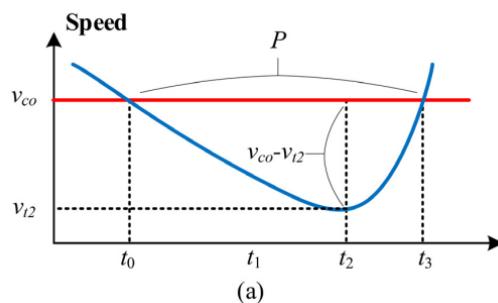
v\_co: cut-off speed

v\_t2: lowest speed in congestion duration

fp: MSR reduction constant

P: congestion duration

s: duration-to-speed reduction elasticity factor



Zhou, Xuesong, Qixiu Cheng, Xin Wu, Peiheng Li, Baloka Belezamo, Jiawei Lu, and Mohammad Abbasi. 2022. "A meso-to-macro cross-resolution performance approach for connecting polynomial arrival queue model to volume-delay function with inflow demand-to-capacity ratio." *Multimodal Transportation* 1 (2):100017

## D.9: Linking time-dependent queue length to time-dependent speed

Avg delay time-dependent delay

$$Q(t) = \frac{\gamma}{4} (t - t_0)^2 \cdot (t - t_3)^2$$

time-dependent queue

$$w(t) = \frac{\gamma}{4\mu} (t - t_0)^2 \cdot (t - t_3)^2$$

time-dependent delay

Generate time-dependent speed

$$v(t) = \frac{L}{\frac{L}{v_{co}} + w(t)}$$

time-dependent speed

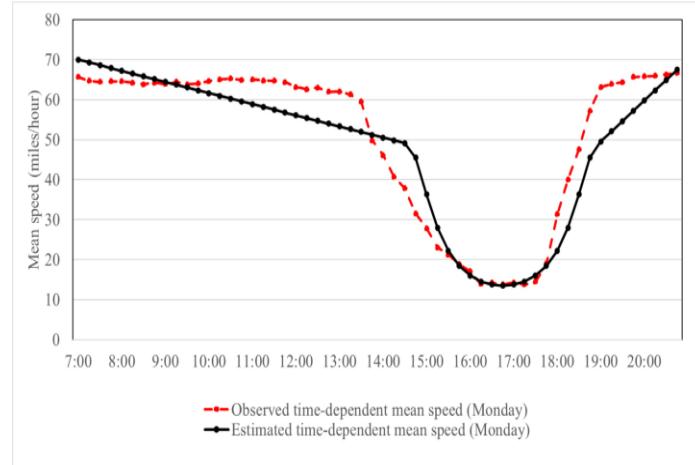


Fig. 14. Observed time-dependent mean speed and estimated speed at detector ID 87 on Monday.

## D.10: Linking avg delay to time-dependent speed

Generate average delay

$$\bar{w} = \frac{\gamma}{120\mu} \cdot P^4 = \frac{\gamma}{120\mu} \cdot \left(\frac{D}{\mu}\right)^4$$

Generate longest delay

$$w_{t_2} = \frac{\gamma}{4\mu} \cdot (t_2 - t_0)^2 \cdot (t_2 - t_3)^2 = \frac{\gamma}{4\mu} \cdot \left(\frac{P}{2}\right)^4 = \frac{\gamma}{64\mu} \cdot P^4$$

Conversion ratio

$$\theta = \frac{\bar{w}}{w_{t_2}} = \frac{64}{120} = \frac{8}{15}$$

Generate average speed

$$\begin{aligned} \bar{v} &= \frac{L}{\frac{L}{v_{co}} + \bar{w}} = \frac{L}{\frac{L}{v_{co}} + \theta w_{t_2}} = \frac{L}{\frac{L}{v_{co}} + \theta \cdot \left(\frac{L}{v_{t_2}} - \frac{L}{v_{co}}\right)} = \\ &= \frac{v_{co}}{1 + \theta \left(\frac{v_{co} - v_{t_2}}{v_{t_2}}\right)} = \frac{v_{co}}{1 + \theta \cdot [f_p \cdot (P)^s]} \\ &= \frac{v_{co}}{1 + \theta [f_p \cdot f_d^s \cdot \left(\frac{D}{C}\right)^{n \cdot s}]} = \frac{v_{co}}{1 + \alpha \left(\frac{D}{C}\right)^\beta} \end{aligned}$$

## Details: Connecting inflow curvature parameter and magnitude of speed reduction

we create a connection between  $w_{t_2}$  (i.e., the longest delay at  $t_2$  compared with  $v_{co}$ ) and  $P$

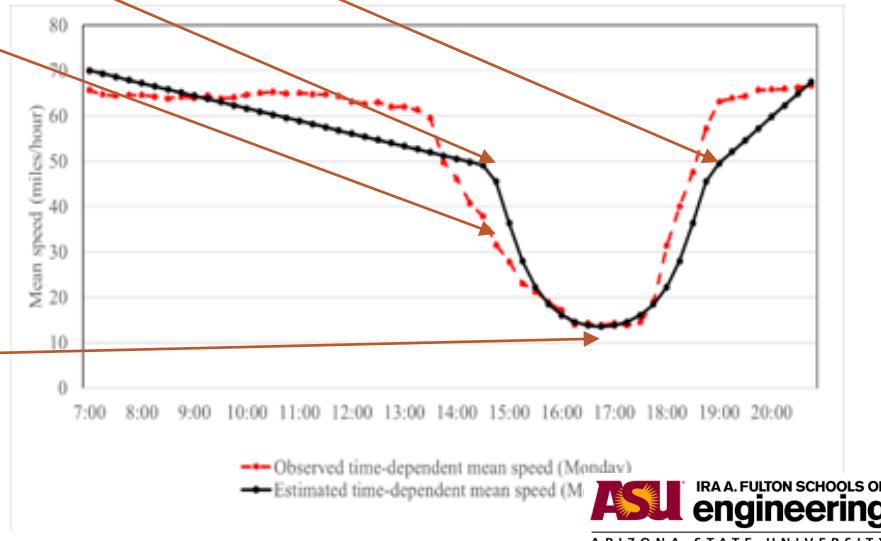
$$w_{t_2} = \frac{L}{v_{t_2}} - \frac{L}{v_{co}} = \frac{L}{v_{co}} \cdot \left( \frac{v_{co} - v_{t_2}}{v_{t_2}} \right) = \frac{L}{v_{co}} \cdot MSR = \frac{L}{v_{co}} \cdot f_p \cdot (P)^s$$

If we use the cubic PAQ model with  $m = 1/2$ , then we have

$$w(t) = \frac{\gamma}{4\mu} \cdot (t - t_0)^2 \cdot (t - t_3)^2.$$

By assuming  $\gamma > 0$  and  $m = 1/2$ , we have  $t_0 = t_2 - P$  and  $t_3 = t_2 + P$  according to **Cheng et al. (2022)**. Then the longest time-dependent delay at  $t_2$  becomes available as follows:

$$\begin{aligned} w_{t_2} &= \frac{\gamma}{4\mu} \cdot (t_2 - t_0)^2 \cdot (t_2 - t_3)^2 \\ &= \frac{\gamma}{4\mu} \cdot \left(\frac{P}{2}\right)^4 = \frac{\gamma}{64\mu} \cdot P^4. \end{aligned}$$



# Details: Time-dependent delay and average speed during congestion duration

If the cubic model is adopted with  $m = 0.5$ , we will have the following time-dependent queue and delay.

$$Q(t) = \frac{\gamma}{4} (t - t_0)^2 \cdot (t - t_3)^2$$

$$w(t) = \frac{\gamma}{4\mu} (t - t_0)^2 \cdot (t - t_3)^2$$

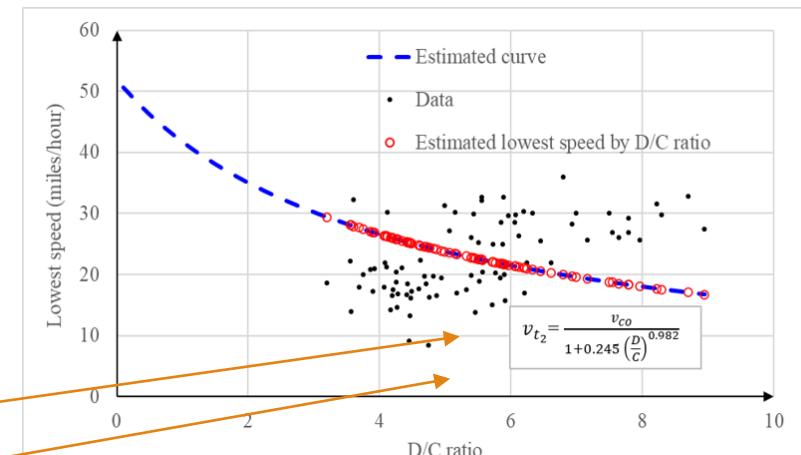
Time-dependent speed can be derived as follows:

$$v(t) = \frac{L}{\frac{L}{v_{co}} + w(t)}$$

With the cubic PAQ model and  $m = 0.5$ , the average delay during a congestion duration can be calculated by the following equation

$$\bar{w} = \frac{\gamma}{120\mu} \cdot P^4 = \frac{\gamma}{120\mu} \cdot \left(\frac{D}{\mu}\right)^4.$$

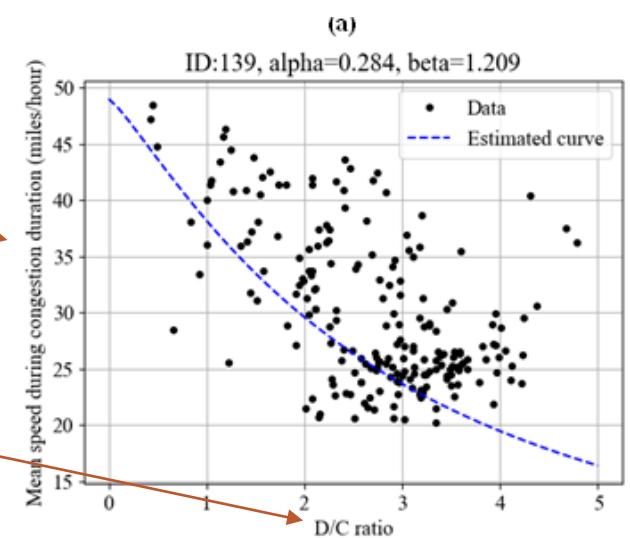
$$\theta = \frac{\bar{w}}{w_{t_2}} = \frac{64}{120} = \frac{8}{15}.$$



# Details: Time-dependent delay and average speed during congestion duration

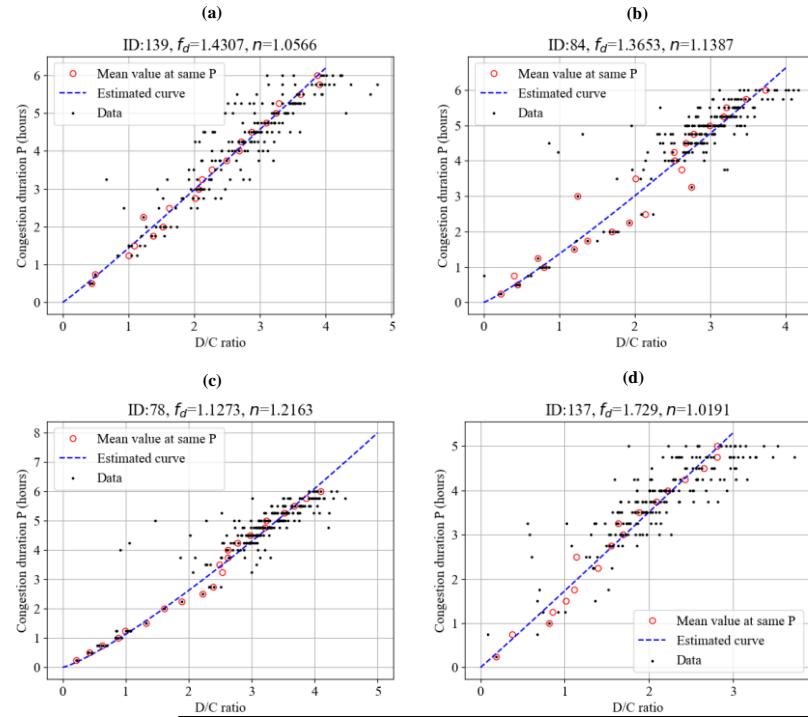
a BPR-like link performance function

$$\begin{aligned}
 \bar{v} &= \frac{L}{\frac{L}{v_{co}} + \bar{w}} = \frac{L}{\frac{L}{v_{co}} + \theta w_{t_2}} = \frac{L}{\frac{L}{v_{co}} + \theta \cdot \left( \frac{L}{v_{t_2}} - \frac{L}{v_{co}} \right)} = \\
 &= \frac{v_{co}}{1 + \theta \left( \frac{v_{co} - v_{t_2}}{v_{t_2}} \right)} = \frac{v_{co}}{1 + \theta \cdot [f_p \cdot (P)^s]} \\
 &= \frac{v_{co}}{1 + \theta \left[ f_p \cdot f_d^s \cdot \left( \frac{D}{C} \right)^{n \cdot s} \right]} = \frac{v_{co}}{1 + \alpha \left( \frac{D}{C} \right)^\beta}
 \end{aligned}$$



where  $\alpha = \theta f_p f_d^s$  and  $\beta = ns$ . An alternative form can be written as  $\bar{w} = \frac{L}{v_{co}} \cdot \alpha \left( \frac{D}{C} \right)^\beta$ .

# Case Study of Phoenix I-10 Corridor

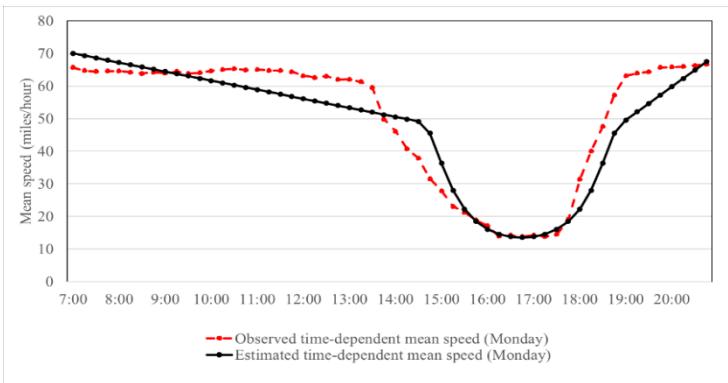


Calibration results of step 1 calibration that links D/C to congestion duration

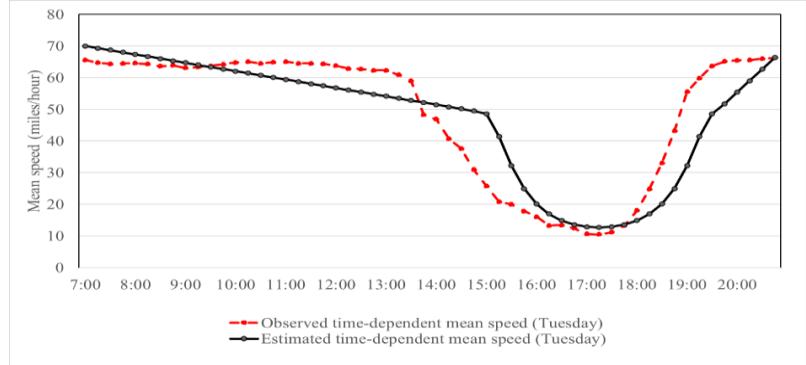
Detector ID	The first step calibration		The second step calibration,		Parameters for BPR-like average speed function	
	$f_d$	$n$	$f_p$	$s$	$\alpha$	$\beta$
139	1.43	1.06	0.35	1.14	0.28	1.21
84	1.37	1.14	0.23	1.64	0.21	1.87
78	1.13	1.22	0.11	2.10	0.07	2.55
137	1.73	1.02	0.21	1.71	0.29	1.75



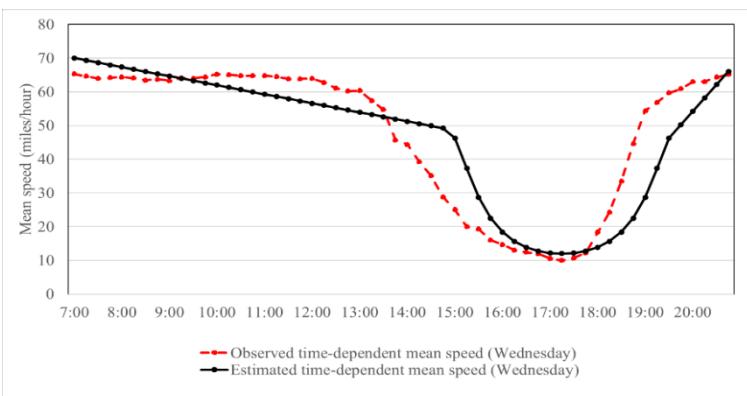
# Case Study



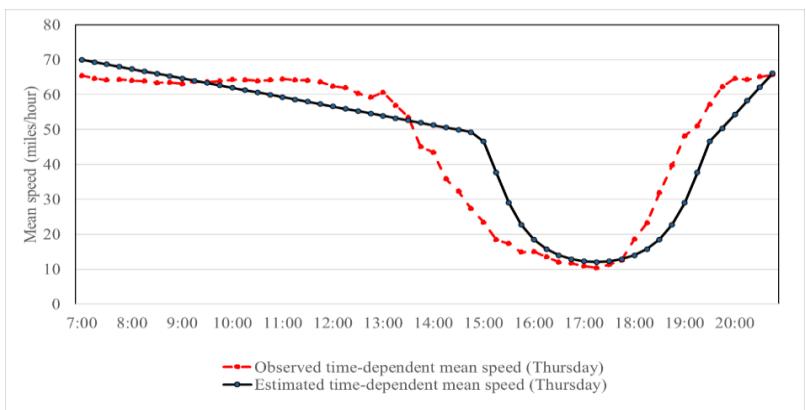
Observed time-dependent mean speed and estimated speed at detector ID 87 on Monday



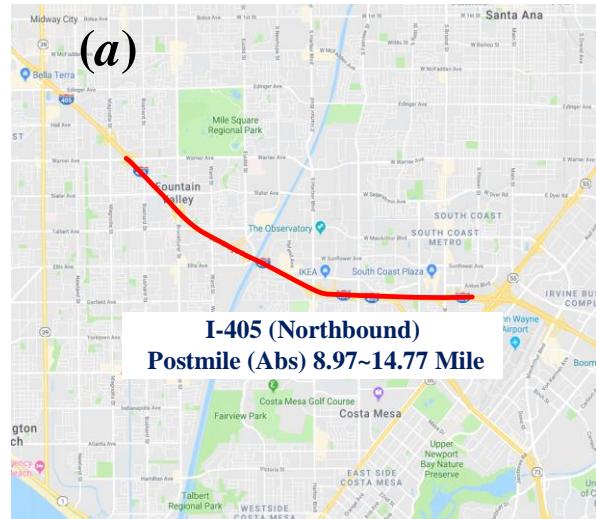
Observed time-dependent mean speed and estimated speed at detector ID 87 on Tuesday



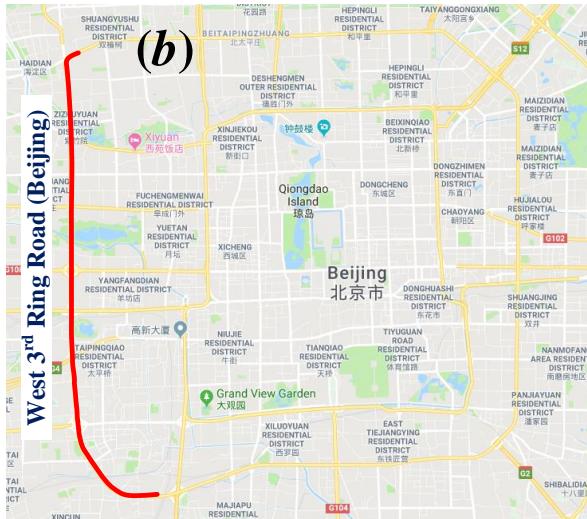
Observed time-dependent mean speed and estimated speed at detector ID 87 on Wednesday



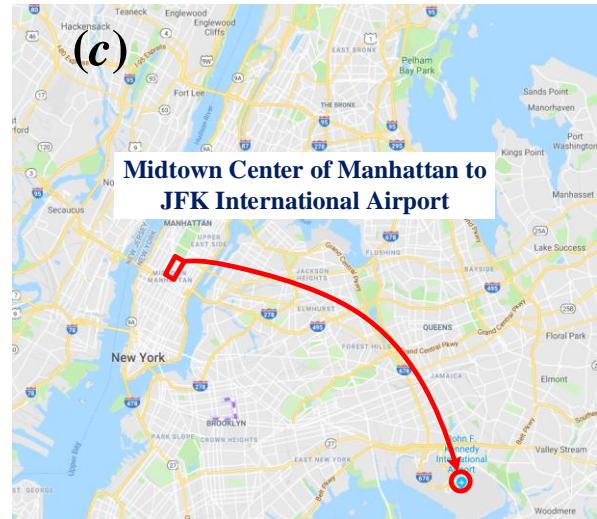
# Case Study of 3 Cities



DS1: Los Angeles



DS2: Beijing



DS3: New York

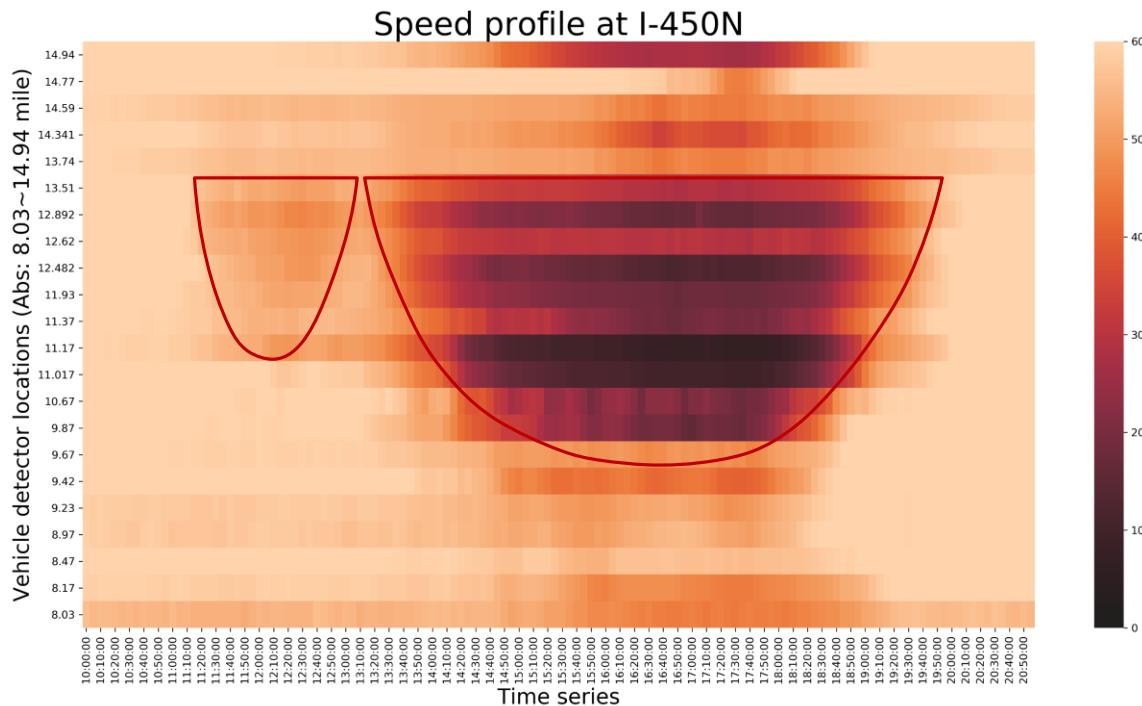
# Description of empirical data sets

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Data set	Queueing system	Object	Data type and information	Collected time and location
DS1	Traffic system	Vehicle	Freeway detector data with traffic flow, speed, and occupancy, etc.	Collected from the Northbound direction of I-405 freeway between absolute postmile 8.97 to 14.77 mile in Los Angeles on April, 2019, from 11:00 a.m. to 20:00 p.m.
DS2	Traffic system	Vehicle	Remote traffic microwave sensor data with the traffic flow, and probe vehicle data with averaged traffic speed, etc.	Collected from the west-third-ring of Beijing City on June 8, 2018, from 6:00 a.m. to 12:00 a.m.
DS3	Transportation system	Taxi	Taxi trip record data with trip ID, pickup time and location, drop off time and location, etc.	Collected from the Midtown Center of Manhattan to the John F. Kennedy International Airport in New York City on October 2018, provided by the New York City Taxi and Limousine Commission

# DS1 Los Angeles

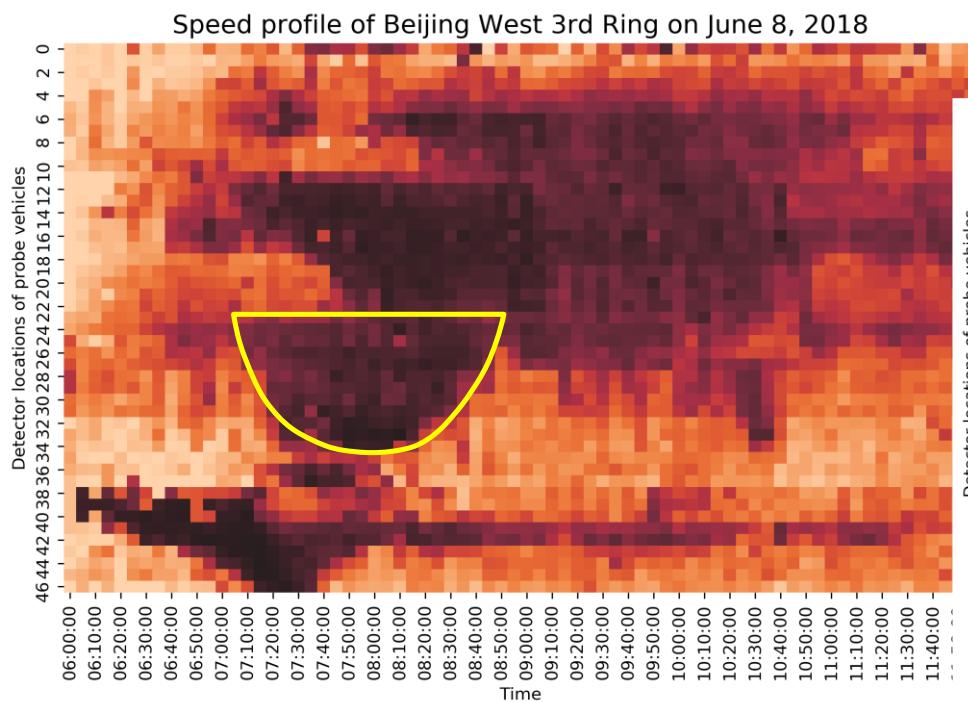
## Bottleneck detection with speed contour map



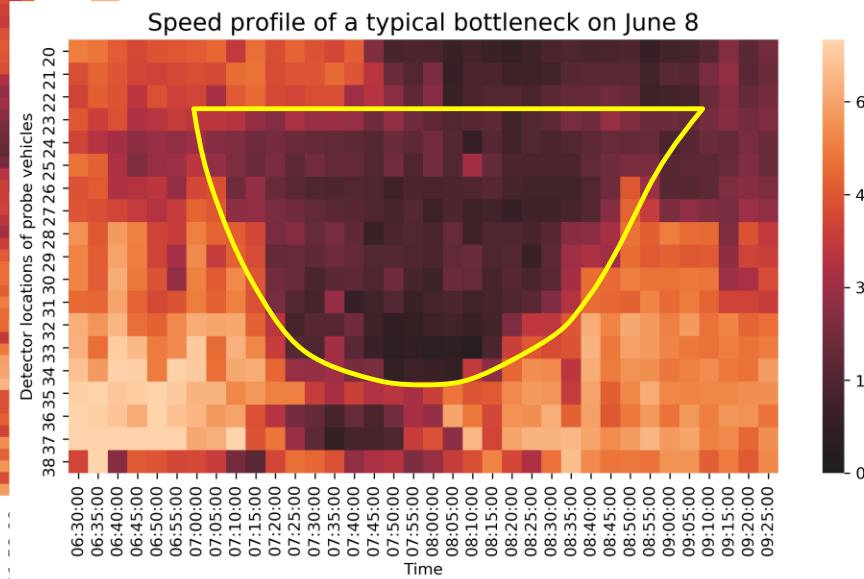
It is clear that the bottleneck is located at  $\text{Abs}=13.51$  mile. In this case, we only analyze one single bottleneck with one peak period from  $t_0 = 13:10$  to  $t_3 = 19:45$ .

# DS2 Beijing

## Bottleneck detection with speed contour map

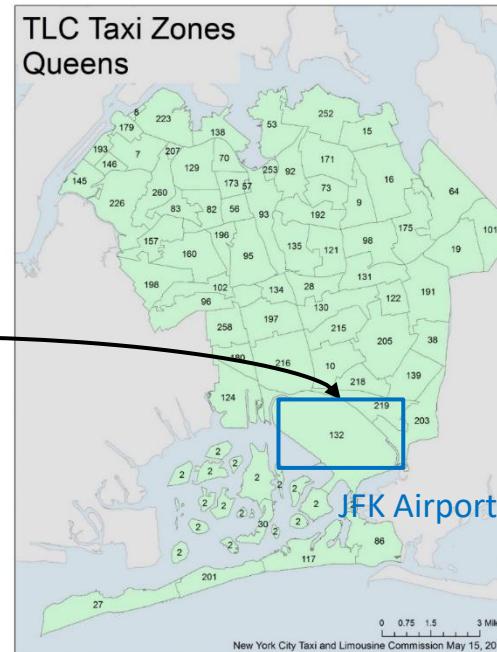


**It exists the spatial queue spillback and the temporal queue connection phenomena.**



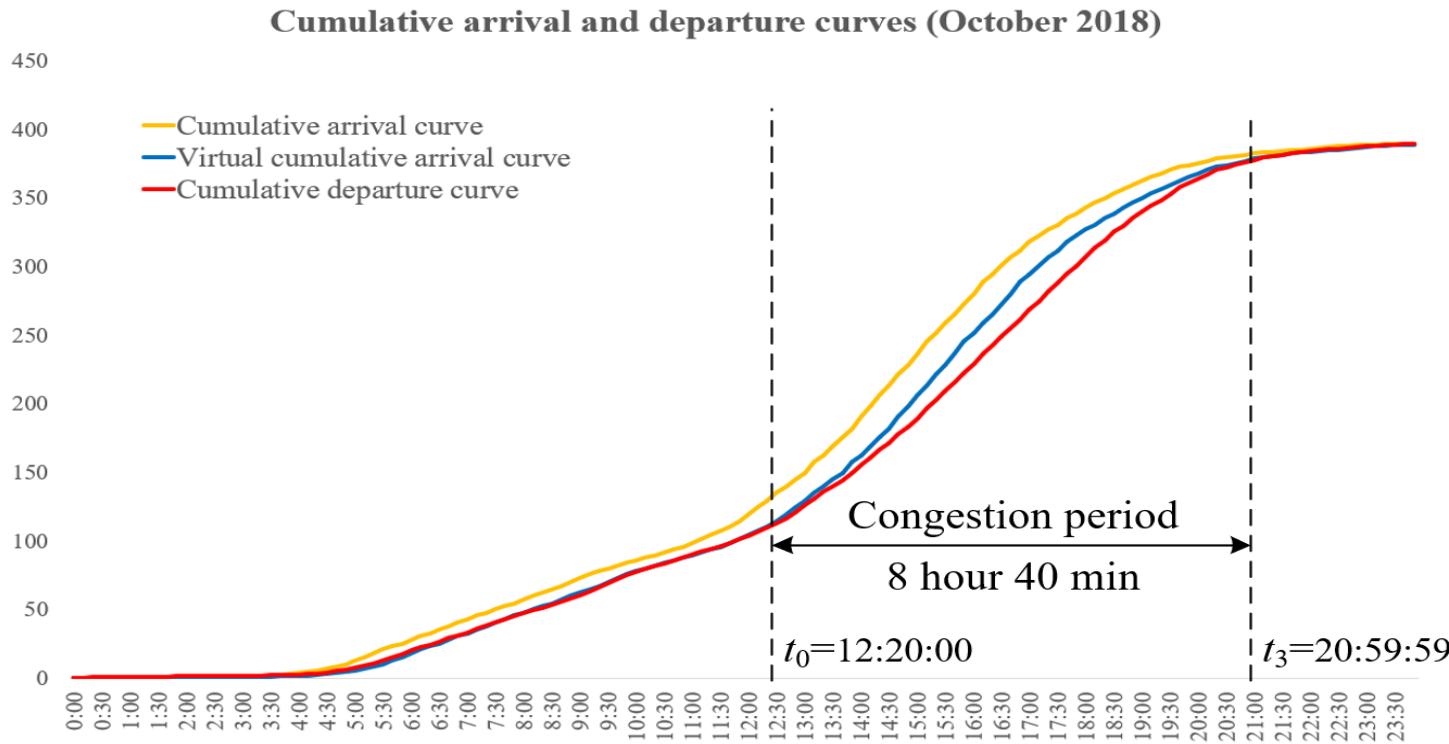
# DS3 New York

## Bottleneck detection



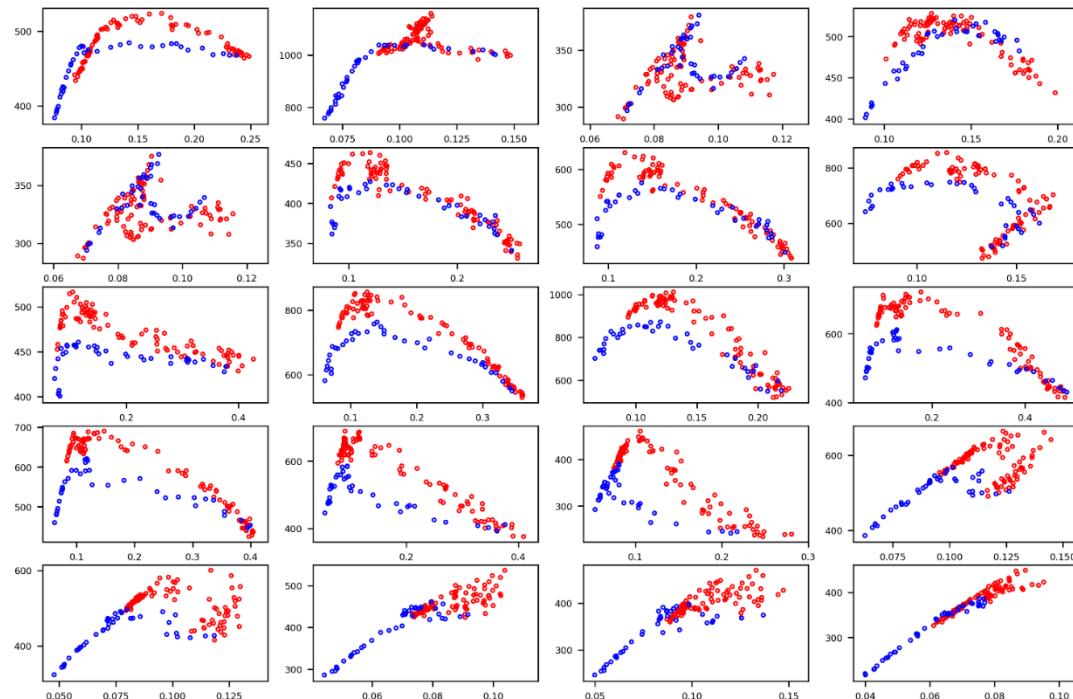
# DS3 New York

## Cumulative arrival and departure curves



# Calibrations: DS1 Los Angeles

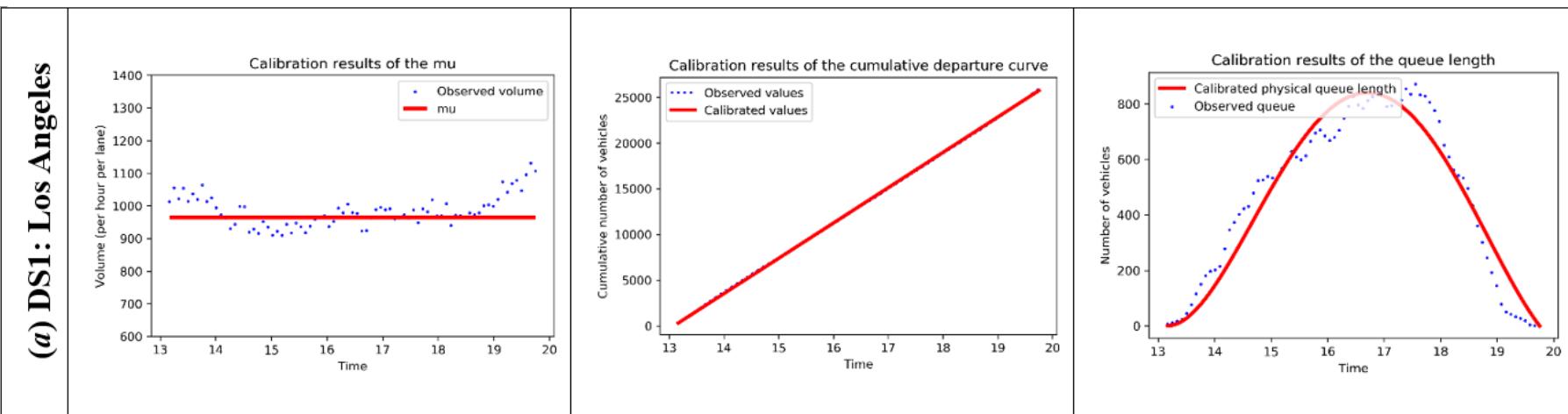
## Flow-occupancy plot



The horizontal axis is the **occupancy**, while the vertical axis is the **flow**. The red dots represent the traffic loading process, while the blue dots represent the traffic unloading process. The bottleneck is located on the second row and second column. We can see that the critical occupancy of the queueing system closes to 0.13, which is used to calculate the observed physical queue length for the parameter calibration.

# Calibrations: DS1 Los Angeles

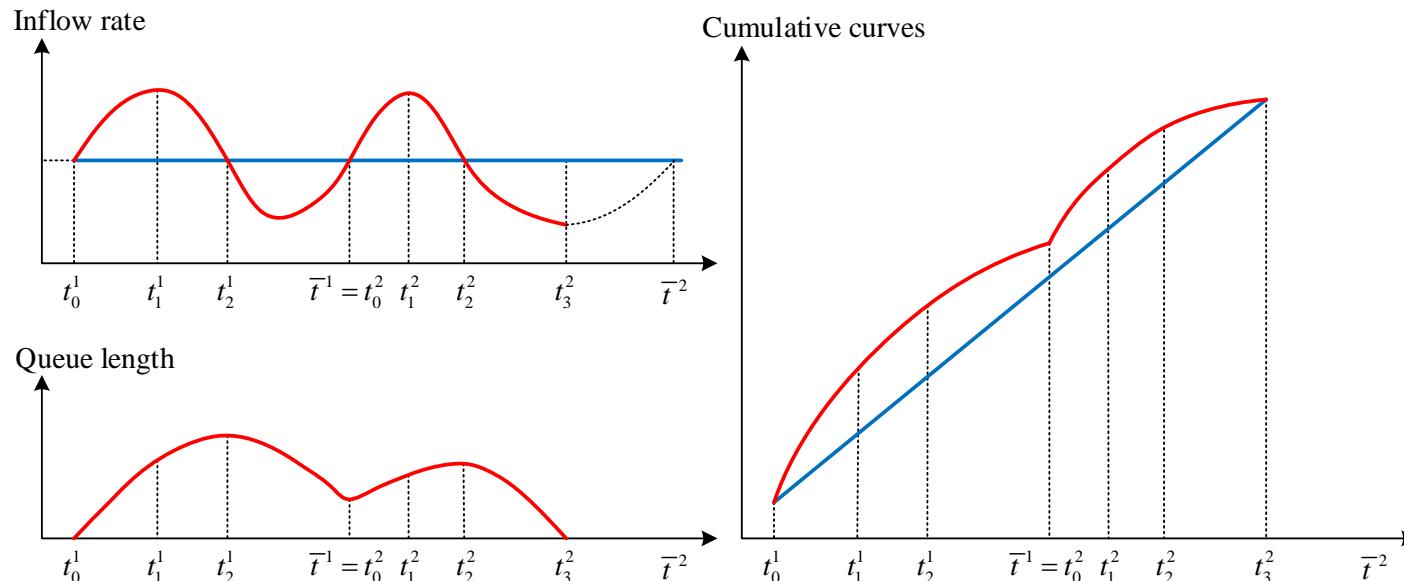
## Results



$$\mu_{DS1} = 3860 \text{ veh/hour} \quad \text{or} \quad \mu_{DS1} = 965 \text{ veh / hour / lane}, \quad \gamma_{DS1} = 13 \text{ veh/hour}^4, \quad m_{DS1} = 0.537$$

# Calibrations: DS2 Beijing

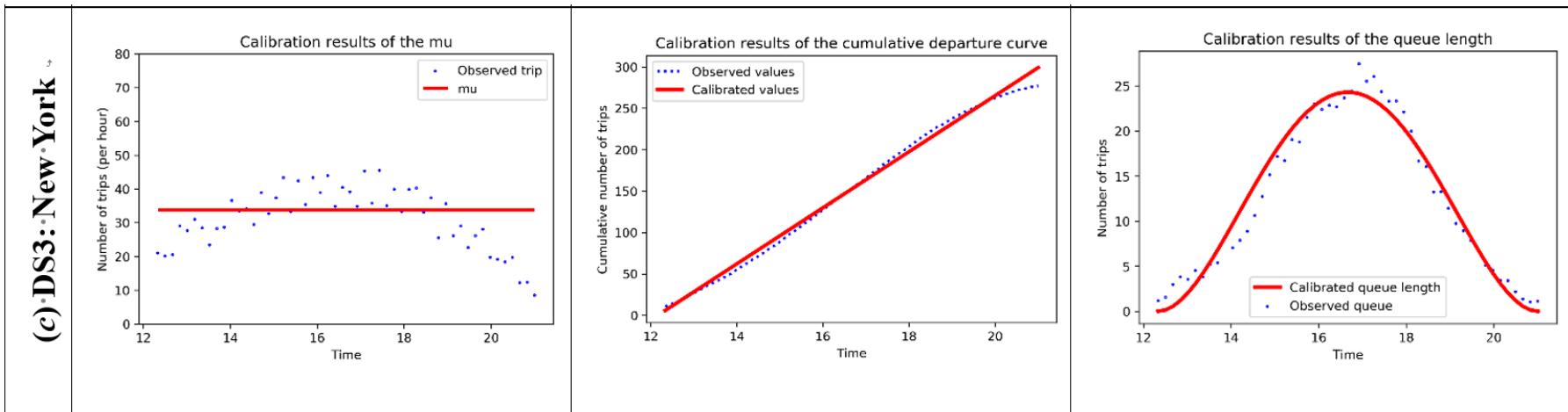
Illustration of the two peak period model



It can be used to solve the temporal queue connection problem.

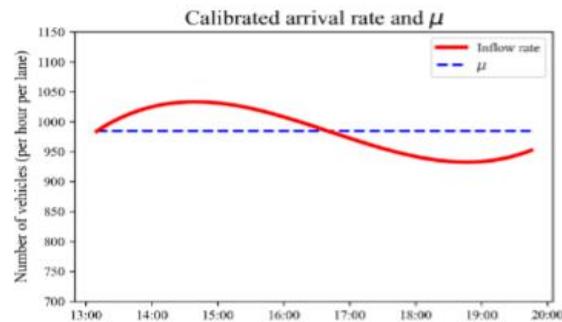
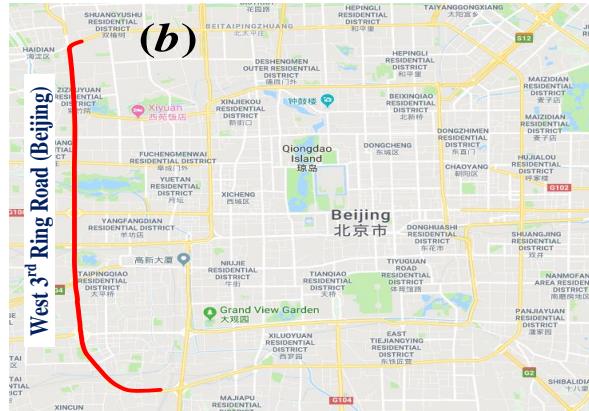
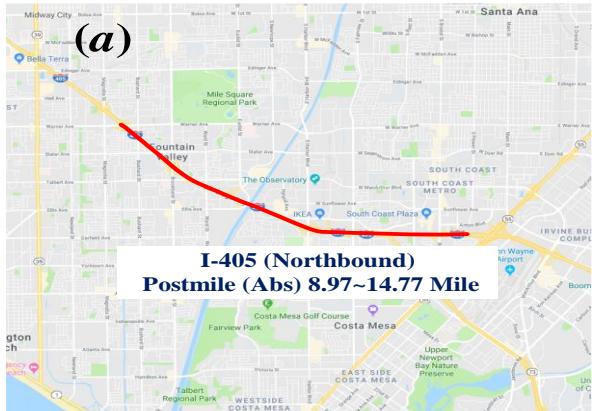
# Calibrations: DS3 New York

## Results

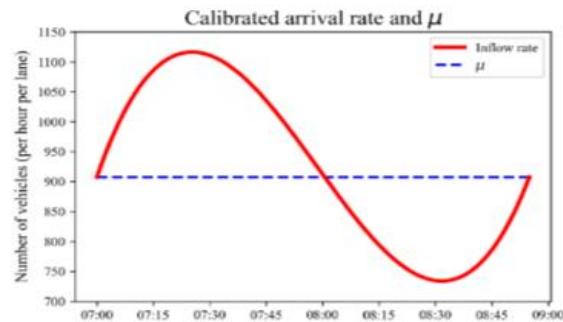


$$\mu_{DS3} = 34 \text{ trip/hour}, \gamma_{DS3} = 0.276 \text{ trip/hour}^4, m_{DS3} = 0.500$$

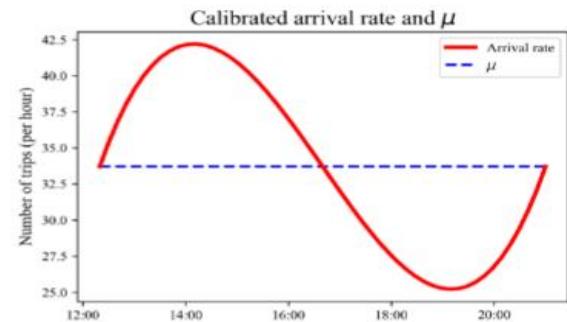
# Final comparison across different cities



(a) DS1



(b) DS2



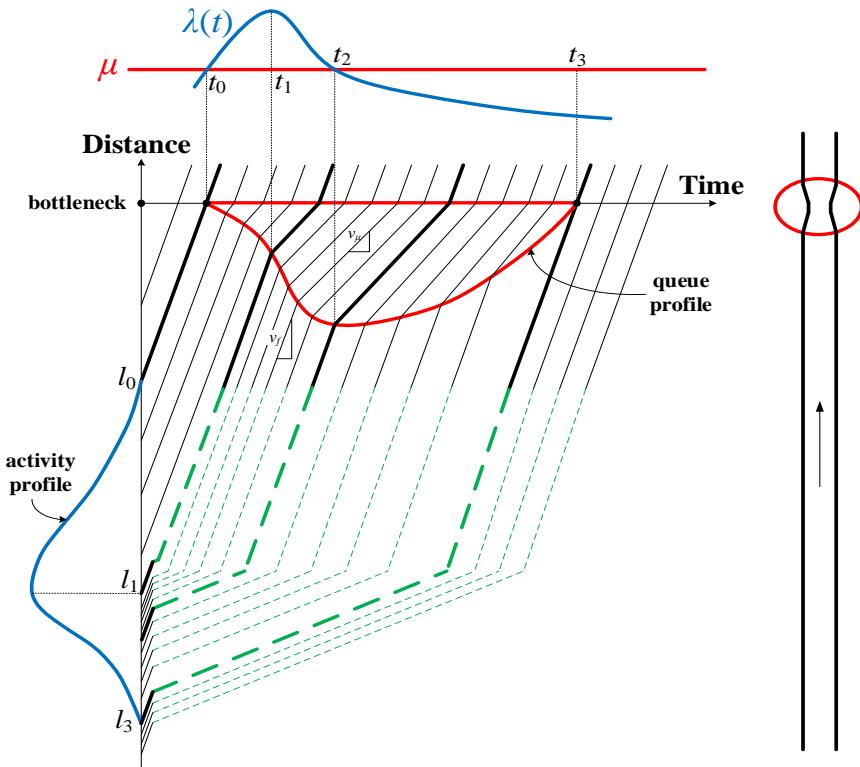
(c) DS3

The ratio of  $\lambda/\mu$  indicates the variation of the demand over supply.

The results of  $\lambda/\mu$  are:  $0.937 \leq (\lambda/\mu)_{DS1} \leq 1.059$ ,  $0.813 \leq (\lambda/\mu)_{DS2} \leq 1.234$ ,

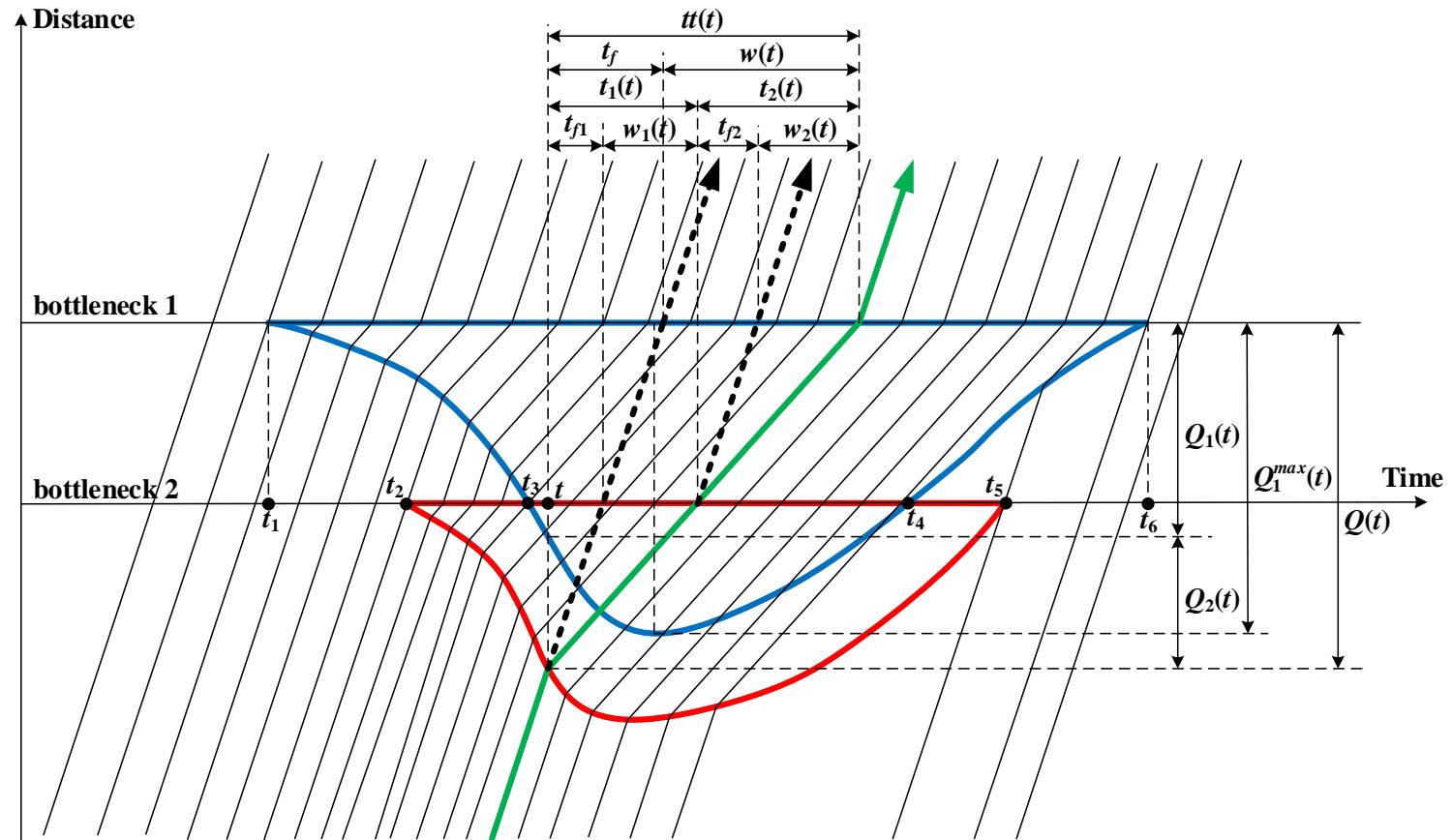
$0.745 \leq (\lambda/\mu)_{DS3} \leq 1.255$ .

## Discussion 1: Activity profile and queue profile



**Illustration of the activity profile and queue profile in an oversaturated queueing system.** Vehicle trajectories shown with blue colors are illustrated by the broken lines due to space limitation; however, they should be straight lines before they encounter with the back of the queue. It reveals that traffic demand of  $\lambda(t)$  and the queue are the consequences of the travel activity upstream the bottleneck.

## Discussion 2: Queue spillback



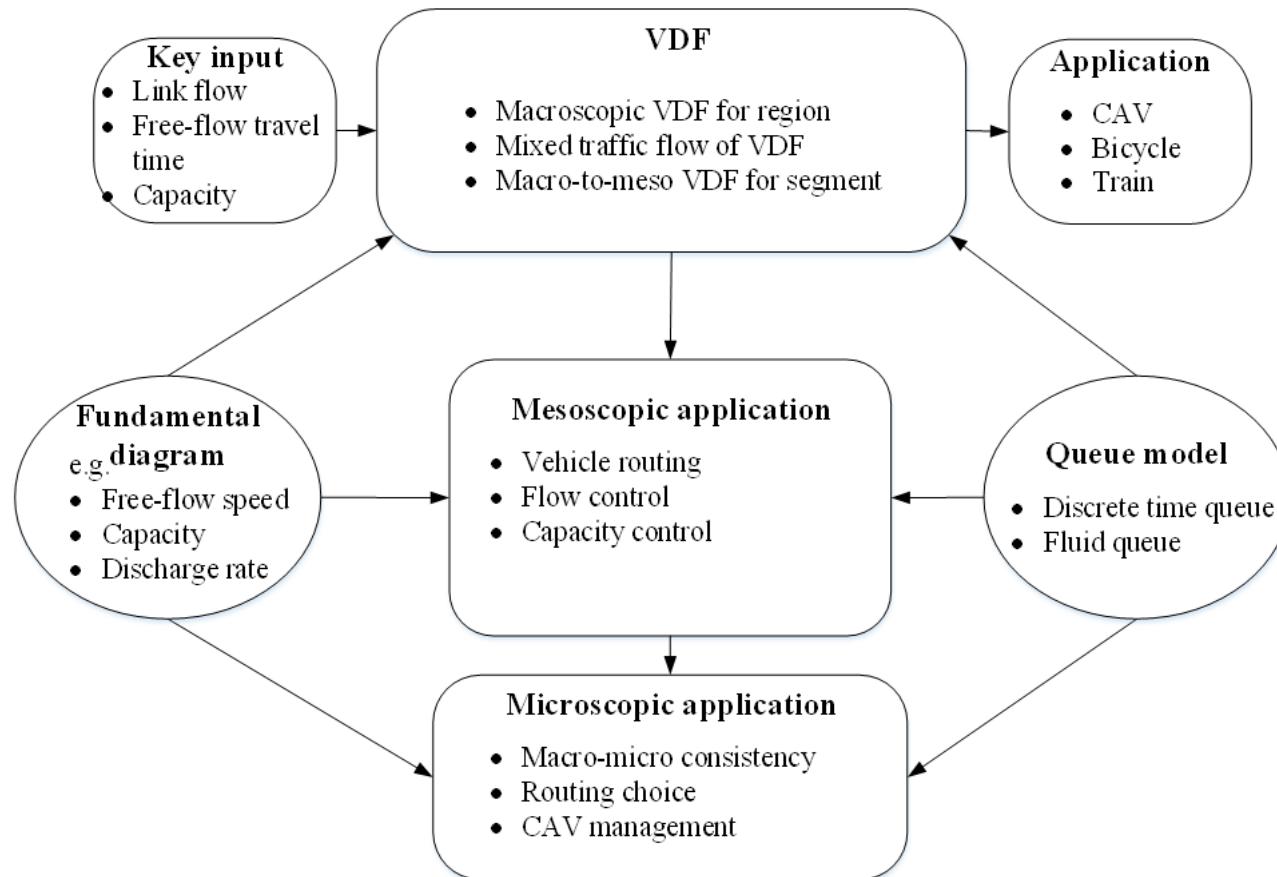
# Queue VDF-related modeling for emerging technologies

Stakeholders	Micro	Meso	Macro
Autonomous vehicles (AVs)	<ul style="list-style-type: none"> <li>link kinematic wave to capacity</li> <li>link car following to reaction time</li> <li>traffic control</li> </ul>	<ul style="list-style-type: none"> <li>control inflow rate</li> <li>link capacity to discharge rate</li> <li>queue evolution</li> </ul>	<ul style="list-style-type: none"> <li>departure time choice</li> <li>incoming flow rate shape and pattern</li> </ul>
Mobility as a service (MaaS)	<ul style="list-style-type: none"> <li>choice pick up service time</li> </ul>	<ul style="list-style-type: none"> <li>point to point travel time</li> <li>time-dependent discharge rate</li> </ul>	<ul style="list-style-type: none"> <li>control inflow rate curves</li> <li>control departure rates curves</li> </ul>
Public transportation and multimodal transit	<ul style="list-style-type: none"> <li>departure time choice</li> </ul>	<ul style="list-style-type: none"> <li>generate time-varying travel time</li> <li>travel distribution</li> </ul>	<ul style="list-style-type: none"> <li>trip pattern</li> <li>rush hour</li> </ul>
Energy	<ul style="list-style-type: none"> <li>vehicle specific power (VSP)</li> </ul>	<ul style="list-style-type: none"> <li>time-dependent speed profile</li> </ul>	<ul style="list-style-type: none"> <li>link average speed from BPR to fluid approximation queue</li> </ul>
Emissions and health	<ul style="list-style-type: none"> <li>control freeway ramp metering</li> <li>arterial traffic signal coordination</li> </ul>	<ul style="list-style-type: none"> <li>dynamic route shifts</li> <li>traffic dynamics scheduling</li> </ul>	<ul style="list-style-type: none"> <li>estimating the spatial distribution of travelers and mode shifts</li> <li>trip patterns</li> </ul>

# Queue VDF-related modeling for multimodal systems

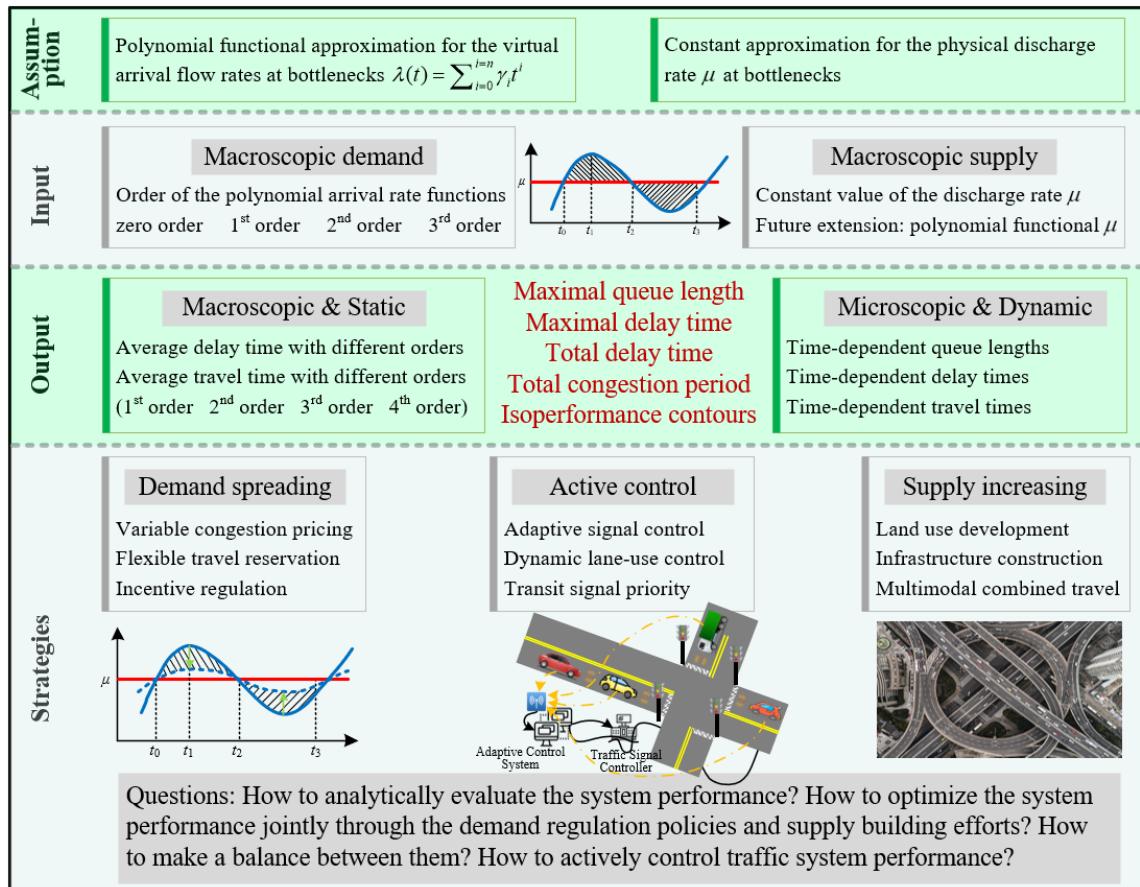
		Underlying Methodology			
Research Problems	Focus	Fundamental diagram	Queueing theory with a stochastic process	Point queues/bottleneck models	Cell transmission model, simplified KW
Multi-modal transportation (rail, air, truck)	Freight flexibility	✓	✓	✓	
	Railway capacity	✓	✓		
Public transportation	Bus accessibility	✓	✓	✓	✓
Bike	Congestion in a single path	✓	✓	✓	
Pedestrians	Bidirectional pedestrian streams	✓	✓		✓
MaaS	Ridesourcing shared transportation	✓	✓		
CAV	Mixed traffic flow	✓	✓		✓

# Summary of today' talk



The primary purpose of applying VDF in emerging technology is to equilibrate assigned traffic volumes over the urban network by estimating the level of free-flow or congestion (i.e., delay) condition at the macroscopic, mesoscopic, or microscopic transportation network analysis level. The classification of research problems of VDF in the emerging application is shown in this figure.

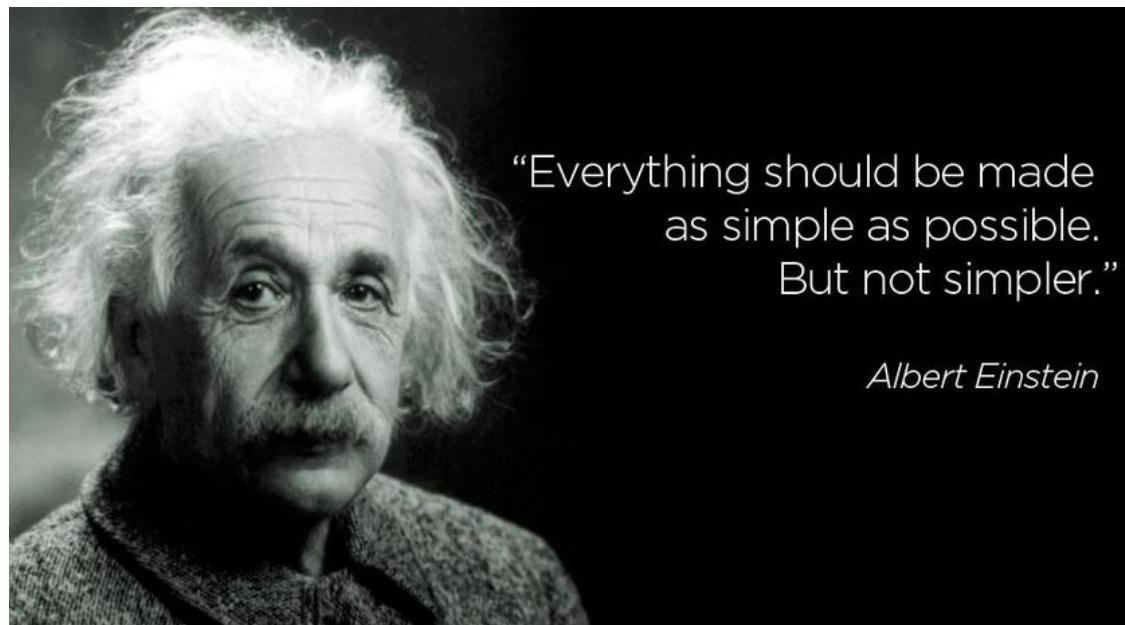
# Applications of Integrated demand-supply models



## Concluding Remarks

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- First principle: simple but not simpler;
- Simplify the complexity but still good enough.



# THANK YOU

<https://github.com/asu-trans-ai-lab/>