

Full-Day Accessibility Evaluation of Transit Systems Using GPS-Based Location Data

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Assistant Professor

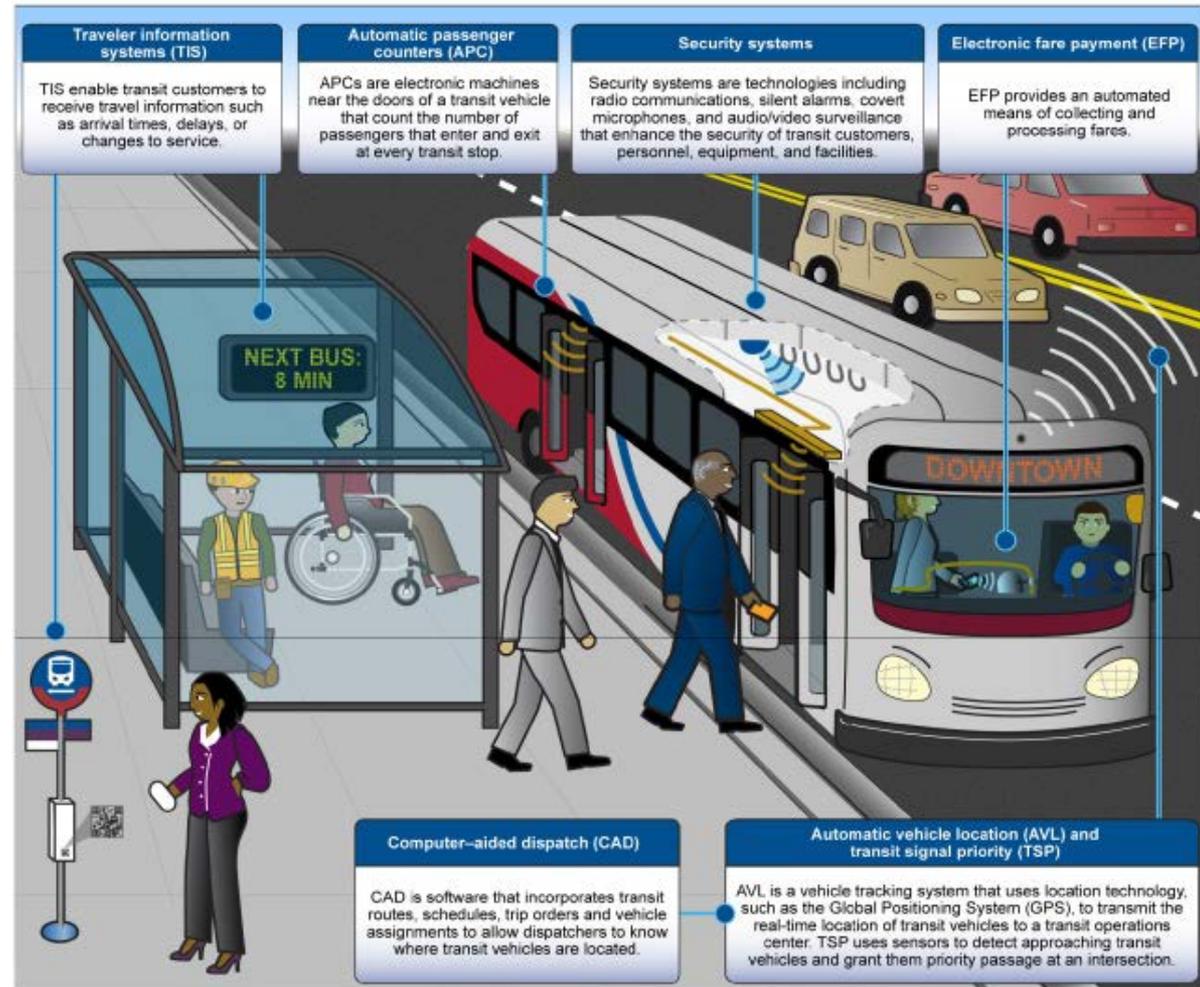
Geography, Environment and Society

University of Minnesota

Background

❖ Transit Intelligent Transportation System (ITS) Technologies

- **Automatic vehicle location (AVL)**
- **Automatic passenger counters (APC)**
- Traveler Information System (TIS)
- Electric fare payment
- Transit signal priority
- Security systems
- Computer-aid dispatch



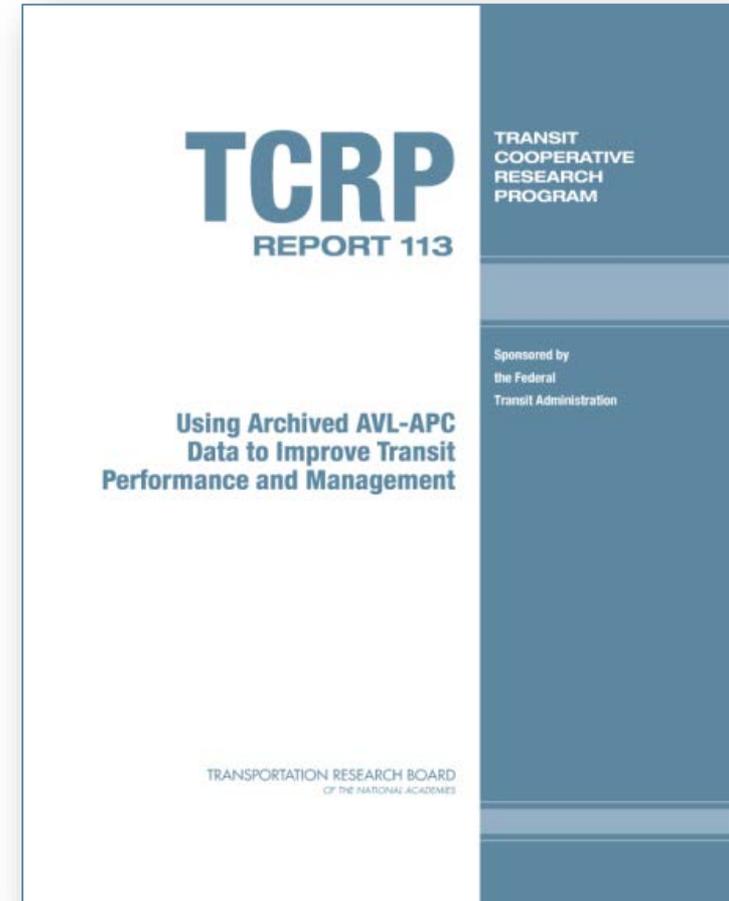
Source: GAO analysis of Department of Transportation documentsGAO-16-638

Background

❖ Using AVL-APC data to improve transit performance & management

- **Automatic vehicle location (AVL)**
 - ✓ primarily for real-time monitoring/control
 - ✓ not typically processed for off-line analysis
- **Automatic passenger counters (APC)**
 - ✓ provide passenger-activity data compatible with AVL operating data
- **Off-line analysis has substantial promise**
 - ✓ provide insights on transit performance
 - ✓ improve service planning and scheduling
 - ✓ promote transit ridership

[reliable services](#) & [accurate information](#)



Board, T.R., (2006) National Academies of Sciences, E., Medicine: Using Archived AVL-APC Data to Improve Transit Performance and Management. The National Academies Press, Washington, DC

Background

- ❖ **Improve transit accessibility**
 - Fast and frequent services
 - Sufficient service coverages
 - First- and last- mile connection
 - **Reliable schedule**
 - ✓ Better scheduled arrival times
 - ✓ Accurate real-time estimations

**TransitTracker™** **Portland TriMet**
TriMet's real-time arrival information system

**CTA Bus Tracker** **Chicago Transit**
Your official source for up-to-the-minute arrival info.



Serving the Minneapolis / St. Paul Area

NexTrip

Get real-time departure times for your routes (?)

A Line - Northbound
Snelling & Larpenteur Station
Stop Number: 50010

NexTrip departs in 3 Min

Current Time: 9:24 am

Route	Departs
A Line Rosedale/Rapid	3 Min
A Line Rosedale/Rapid	7 Min
A Line Rosedale/Rapid	18 Min
A Line Rosedale/Rapid	9:51
A Line Rosedale/Rapid	10:01
A Line Rosedale/Rapid	10:11

• Real Time • Scheduled

[← BACK TO STOP](#) [SHOW MY BUS →](#)

only at time stops (TS); accurate?

Methods

❖ Model Components

1. Linear-referencing

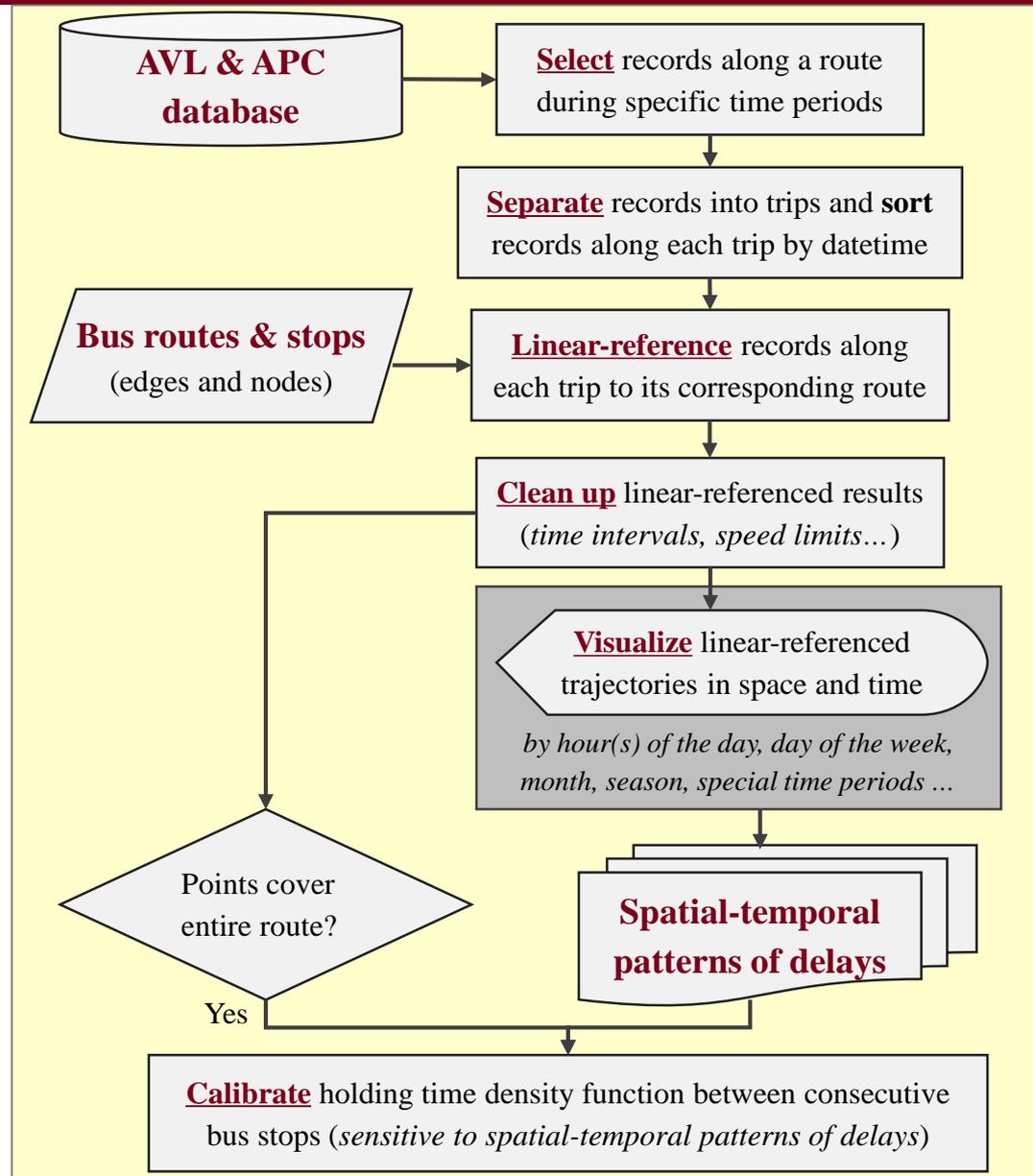
- ✓ allocate AVL and APC data along transit routes

2. Visual exploration

- ✓ spatio-temporal patterns of bus delays along routes

3. Semi-Markov process

- ✓ model vehicle movements as “jumping” processes from one stop to its next stop
- ✓ calibrate “holding times” for the “jump” between stops



❖ Method Highlights

1. Linear-referencing

- ✓ discrete GPS coordinates ⇒ **an ordered sequence of placement along route**
- ✓ eliminate artificial movements due to device precision etc.

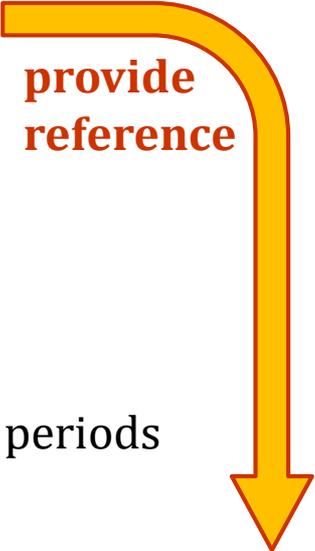
2. Visual exploration

- ✓ where and when are the delays ⇒ **at and between bus stops**
- ✓ traffic congestions, dwelling times, ...

3. Semi-Markov process

- ✓ recognize **ripple impacts of delays** along the trip
- ✓ sensitive to **traffic situation and ridership** during specific time periods

 **support**

 **provide reference**

Use AVL and APC data to calibrate delays between each pair of bus stops.

A Pilot Study

❖ Original Dataset

• A-Line

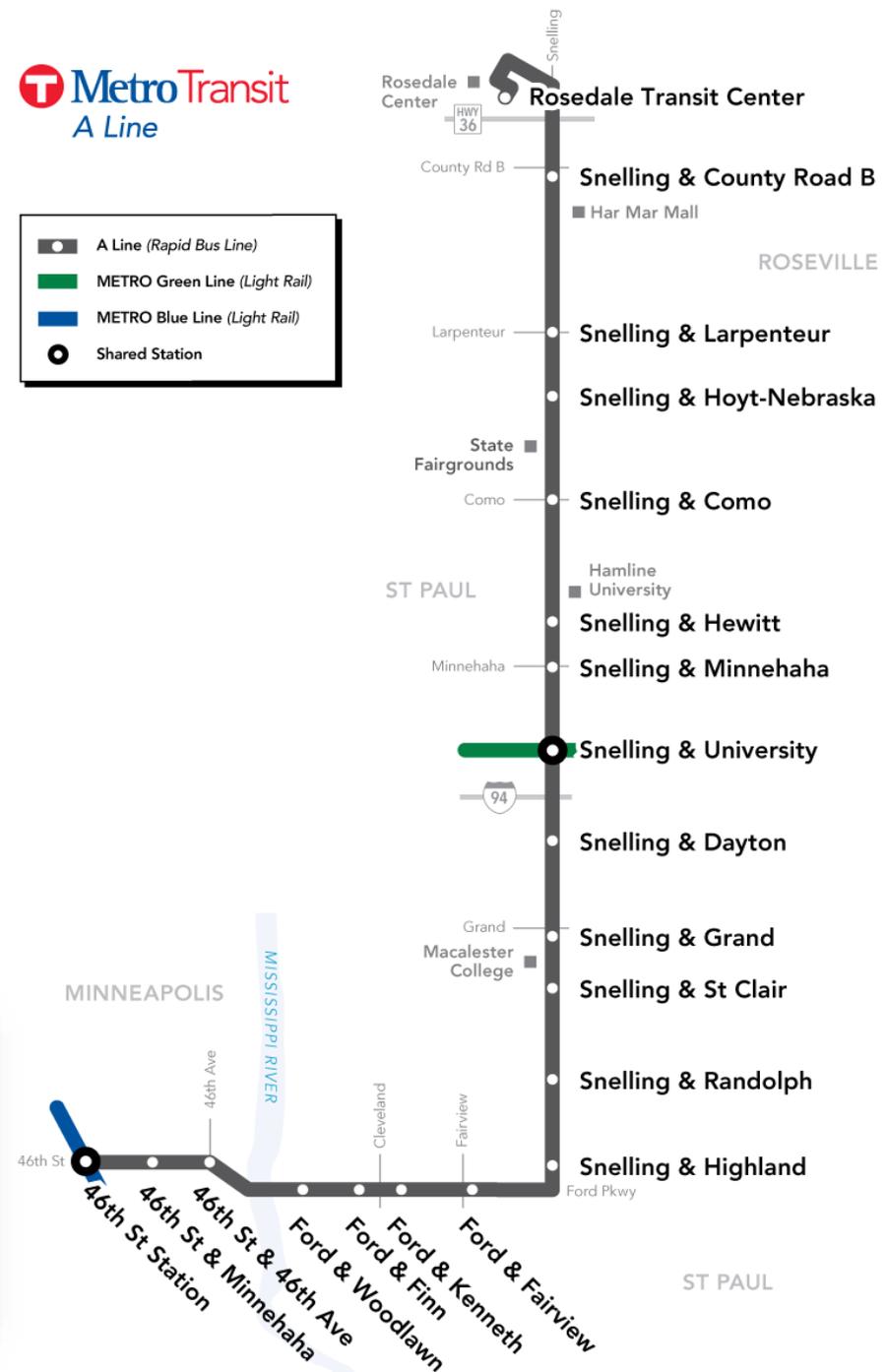
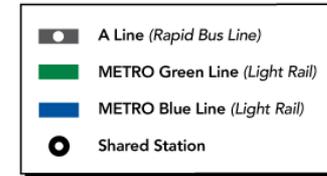
- ✓ every 10 minutes
- ✓ pay ticket before boarding

• AVL and APC data

- ✓ Oct. 1 to Oct. 8, 2016
- ✓ 12 vehicles
- ✓ 136,883 messages

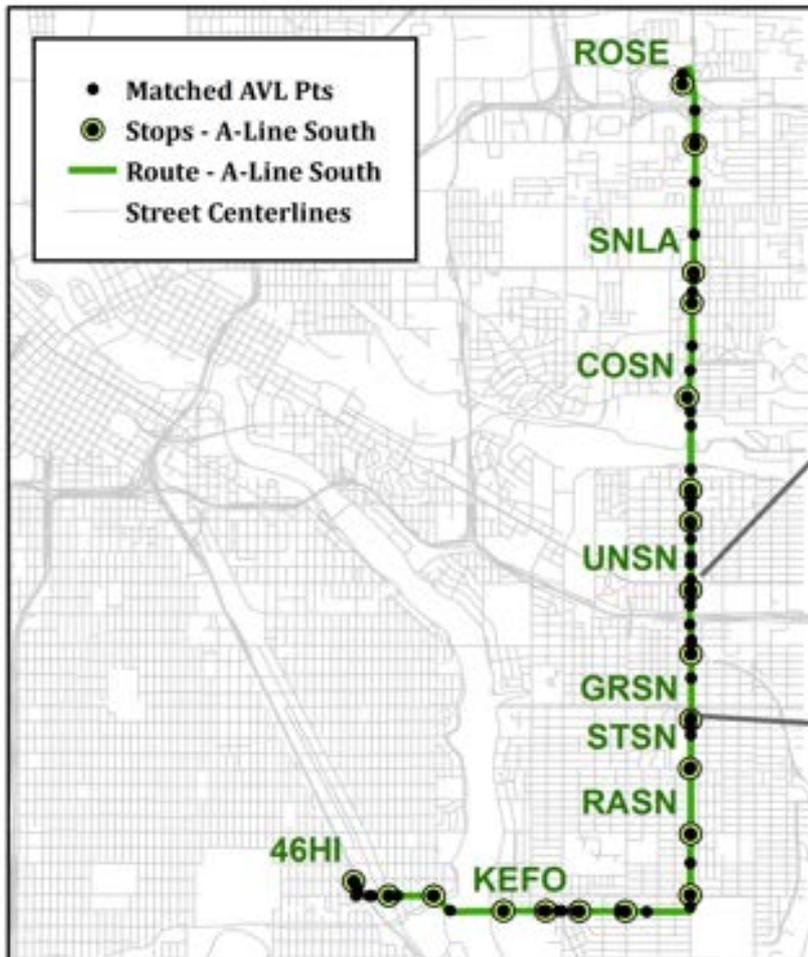


 Metro Transit
A Line



1. Linear Referencing

(longitude, latitude, timestamp, ...) ⇒ (route, distance from first stop, time stamp, ...)

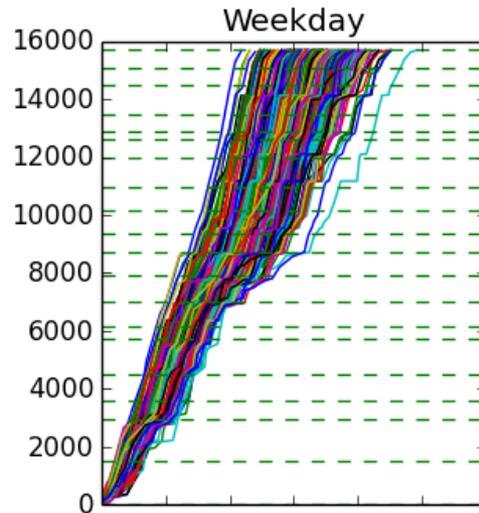


deal with
“back-forth”
movements

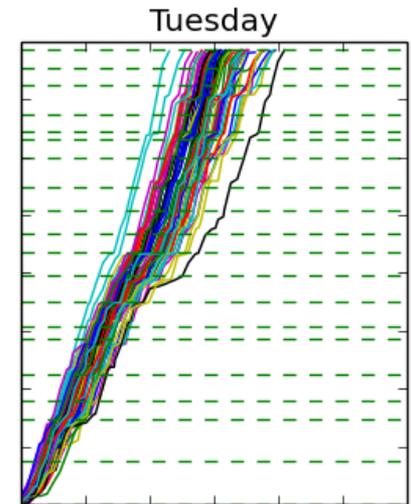
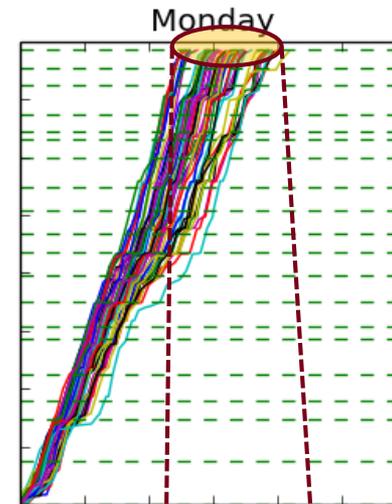
Results

2. Visual exploration – day of the week

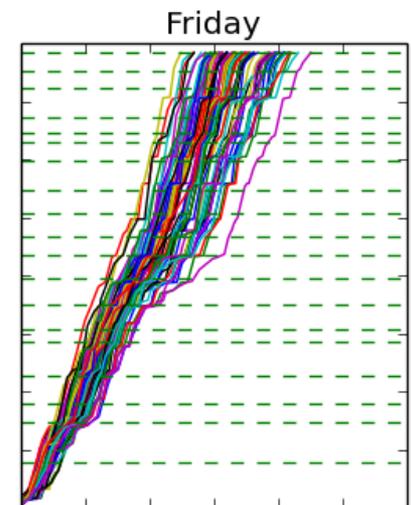
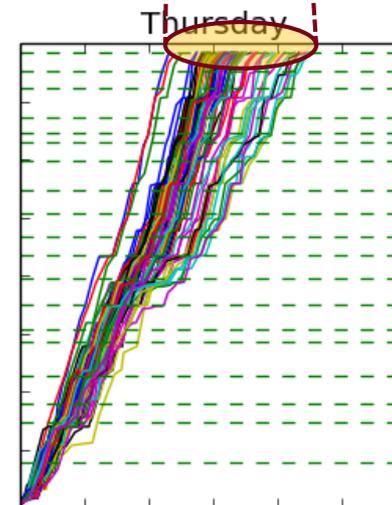
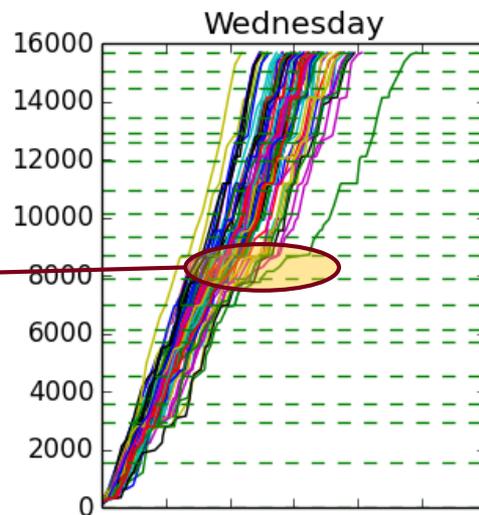
less variation
less delays
⇒
more reliable



different variations

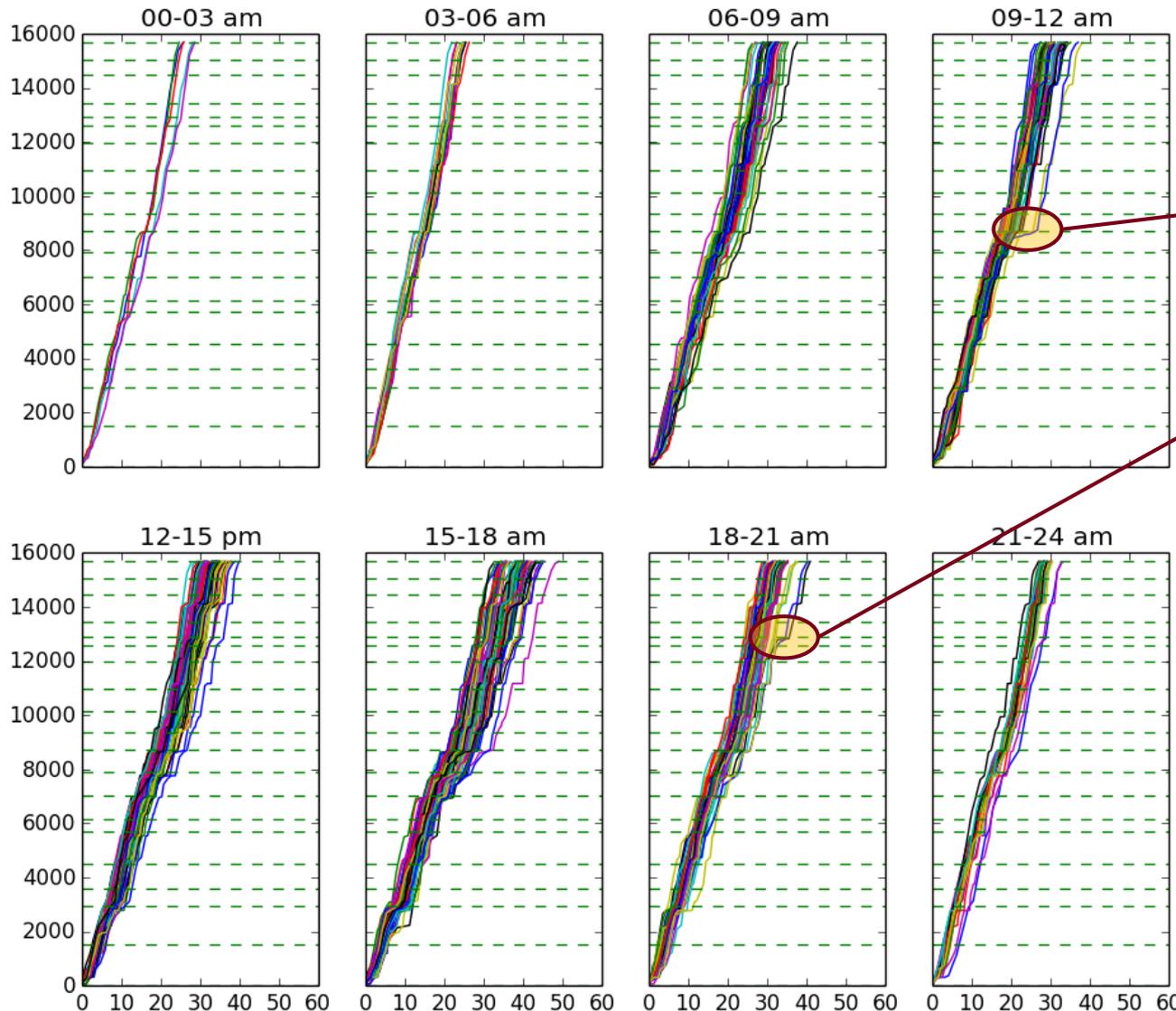


where lead to
more delays



Macalester College
& University Ave W

2. Visual exploration – hour of the day



**where & where
has more delays**

**Macalester College
& University Ave W
9am – 12pm**

**Near Cleveland Ave
S & Ford Pkwy cross
6pm – 9pm**

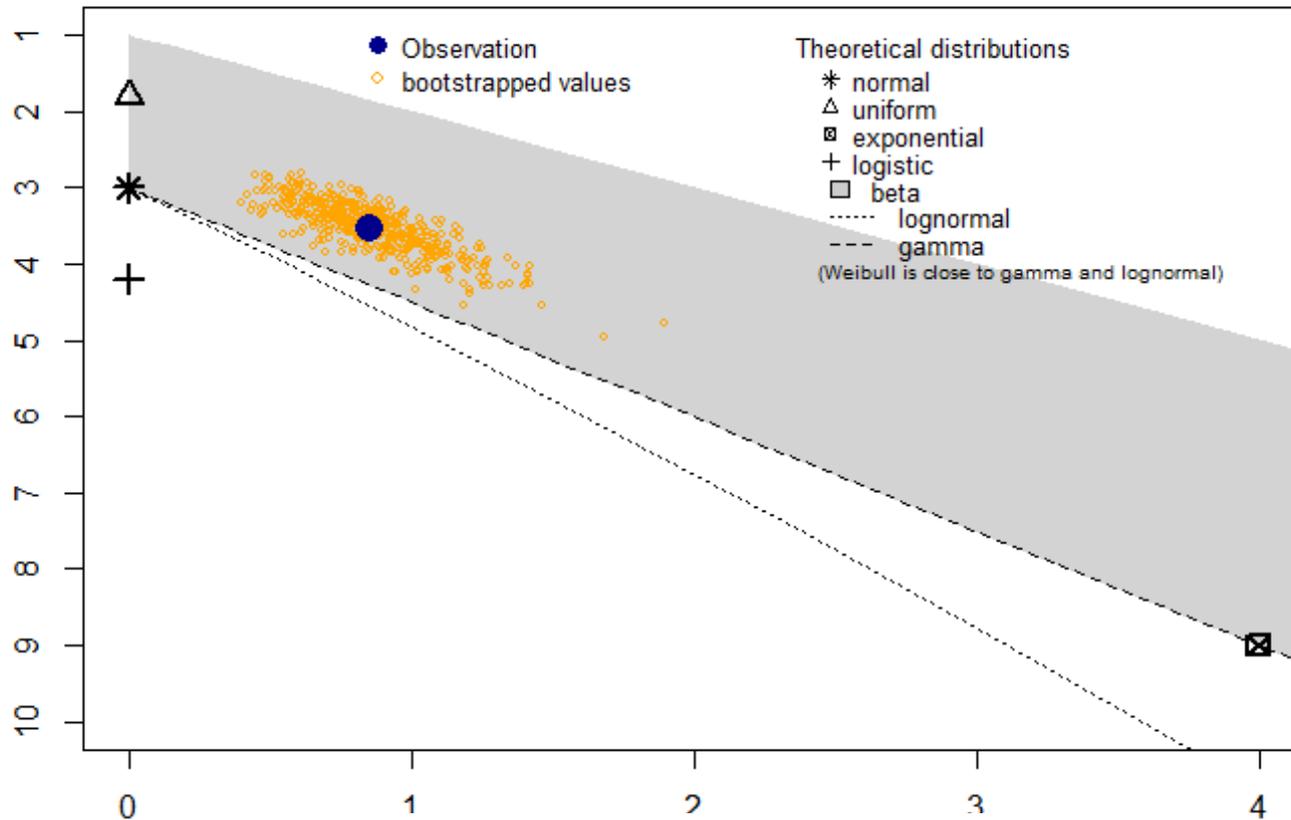
**consistent with
afternoon peak
hours for traffic**

3. Calibrate delays between stops

1) kurtosis (variability) and squared skewness (location) of empirical delays

Kurtosis - measure of heavy-tailed

Cullen and Frey graph



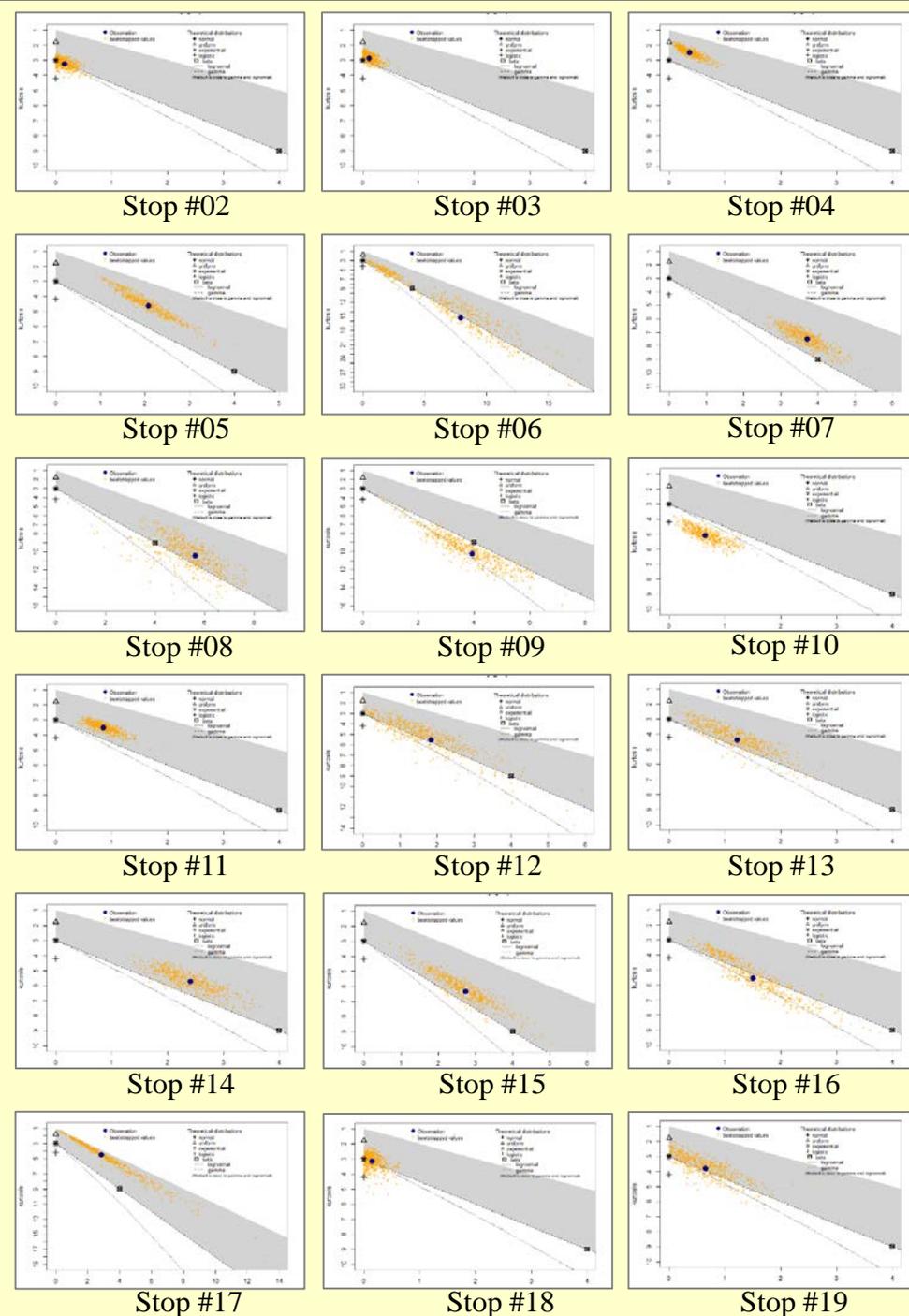
Skewness - measure of symmetry

Results

3. Calibrate delays between stops

Lognormal & Gamma distribution

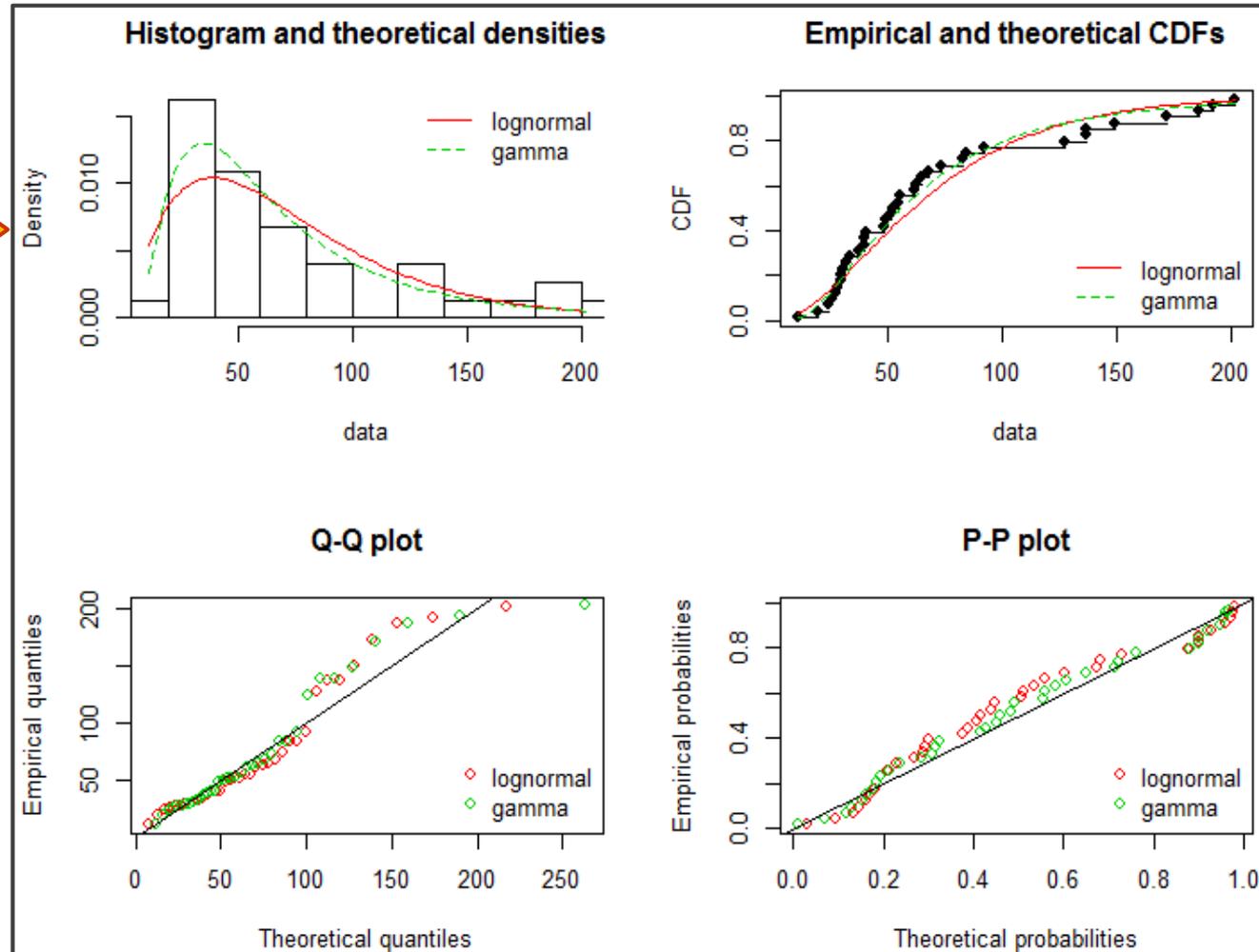
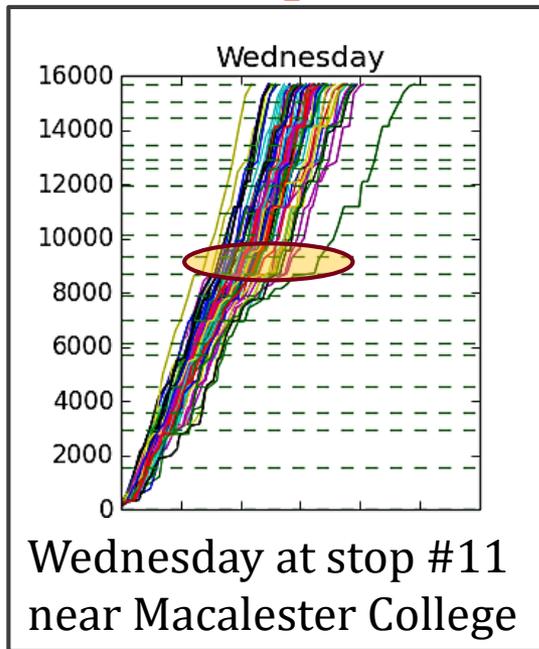
- Patterns are slightly different along trips between two stops
- **Beginning and end** of the trip tend to have **low kurtosis**
⇒ less outliers (long delays)
- **Towards the middle** of the trip, the **skewness is larger**
⇒ not symmetrically distributed



3. Calibrate delays between stops

2) Compare empirical and theoretical fitted distribution <maximum likelihood>

Mean = 56.26 seconds
Std. = 2.01 seconds



❖ Results Highlights

- Visualizing linear referenced trips can reveal **spatial-temporal patterns of vehicle delays** along the routes, and **identify locations and/or times** that lead to delays
- Along A-Line route, delays during weekdays fit best with **lognormal and gamma distribution** for movements between any two consecutive stops; yet the patterns are slightly difference with respect to **location and variability** of empirical delays

❖ Practical Merits

- **Adjust bus schedules** to reflect the revealed delay patterns
- Provide **more accurate arrival time estimations** to transit users that can
 - recognize that the delays between two stops may not be the same
 - capture the ridership and traffic “congestions” during different time periods
- **Refine current transit accessibility measures** to consider expected/mean delays at stops in addition to the scheduled service time and frequency

Thank You!

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*This material is based upon work supported by
CTS Seed Fund*



The supplementary data is provided by Metro Transit

1. Jump Rate

$$p_{ij} = \begin{cases} 1 & t_{ij} < t_j^+ - t_i^- \\ 0 & \text{otherwise} \end{cases}$$

2. Holding Time– reach e_{ij} from v_i at time t

$$p_i(t) = \begin{cases} 0 & , \quad t \in [t_0, t_i^-) \\ \int_{t_i^-}^t \lambda e^{-\lambda\tau} d\tau = e^{-\lambda t_i^-} - e^{-\lambda t} & , \quad t \in [t_i^-, t_j^+ - t_{ij}) \\ \int_{t_i^-}^{t_j^+ - t_{ij}} \lambda e^{-\lambda\tau} d\tau = e^{-\lambda t_i^-} - e^{-\lambda(t_j^+ - t_{ij})}, & t \in [t_j^+ - t_{ij}, t_D] \end{cases}$$

3. State probability

$$Prob(e_{ij}, t) = p_{ij} * p_i(t) * p_j(t)$$

Root Mean Square Error (“fitdistrplus” package in R)

Based 1: the estimated NTP

$$RMSE^1(t) = \frac{\sum_{\{e_{ij}\}^t} \left(EmpP_{(e_{ij},t)} - SimP_{(e_{ij},t)} \right)^2}{Cnt(t)}$$

$\{e_{ij}\}^t$: edges within NTP at time t

$\{SimP_{(e_{ij},t)}\}$ simulated visit probabilities

$\{EmpP_{(e_{ij},t)}\}$ empirical visit probabilities

Based 2: the empirical NTP traces

Further select from $\{e_{ij}\}^t$: edges have been used by at least one GPS trace