# Using AVL-equipped Buses to Detect Recurring Traffic Congestion on Surface Streets

Mark M. McCord,\* Prem Goel,\*\* Rabi G. Mishalani,\*
Cheng Chen,\* and Jinguo Gao\*\*

\*Dept. of Civil and Environ. Engineering and Geodetic Science

\*\*Department of Statistics

The Ohio State University

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## "Buses as probes"

- Many transit buses are presently equipped with AVL systems (low marginal cost)
- Buses regularly and frequently cover surface streets
- Let's use the "infrastructure" to help identify traffic conditions
- Any AVL equipped public fleet would do?

#### Buses as probes on surface streets

- Traditional difficulty: Buses behave differently than prevailing traffic
- Approach used in this effort
  - "Compare buses to buses"
  - Find systematic time-of-day differences in bus speeds to infer time-of-day differences in traffic conditions
  - Recurring congestion: More appropriately, time-of-day reductions in bus speeds to *indicate* possible recurring congestion

### Overview of remaining presentation

- Location specific concept
- Route-based concept
- Empirical study and results
- Comments with respect to classical approaches
- Ongoing and future work

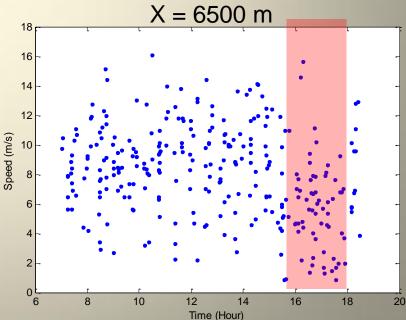
#### Location-specific concept



At a location X on a route

Group bus speeds over several days assumed homogeneous in bus operations (e.g., weekdays)

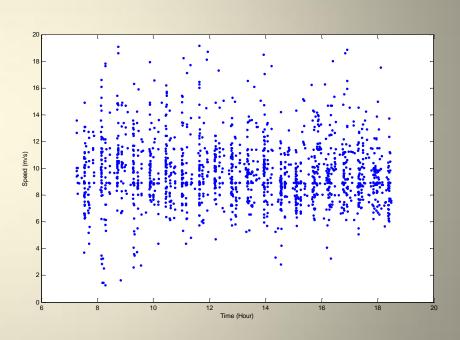
Identify time periods where speeds are markedly lower than the speeds at other time periods



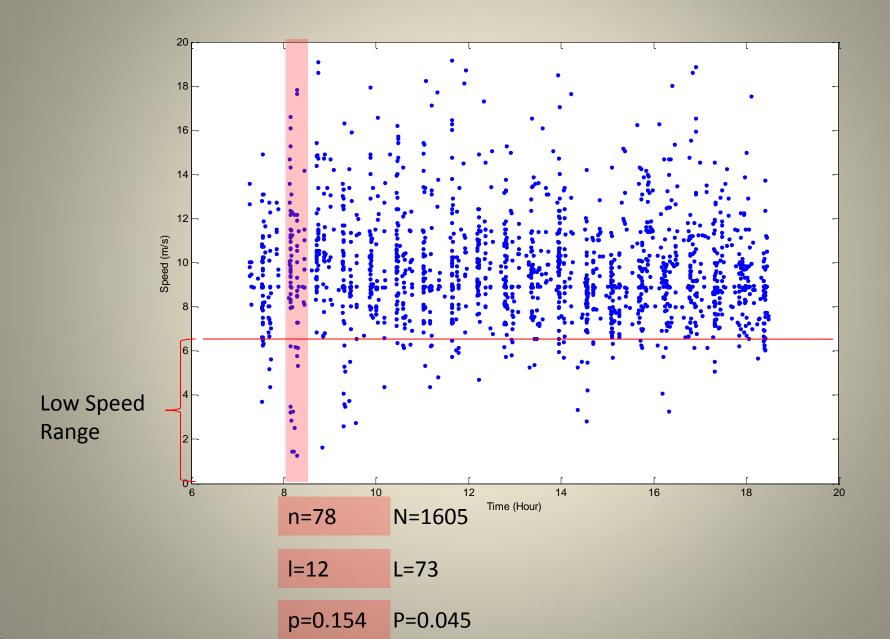
# Congestion Identification at Location X: "Low-speed" Method

- Identify low speed range (e.g., 0.35 x 95%-ile of speeds)
- Determine:
  - N: # overall speed observations in data set
  - L: # overall low speed observations data set
- Divide day into time intervals
- For a time interval, determine:
  - n: # speed observations in selected time interval
  - I: # low speed observations in selected time interval

Location X



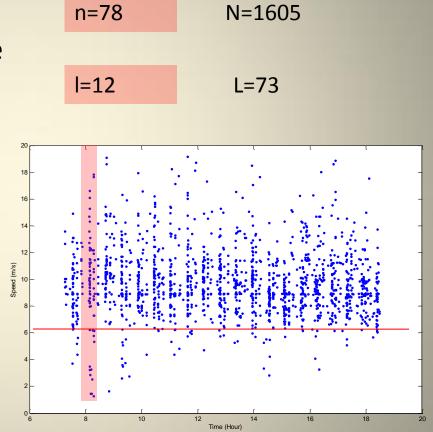
#### Illustration of Low-speed Method at Location X = 2000



# Congestion Identification at Location X: "Low-speed" Method

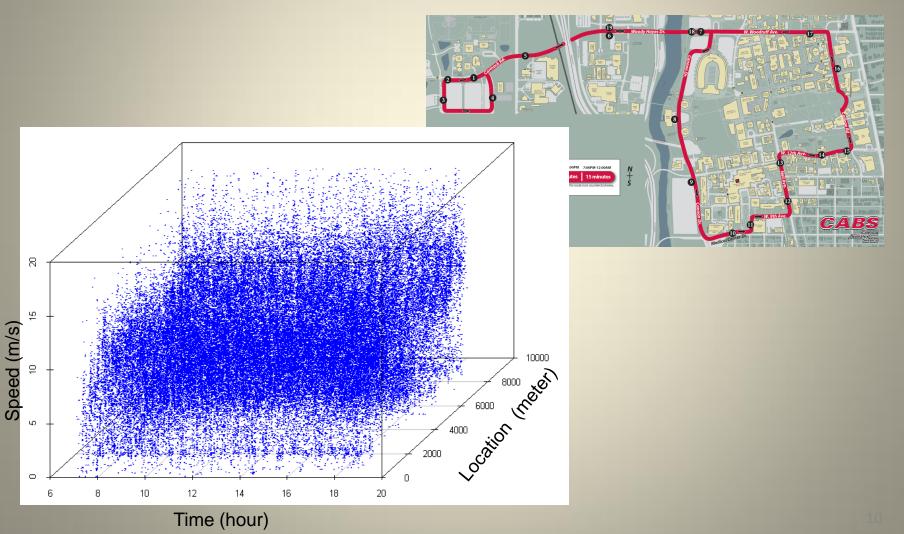
#### Given (*N,L*; *n,l*)

- Test: Is it unlikely to observe
   > I low speed observations
   in n trials when considering
   a distribution (N, L)?
  - If yes, time interval indicates recurring congestion
  - If no, time interval does not indicate recurring congestion (\*\*is not same as indicating no congestion\*\*)
- Repeat for all time intervals

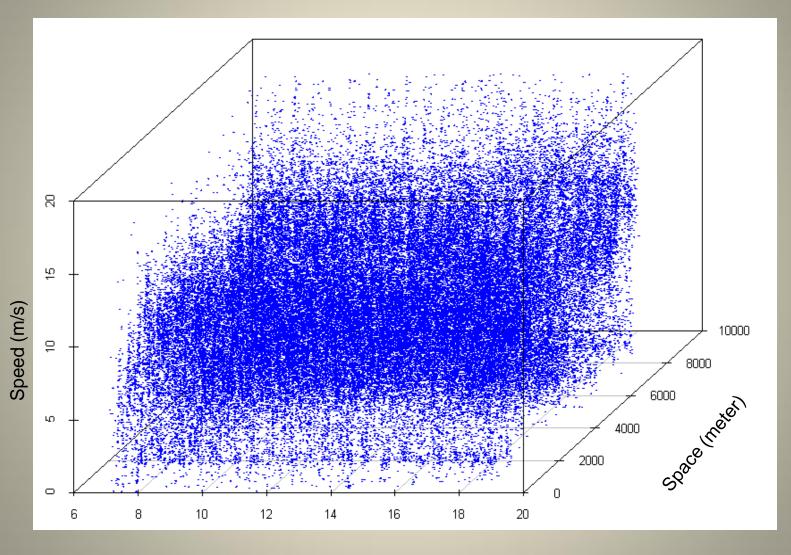


#### Illustration of Route-based Approach

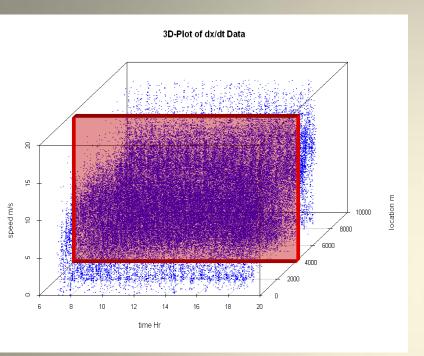
Speed data by (x,t) collected over homogeneous days



#### Illustration of Route-based Approach

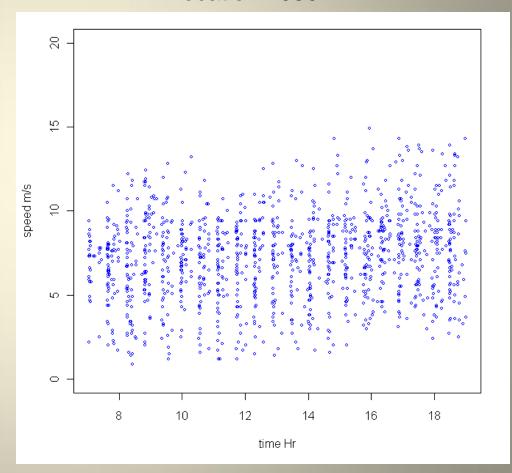


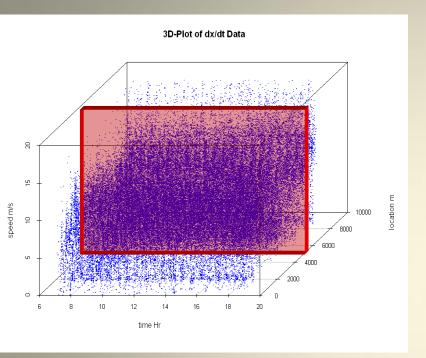
Time (hour)





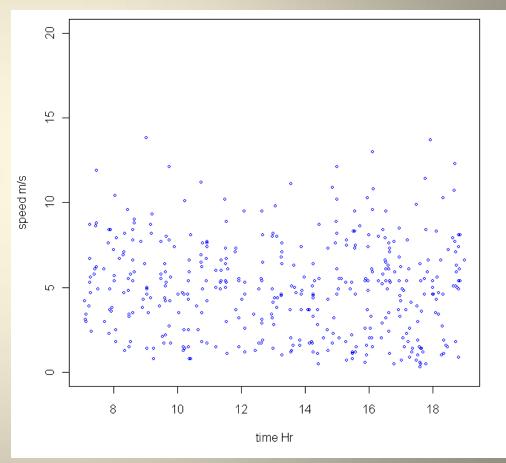
#### Location 4050 m

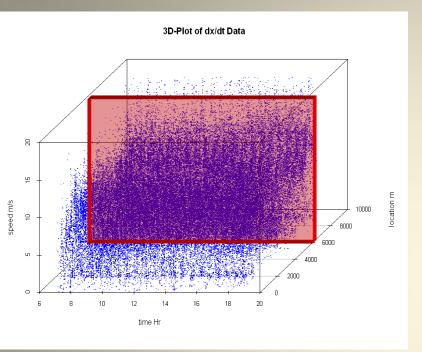






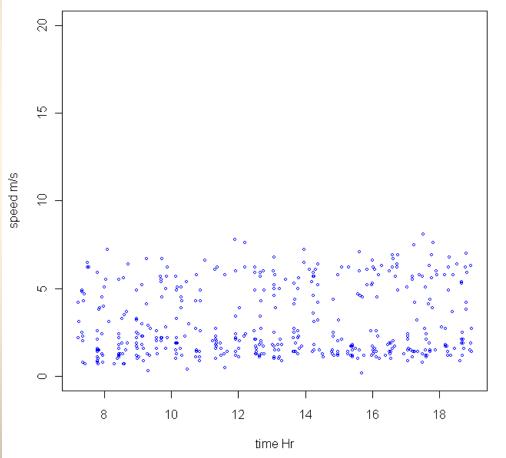
#### Location 5050 m





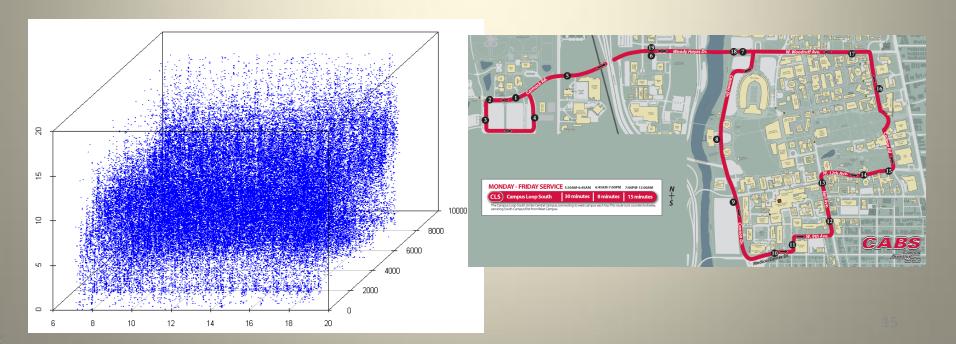


#### Location 6050 m



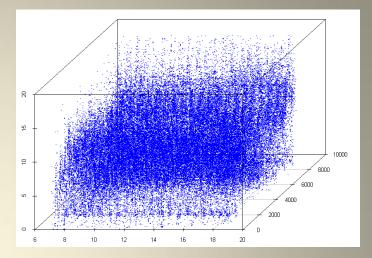
### Route-based Approach

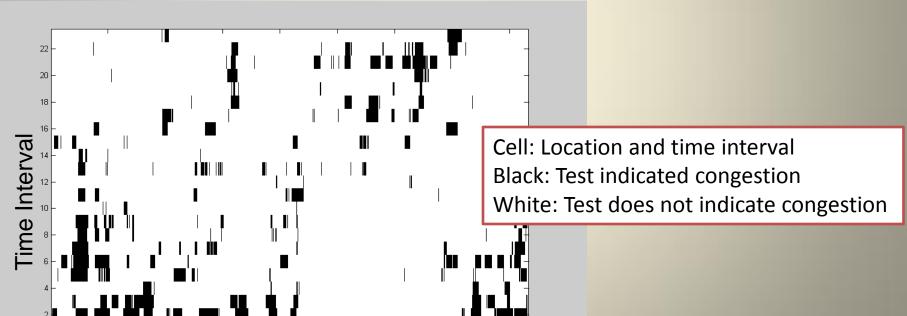
- Use speed data from multiple homogeneous days
- Start at location X = Xo near beginning of route
- Gather speed data in [X INC, X + INC]
- Apply "low speed" method to identify time intervals indicating congestion
- Move to next location  $X+\Delta$ ; repeat until end of route



#### Location x Time Binary Matrix





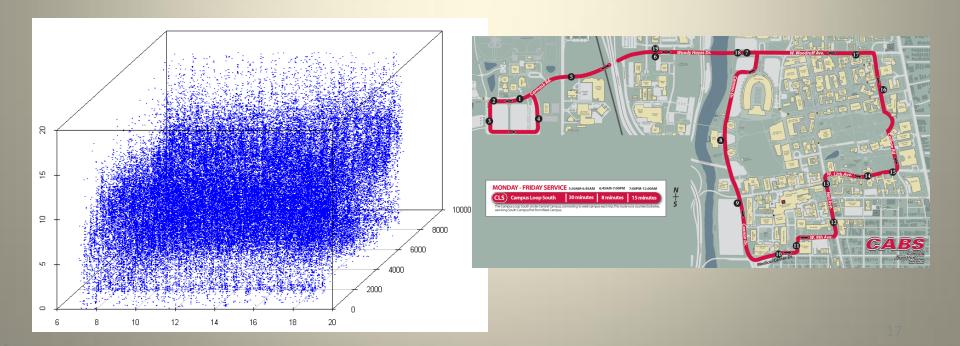


600

Location

### **Empirical Study**

- OSU CABS Campus Loop South route (8.3 km, 19 stops)
- 99,072 weekday speed observations April June, 2004
- Arc speed calculated based on two consecutive spacetime AVL points

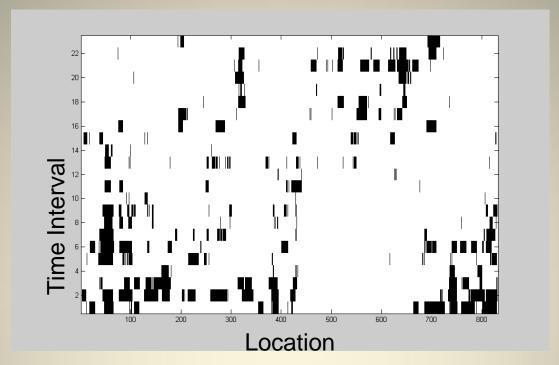


#### **Empirical Study Parameters**

- Time intervals: 23 non-overlapping, 30-min intervals from 7:00 am to 6:30 pm
- Low speed threshold at X: 0.35 x 95-percentile of speed distribution at X
- Gather speed data in [X-INC, X+INC]: INC = 50 m
   (100m space window to allow sufficient data)
- Apply "low speed" method:  $\alpha = 0.05$
- Move to  $X + \Delta$  and repeat:  $\Delta = 10$  m (cover entire route; 833 spatial windows)

Binary location-time matrix: 833 x 23 cells

## Spatial "Rule of 3 Filter"

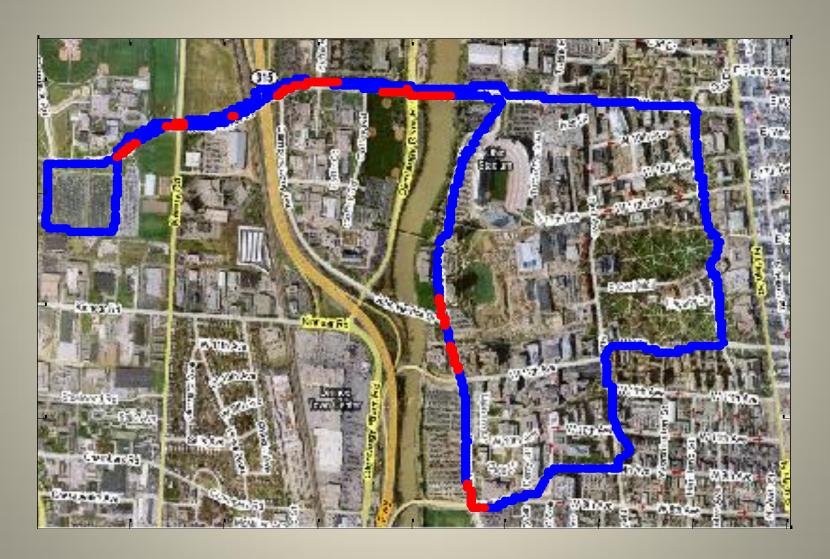


#### Motivation:

- Filter for "non-causal positives"
- Acknowledges overlapping data and spatial extent of congestion
- 3 consecutive location cells must be "positive" to be considered as indicating congestion
- 3 cells "cover" 30 meters (includes 120m of data, with 80m common to all cells, 90 m common to 2 cells)

# **Empirical Route-wide Results**

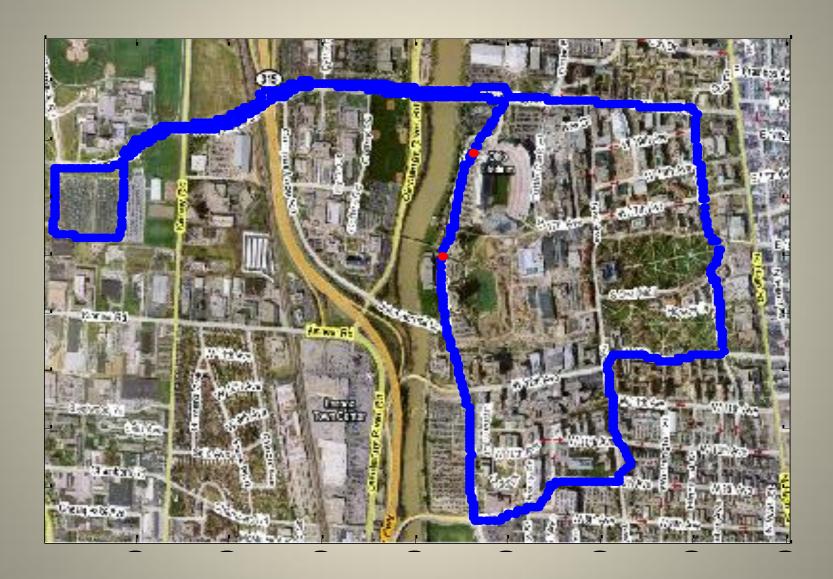
## 8:00-8:30 am



# 9:00-9:30 am



## 10:30-11:00 am

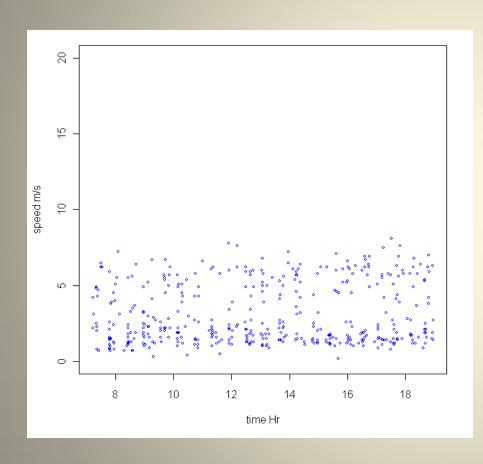


#### A Few Comments and Observations

# Congestion Identification: Classical Statistical Methods

- Mean based approach (ANOVA-based test)
  - test if the mean speed in a certain time interval is lower than the others
  - assume normality
- Median based approach (Rank sum test)
  - test if the median speed in a certain time interval is lower than the others
  - assume same shape distribution

### Deficiency of Classical Statistical Methods



Necessary assumptions violated (normality violated in over 70% of cells)

High variance because of bi-modality

# Ongoing/Future Work

- Systematic investigation of reasonableness of indicated results
- Contribution of/Integration with other (classical) methods
- Impacts of
  - space window and time interval size
  - overlapping space window
  - overlapping time intervals
  - "rule of 3"

## Ongoing/Future Work

- Use of arc speeds vs. instantaneous GPS speed
- Impact of data frequency
- Investigations with OSU CABS smart bus system
  - Replications
  - Investigations of assumptions
- Validate with other transit systems