

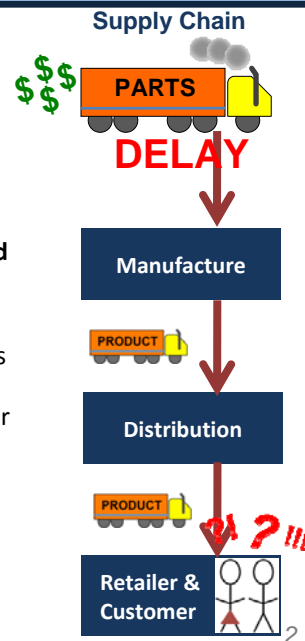
Techniques for Mining and Complementing Truck Traffic Data to Analyze the Impacts of Congestion



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Introduction

- **Projected Growth will Increase Delay and Congestion**
- **Delay and Congestion Impact Freight Industry**
 - Timely deliveries
 - Increase in emission
 - Increase in cost
 - Difficulty scheduling
- **Freight Performance Measures (FPM) not widely used**
 - Early focus on passenger vehicles (Schofield and Harrison, 2007)
 - Current measures may not meet needs of all users (e.g., loop sensors)
 - Better understanding of Freight reliability = Better planning/engineering
- **OBJECTIVE: Develop tools for measuring impacts of congestion on freight vehicles**



Outline

- Background/Context of problem
- Description of Data Sources
- Methodology
- Case Study Results
 - Recurring Congestion
 - Non-Recurring Congestion
- Conclusions



Background

Performance Measures

- Tools to evaluate current/future needs
- Travel time, speed, travel time reliability
(NCHRP, 2006)

GPS Technologies

- Travel Time Reliability & Bottlenecks
(ATRI, 2005), (ATRI, 2009), (Kamran and Hass, 2007)
(McCormack et al., 2010)

Electronic Truck Transponders (Weigh-in-Motion)

- Many freight vehicles needed
(McCormack and Hallenbeck, 2006)
- Rural areas in Oregon
(Monsere et al., in Progress at PSU)

PSU direct access to Loop Sensor data from Oregon DOT

- Recurring, Non-Recurring studies
(Monsere et al., 2006), (Bertini et al., 2005),
(Wieczorek et al., 2009)



Background cont.

Unique Contributions of this Work

•Combining Multiple Data Sources

- GPS data from commercial trucks
- Loop sensor data (Oregon DOT)
- Incident data (ODOT ATMS)

•Create unbiased FPM

- Separating trucks experiencing congestion vs rest/refuel
- Develop alternatives to current PM

•Useful to Public Agencies

•Oregon Freight Data Mart

- in Development at PSU (Figliozzi and Tufte, 2009)
- Web-based platform
- Available to commercial & private carries



Description of Data Sources Available

•PORTAL (SEE: <http://portal.its.pdx.edu>) Northbound I-5 Loop Stations

- Loop Sensor Data from ODOT

•Incident Data

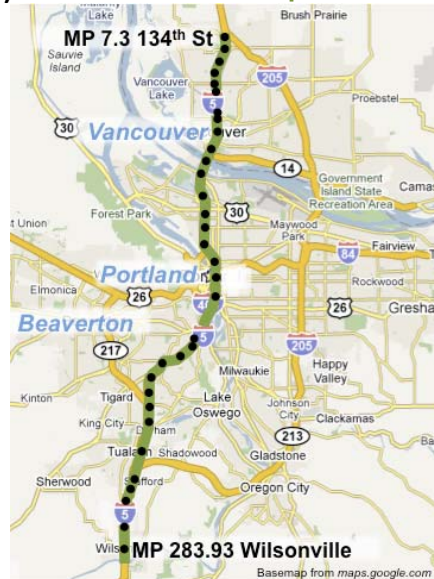
- Type, Severity, Approximate location, Start/End time (duration)

•GPS Truck Data

- TruckID number, Date, Time
- Position (Latitude/Longitude)

•Data Challenges

- No common gap time btw readings
- Multiple trips on same day
- Different truck types (travel behavior)



Methodology

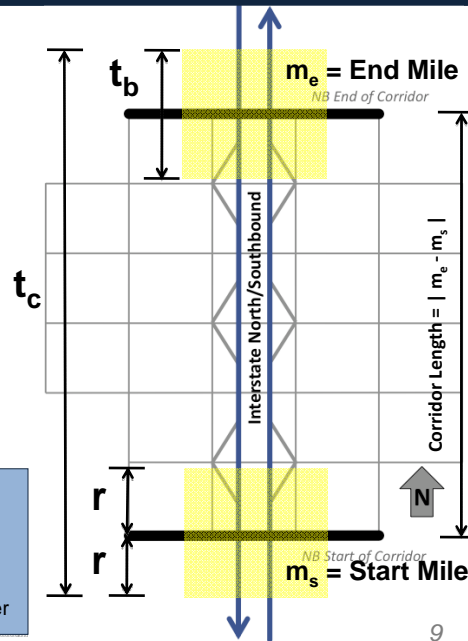
•Purpose of Filter: To Identify Through Trucks for analysis

•Two Step Process:

- Filter Process 1: Matching GPS Readings to Identify Potential Through Trucks
- Filter Process 2: Comparison to PORTAL Average Travel Times
- Integrates available data sets and ensures no stops midway for rest/refuel

Filter Parameters

- m_s = Start Milepost
- m_e = End Milepost
- r = Buffer radius
- t_b = Threshold to clear buffer
- t_c = Threshold to clear corridor and buffer

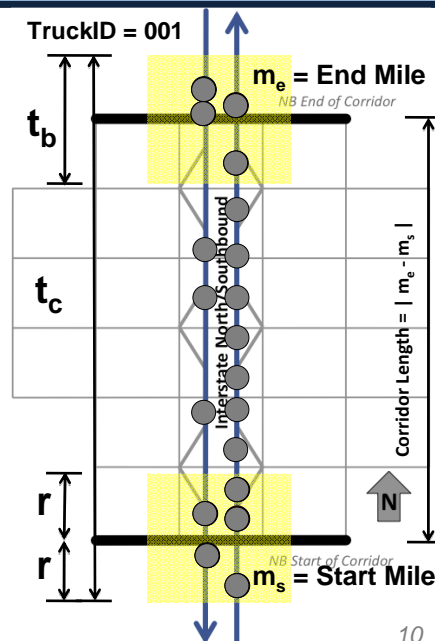


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Methodology cont.

Filter Process 1: Matching GPS Readings to Identify Potential Through Trucks

1. Obtain milepost measures using ArcGIS
2. Determine m_s and m_e
3. Look at points falling in buffer ranges
4. Distinguish individual trips by each truck using time thresholds t_c and t_b and identify the "start" and "end" points of each trip
5. Match readings in "start" buffer to downstream reading in "end" buffer occurring within t_c

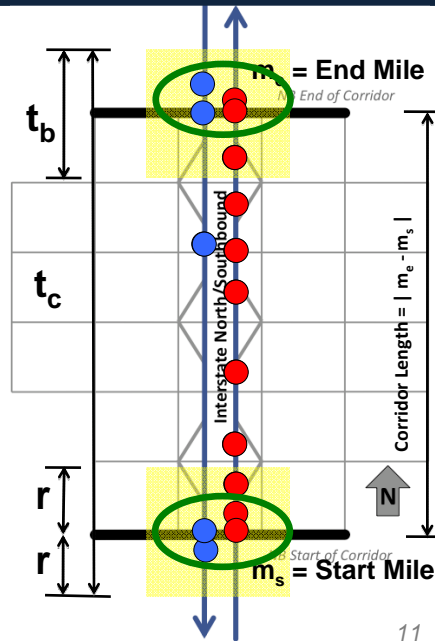


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Methodology cont.

Filter Process 1: Matching GPS Readings to Identify Potential Through Trucks

- Find all intermediate readings for a truck ID falling between the trip “start” and “end” readings
- Adjust the “start” and “end” reading timestamp and milepost to begin at m_s and m_e using speeds obtained from the next closest reading
- Obtain the travel time and speed through the corridor, and identify trip direction using milepost data



Methodology cont.

Filter Process 2: Comparison to PORTAL Average Travel Times

- Data sorted by the “start” reading timestamp into time bins of 15 minute intervals.
- Deviation Index* is calculated using the PORTAL:

For a 15 minute time bin t ,
Then the Deviation Index g_k is

defined as

$$g_k = |a_t - T_k| / \sigma_t$$

Where:

a_t = PORTAL average travel time at time bin

t
of travel time

σ_t = PORTAL day-to-day standard deviation

trip k

T_k = the corridor average travel time for truck

- Truck trip is too far from the expected average if:

$$g_k > m * \sigma_t$$

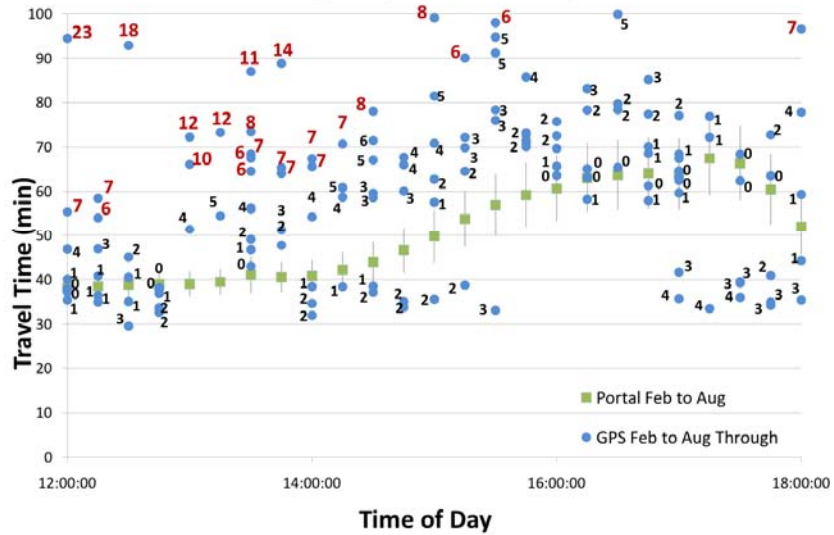
Where:

Methodology cont.

Filter Process 2: Comparison to PORTAL Average Travel Time

Through Truck vs PORTAL Corridor Average Travel Time

*Results following filter process 1, showing Deviation Index

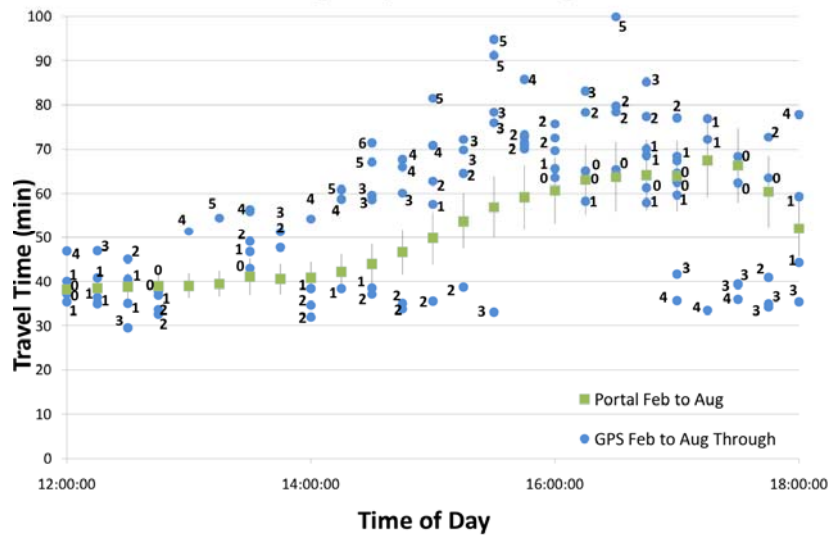


Methodology cont.

Filter Process 2: Comparison to PORTAL Average Travel Time

Through Truck vs PORTAL Corridor Average Travel Time

*Results following filter process 2, showing Deviation Index

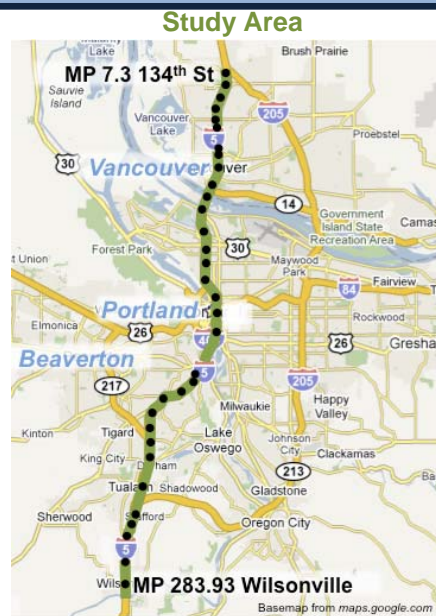


Recurring Congestion Analysis

Case Study: Recurring Congestion

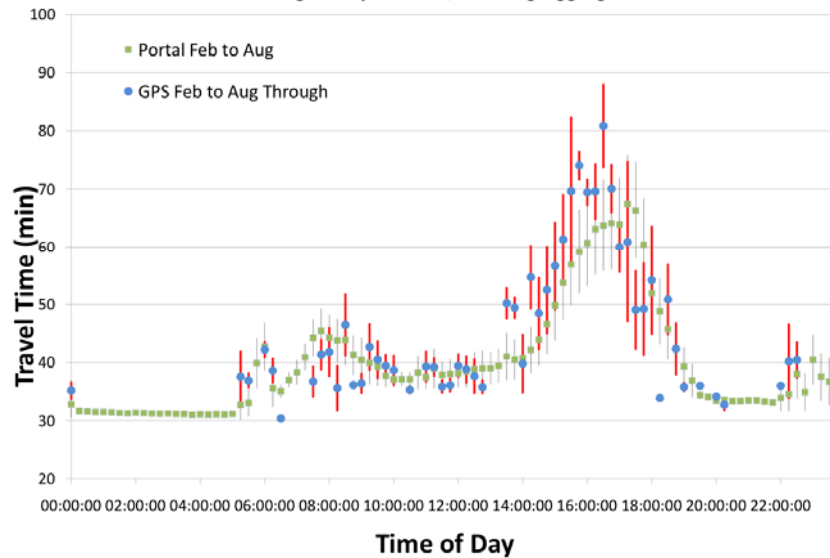
Recurring Congestion

- I-5 NB
- Wilsonville, OR to Vancouver, WA
- 31.75 miles
- Feb-Aug, 2007 (weekdays)



Results: Recurring Congestion

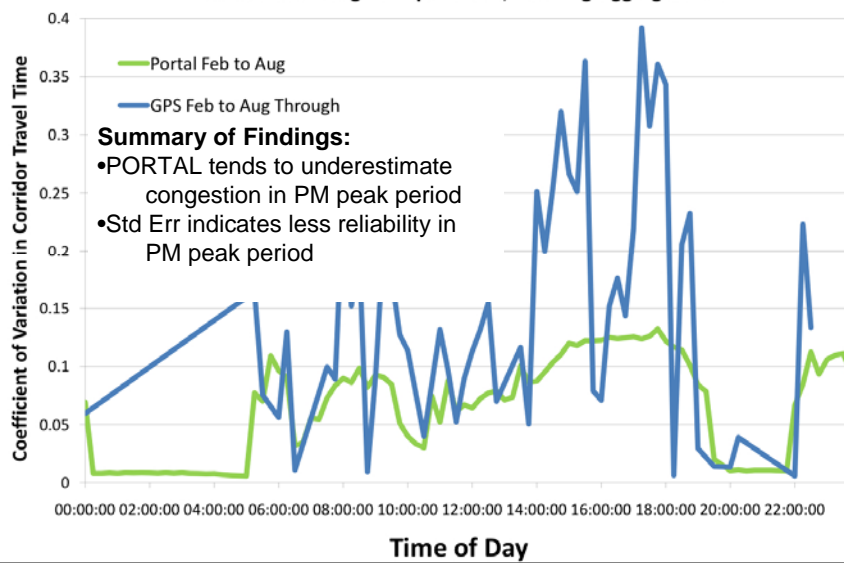
Through Truck vs PORTAL Corridor Average Travel Time
 *Results following filter process 2, showing Aggregated Data and Std Err



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Results: Recurring Congestion

Through Truck vs PORTAL CV in Corridor Average Travel Time
 *Results following filter process 2, showing Aggregated Data



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Non-Recurring Congestion Analysis

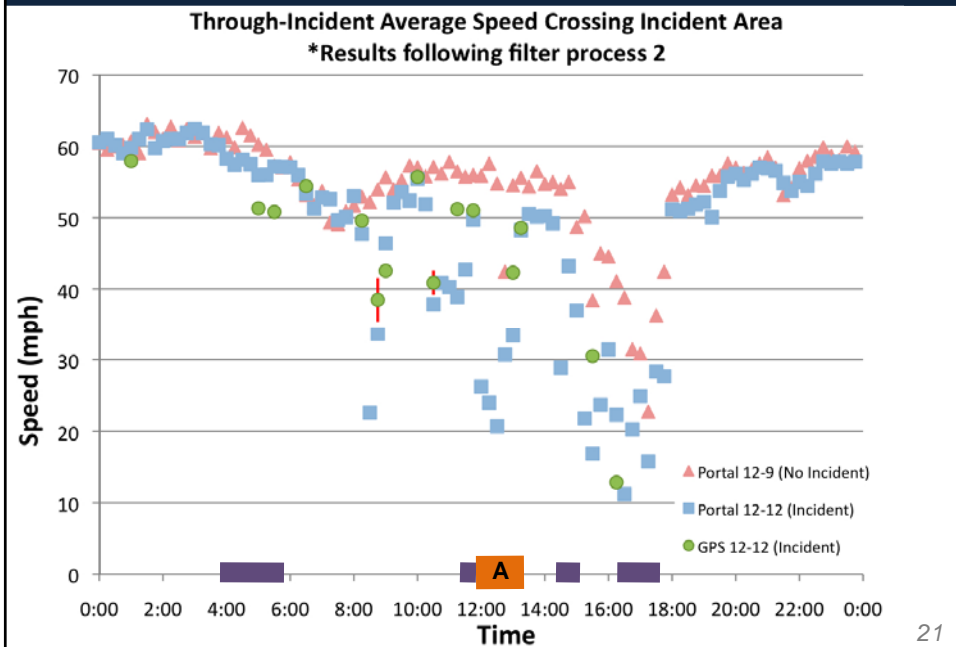
Case Study: Non-Recurring Congestion

Non-Recurring Congestion

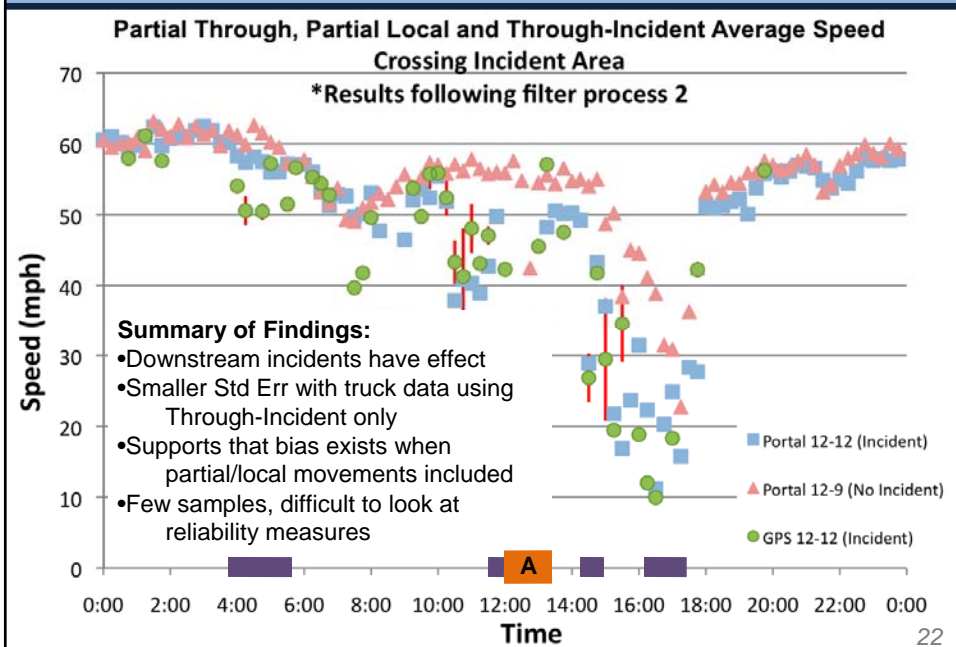
- Similar Methodology,
- *5 mile segment near incident "A"*
- Incidents Downstream of "A"



Results: Non-Recurring Congestion

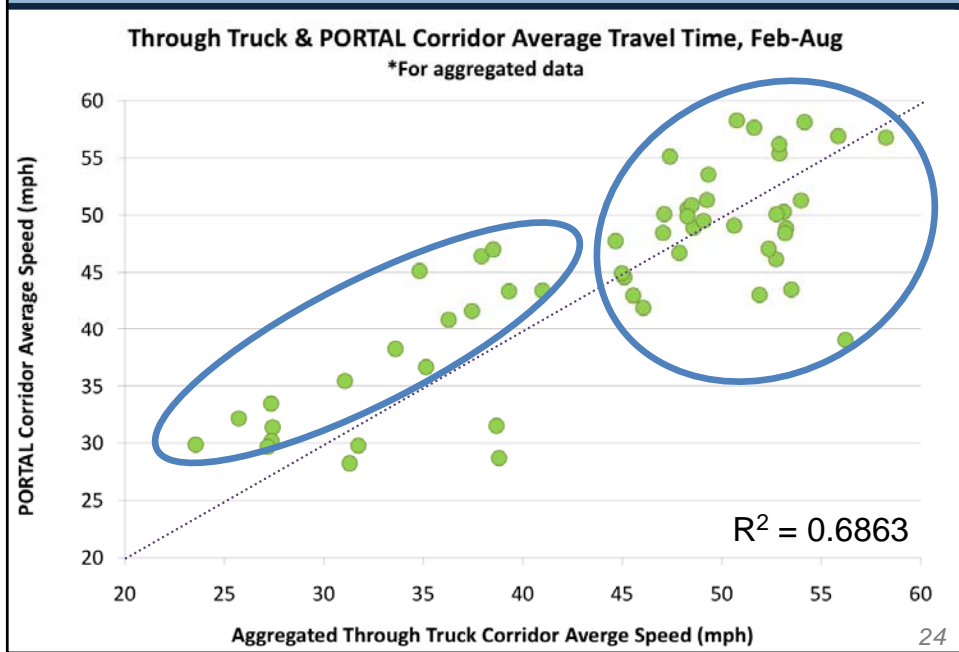


Results: Non-Recurring Congestion



Summary of Findings

Truck Data vs PORTAL Comparison



Research Next Steps

Work to-date is a good foundation...

- Improving complexity of programming
- Using PORTAL data dynamically
- Statistical and Sensitivity Testing
- Quantifying Costs, Emissions and Health Impacts



Conclusions

- Integrated GPS, loop sensor and incident data
- New methodology to identify local and through trucks
 - Remove bias of trucks resting/refueling
 - Through trucks best indicator of congestion
 - Indications that loop sensor data may underestimate congestion in PM peak
- Performance data useful to public agencies
 - Expand to look at bottlenecks
 - Study greater time periods
 - Local/freeway transitions
- Could be incorporated into Oregon Freight Data Mart or other web based platform

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Questions??



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