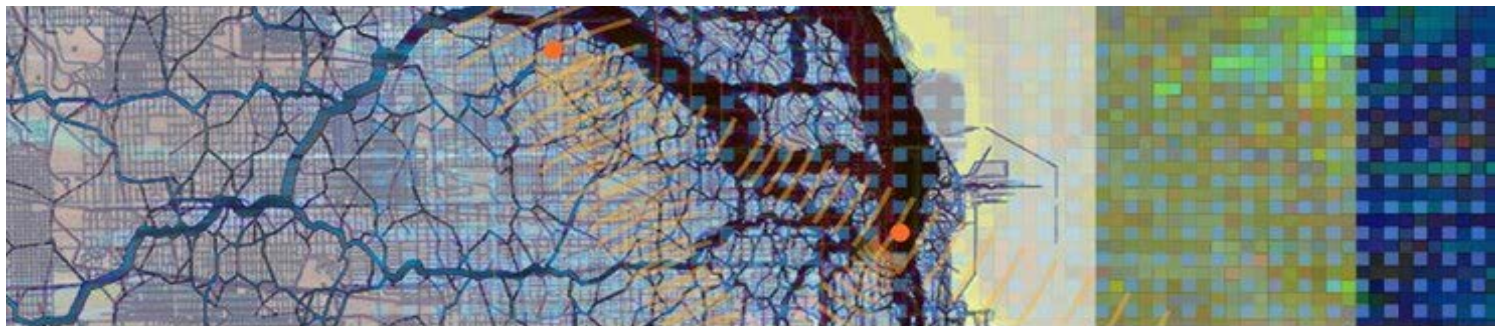




Leveraging Sensor Technologies for Smarter Mobility

Hani Mahmassani



Smarter Cities/Smarter Mobility

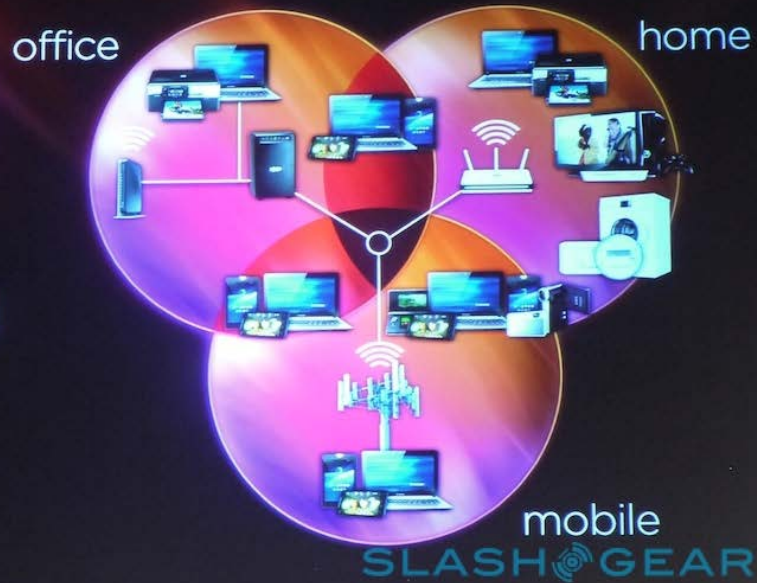
Outline

1. THE CONTEXT: MOBILITY AS PROCESS IN CONNECTED SYSTEMS
2. SYSTEM INTELLIGENCE THROUGH PREDICTIVE ANALYTICS
3. PREDICTION AND REAL-TIME TRAVELER INFORMATION
4. PREDICTIVE CONTROL: PRICING
5. WEATHER-RELATED TRAFFIC MANAGEMENT
6. LOGISTICS OPERATIONS IN CONGESTED URBAN ENVIRONMENTS
7. TAKEAWATS

I.

The Context:
Mobility as *Process*
in Connected Systems

the internet
of everything



THE **USER** IS AT THE
CENTER OF THIS WEB
OF **CONNECTIVITY** AND
“ALWAYS AWARE”
SYSTEMS AND DEVICES



Source: Qualcomm

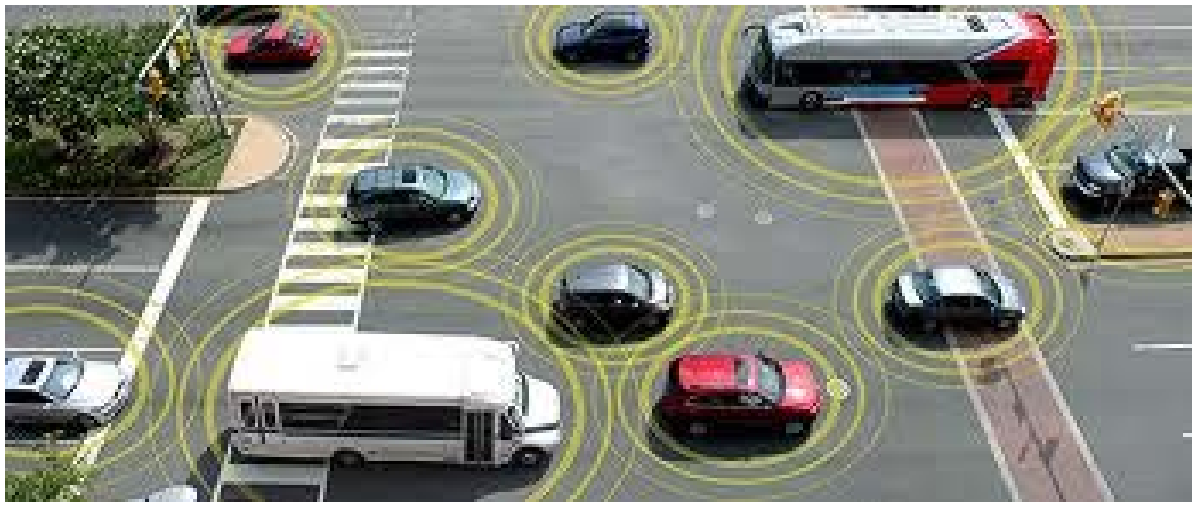
WHY IS THIS RELEVANT TO TRANSPORTATION?

WHY IS THIS RELEVANT TO TRANSPORTATION?



SEAMLESS CONNECTIVITY

TRANSPORTATION DELIVERS PHYSICAL MOBILITY IN A VIRTUALLY CONNECTED MOBILE ENVIRONMENT



Everybody is talking about it

The real value of the **Internet of Everything** lies in the value of connections among **people, process, data, and things**, not simply in the sheer number of things that are connected.

When your car becomes connected to the **Internet of Everything**...



...**more numerous, valuable, and relevant connections** with other cars, stop signs, your home, and even the road itself will make your driving experience safer, more fun and informed, and even more efficient.

It's the **connections** that matter most.

The Internet of
EVERYTHING

#InternetofEverything
#IoE



KEY TECHNOLOGY ENABLERS



Canon CMOS sensor, EOS 400D

Sony CCD sensor, Nikon D2X

Foveon X3 sensor, Sigma SD14



NETWORKS

Peer to peer

Wide area wireless

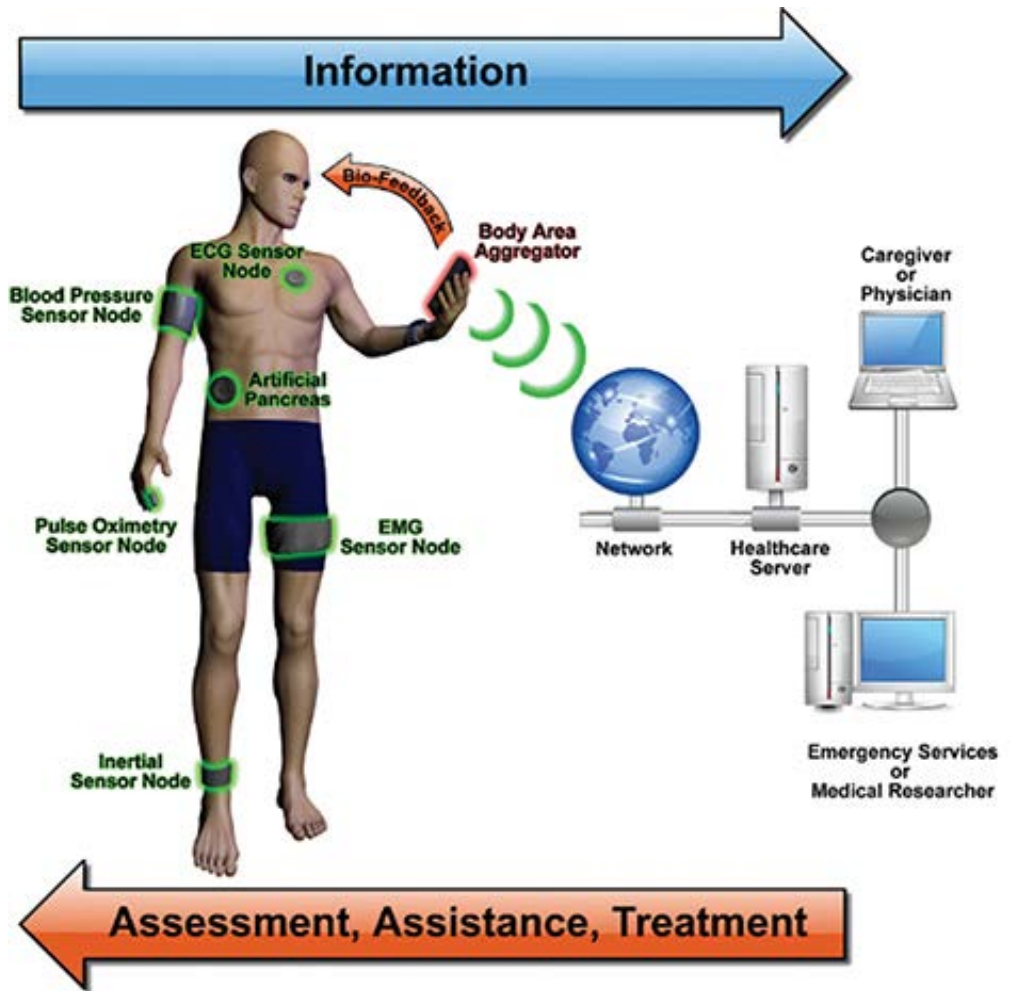
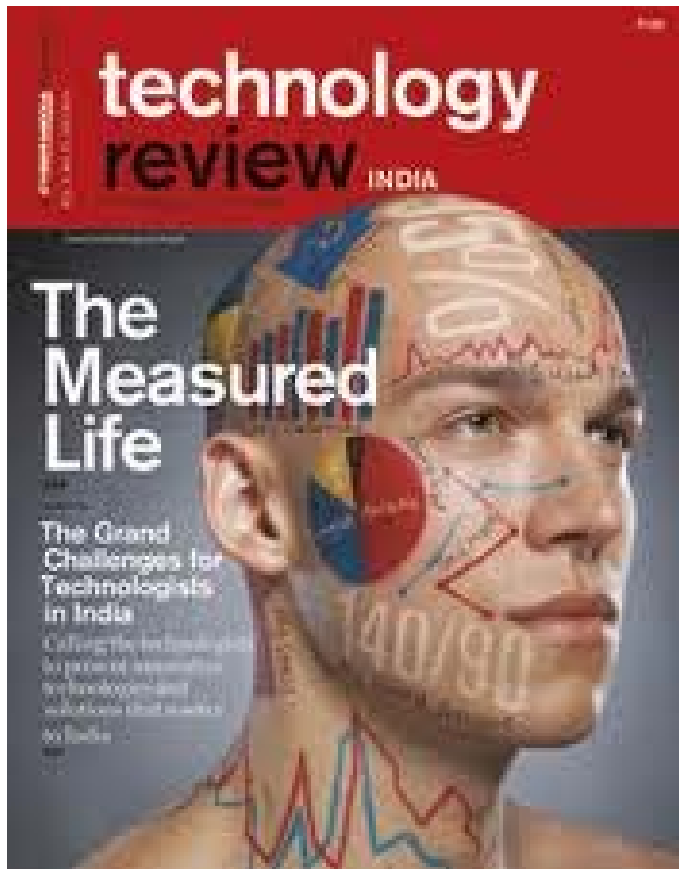
Backbone

SERVERS AND DATA STORAGE



Tracking your phone, Family & Friends

NOT LIMITED TO CARS AND OBJECTS



II.

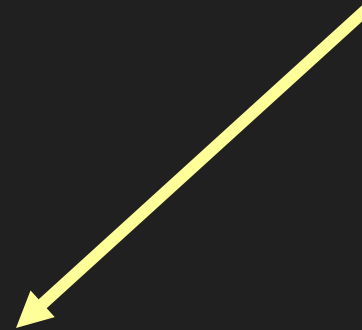
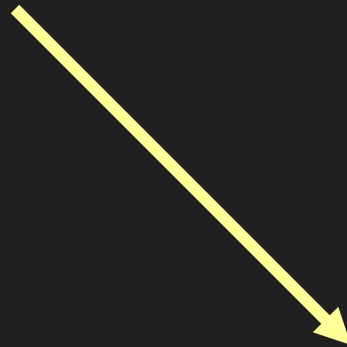
System Intelligence through Predictive Analytics

Mobile units +
wireless internet:

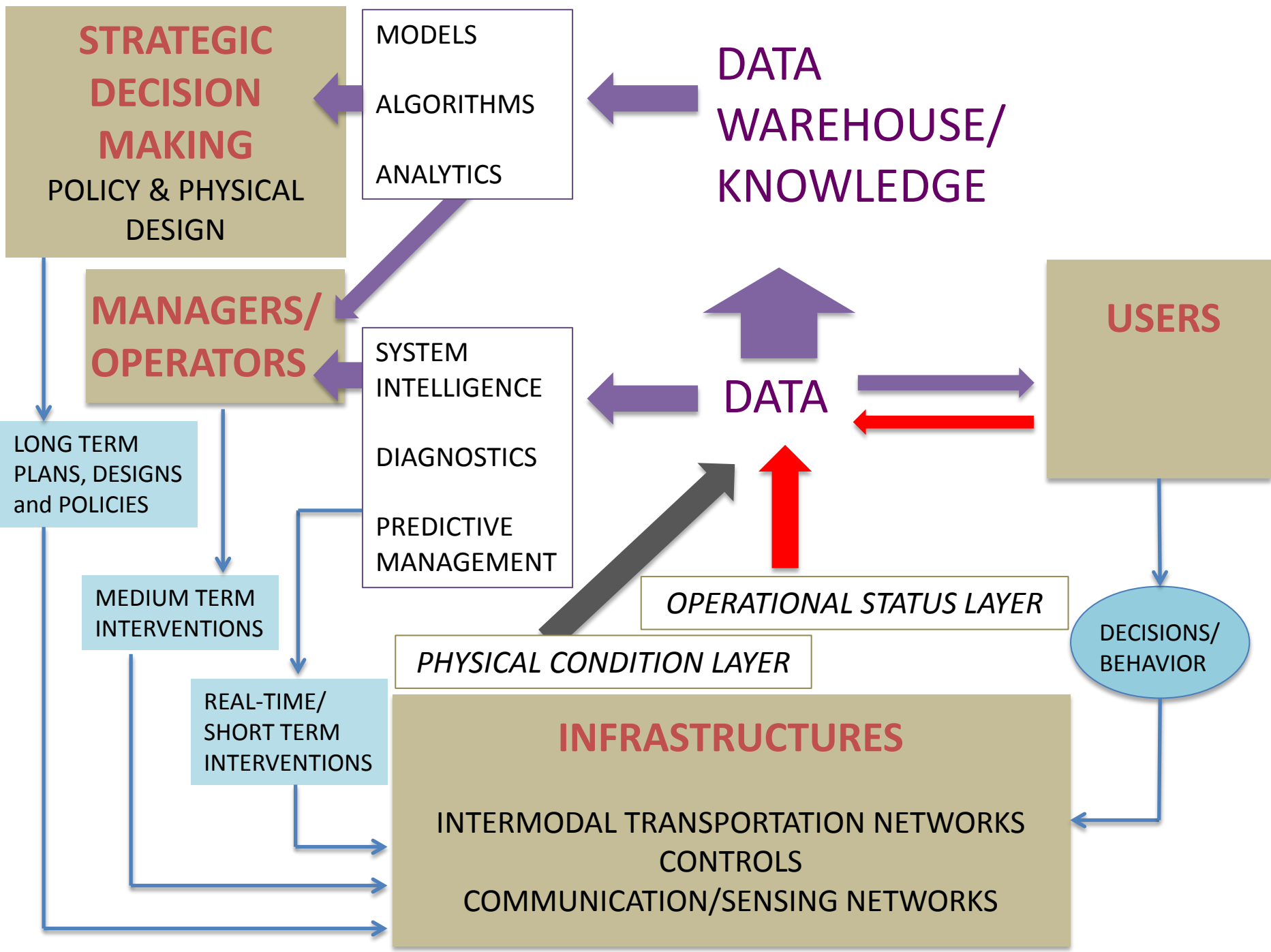
Provides particle
(user-centric)
views of system

Inexpensive
wireless sensors

Provides view from
perspective of
infrastructure or
fixed assets



REAL-TIME INFORMATION



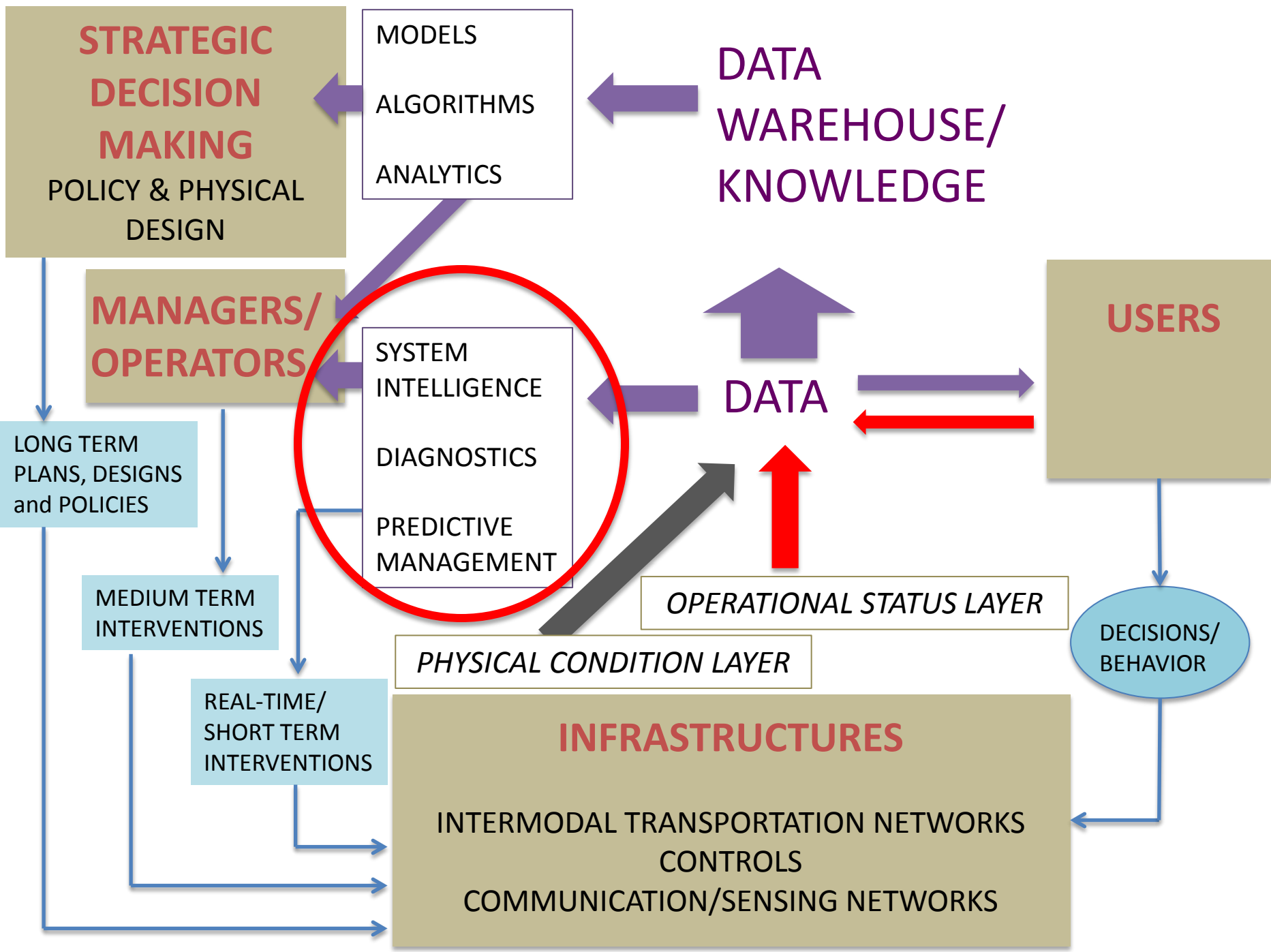
ACID TEST

How is more data allowing me to

Do things differently (better– faster, cheaper, safer, higher impact, customer-pleasing...)

Do different things (grow activities, revenue, improve image, employee retention...)

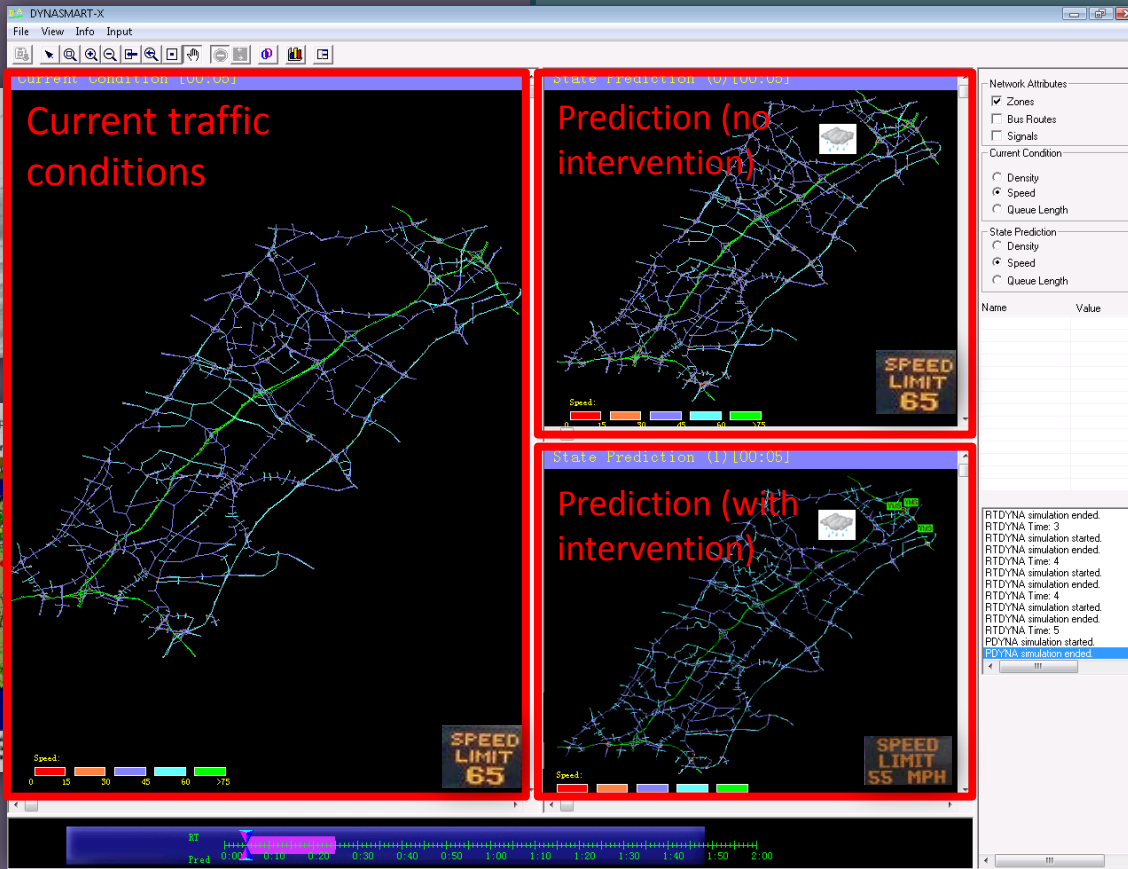




Traffic Estimation and Prediction System (TrEPS)

DYNASMART-X

ESTIMATION



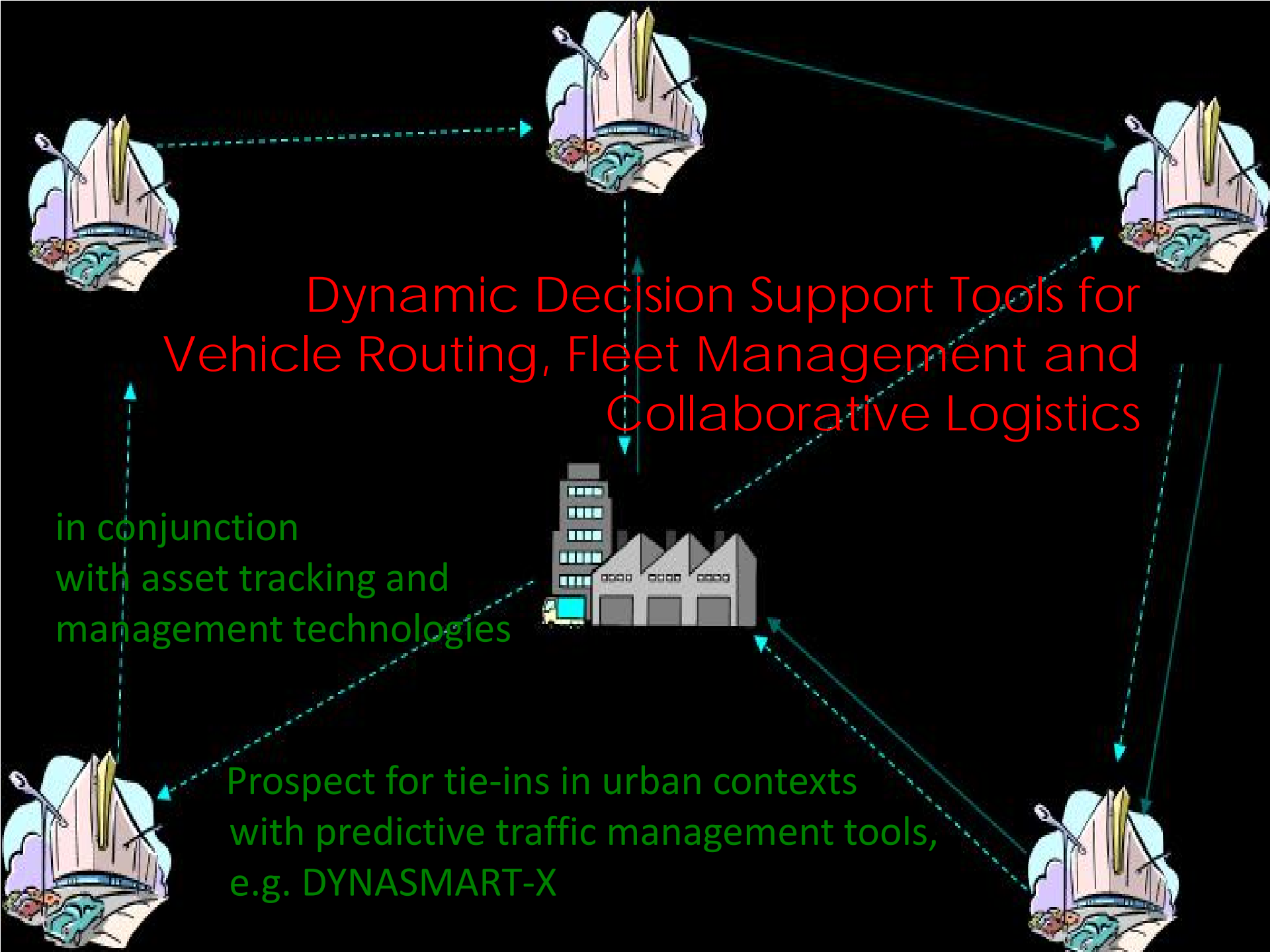
The screenshot displays the DYNASMART-X interface with three main simulation panels:

- Current traffic conditions:** A network map showing traffic flow with a speed limit of 65.
- Prediction (no intervention):** A network map showing predicted traffic flow with a speed limit of 65.
- Prediction (with intervention):** A network map showing predicted traffic flow with a speed limit of 55 MPH.

Additional interface elements include a menu bar (File, View, Info, Input), a toolbar, a status bar at the bottom, and a right-hand sidebar with settings for Network Attributes, Current Condition, State Prediction, and a Name/Value table.

PREDICTION





Dynamic Decision Support Tools for Vehicle Routing, Fleet Management and Collaborative Logistics

in conjunction
with asset tracking and
management technologies

Prospect for tie-ins in urban contexts
with predictive traffic management tools,
e.g. DYNASMART-X

Descriptive conditions;
PREDICTION

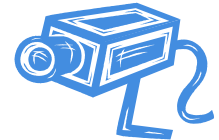
Anticipatory
information
control
pricing

Traffic
Management
Center

Guidance (VMS,
Info to users),
Signal control
Prices

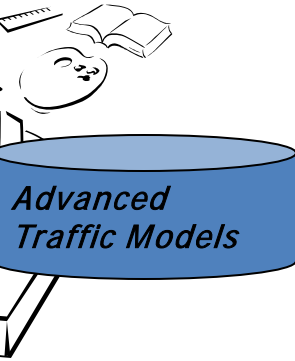


Network



Sensor
systems

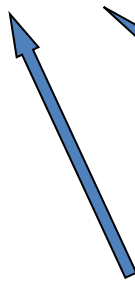
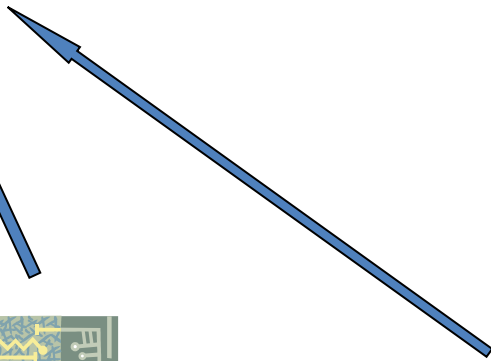
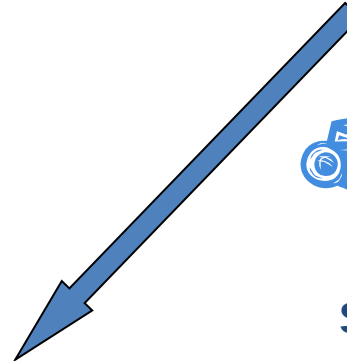
- Fundamental core
- *Flow Models*
 - *Behavior*
 - *Algorithms*



Historical
data



Real-time
traffic
data



Real-time Traffic
Estimation /
Prediction System

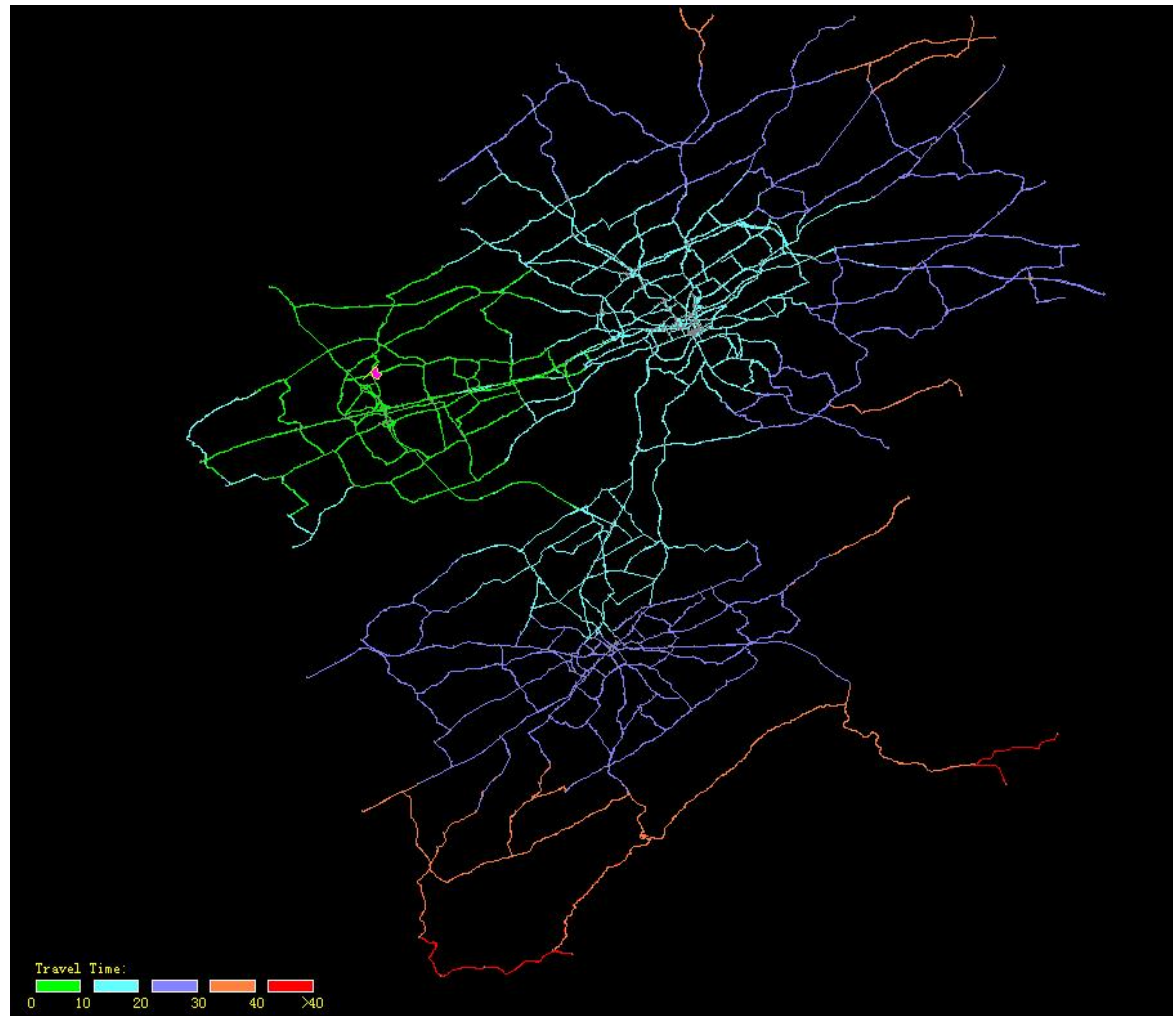
III.

Prediction and Real-time Traveler Information

Consistent Anticipatory Travel Time Information



(Reference: Dong and Mahmassani, 2010)



Consistent Anticipatory Travel Time Information

WHAT WE KNOW

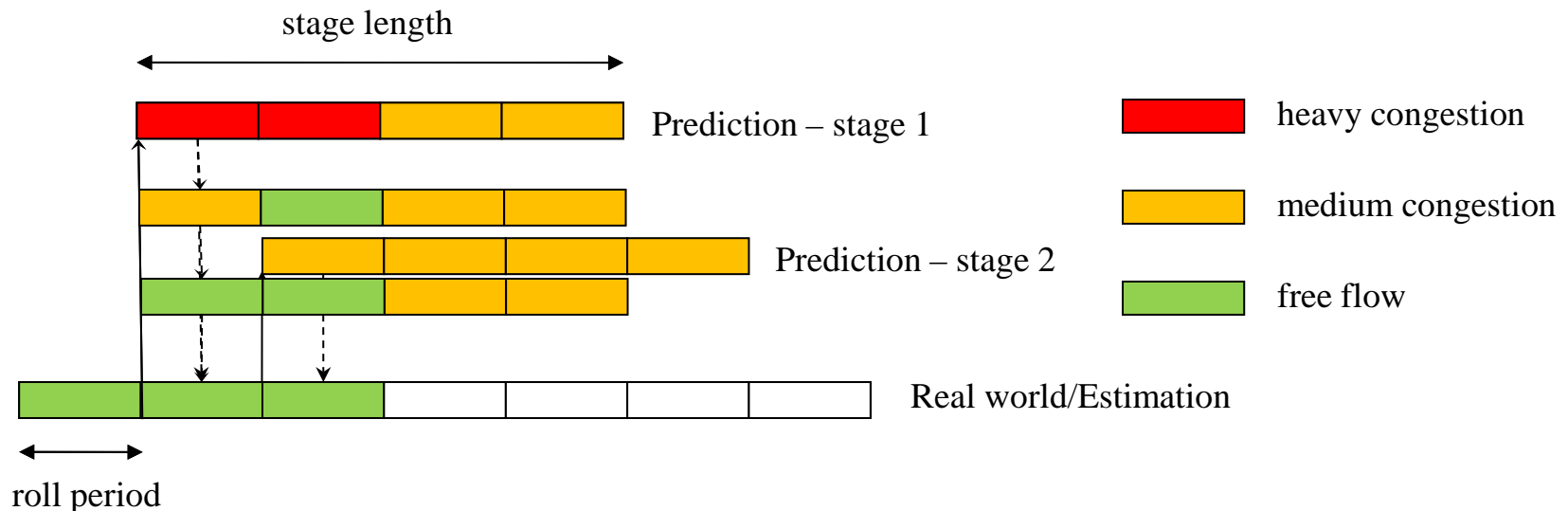
Information on currently prevailing conditions may not be effective: *overreaction, time lags, stochastic and dynamic variation*

Anticipatory information effective, but poses three challenges:

- capturing user responses to provided information:
CONSISTENCY
- users care about reliability of information
- computation for large networks

Closed-loop Rolling Horizon Framework

- RH approach is a practical method for generating and implementing solutions to dynamic programming problems.
- Closed-loop structure allows the control policies obtained in traffic prediction model to be implemented in real world and transferred to state estimation model.



The Test Bed Network : Irvine



■ Network

- Freeways I-405, I-5, state highway 133
- 326 nodes
- 626 links
- 61 TAZs
- 57 road detectors

■ Demand

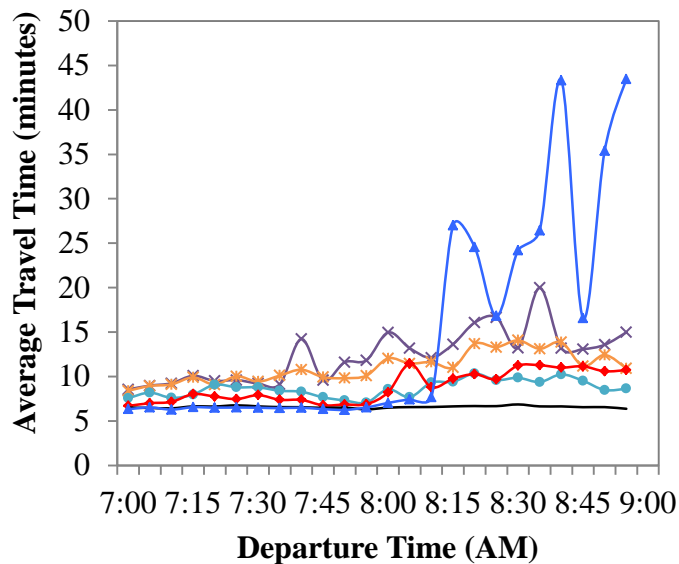
- Two hours morning peak
- 15min warm-up period + 45 min clearance time

■ Parameters

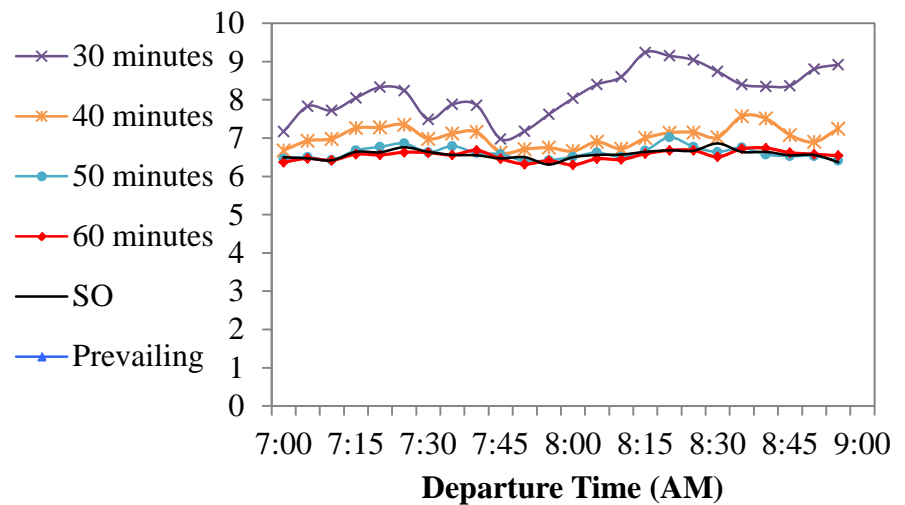
- Roll period: 5 minutes
- Prediction horizon varying from 30 to 60 minutes

Sensitivity to *Prediction Horizon*

look-ahead algorithm

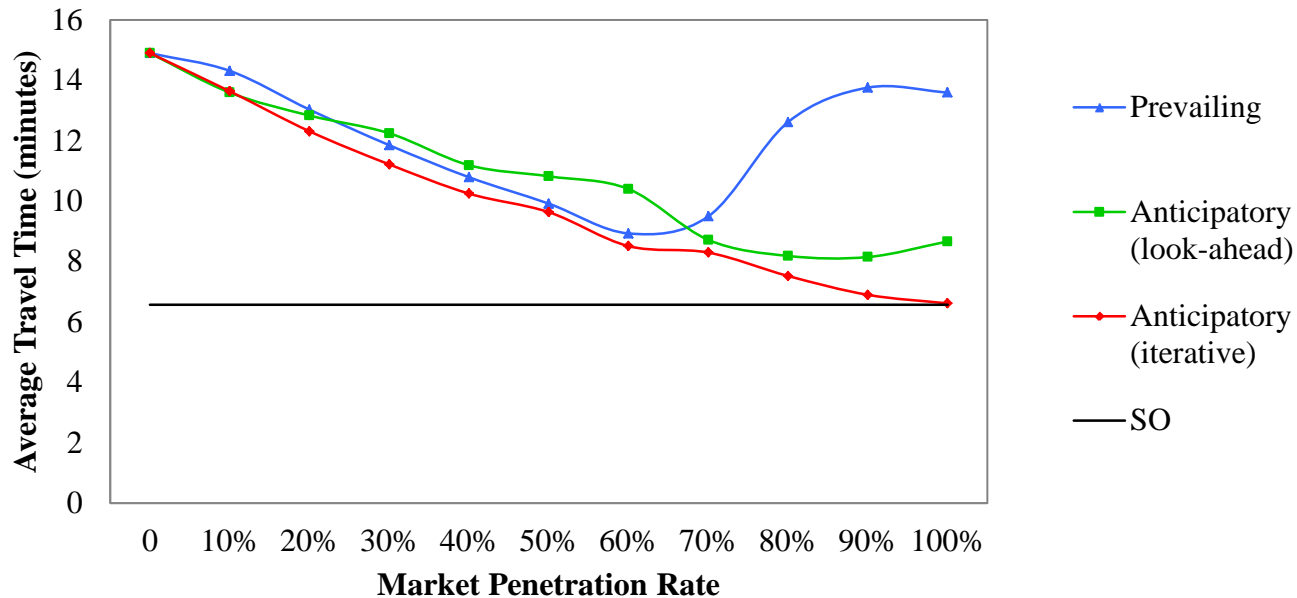


recursive averaging algorithm



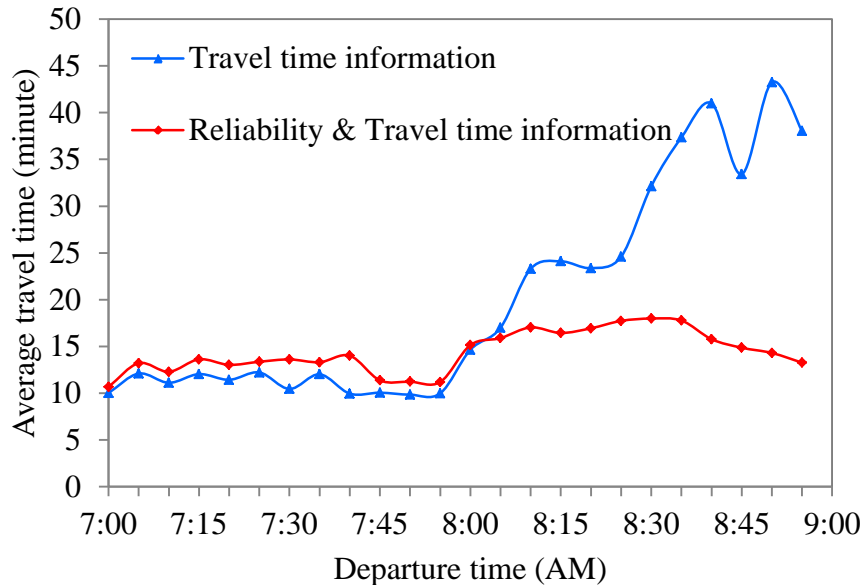
- **Anticipatory** information works better than **prevailing** information
- Longer prediction horizon provides better performance

Sensitivity to *Market Penetration Rate*

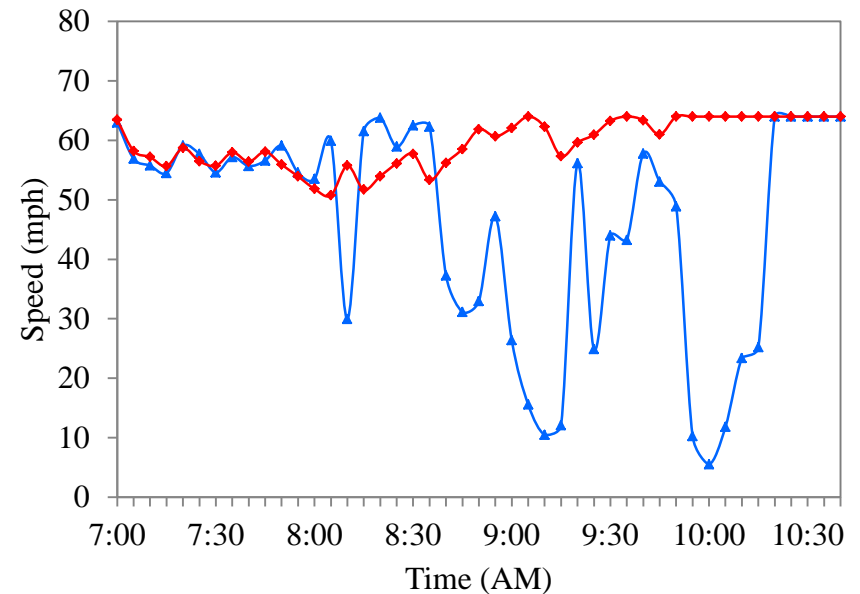


- Provision of **anticipatory travel time information** improve the overall network performance
- Solve the overreaction problem caused by providing **prevailing (instantaneous) information**

Average Travel Time



Freeway Speed



Scenarios:

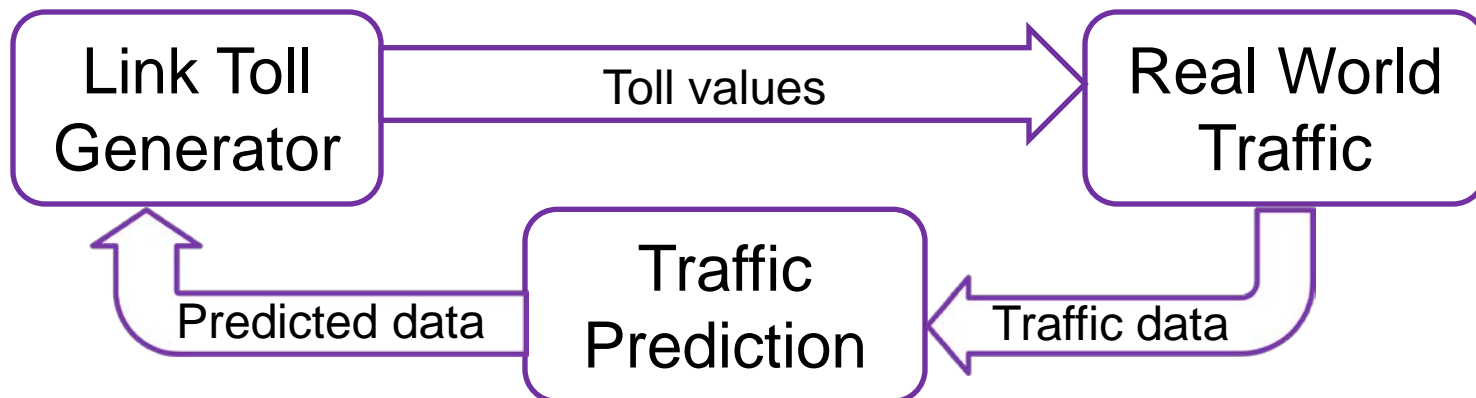
- only anticipatory travel time information is provided
- both anticipatory travel time and reliability information provided
- Significant time savings are observed when **travel reliability information** is provided in addition to **travel time information**
- Providing travel reliability information contributes to *delaying the onset of breakdown* and alleviating its extent, with higher and more stable flow indicating an increase in freeway's utilization

IV.

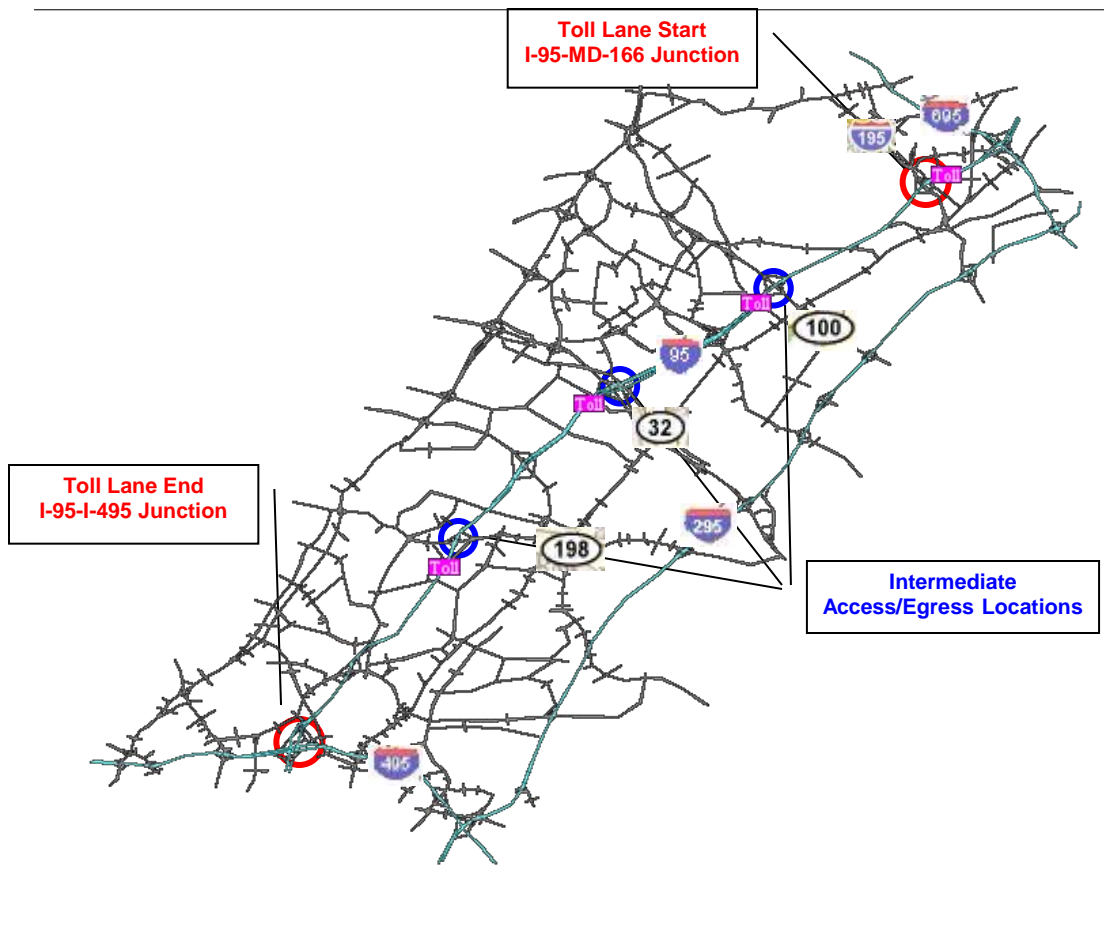
Predictive Control: Pricing

Anticipatory Pricing Strategy for Managed Lane Operation

- What differentiates anticipatory from reactive pricing?
 - Network state prediction
 - Use **predicted** traffic conditions
 - Calculate link toll within the **prediction horizon** and implement it in real time



The Test Bed Network: CHART



- I-95 corridor between Washington, DC and Baltimore, MD, US
- 2 toll lanes
- 2241 nodes
- 3459 links
- 111 TAZ zones
- 2 hours morning peak demand

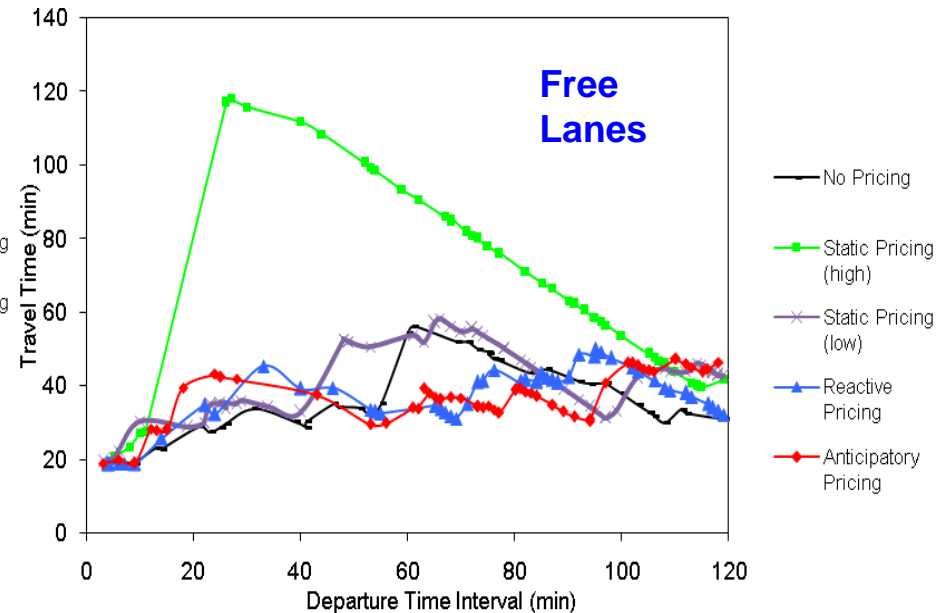
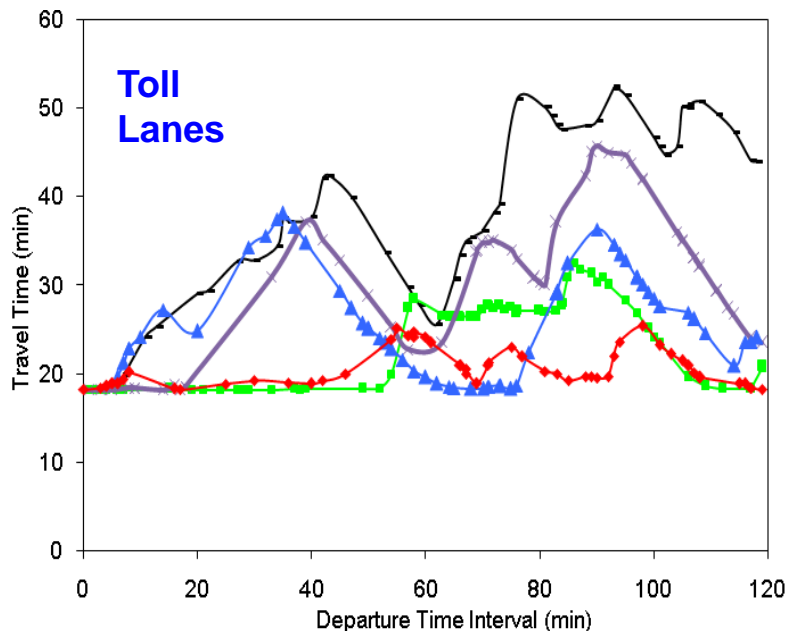
Pricing Strategies Compared

- No pricing (base case)
- Static pricing
 - Predetermine the time-varying link tolls based on the historical information
- Reactive pricing
 - Set time-varying link tolls based on prevailing traffic conditions
- Anticipatory pricing
 - Set time-dependent link tolls based on predicted traffic conditions

OBJECTIVE: AVOID BREAKDOWN– optimize throughput, reliability, under economically efficient allocation

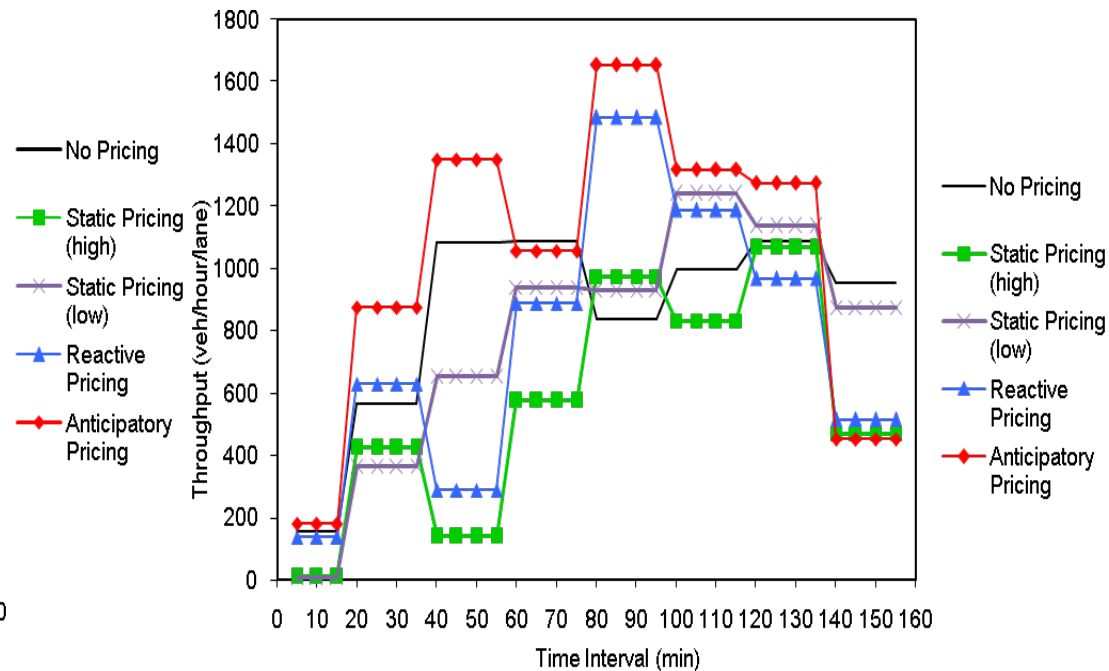
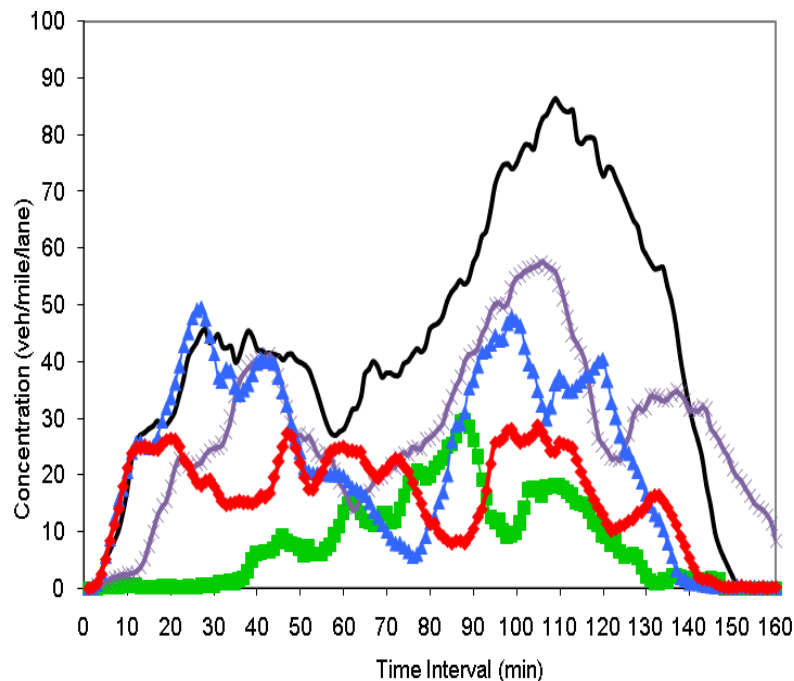
Illustrative Results – Travel Time

- Warm-up period: increase in travel time at the beginning
- With the anticipatory pricing strategy, the travel times become steady after 1 hour (free flow condition)
- Static pricing strategy provides free flow condition on the toll lanes, but reduces the LOS on the alternative freeway lanes



Illustrative Results – Traffic Measures

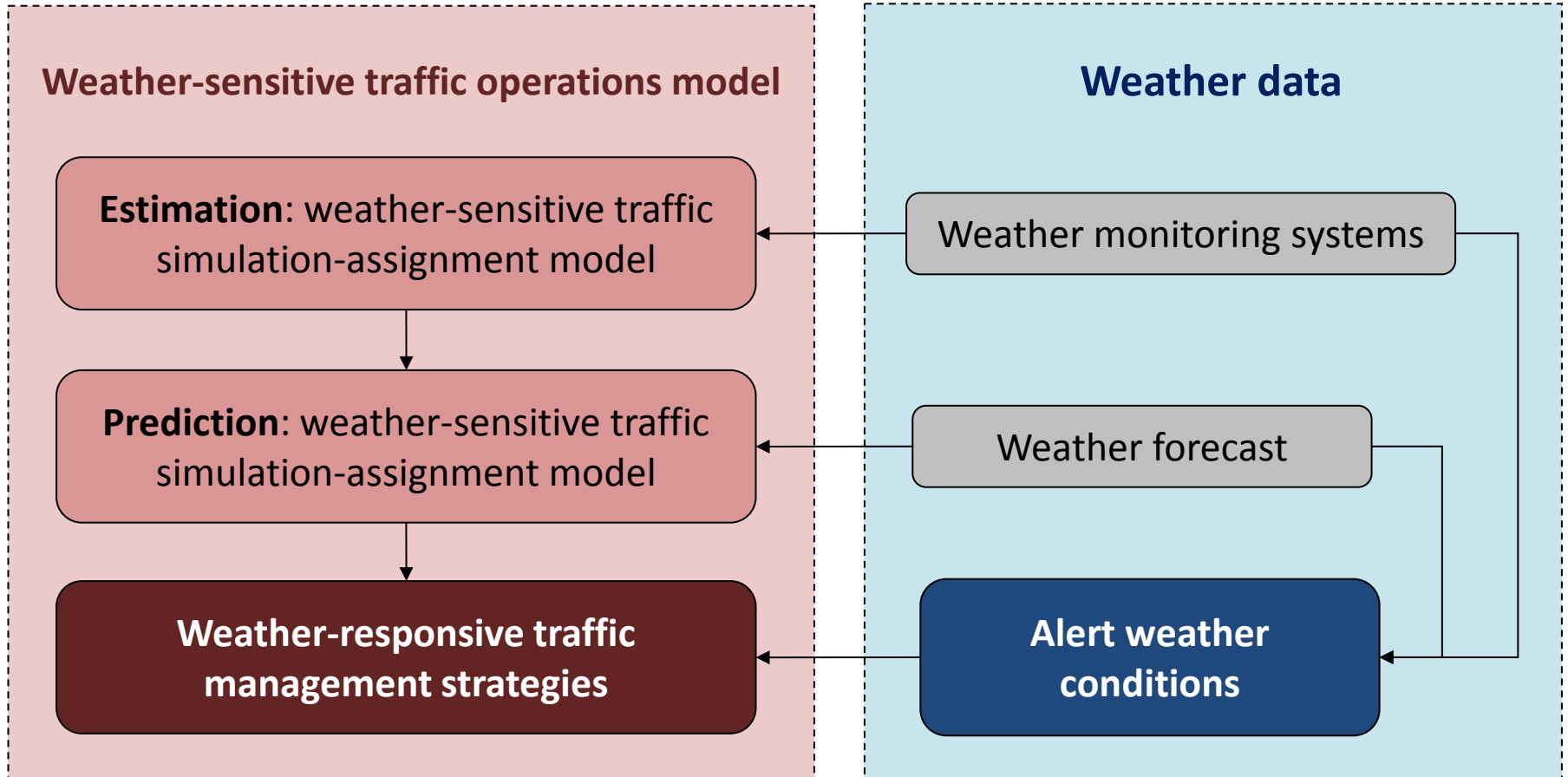
- Concentrations averaged over links along the congested portion of toll road, weighted by the link length
- Throughputs measured at downstream of where traffic breaks down in base case (no pricing)
- Anticipatory pricing strategy can provide higher throughput while maintaining lower concentration (steady traffic flow)



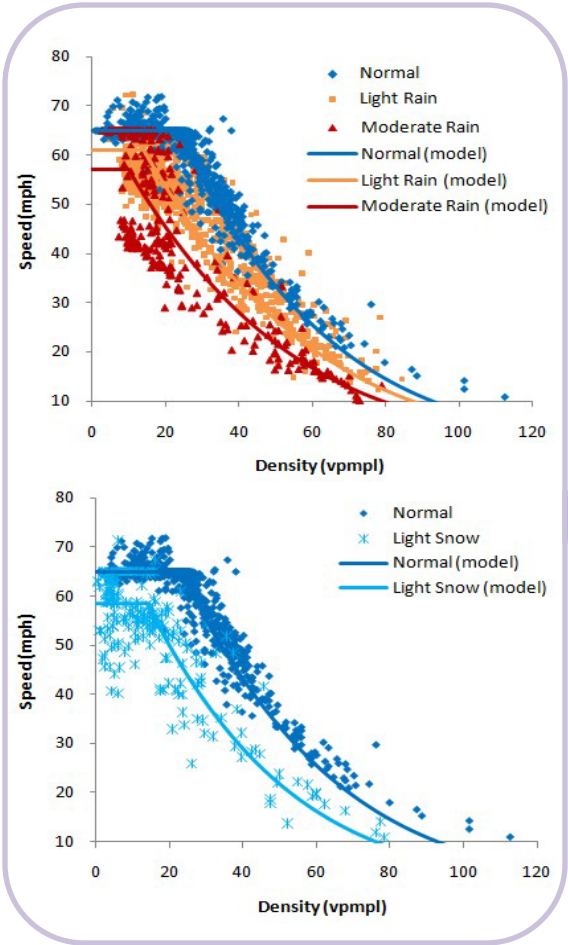
V.

Weather-Related Traffic Management (WRTM)

Weather-sensitive Traffic Estimation and Prediction System (TrEPS)



Model impacts of adverse weather on transportation networks



TrEPS
DYNASMART

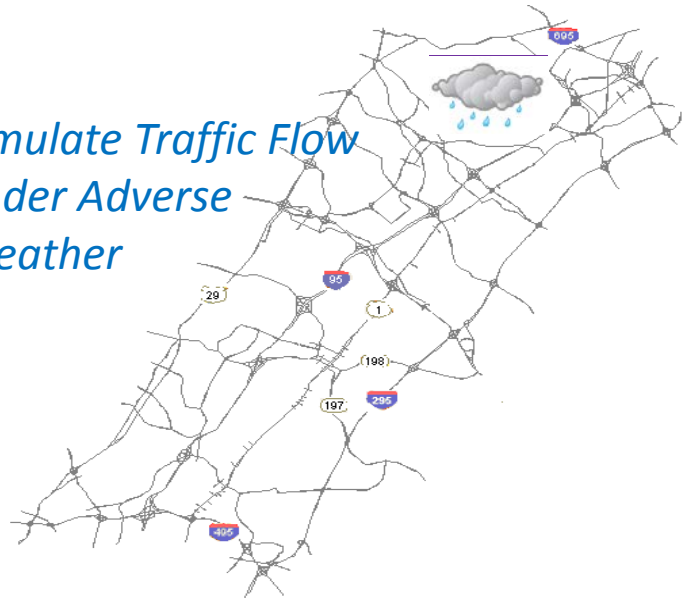
Supply-side
Parameter
Calibration
Weather Adjustment
Factor (WAF)

- Free-flow speed,
- Saturation flow rate,
- Section capacity,
- etc.

Weather Scenario Specification

- Rain intensity (r)
- Snow intensity (s)
- Visibility (v)

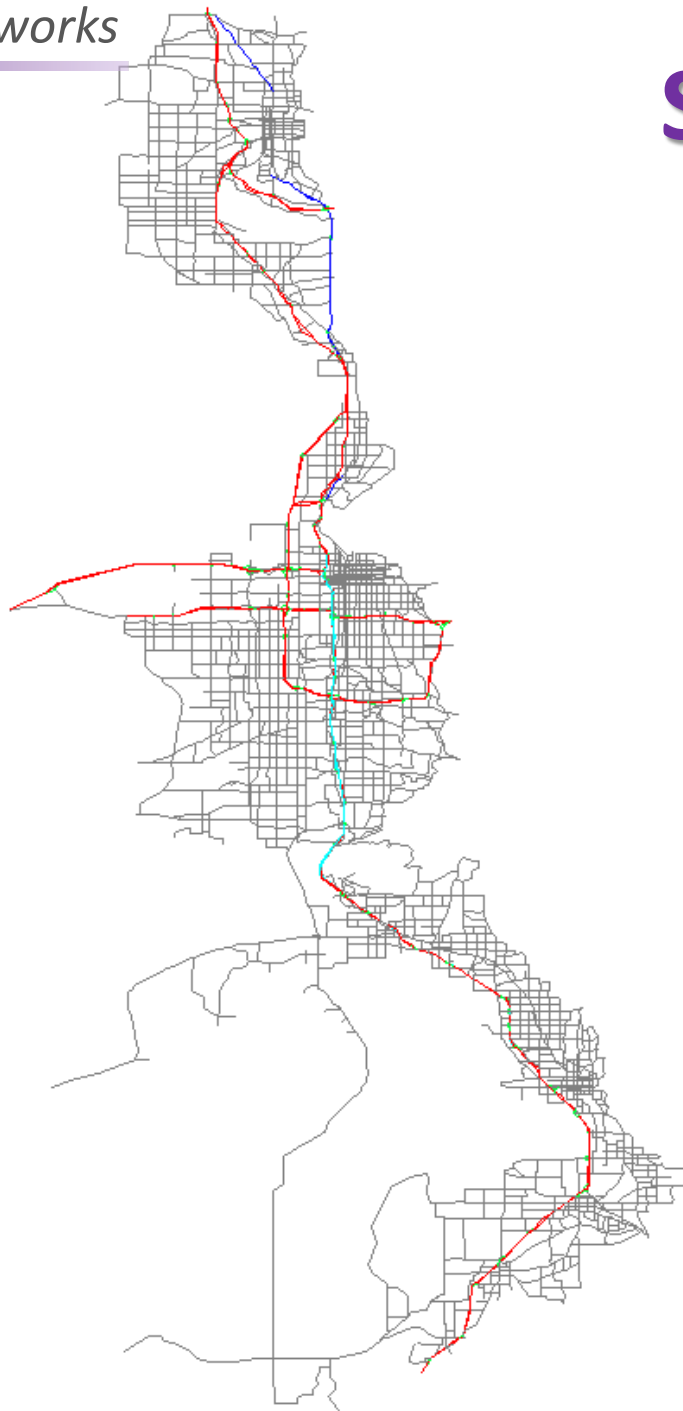
Simulate Traffic Flow
under Adverse
Weather



Chicago

- 40443 links
 - 144 links are tolled
 - 1400 freeways
 - 201 highways
 - 2120 ramps
 - (96 of them are metered)
 - 36722 arterials
- 13093 nodes
 - 2155 signalized intersections
- 1961 zones
 - 1944 internal
 - 17 external
- Demand period
 - 5am -10am hourly demand
 - 355 unique link counts
 - Observation Interval: 5 min





Salt Lake City

- 2,250 zones
- 17,947 links
 - 16,293 arterials
 - 576 ramps
 - 136 highways
 - 791 freeways
 - 151 HOV lanes
- 8,309 nodes
 - 1,134 signalized intersections
- Demand horizon
 - 6am – 9am
- Simulation horizon
 - 6am – 10am

Long Island



- 1,431 zones
- 21,790 links
 - 17,942 arterials
 - 2,059 ramps
 - 31 highways
 - 1,588 freeways
 - 170 HOV lanes
- 9,402 nodes
 - 4,691 signalized intersections
- Demand horizon
 - 5am – 10am
- Simulation horizon
 - 5am – 11am

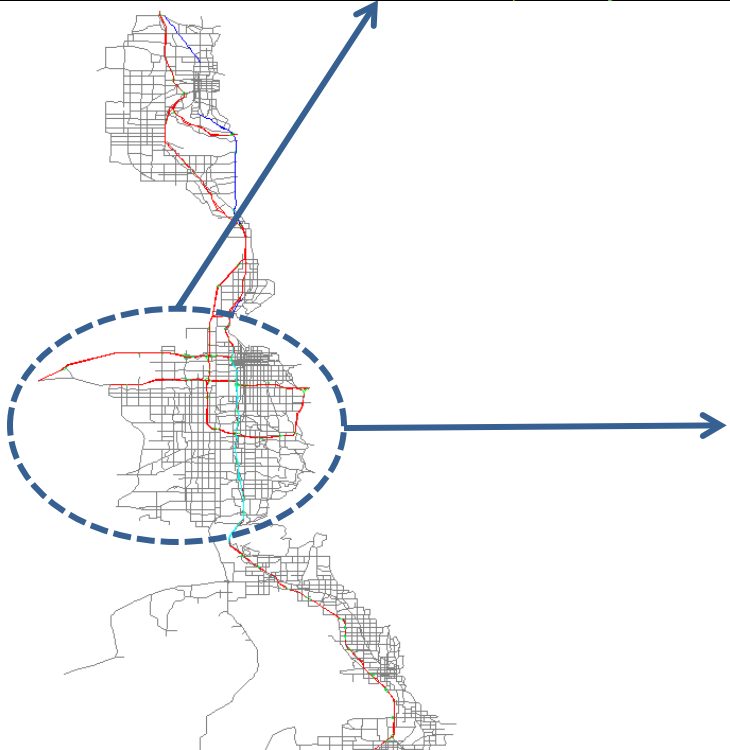
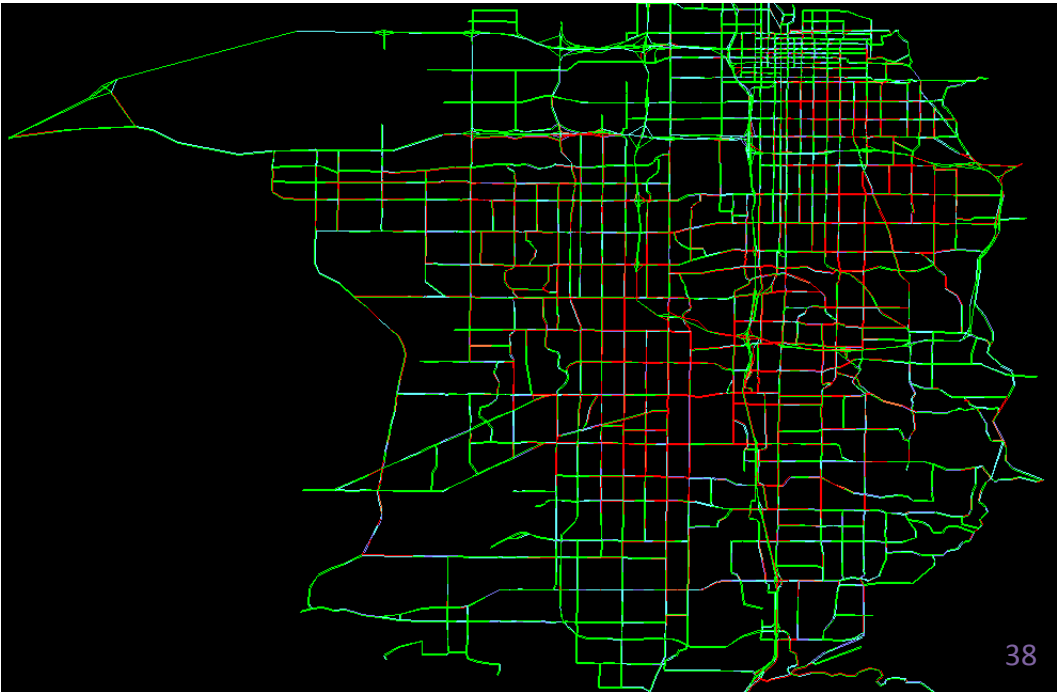
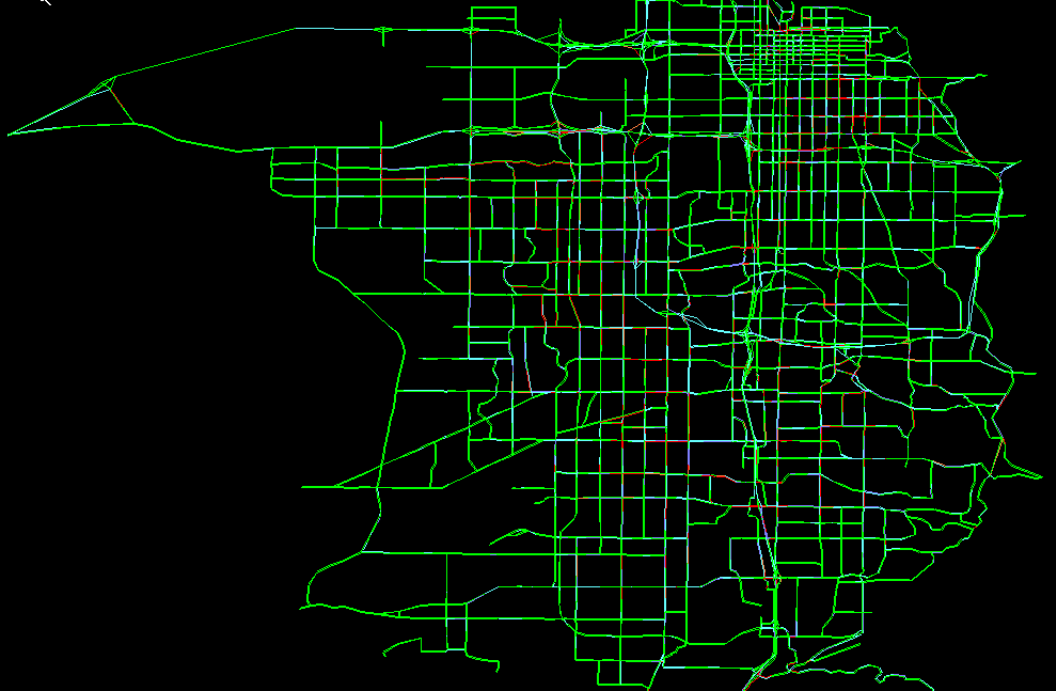
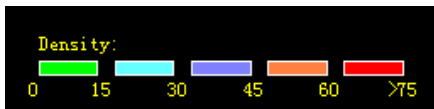
Off-line Implementation : Effectiveness of VSL/VMS Strategies

- Test Scenarios
 - **Clear Day**: Maximum visibility with zero precipitation.
 - **Snow**: Visibility ranges from 10 to 1.75 miles, snow intensity ranges from 0.01 to 0.15 inches per hour network-wide.
 - **Snow with VMS – Variable Speed Limit**: Speed reduction strategies are implemented on freeway corridors.
 - **Snow with VMS – Mandatory Detour**: Vehicles are detoured from some heavily impacted links to alternative routes.

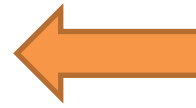
Weather Impact

← Clear Day

↓ Snow

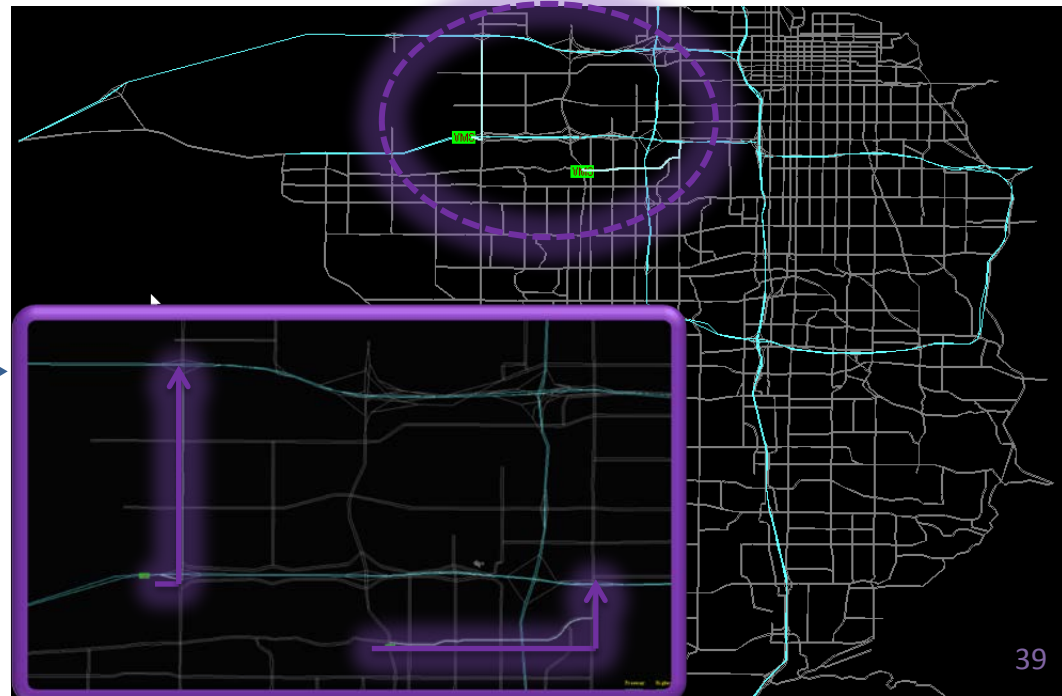
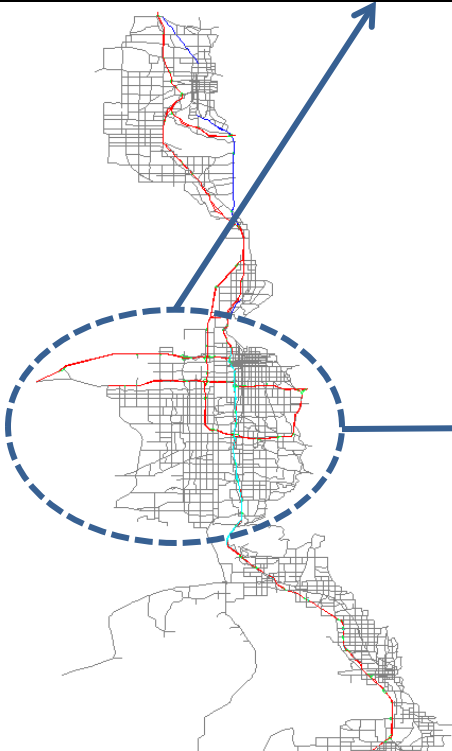
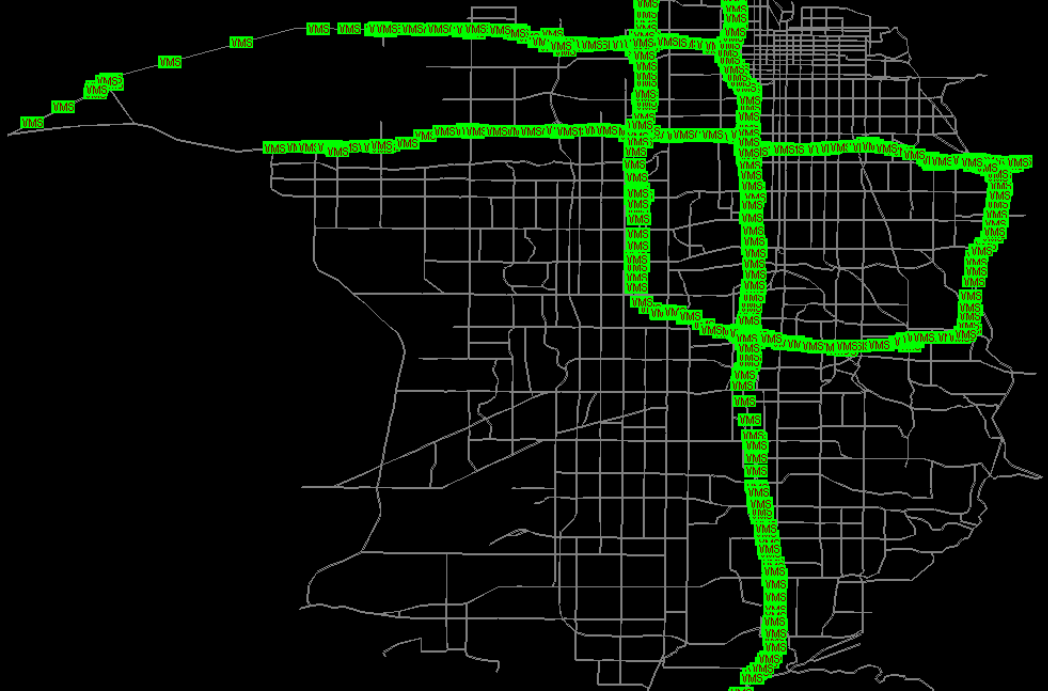


WRTM Strategies

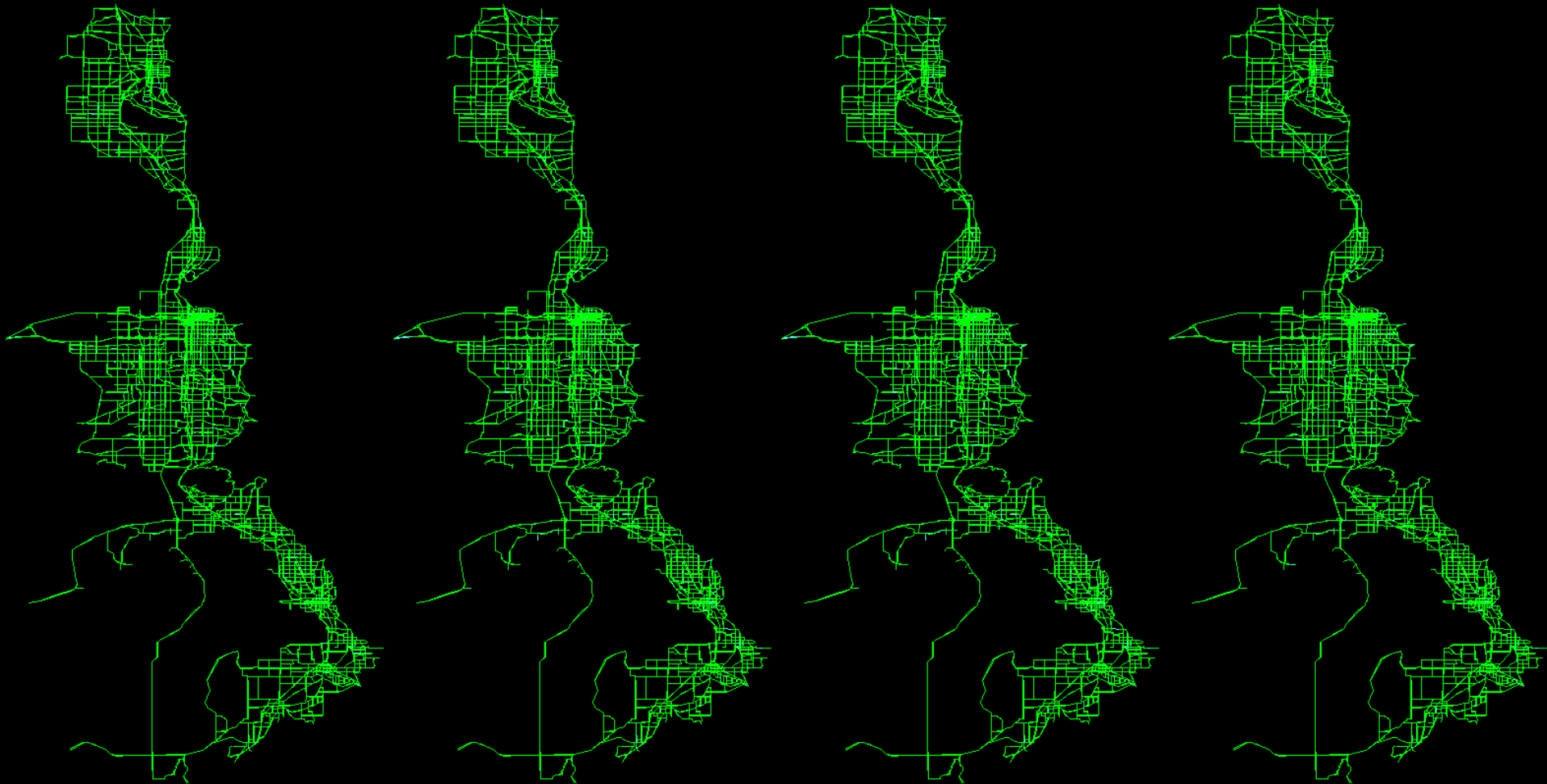
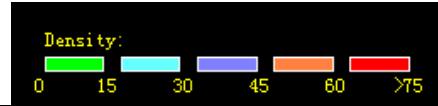


VMS - VSL

VMS - Detour



6:00 am



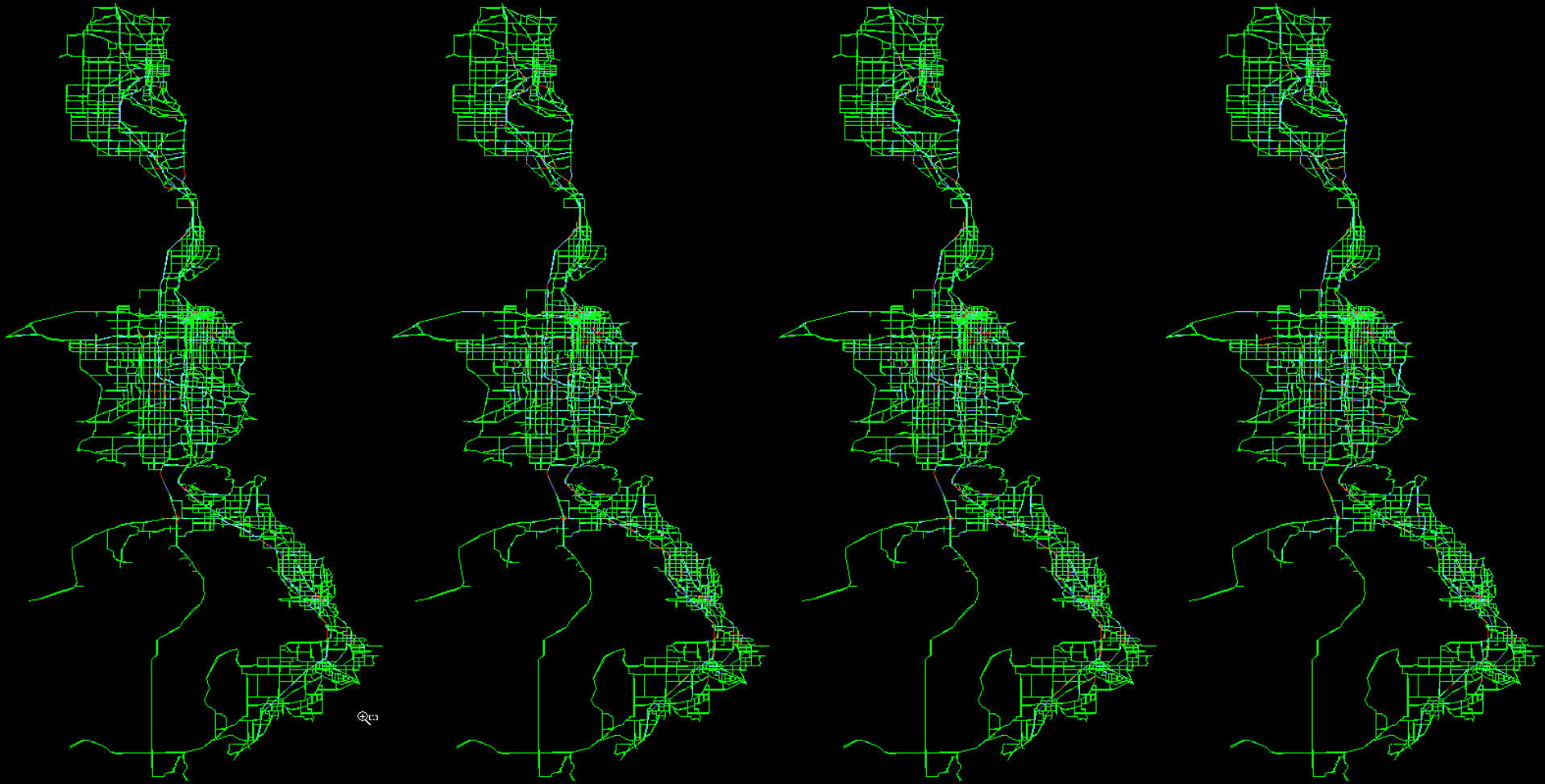
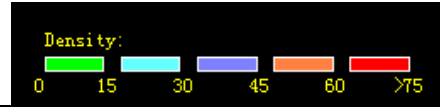
01-00.Regular

01-01.NoWRTM

01-02.VSL7

01-03.VMS2

6:30 am



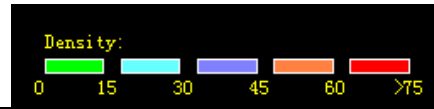
07-00.Regular

07-01.NoWRTM

07-02.VSL7

07-03.VMS2

7:00 am



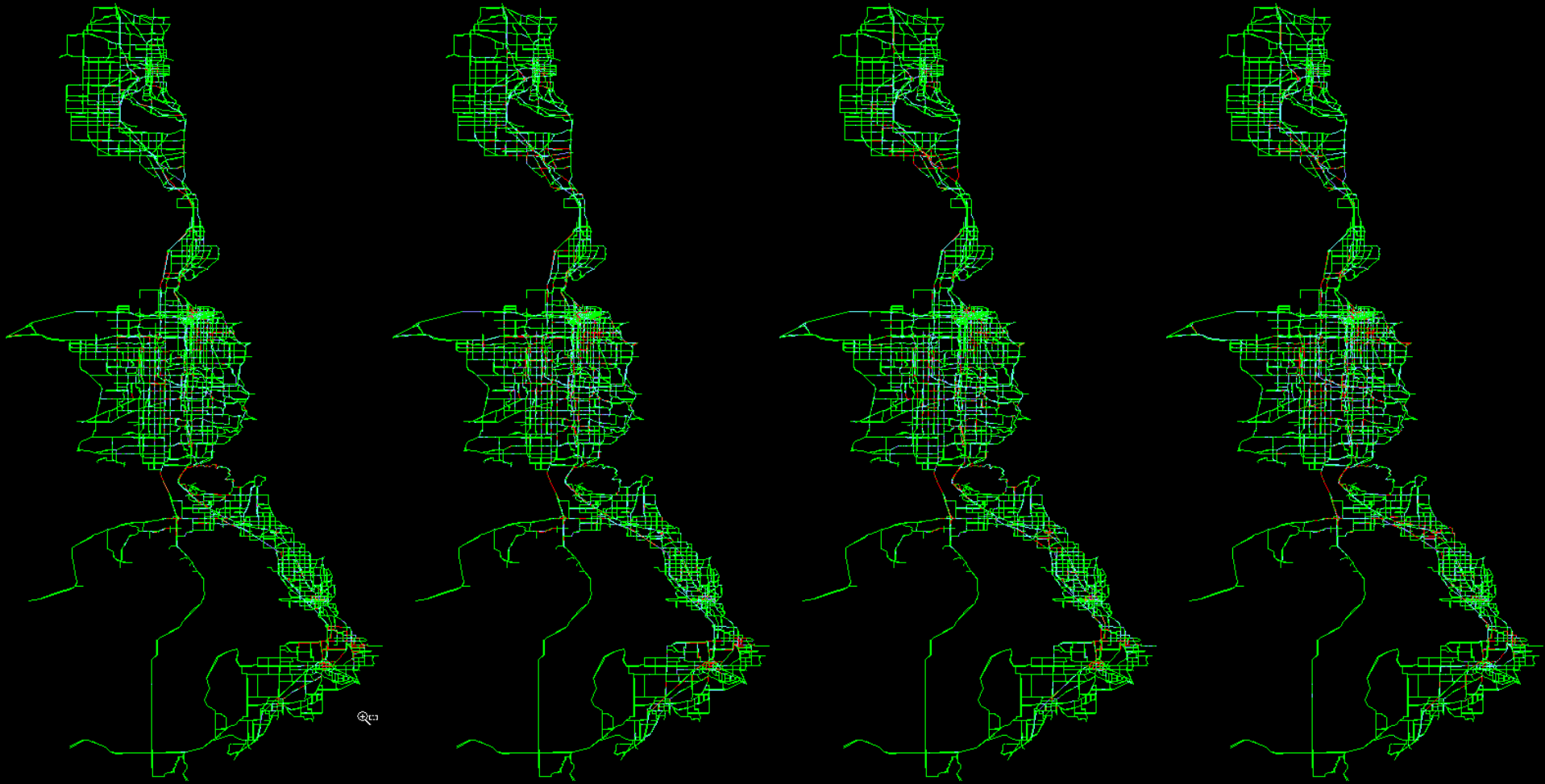
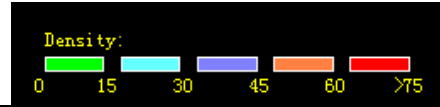
13-00.Regular

13-01.NoWRTM

13-02.VSL7

13-03.VMS2

7:30 am



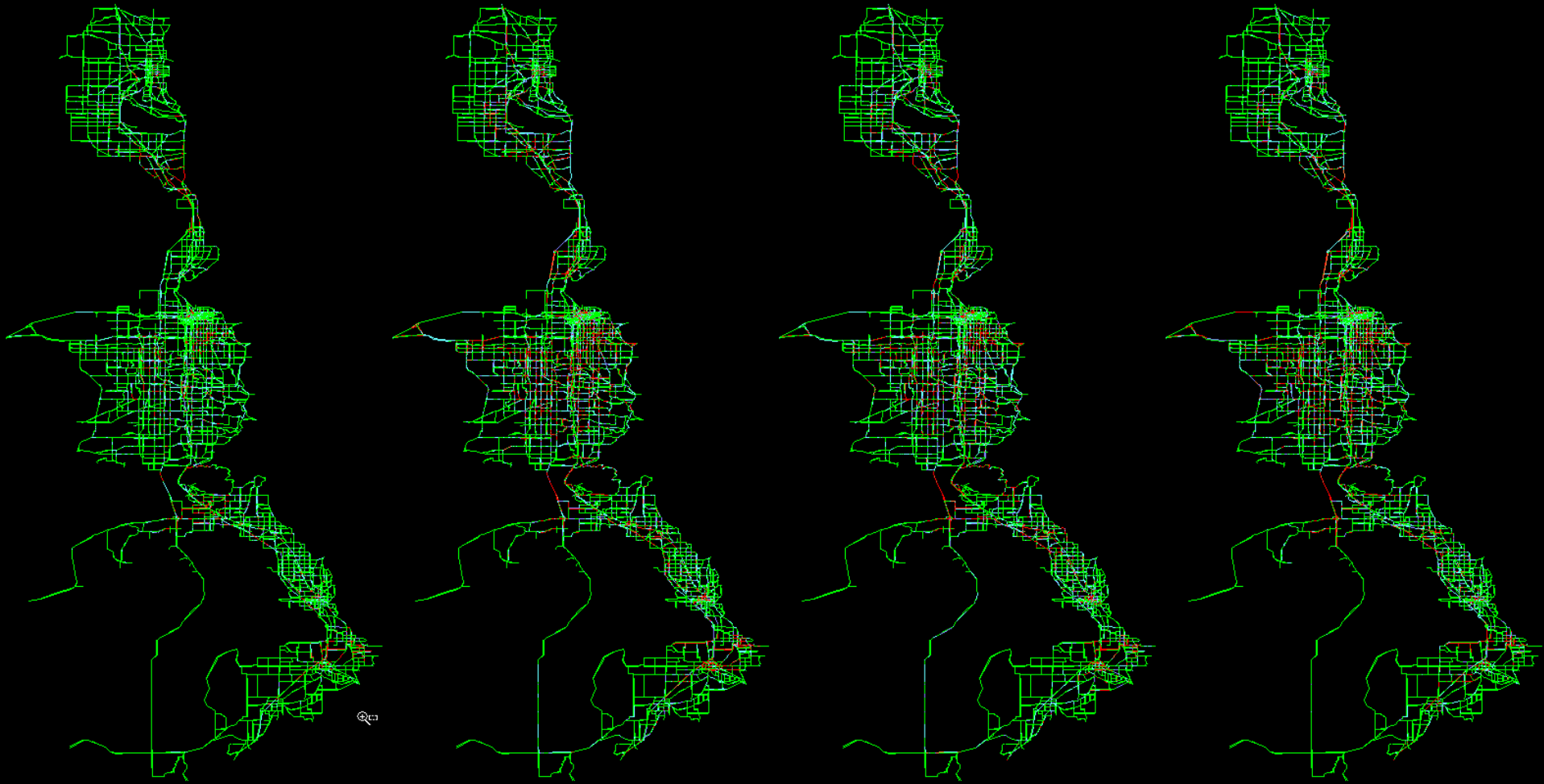
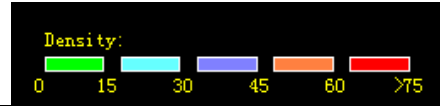
19-00.Regular

19-01.NoWRTM

19-02.VSL7

19-03.VMS2

8:00 am



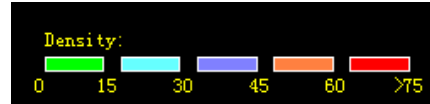
25-00.Regular

25-01.NoWRTM

25-02.VSL7

25-03.VMS2

8:30 am



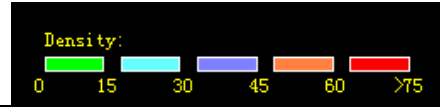
31-00.Regular

31-01.NoWRTM

31-02.VSL7

31-03.VMS2

9:00 am



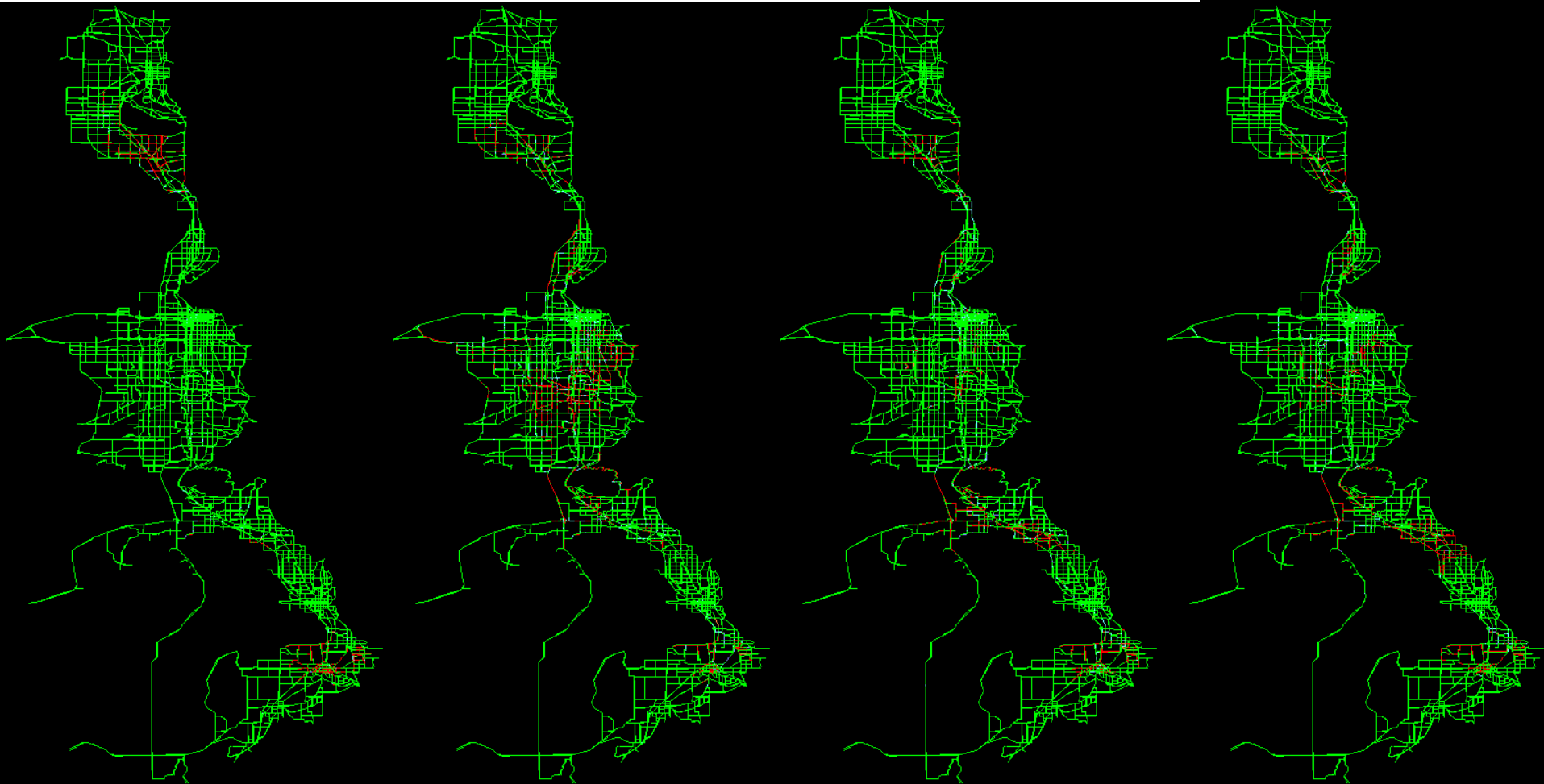
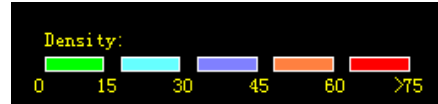
37-00.Regular

37-01.NoWRTM

37-02.VSL7

37-03.VMS2

9:30 am



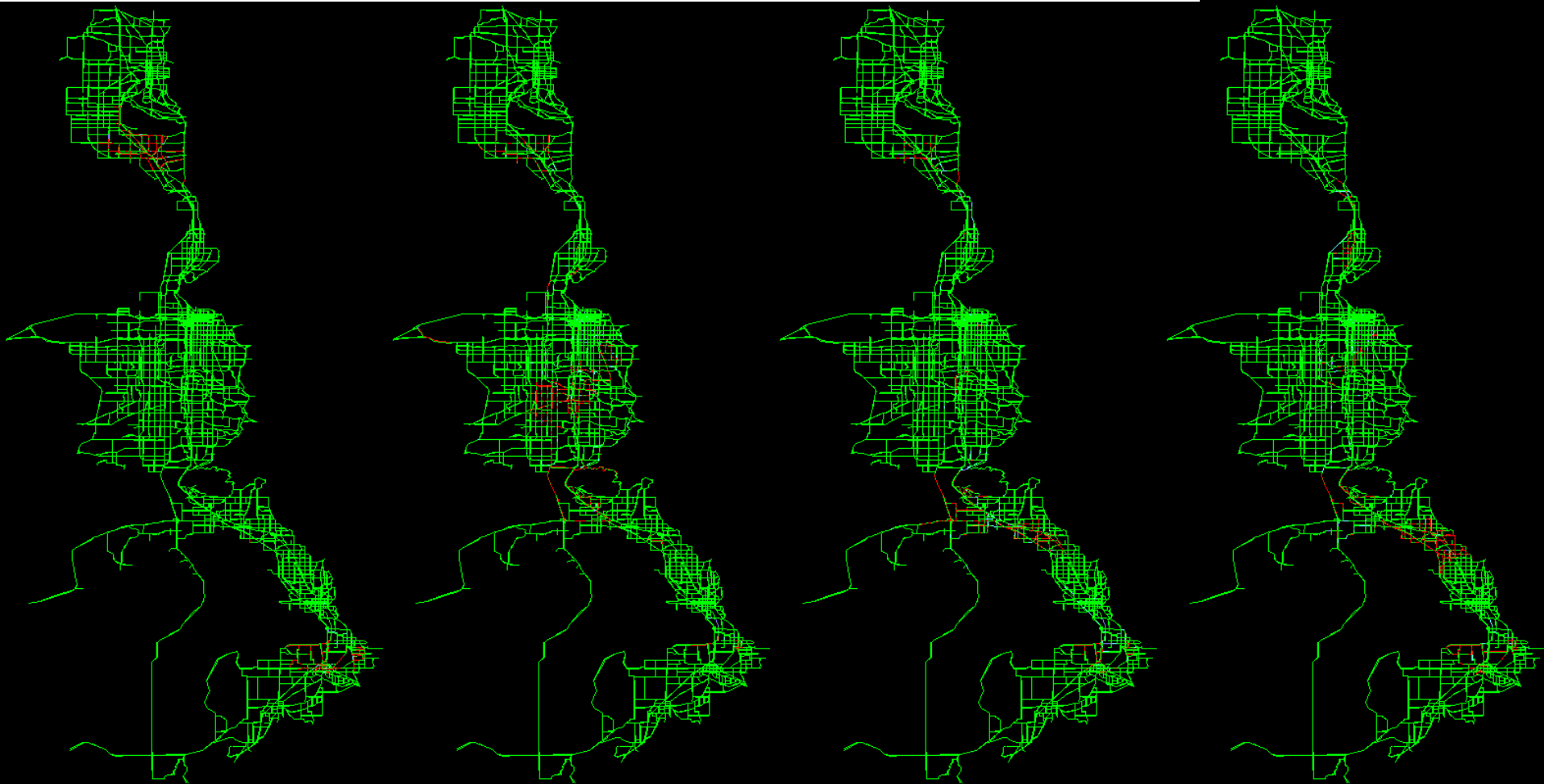
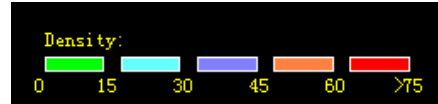
43-00.Regular

43-01.NoWRTM

43-02.VSL7

43-03.VMS2

10:00 am



49-00.Regular

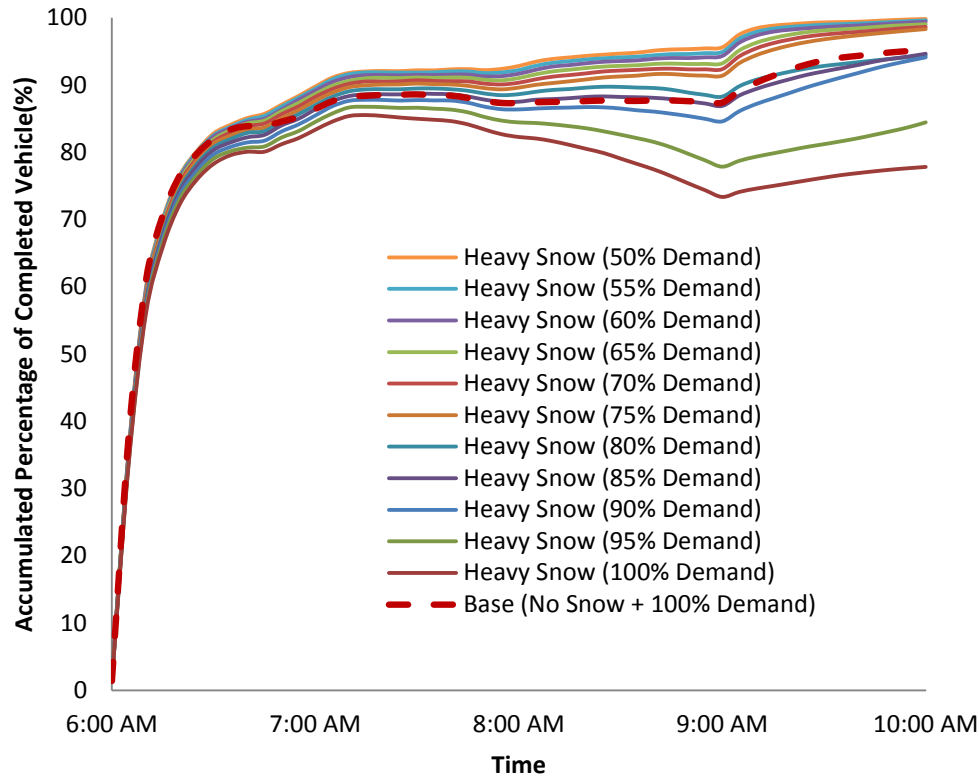
49-01.NoWRTM

49-02.VSL7

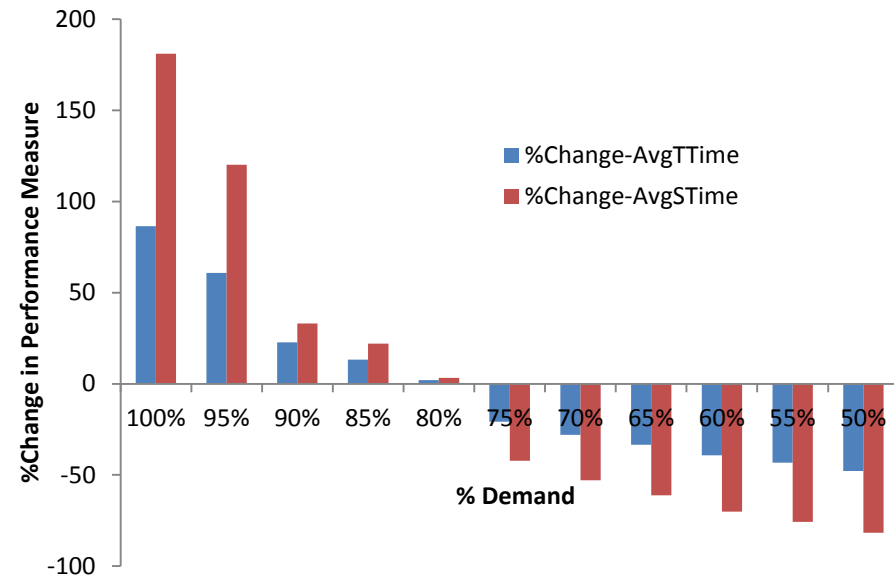
49-03.VMS2

Off-line Implementation (Salt Lake City)

- Demand Management
 - Analysis Results



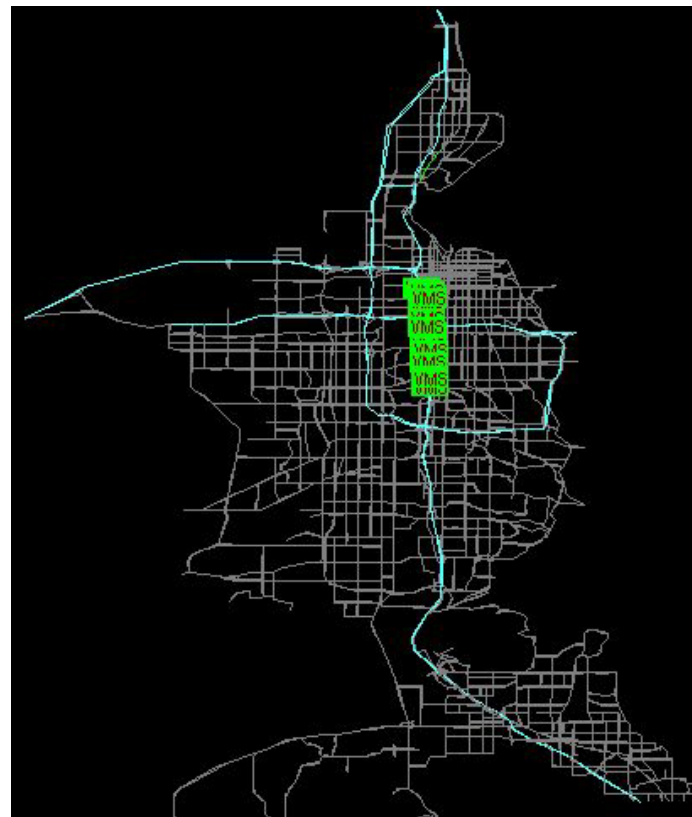
(a) Time-dependent network throughput measure



(b) %Change in performance measures for different demand levels relative to base-case

On-line Implementation (Salt Lake City)

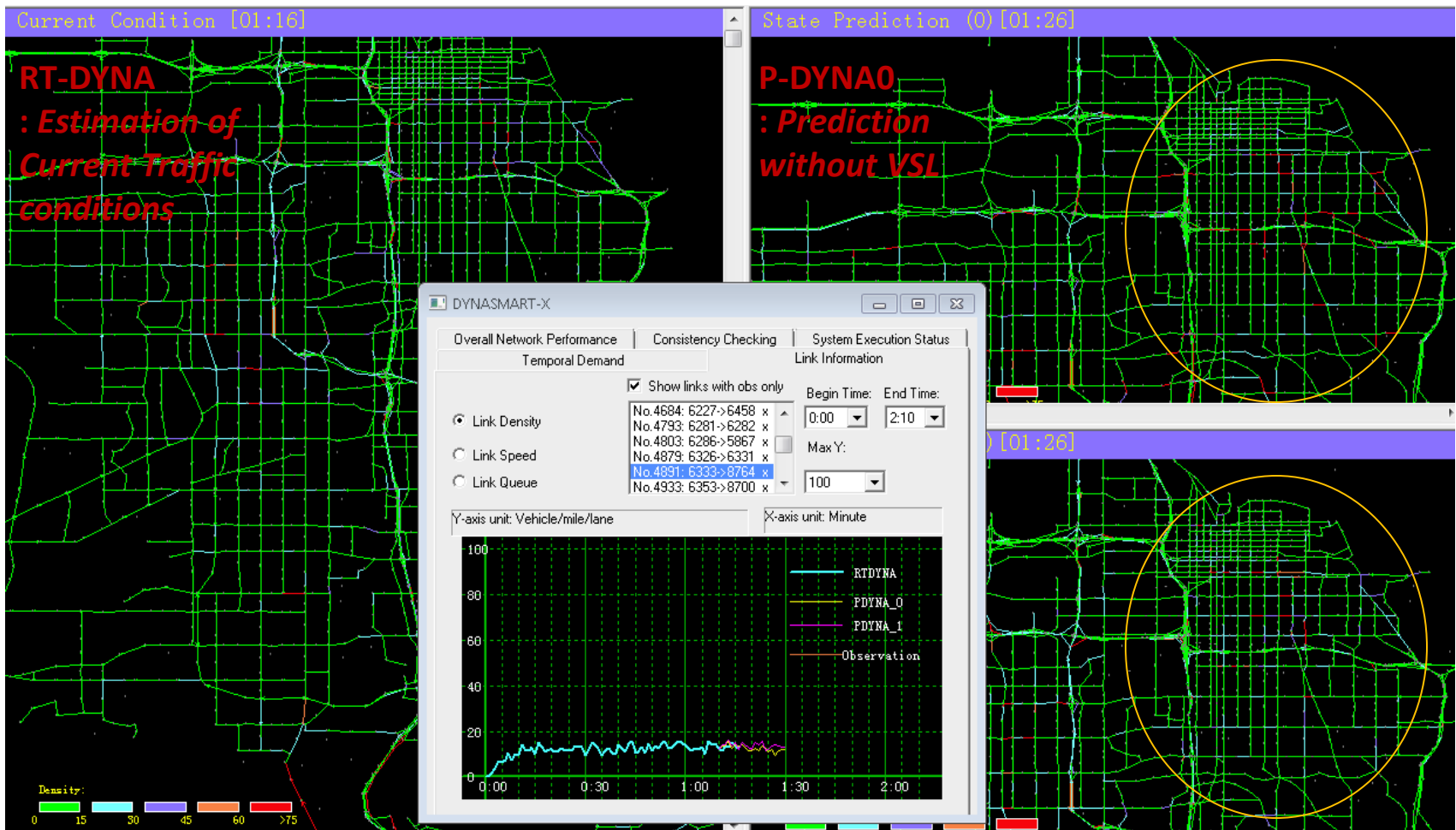
- Target weather event :
 - Snow on April 6, 2012
- Before the event
 - Retrieved a set of VSL strategies from the WRTM strategy repository.
 - Performed the off-line simulation analysis to select the best strategy given the predicted weather scenario.
 - Selected VSL strategy
 - **Deploy VSL on Veterans Memorial Highway (Southbound)**



Selected VSL strategy under the given snow scenario

On-line Implementation (Salt Lake City)

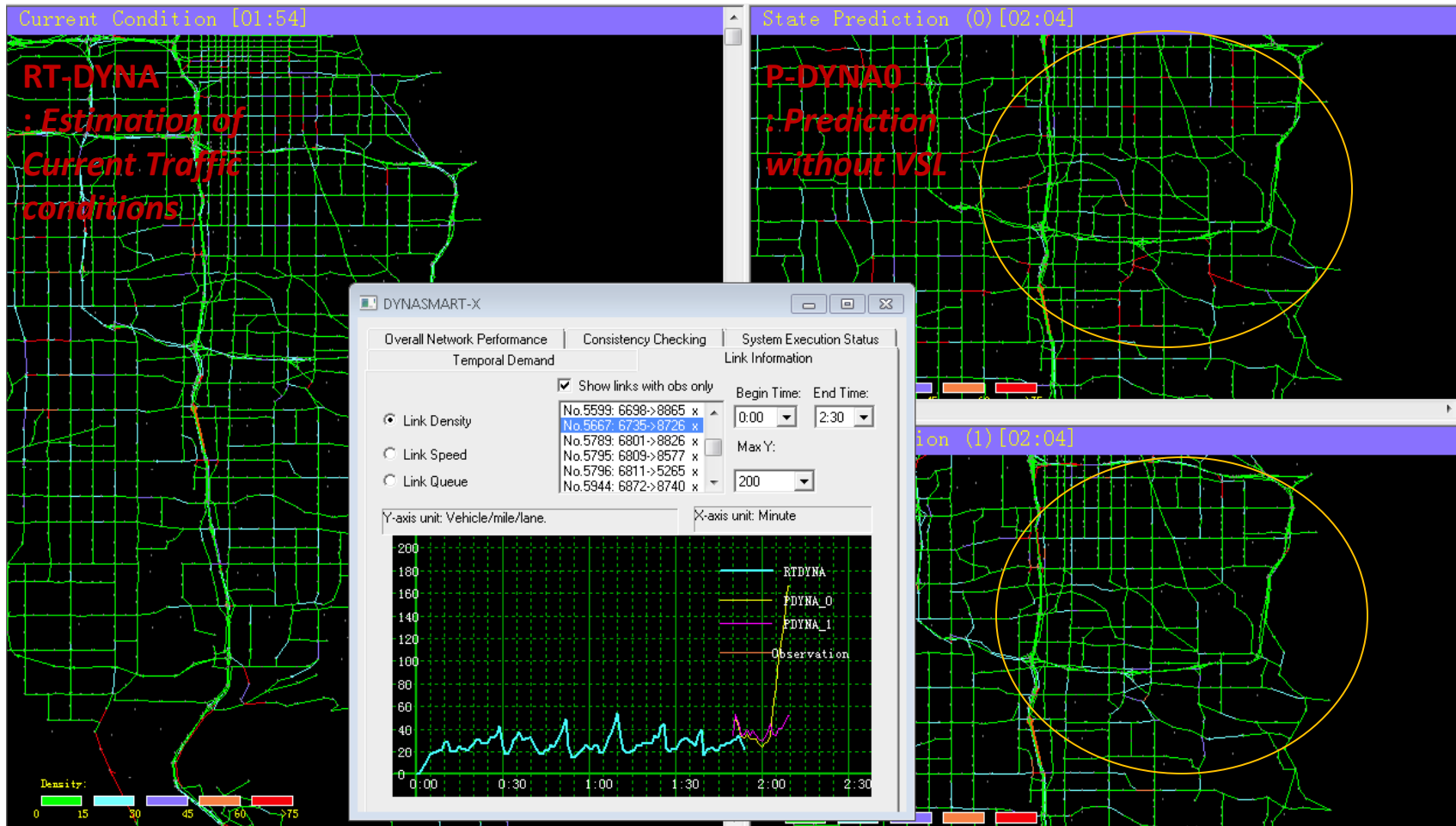
At 7:16AM, predicted traffic states for 7:26AM



P-DYNA1
: Prediction with VSL

On-line Implementation (Salt Lake City)

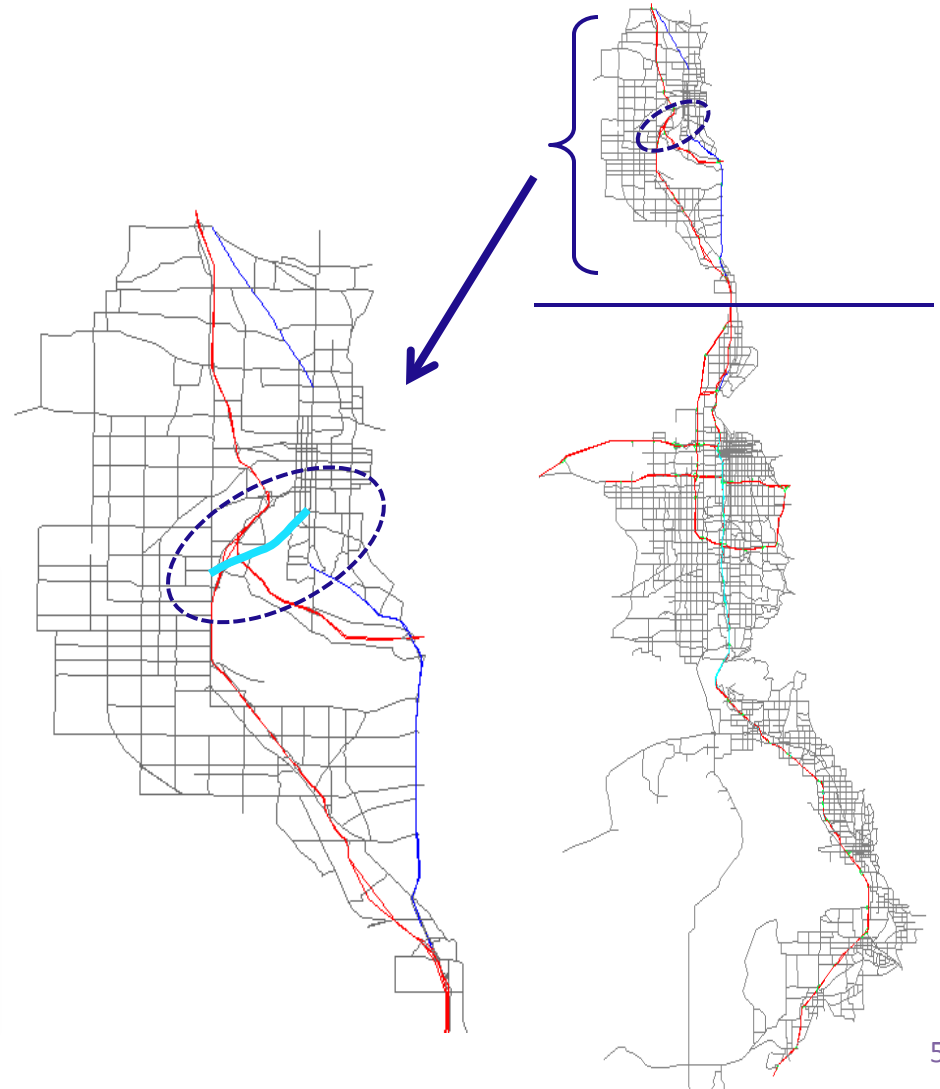
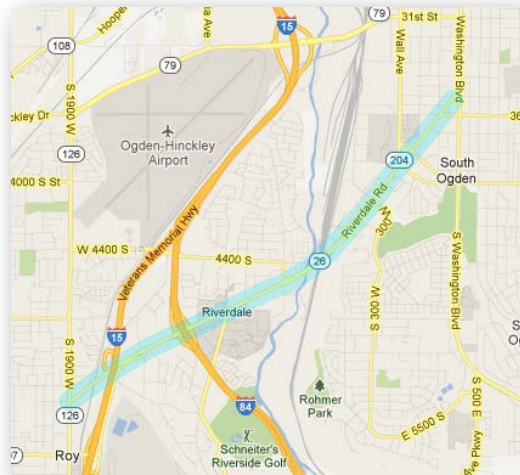
At 7:54AM, predicted traffic states for 8:04AM



P-DYNA1
: Prediction with VSL

ONGOING PROJECT WITH FHWA and UDOT in Salt Lake City

Deploy and evaluate calibrated TrEPS for an arterial corridor(RIVERDALE) to support WRTM interventions, especially signal control strategies



VI.

Logistics Operations in Congested Urban Environments

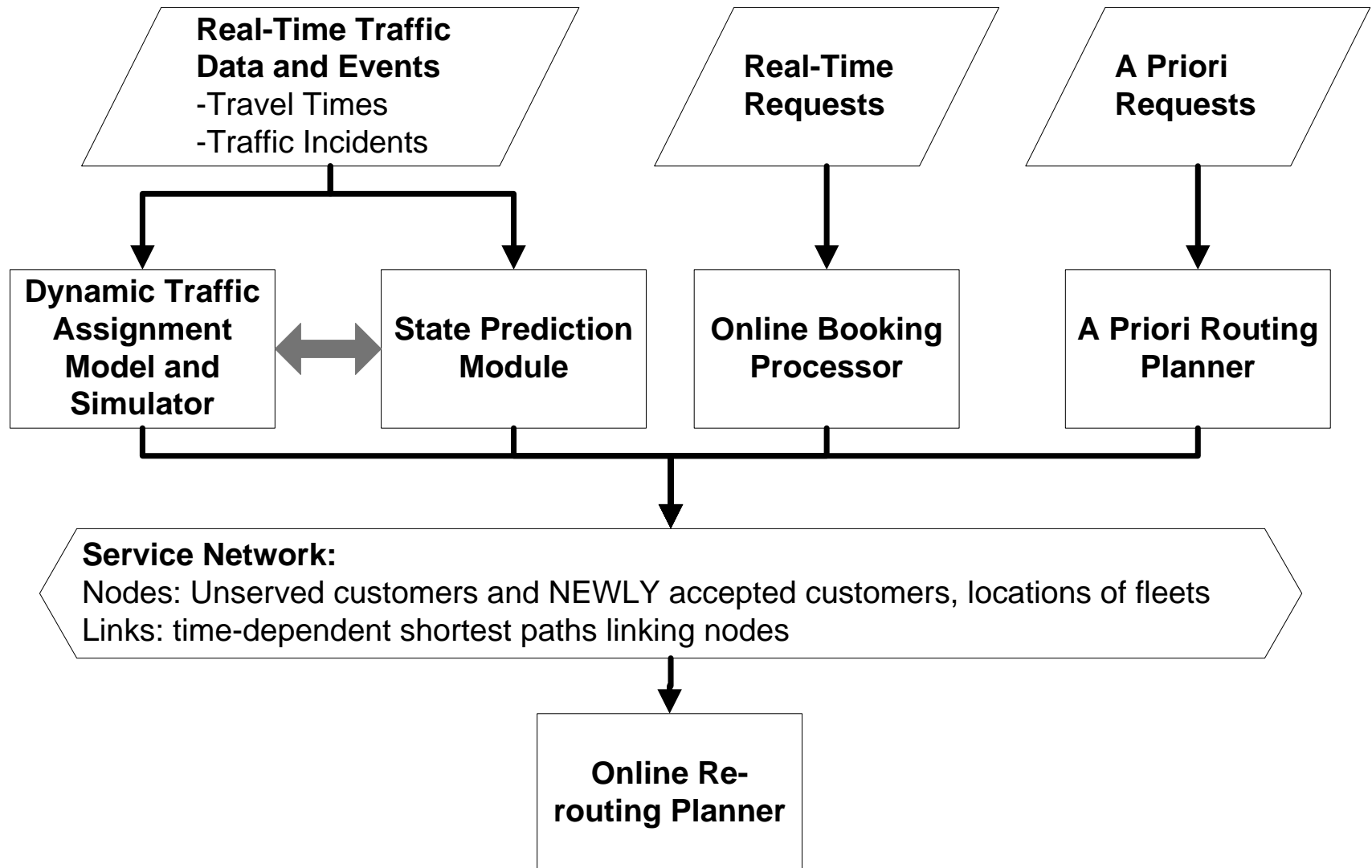
Challenges for City Logistics Carriers

- Deliveries in urban areas suffer from time-varying congestion, and various traffic events, such as lane-closure, accidents, construction, weather etc.
- Real time customer requests.
- Customers expect on-time deliveries within service time windows.

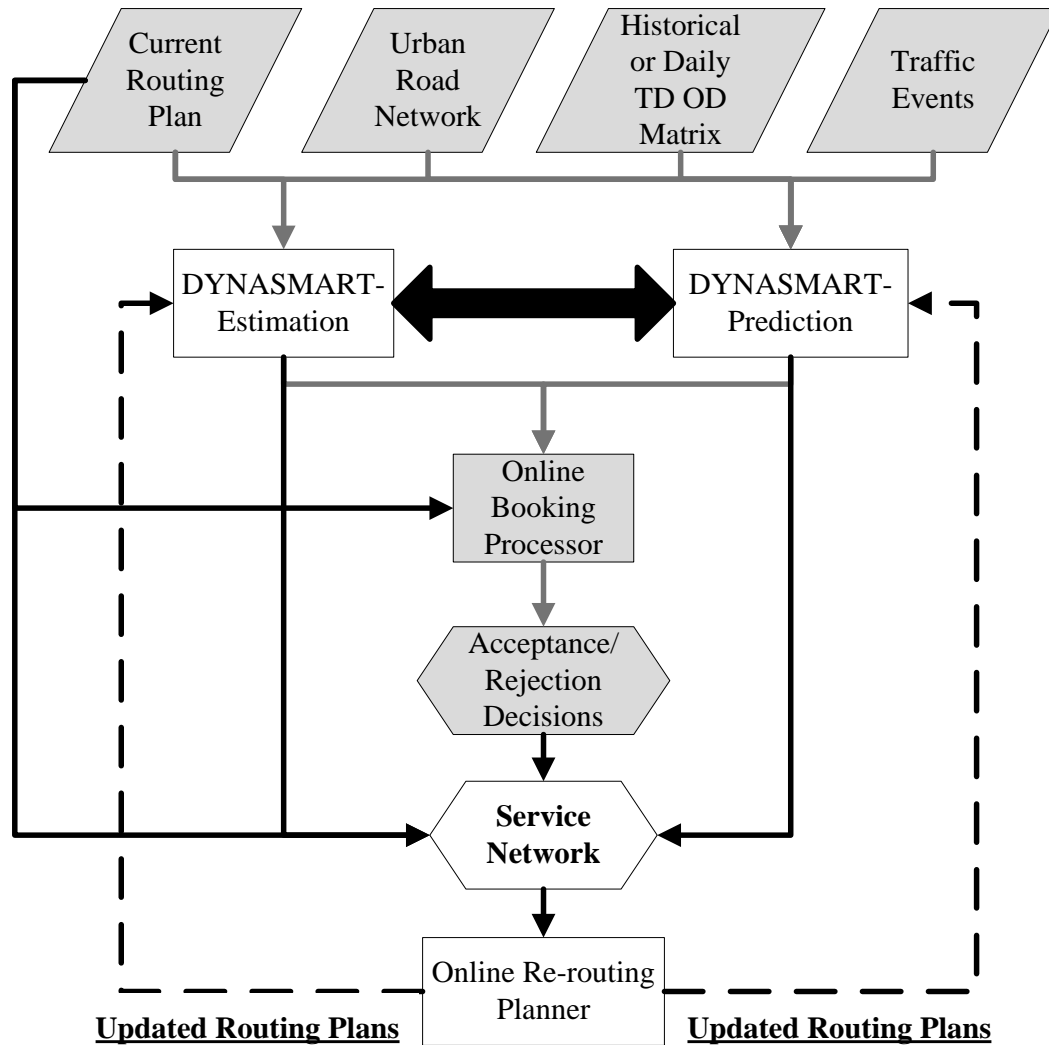
Research Objective

- To develop an integrated system which has the following features:
 - Capable of mapping real-life operational components into analytical VRP models.
 - Respond to real-time customer requests.
 - Consider traffic variations on road networks (including effect of weather, incidents, special events, etc...)
 - Applicable to problems of practical sizes.

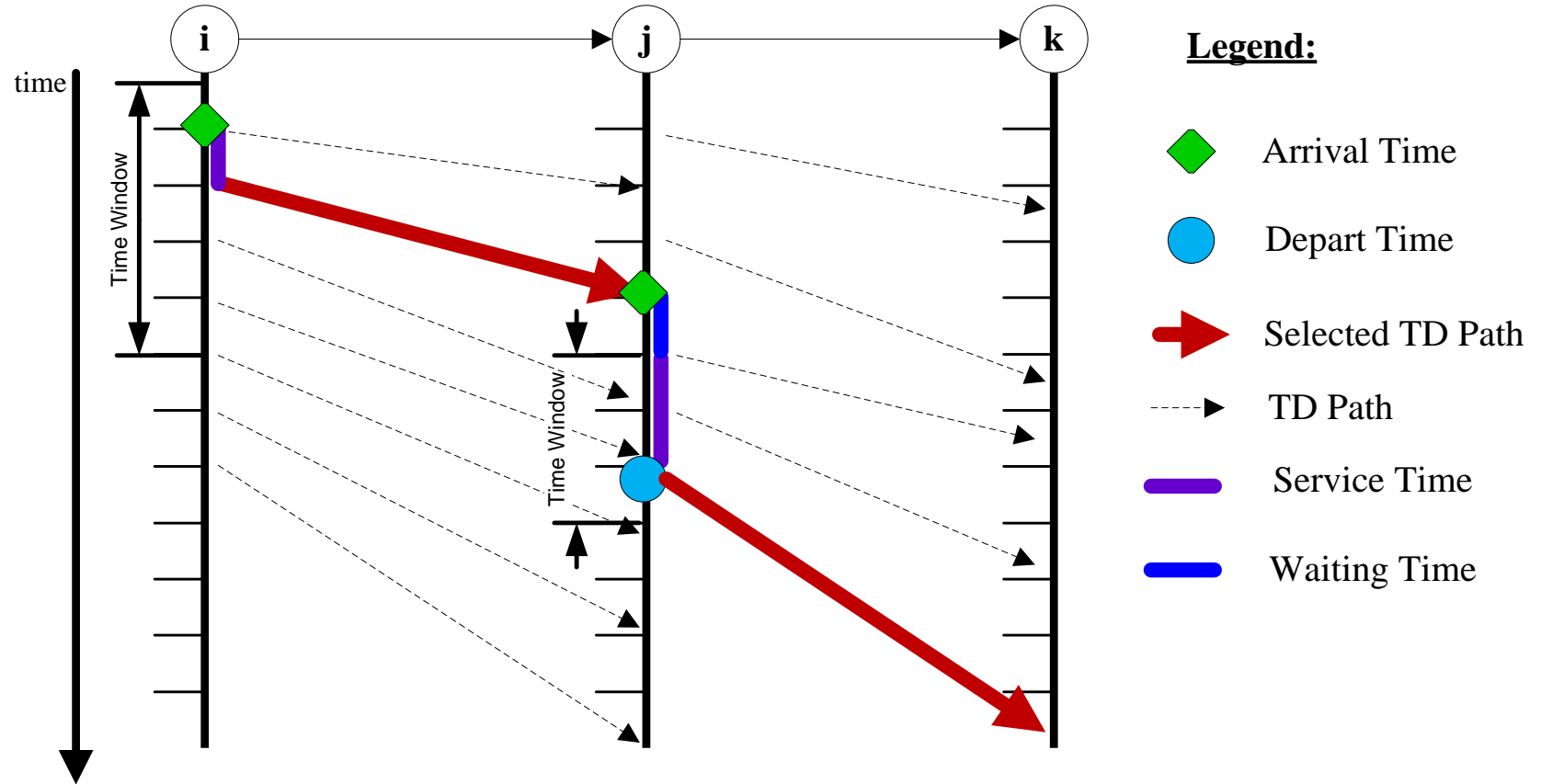
Overall Architecture



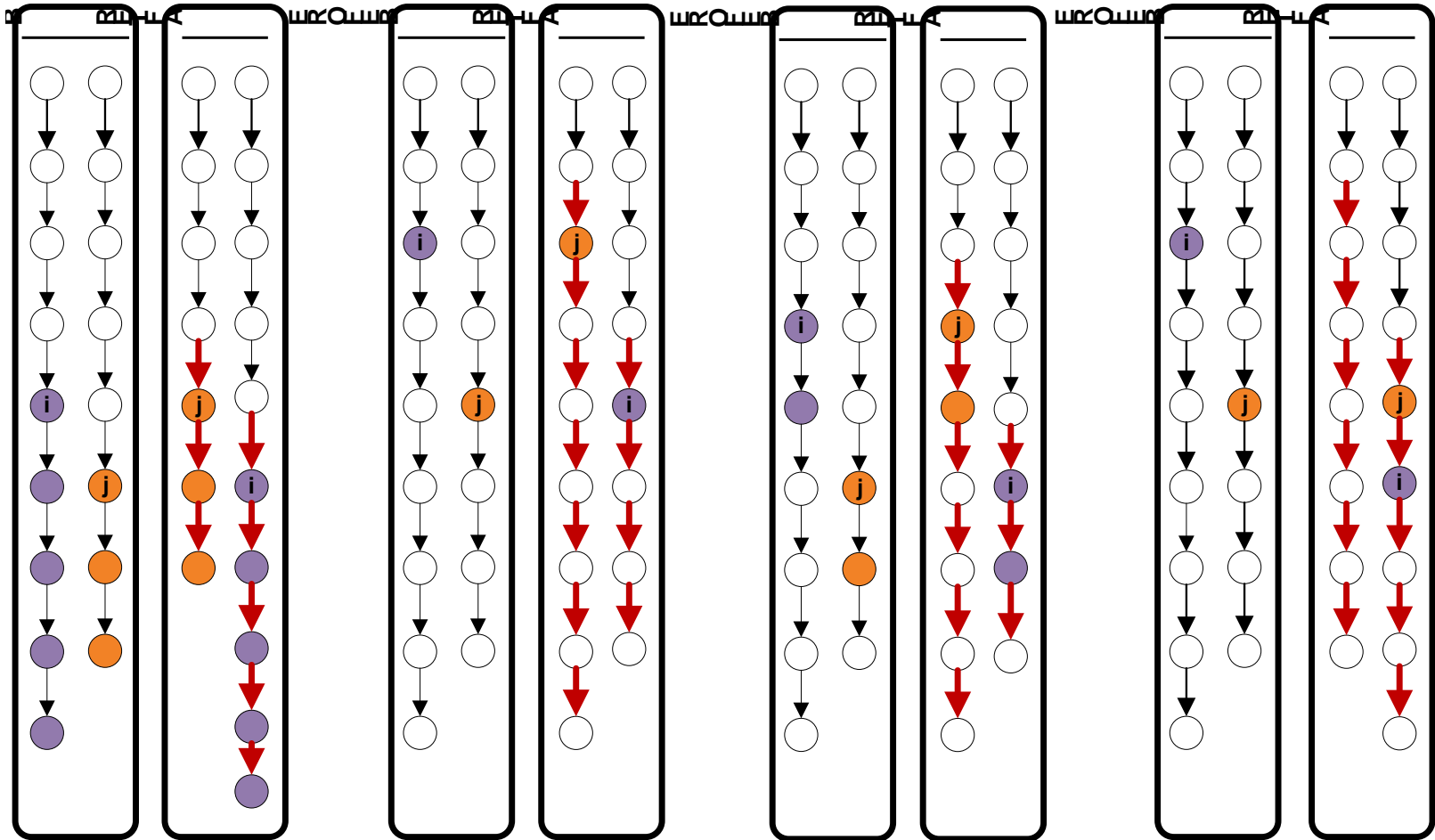
Online Booking Processor & Re-routing Planner



An Illustration of VRP with TDDT



Heuristic Algorithm: Local Search Operators



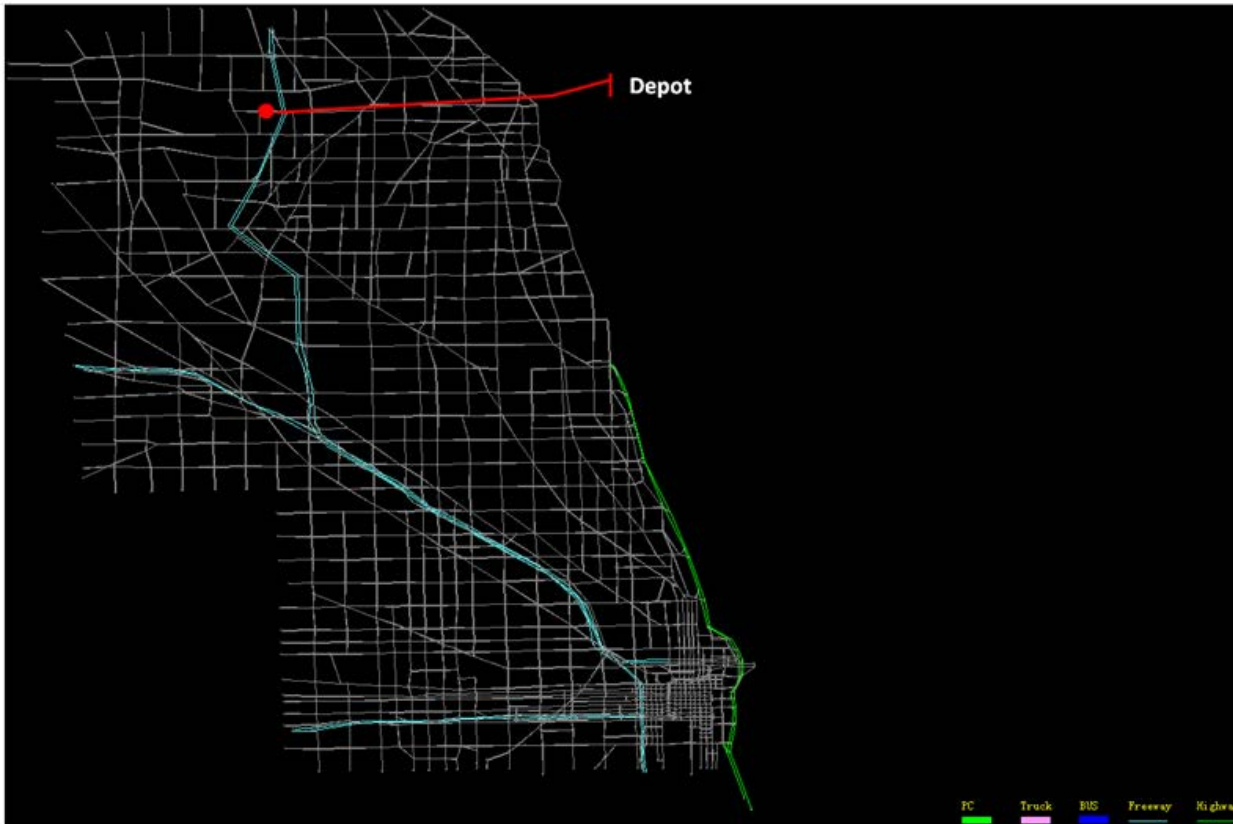
a) 2-opt*

b) Exchange

c) Segment Exchange

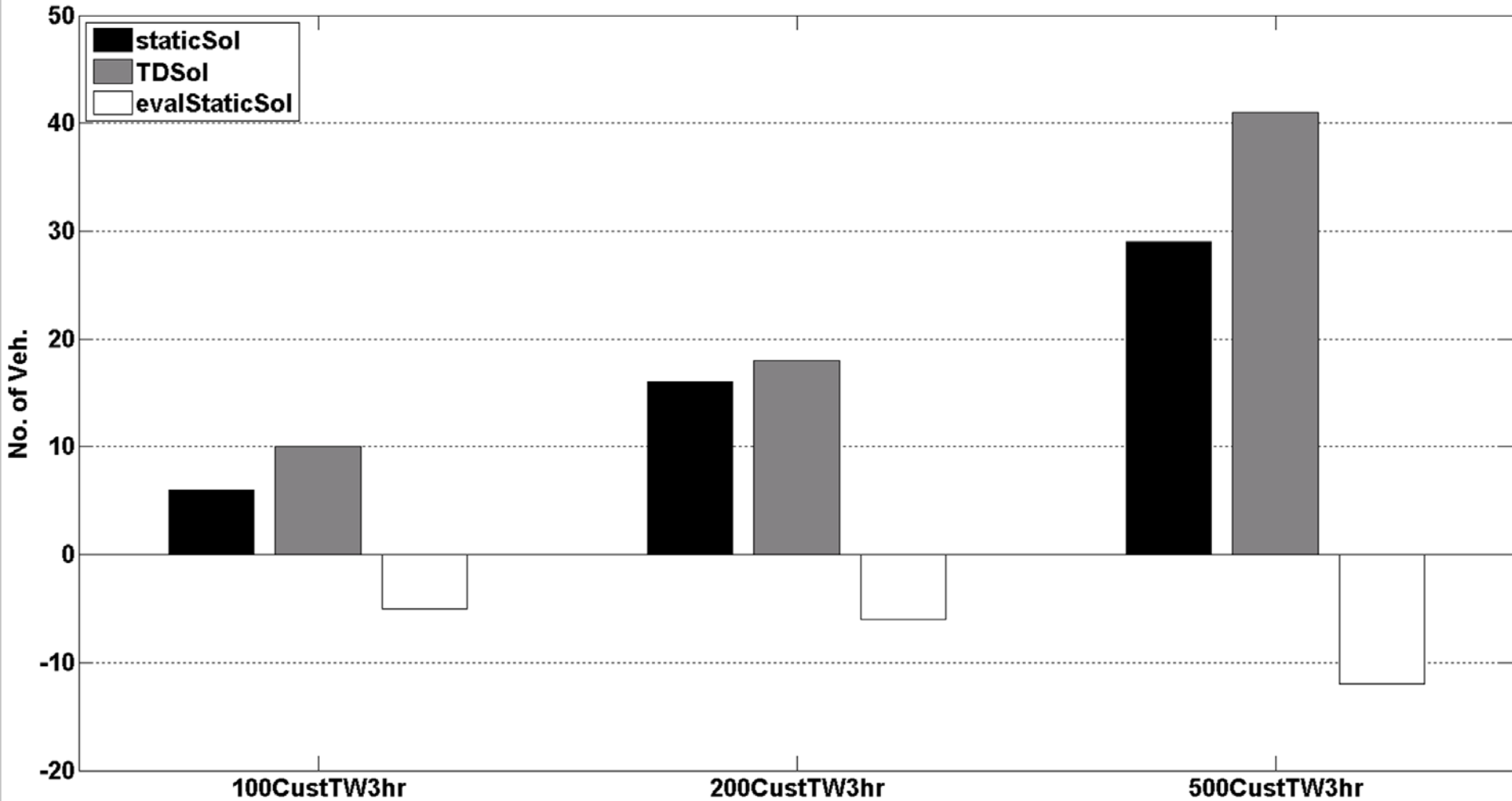
d) Insertion

Case Study: Chicago Network



- Nodes: 1,578
- Links: 4,805
- TAZ: 218
- TD OD: 16hr (5am-9pm), ~1.6mil vehicles

Numerical Results: Feasibility



VII.

Takeaways



PREDICTION essential in real-time traffic management and urban logistics

Considerable opportunities: new sources of personal information, emerging technologies

Computational challenges remain

User behavior: will remain moving target, because users will adapt hence need for adaptive schemes

Growing role of private sector as business models become more compelling

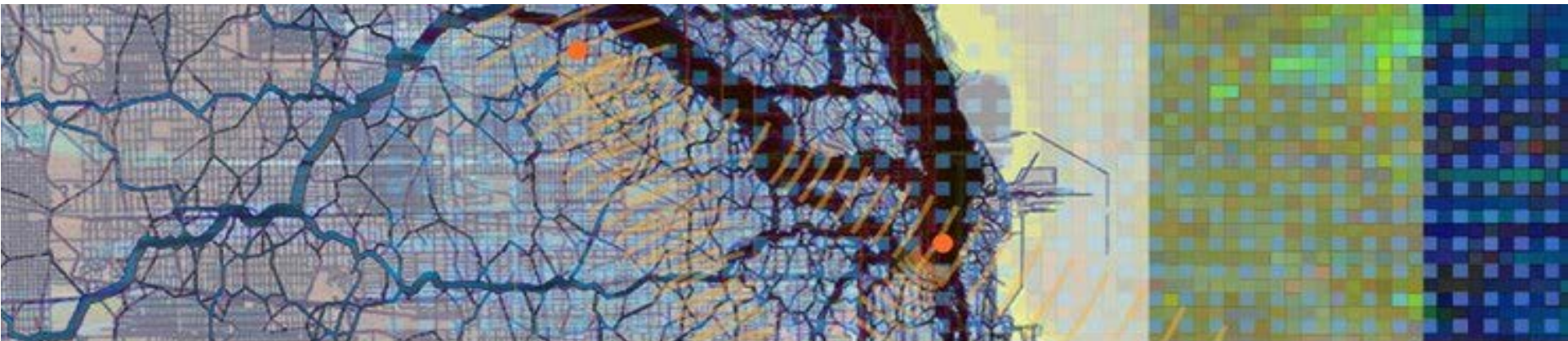
THE SWEET SPOT FOR SYSTEM MANAGEMENT

Leverage system state information and individual characteristics (and preferences) in generating interventions that are

- dynamic (timely)
- localized (consider network and non-network factors)
- anticipatory (consider predicted events and system evolution)
- adaptive (learn about individual responses and system impacts)
- distributive (across modes, times of day, user groups)
- economically efficient (e.g. consider value of time distribution)



Thank you



Smarter Cities/Smarter Mobility