



Developments in Integrated Modeling of Activity-Travel Demand and Dynamic Network Flows

Hani Mahmassani

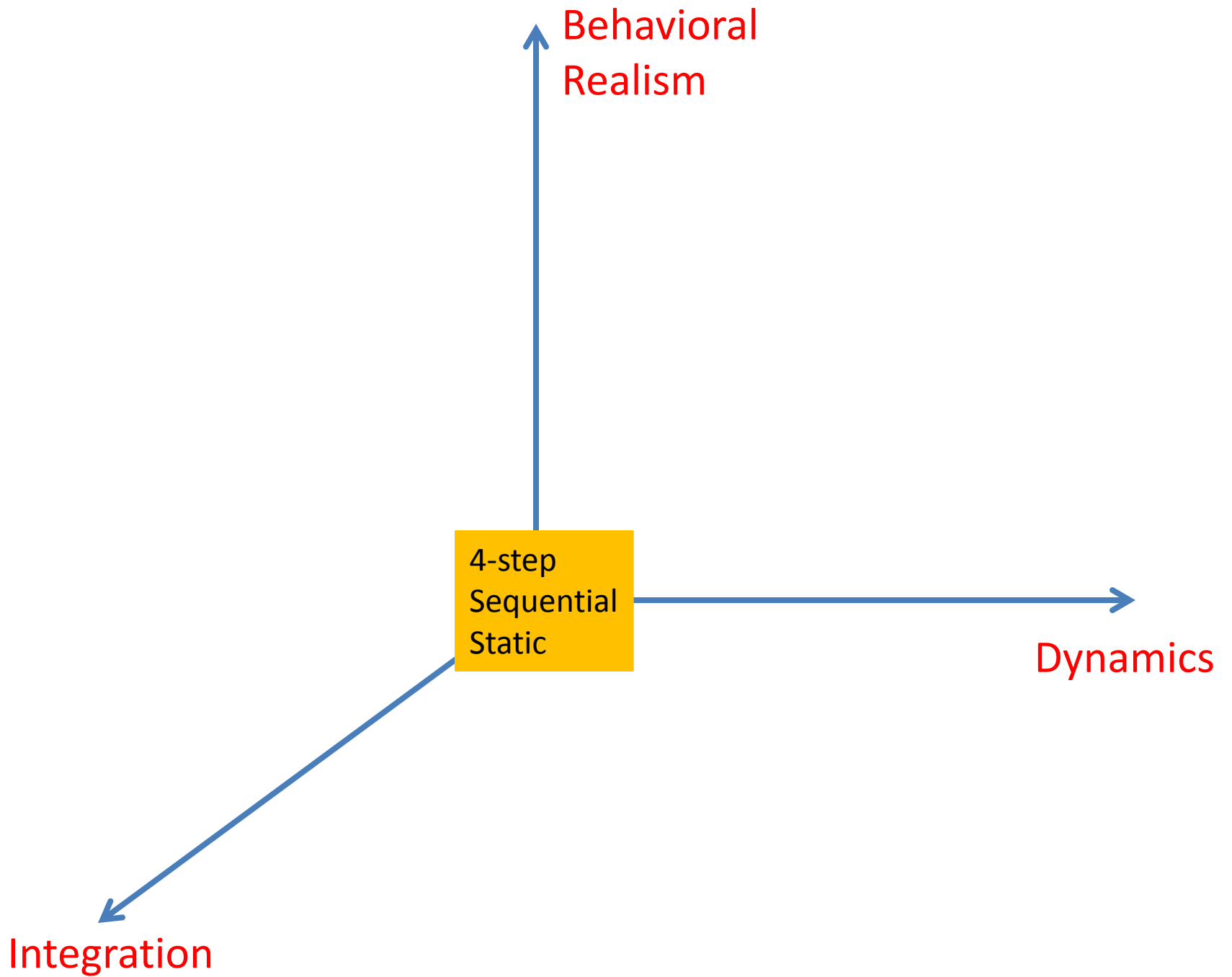


EXECUTIVE SUMMARY

- Existing static assignment tools inadequate for incorporating user responses (e.g. to dynamic prices, reliability) and activity models: require time-varying representation of flows in networks
- Simulation-based DTA methods provide appropriate platform for integrating advanced user travel-activity behavior models
- DTA tools used in practice still lack several key features
 - Limited to route choice as only user choice dimension
 - Do not capture user heterogeneity
 - Cannot generate travel time reliability measures as path LOS attributes
 - Do not produce distributional impacts of contemplated projects/ measures (social justice)
 - Limited applicability of dynamic equilibrium procedures to large-scale regional networks

EXECUTIVE SUMMARY II

- Recent SHRP-2 projects (e.g. C04, L04) have developed the methodologies to integrate user response models in network simulation procedures, for application over the near, medium and long terms
- The algorithms solve for a multi-criterion dynamic stochastic user equilibrium with heterogeneous users in response to dynamic prices, and congestion-induced unreliability
- The integrated procedures are demonstrated on the New York regional network, using advanced demand models developed in Project SHRP-2 C04 on the basis of actual data, coupled with the algorithmic procedures developed and adapted for large-scale network implementation.



Activity-scheduling,
real-time response to information

Activity-based models

Trip chains

Disaggregate, choice models

**Behavioral
Realism**

Prospect theory, Cumulative PT

Learning dynamics

Bounded rationality, thresholds, heuristics,
Computational process models

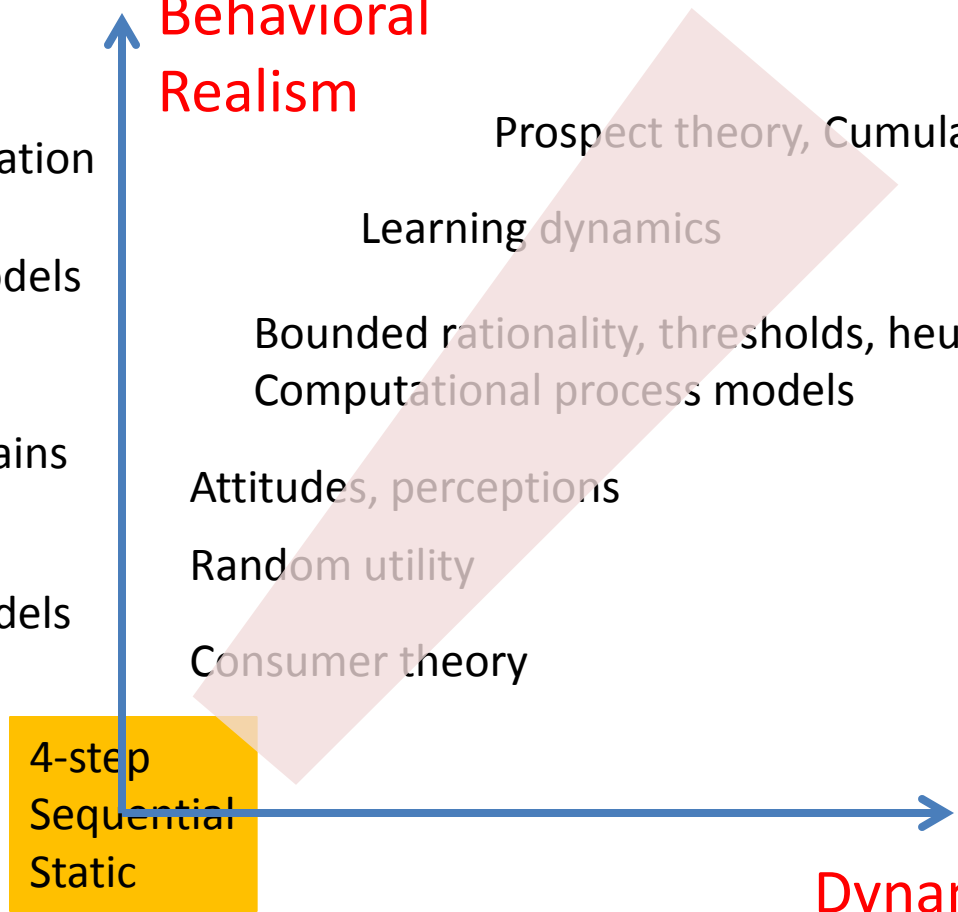
Attitudes, perceptions

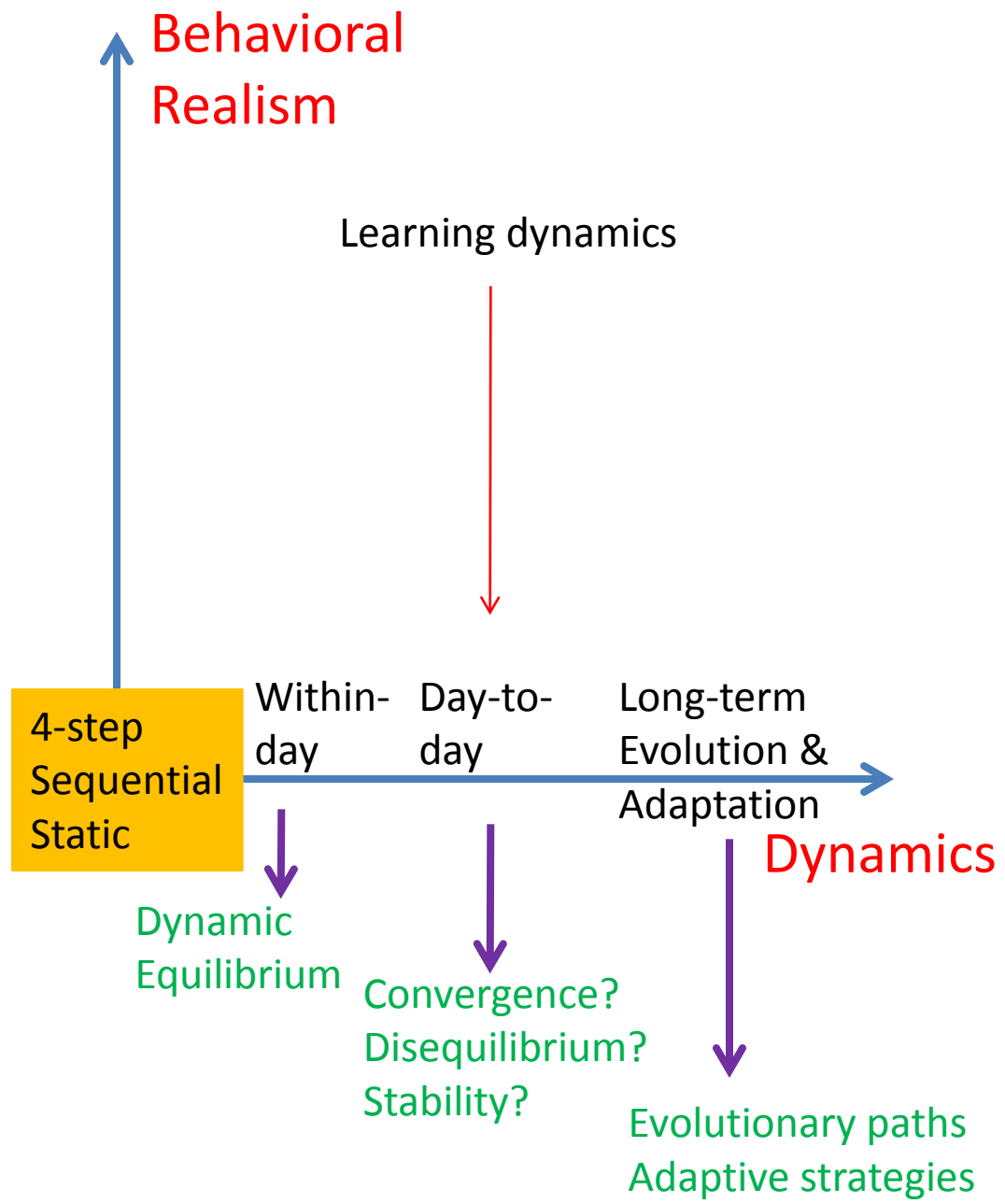
Random utility

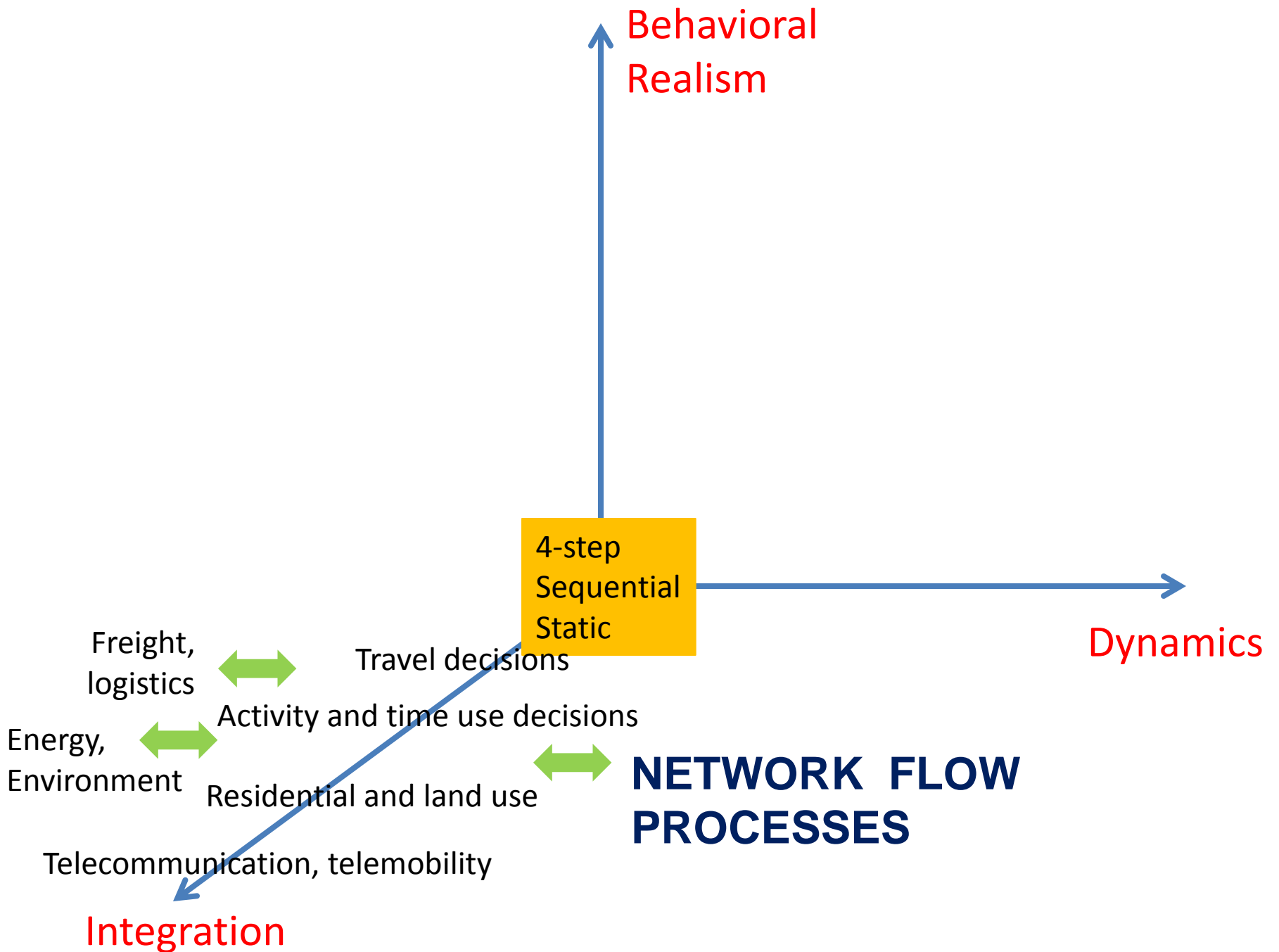
Consumer theory

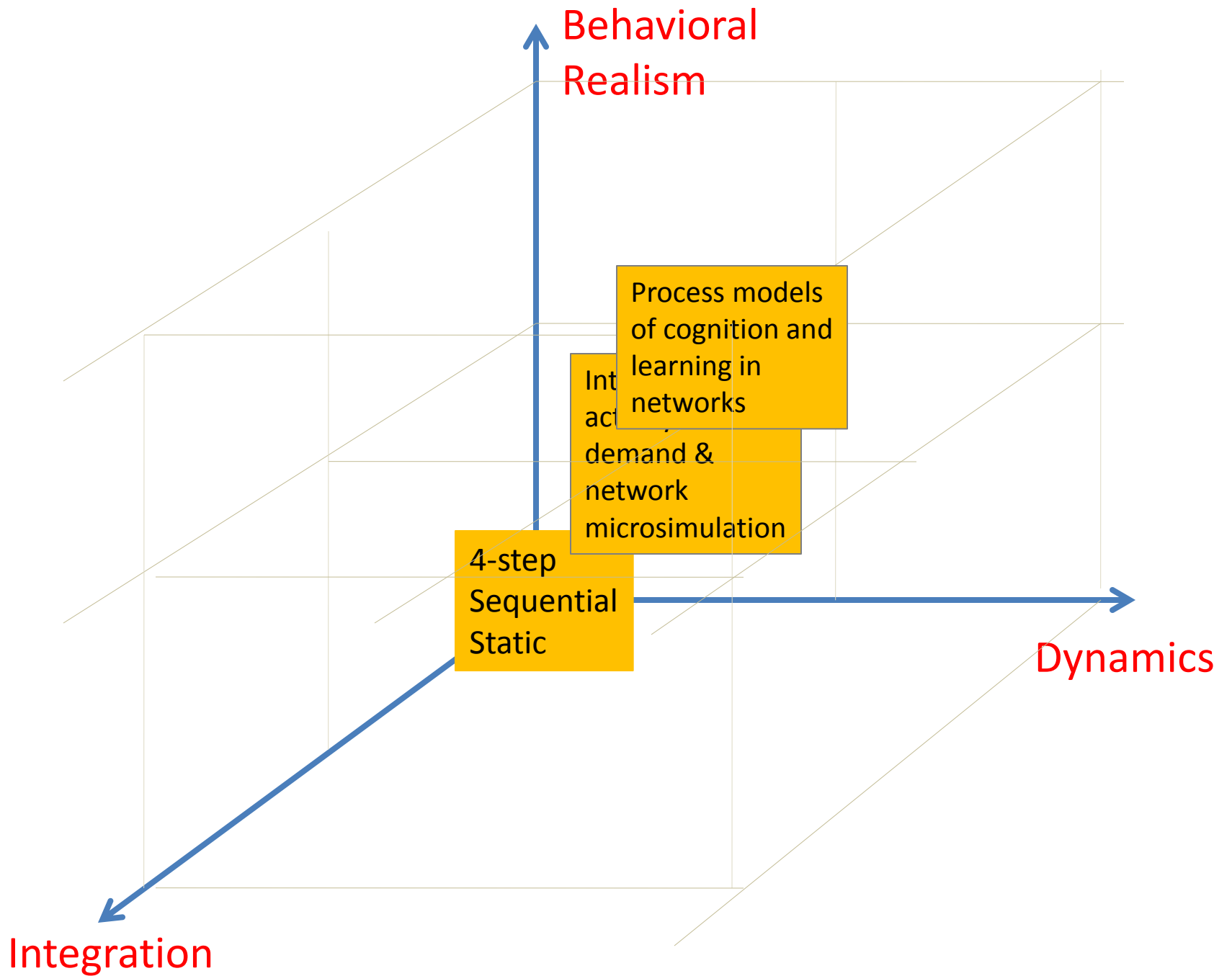
4-step
Sequential
Static

Dynamics









State of Practice in Network Modeling

1. Most agencies use static assignment models, often lacking formal equilibration, with very limited behavioral sensitivity to congestion-related phenomena (incl. reliability)
2. Some agencies use traffic microsimulation models downstream from assignment model output, primarily for local impact assessment
3. Time-dependent (dynamic) assignment models continuing to break out of University research into actual application— market growing, still fragmented, with competing claims and absence of standards:
 - existing static players adding dynamic simulation-based capabilities,
 - existing traffic microsimulation tools adding assignment (route choice) capability, often in conjunction with meso-simulation
 - standalone simulation-based DTA tools

State of Practice in Network Modeling (ctd.)

4. Applications to date complementary, not substitutes, for static assignment; primary applications for operational planning purposes: work zones, evacuation, ITS deployment, HOT lanes, network resilience, etc... Still not introduced in core 4-step process, nor integrated with activity-based models
5. Existing commercial software differs widely in capabilities, reliability and features; not well tested. So-called open source is illusion for practice – no QA, nor accountability.
6. Equilibration for dynamic models not well understood, and often not performed
6. Dominant features, first introduced by DYNASMART-P in mid 90's:
 - Micro-assignment of travelers; ability to apply disaggregate demand models
 - Meso-simulation for traffic flow propagation: move individual entities, but according to traffic flow relations among averages (macroscopic speed-density relations): faster execution, easier calibration
 - Ability to load trip chains (first tool with this capability, essential to integrate with activity-based models)

Responses to Pricing, in Existing Network Models

1. Route choice main dimension captured; replace travel time by travel cost in shortest path code, assuming constant VOT.
2. When multiple response classes recognized, discrete classes with specific coefficient values are used; number of classes can increase rapidly; not too common in practice.
3. Reliability is almost never considered.

DELIVERING THE METHODS: SIX KEY CHALLENGES

- ADVANCED BEHAVIOR MODELS C04
- HETEROGENEOUS USERS C04, C10?
- INTEGRATION WITH NETWORK MODELS:
THE PLATFORM– SIMULATION-BASED MICRO-
ASSIGNMENT DTA C04, L04, C10
- GENERATE THE ATTRIBUTES: RELIABILITY IN
NETWORK LEVEL OF SERVICE L04
- CONSISTENCY BETWEEN BEHAVIOR (DEMAND) AND
PHYSICS (SUPPLY): EQUILIBRATION C04, C10?
- PRACTICAL LARGE NETWORK APPLICATION:
INTELLIGENT IMPLEMENTATION C10?



User Heterogeneity



User Heterogeneity

- Trip-makers choose their paths based on many criteria, including **travel time**, **travel reliability** and **out-of-pocket cost**, and with heterogeneous perceptions.
- Empirical studies (e.g. Hensher, 2001; Cirillo et al. 2006) found that the **VOT** varies significantly across individuals.
- Lam and Small (2001) measured the **value of reliability (VOR)** of \$15.12 per hour for men and \$31.91 for women based on SP survey data.



Beyond Value of Time...

User Heterogeneity

- Present in valuation of key attributes, and risk attitudes
 - Value of schedule delay (early vs. late, relative to preferred arrival time), critical in departure time choice decisions.
 - Value of reliability.
 - Risk attitudes.

Causes significant challenge in integrating behavioral models in network simulation/assignment platforms

Estimation Results Route Choice

Model NYC Area

Model			Lognormal [-1.00,1.00]	
Description	Congested Time, Cost, Toll Bias and Std. Dev.		Congested Time, Cost, Toll Bias and Std. Dev.	
Number of Observations	1694		1694	
Likelihood with Zero Coefficients	-1174.1913		-1174.1913	
Likelihood at Convergence	-1017.4036		-1015.6495	
Parameter	Coefficient	T-Statistic	Coefficient	T-Statistic
Contant for Toll Route	-1.0155	-11.794	-1.0512	-14.041
Highway Cost (Dist*16+Tolls, cents) by Occupancy	-0.0010	-2.058	-0.0010	-2.350
Congested Time (minutes)	-0.0430	-5.569	-3.1732	-18.155
Congested Time on Highways (minutes)	---	---	---	---
Congested Time on Non-Highway Roads (minutes)	---	---	---	---
Congested Time on Roads with v/c => 0.9 (minutes)	---	---	---	---
Congested Time on Roads with v/c < 0.9 (minutes)	---	---	---	---
Standard Deviation - Congested Time per Mile	-0.7344	-0.650	-0.7333	-1.312
Error Term Parameters				
Varince log-Beta-Congested Time	---	---	1.0142	6.357
Values of Time (\$/hr)				
Mean Based on Congested Time	25.80		28.92	
Standard Deviation Based on Congested Time	---		15.42	

Dealing with Heterogeneity in Existing Network Models

1. Ignore: route choice main dimension captured; replace travel time by travel cost in shortest path code, assuming constant VOT.
2. When multiple response classes recognized, discrete classes with specific coefficient values are used; number of classes can increase rapidly; not too common in practice.
2. Recent developments with simulation-based DTA:

Heterogeneous users with continuous coefficient values; made possible by

Breakthrough in parametric approach to bi-criterion shortest path calculation.

Include departure time and mode, in addition to route choice, in user responses, in stochastic equilibrium framework

Efficient implementation structures for large networks: Application of integrated model to New York Regional Network.

Selected Developments in Flow Simulation for Network Application

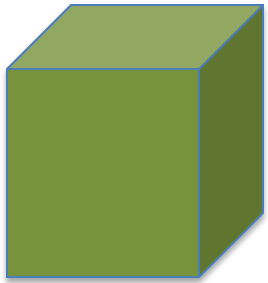
- Capturing **user heterogeneity**
- Convergence of micro and meso level models → *particle-based* models
- Incorporating sources of variability in both micro and meso levels
- **Vehicle trajectories as unifying concept** for output processing, measurement, and tying theoretical development to empirical validation
- Modeling **flow breakdown**: micro mechanisms, collective phenomenon



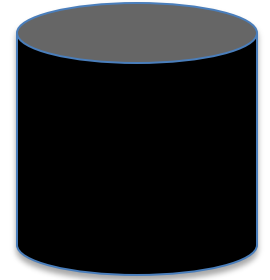
Integration Issues

Integration Issues

- As demand models reflect greater behavioral realism, supply side simulation models need to incorporate these improvements as well.
- Current travel choice models reflect the following:
 - Random heterogeneity and taste variations
 - Serial correlation among repeated choices
 - Non-IIA substitution pattern among alternatives; general error structures
 - Process models for activity choice and scheduling
- Incorporating these behavioral extensions into supply-side (network) models requires producing the attributes included in the estimated choice models
→ implications for core algorithms (e.g. path finding) and consistency-seeking (equilibrium) procedures.

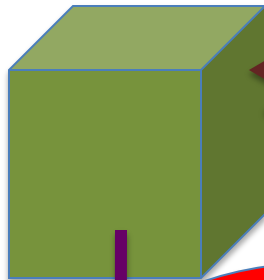


DEMAND

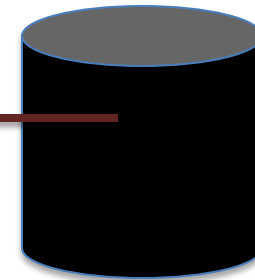


SUPPLY

~~INTEGRATE ?~~



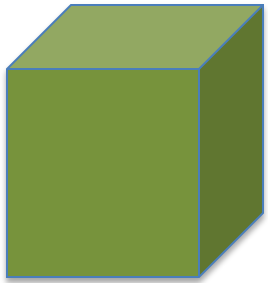
CONVERT



CONVERT

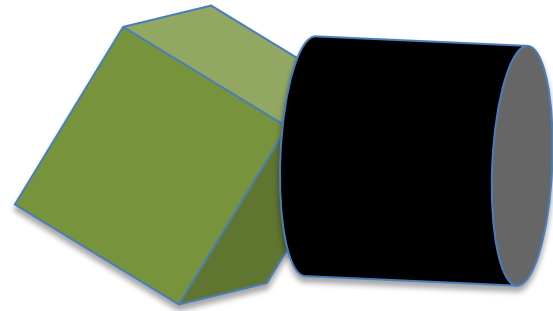
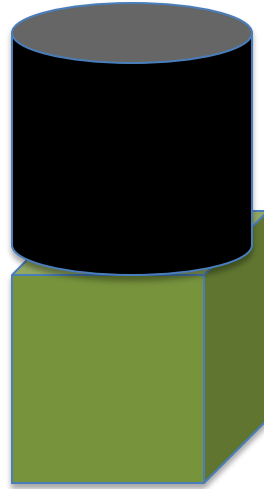
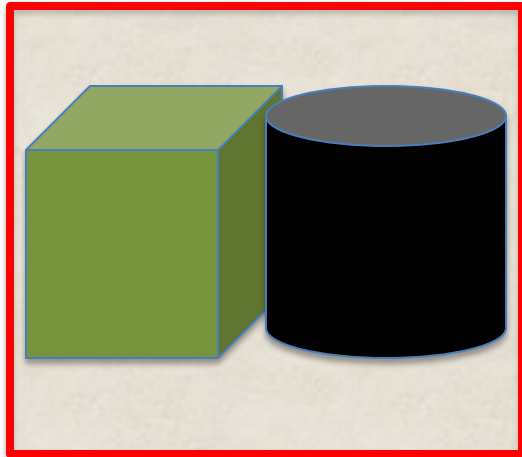
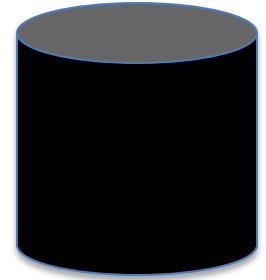
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INTERFACE



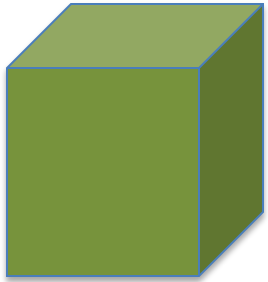
DEMAND

SUPPLY



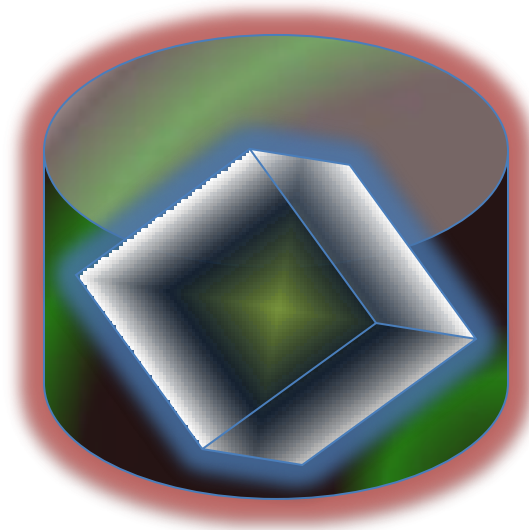
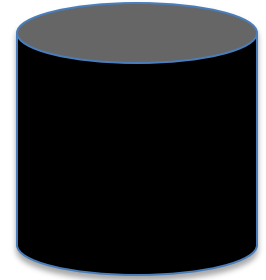
~~INTEGRATE ?~~

JUXTAPOSE



DEMAND

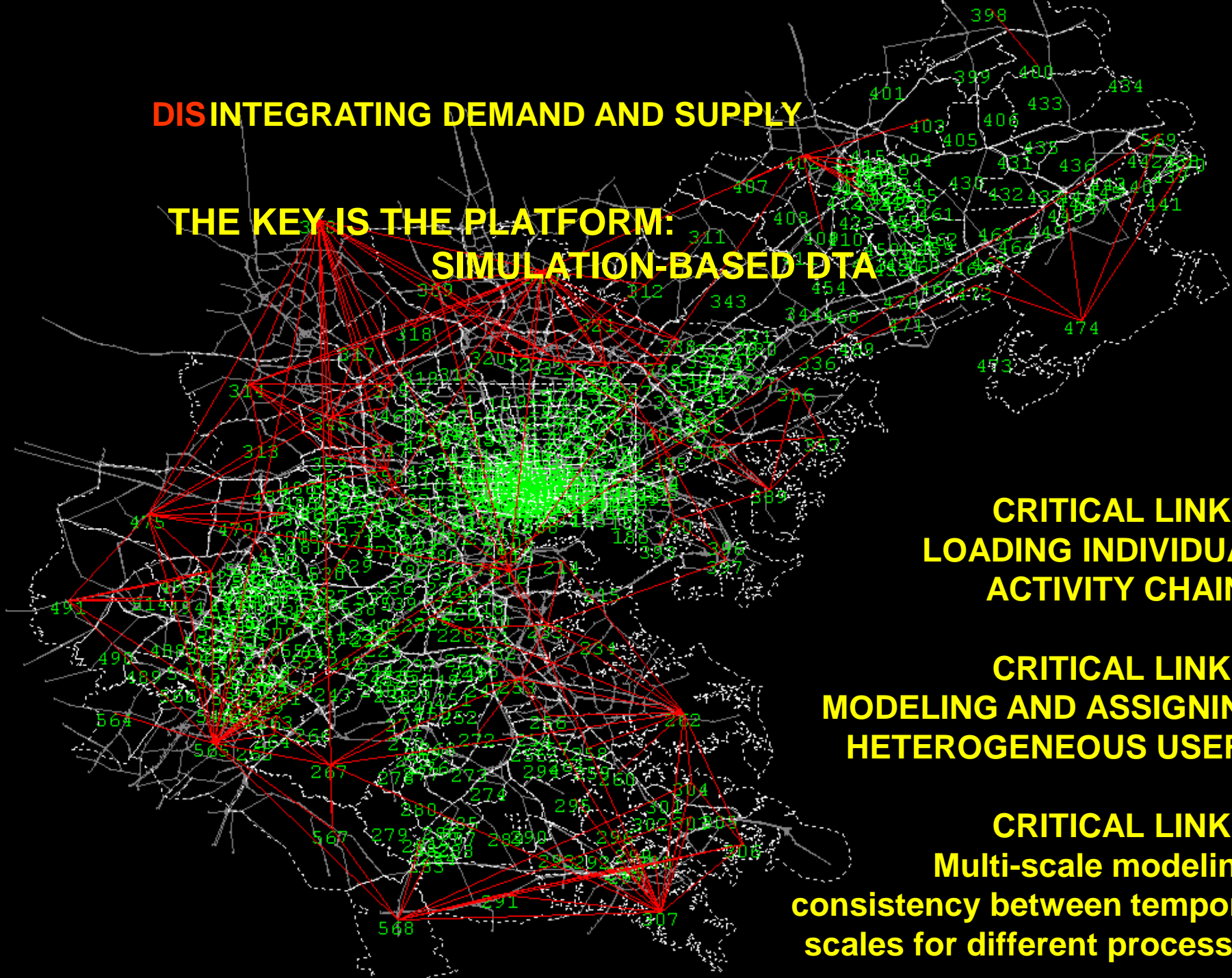
SUPPLY



INTEGRATE?

DISINTEGRATING DEMAND AND SUPPLY

**THE KEY IS THE PLATFORM:
SIMULATION-BASED DTA**



**CRITICAL LINK 1:
LOADING INDIVIDUAL
ACTIVITY CHAINS**

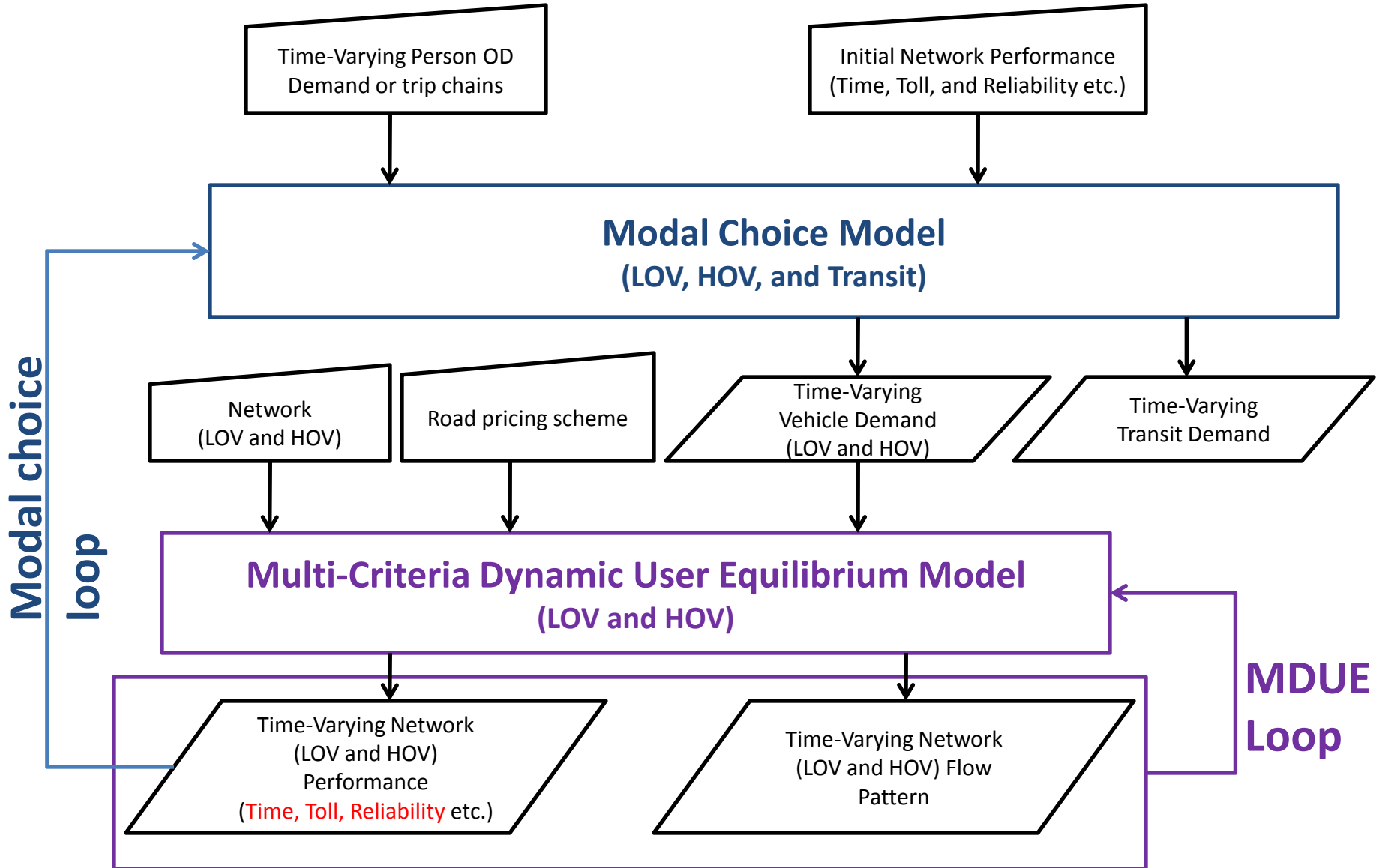
**CRITICAL LINK 2:
MODELING AND ASSIGNING
HETEROGENEOUS USERS**

**CRITICAL LINK 3:
Multi-scale modeling:
consistency between temporal
scales for different processes**

Example: Mode choice and multi-criteria dynamic user equilibrium model

- Assumptions:
 - Given network with discretized planning horizon
 - Given time-dependent OD person demand
 - Given calibrated mode choice model (LOV, HOV, and Transit)
 - Given VOT distribution
 - Given road pricing scheme
- Solve for:
 - Modal share for each mode (e.g., LOV, HOV, and Transit)
 - Assignment of time-varying travelers for each mode (LOV, HOV) to a congested time-varying multimodal network under multi-criteria dynamic user equilibrium (MDUE) conditions
- Methodology:
 - Descent direction method for solving the modal choice problem
 - Simulation-based column generation solution framework for the MDUE problem

Modeling framework



Model implementation

- Short-term Integration
 - Mode choice loop integrated in model framework
 - MNL, GEV, and Mixed Logit (random coefficients) Mode Choice model
- Medium-term Integration
 - Departure time choice dimension; activity-based models
 - MNL, GEV, Mixed Logit (Random coefficients), and Mixed Logit (Serial Correlation) Choice Model
- Long-term Integration
 - Activity scheduling models, time use, process models

Solution Algorithm for MDUE– UE with random VOT and VOR

For short-term integration: incorporate
MNL/GEV mode choice dimension and
heterogeneous users for mode and
route choices

Generalized Cost

- Generalized cost is defined as a summation of travel monetary cost (TC), travel time (TT) and travel time variability/reliability (TV).



- VOT is considered as a continuous random variable distributed across the population of trip-makers with the density functions:



- $VOR \beta$ is considered as a constant for all trip-makers

Input
OD demand, link tolls, VOT distribution, VOR and initial paths and path assignment

1. Initialization Set $k = 0$
Perform a MDNL by traffic simulation to evaluate initial path assignment and obtain experienced path travel time (TT), and travel cost (TC)

2. PAM
Obtain the set of time-dependent extreme efficient path, breakpoints of VOT and their generalized costs to define the multi-user classes; **augment the path set if new paths are found**

3. Convergence Checking
(a) no new path
(b) $k = K_{max}$

Stop and output solution r^k

**Outer Loop:
Path Generation**

4. Initialization Set $l = 0$
Read output of Step 2 from PAM: **current path set and path assignment r^l**

5. Update Path Assignment
Determine path assignments r^{l+1} by multi-class path flow updating/equilibrating. Set $l = l + 1$

6. Multi-class Dynamic Network Loading
Perform a MDNL by the traffic simulator to evaluate new path assignment r^l and obtain TC and TT.

**Inner Loop:
Solve RMDUE**

Return to outer loop with current link travel times.
Set $k = k + 1$

7. Convergence Checking
(a) GAP
(b) $l = l_{max}$

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OD demand, link tolls, VOT distribution, VOR and initial paths and path assignment

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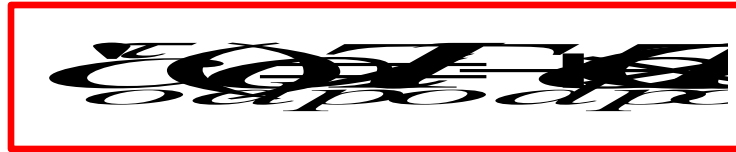
RELIABILITY

Return to outer loop with current link travel times. Set $k = k + 1$

Parametric Analysis Method (PAM)

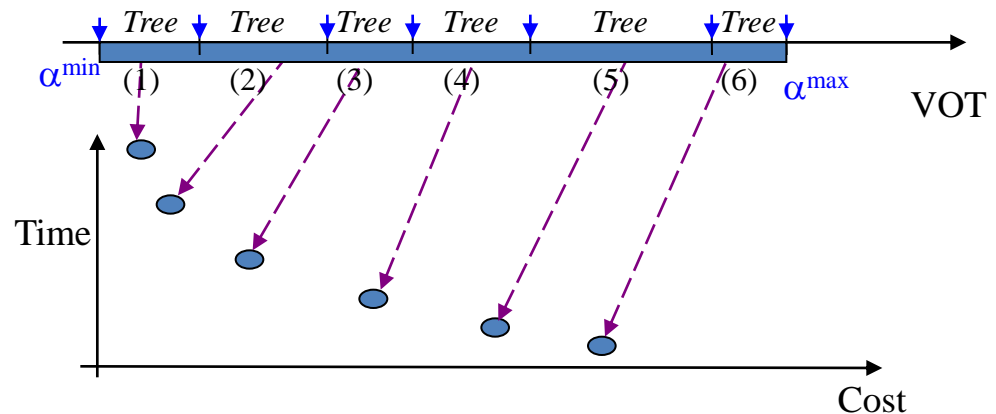
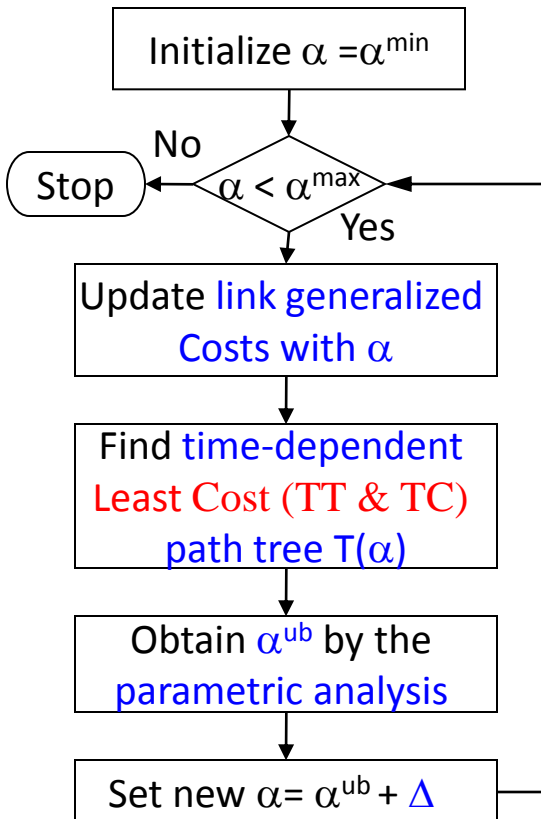
Input: from traffic simulator

- Time-dependent travel time (TT)
- Time-dependent travel cost (TC)



Output: for each dest. j

- A path tree
- VOT Breakpoints



Parametric Analysis Method (PAM)

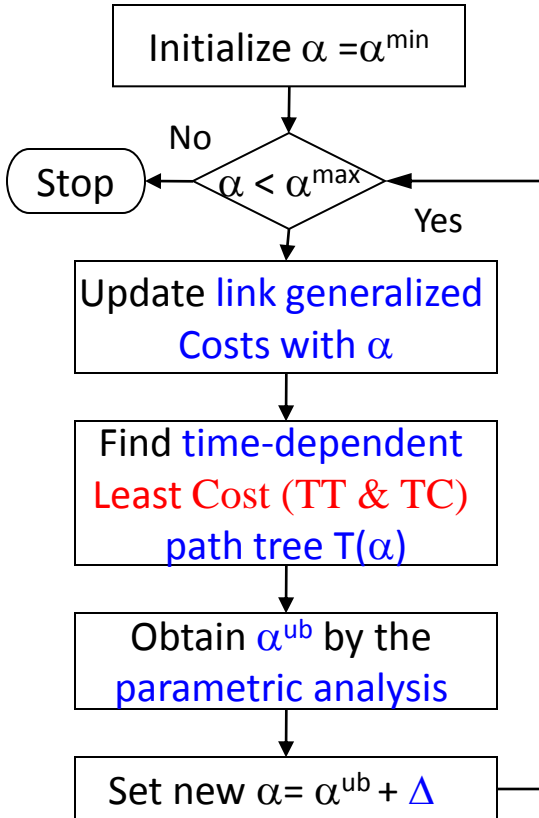
Input: from traffic simulator

- Time-dependent travel time (TT)
- Time-dependent travel cost (TC)



Output: for each dest. j

- A path tree
- VOT Breakpoints



Read VOT break points and path set for every (i,j,t)

Compute TV_{odp} for each path in the path set

Start with the first VOT

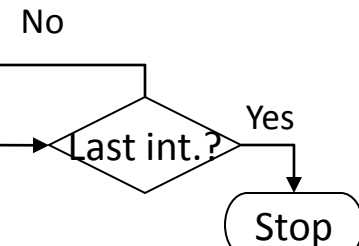
Find time-Dependent Least Generalized Cost Path And move to next interval

No

Last int.?

Yes

Stop



Parametric Analysis Method (PAM)



Output: for each dest. j

- A path tree
- VOT Breakpoints

Read VOT break points and path set for every (i,j,t)

Compute TV_{odp} for each path in the path set

Start with the first VOT

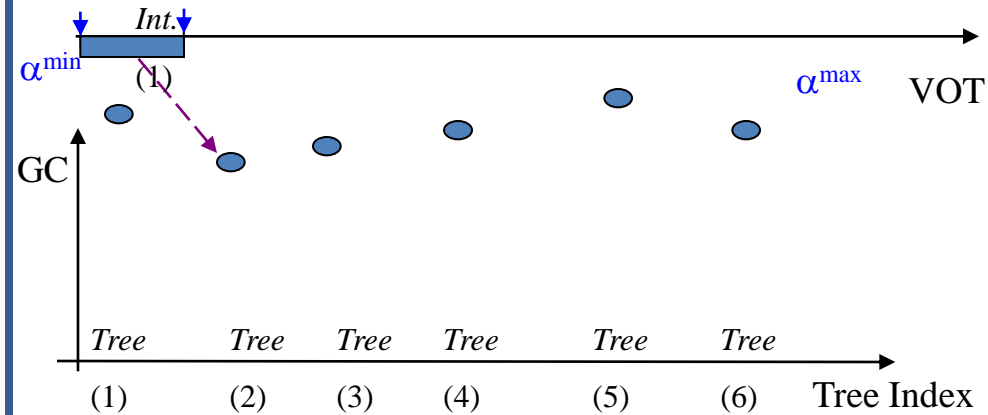
Find time-Dependent **Least Generalized Cost** Path
And move to next interval

No

Last int.?

Yes

Stop



Parametric Analysis Method (PAM)



Output: for each dest. j

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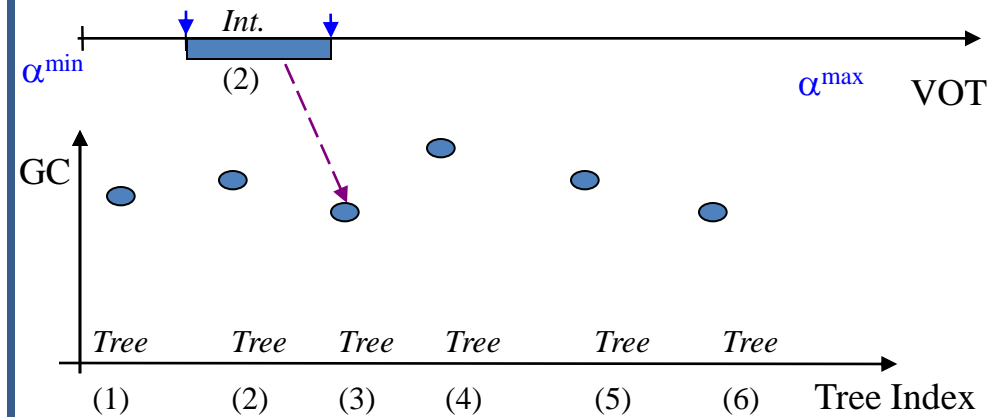
Find time-Dependent **Least Generalized Cost** Path
And move to next interval

No

Last int.?

Yes

Stop



Parametric Analysis Method (PAM)



Output: for each dest. j

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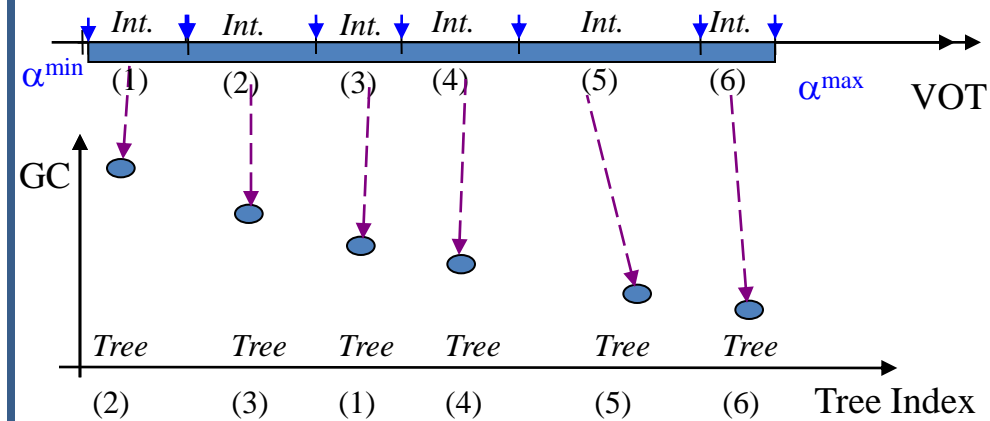
Find time-Dependent **Least Generalized Cost** Path
And move to next interval

No

Last int.?

Yes

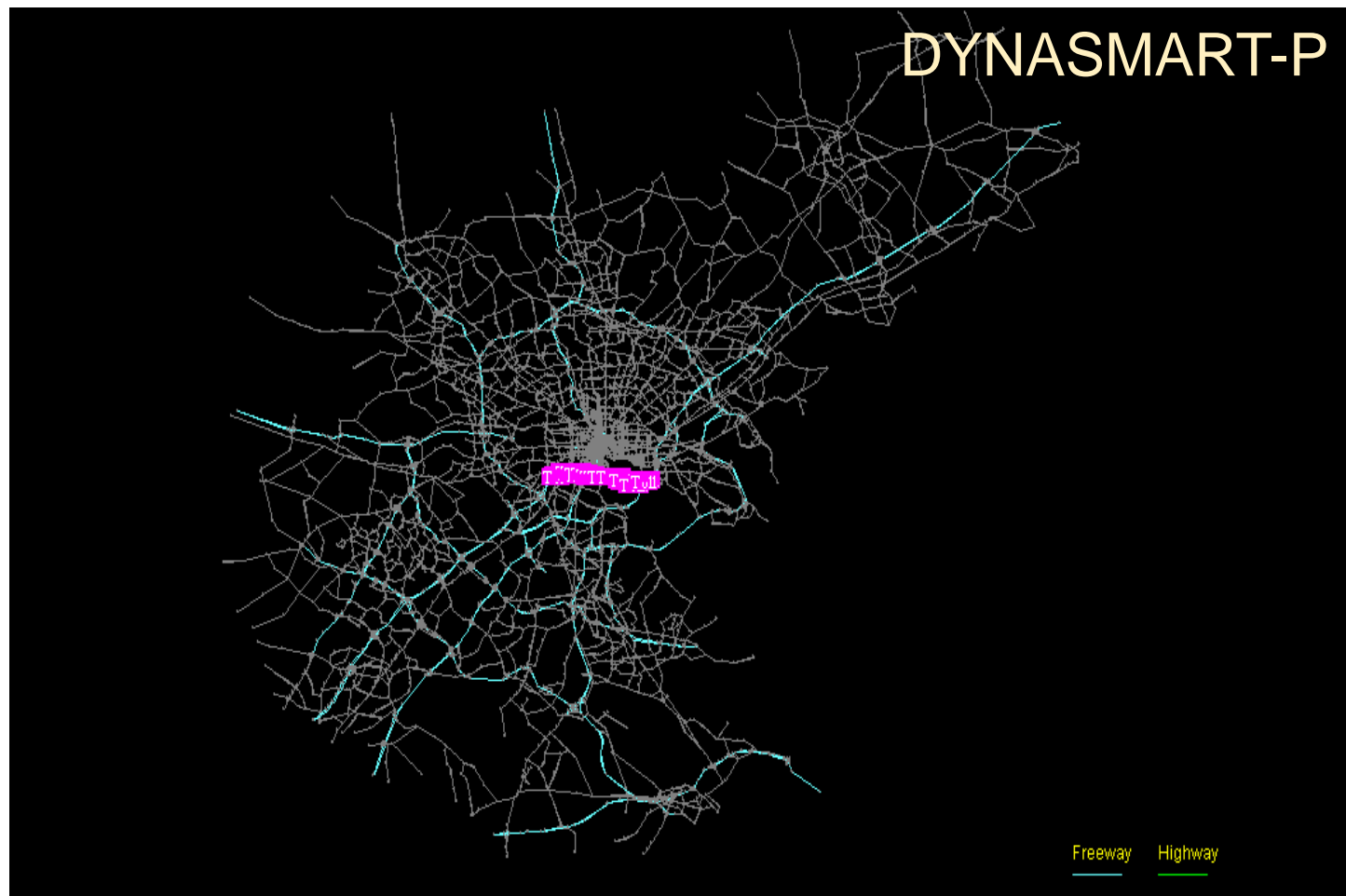
Stop



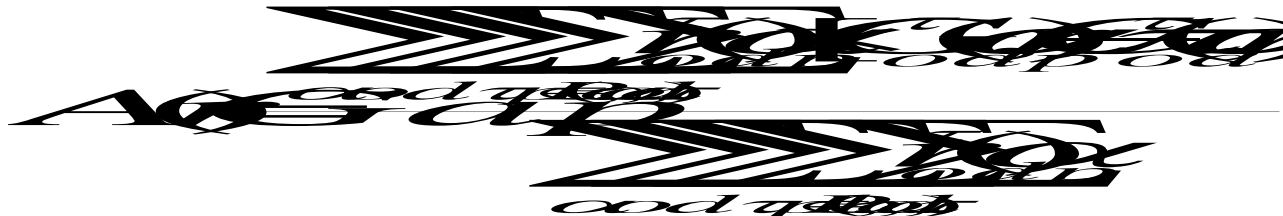
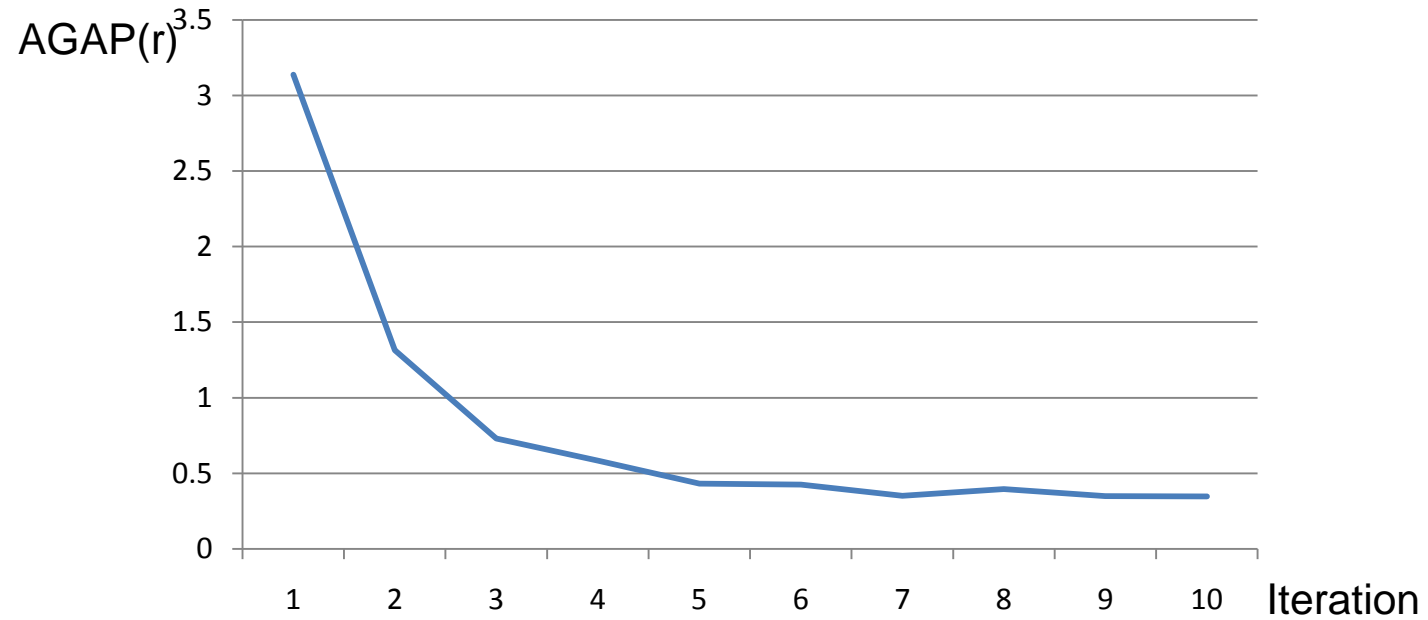
Numerical Results: Baltimore Network

Application of MDUE Procedure with Heterogeneous Users

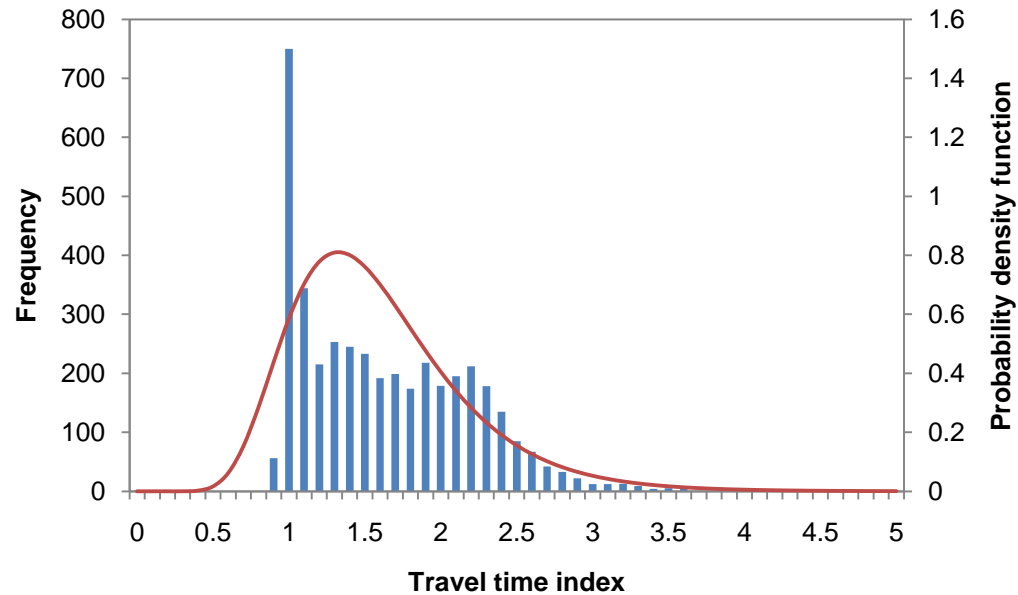
- 6,825 nodes
- 14,317 links
- 570 zones
- Dynamic toll on I-95
- 2-hour (7-9Am) morning peak time-varying OD demand with 898,878 vehicles



Convergence Pattern



Generate Reliability as Network LOS



Challenges in Characterizing Network Variability and Correlations

- Representation of the travel time variability through the network's links and nodes
 - Variability of link travel times
 - Variability of delays associated with movements through the intersections, particularly left-turns
- Strong correlation between travel times in different parts of the network
 - Adjacent links are more likely to experience high delays in the same general time period than unconnected links
 - Difficult to capture these correlation patterns when only link level measurements are available
 - Difficult to derive path-level and OD-level travel time distributions from the underlying link travel time distributions

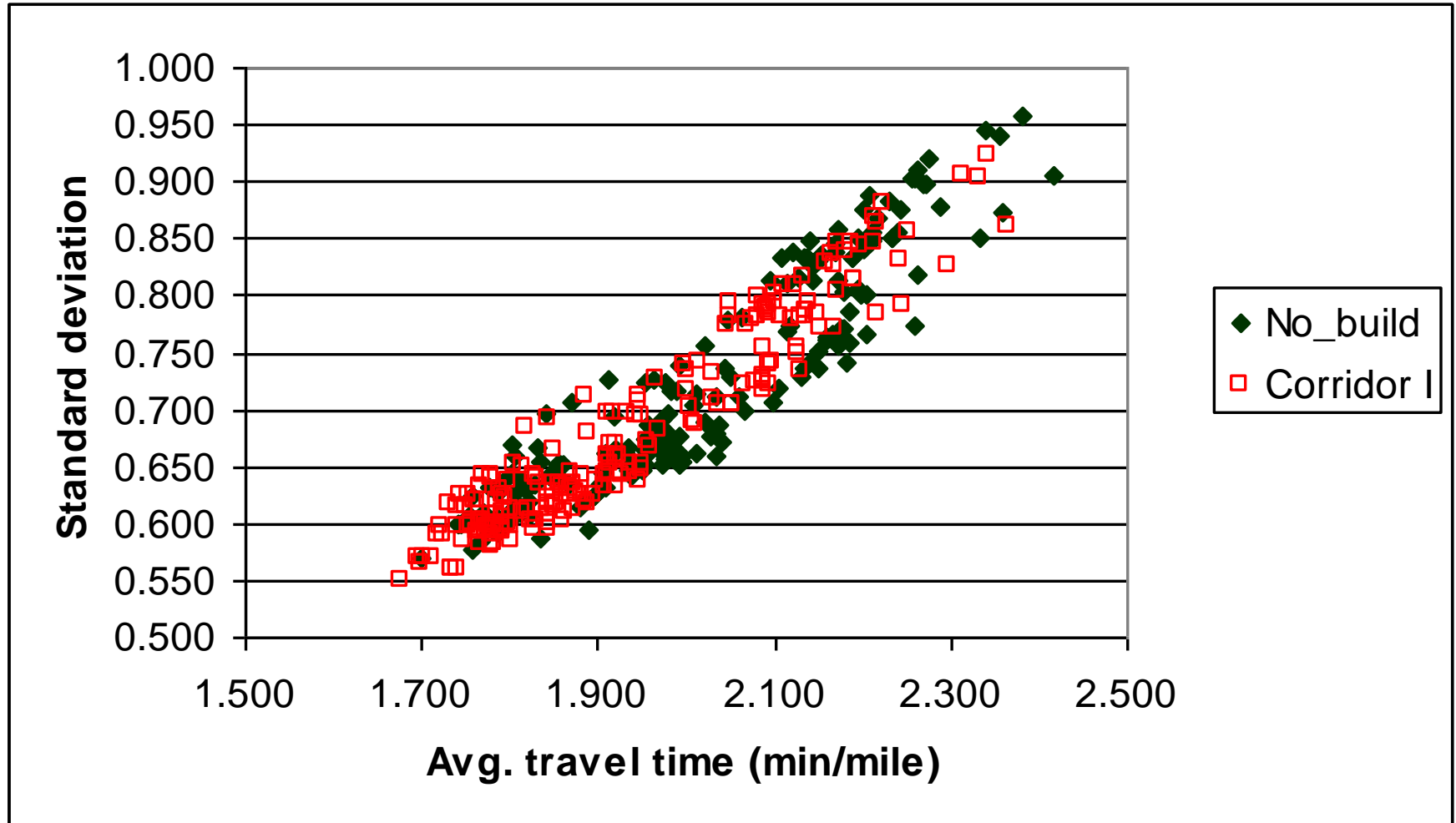
Travel Reliability Measure

- Given a path set for each (i,j,τ) for a given possible VOT range by PAM, we re-evaluate the path generalized cost by adding a travel time reliability measure $TV_{i,j}^\tau$
- In current implementation, exploit relation between std dev per unit distance and mean time per unit distance at network level
- In future work, could estimate std dev per unit distance and mean time per unit distance for specific O-D's and paths from simulation results

Travel Time Reliability

Standard Deviation vs. Average Travel Time (per mile)

(Greater Washington, DC network: OD level variability)



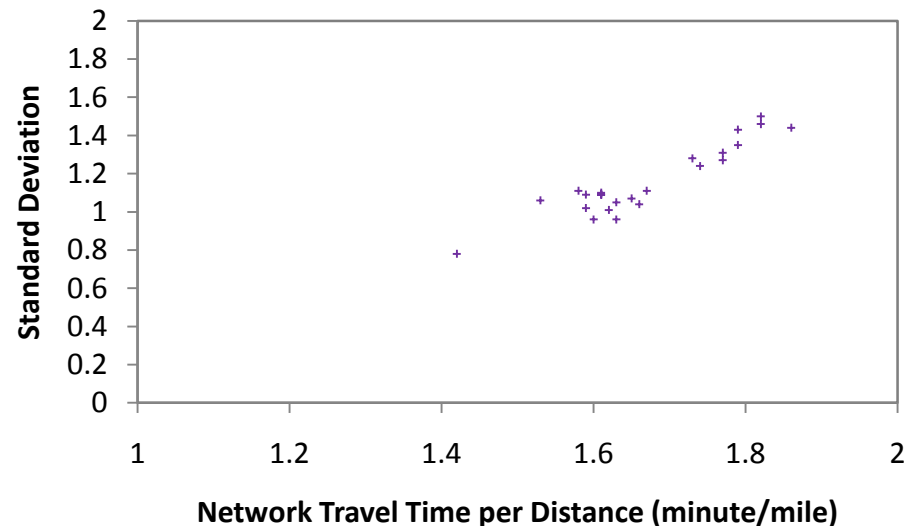
Irvine Network



- Network
 - Freeways I-405, I-5, state highway 133
 - 326 nodes
 - 626 links
 - 61 TAZs
- Demand
 - Two hours morning peak (7-9AM)

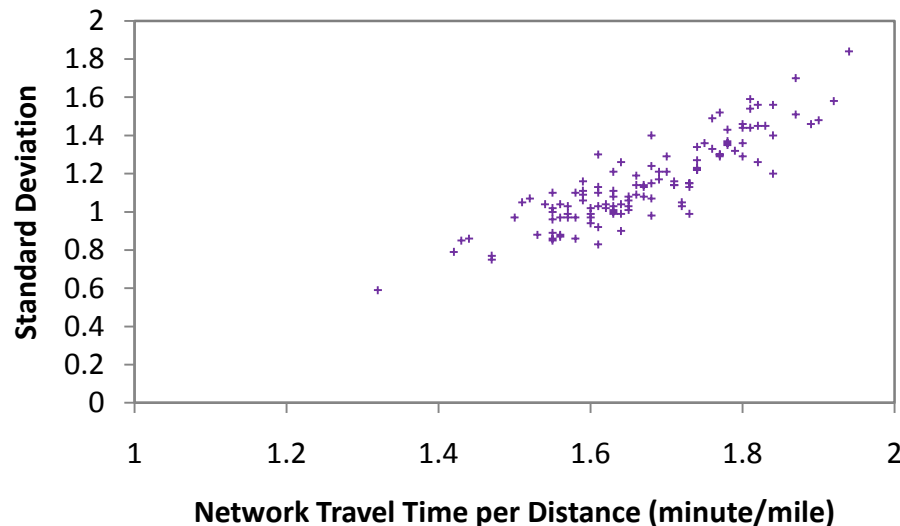
Network Travel Time per Unit Distance and Standard Deviation (5 minute interval)

- Each data point represents the mean and standard deviation of travel times **per mile** for all vehicles departing in 5-minute interval.
- 24 data points for 2-hour demand



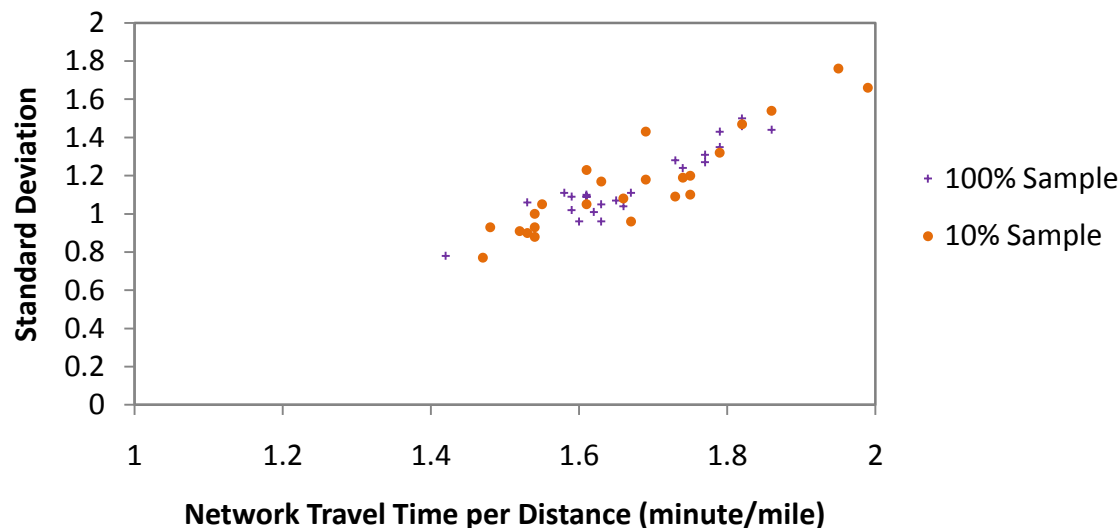
Network Travel Time per Distance and Standard Deviation (1 minute interval)

- Each data point represents the mean and standard deviation of travel times **per mile** for all vehicles departing in 1-minute interval.
- 120 data points for 2-hour demand



Network Travel Time per Distance with Sampling Vehicles

- Each data point represents the mean and standard deviation of travel times **per mile** for all vehicles departing in 5-minute interval.
- 24 data points for 2-hour demand



Vehicle Trajectories: Unifying Framework for Micro and Meso Simulation

- Vehicle trajectory contains the traffic information and itinerary associated with each vehicle in the transportation network, including
 - a set of nodes (describing the path)
 - the travel time on each link along the path
 - the stop time at each node
 - the cumulative travel/stop time
 - possibly lane information

```

**** Output file for vehicles trajectories ****
=====
This file provides all the vehicles trajectories
Veh # 16645 Tag= 2 OrigZ= 5 DestZ= 9 Class= 5 UstmN= 103
DownN= 102 DestN= 11 STime= 70.20 Total Travel Time=
8.49 # of Nodes= 18 VehType 1 LOO 1
102 160 102 103 151 97 89 4 3
24
5 27 28 32 35 39 40 11
==>Node Exit Time Point
0.80 0.90 1.60 2.20 3.00 3.40 3.80 5.00 5.50
5.90
6.00 6.30 6.70 7.10 7.30 7.60 8.20 8.40
==>Link Travel Time
0.80 0.10 0.70 0.60 0.80 0.40 0.40 1.20 0.50
0.40
0.10 0.30 0.40 0.40 0.20 0.30 0.60 0.20
==>Accumulated Stop Time
0.60 0.60 1.20 1.36 1.42 1.44 1.47 2.22 2.57
2.57
2.57 2.57 2.57 2.57 2.57 2.57 2.57 2.57

```


Obtain Vehicle Trajectories from Simulation Models

- Vehicle trajectories could be obtained from all particle-based simulations, regardless of whether the physics underlying vehicle propagation and interactions are captured through microscopic maneuvers or through analytic forms
 - Microscopic simulation models move traffic by capturing individual driver maneuvers such as car following, overtaking, lane changing and gap acceptance decisions.
 - Mesoscopic simulation models move vehicles as individual particles, albeit according to (macroscopic) relations among average traffic stream descriptors (e.g. speed-density relations).
- The realm between micro and meso has narrowed considerably over time—and will continue to do so.
- Trajectories could also be obtained from direct measurement in actual networks: video camera, cell-phone/GPS probes, etc...
- **This enables consistent theoretical development in connection with empirical validation (for L04)**

Application of Integrated Procedures to New York Regional Network

Apply demand and user response models developed
In SHRP-2 Project C04 (w. P. Vovsha, PB Inc.) for NY Metro network:

- route choice model includes time-varying prices, and travel reliability measure
- random value of time (distributed across users)
- mode choice and departure time choice models

in conjunction with

MDUE (multi-criteria Dynamic User Equilibrium and
heterogeneous users to very large scale network

~30,000 Nodes
95,000 Links
3,700 Zones

6-hour AM peak period
5.2 M simulated vehicles



New York Region General View

Freeway Highway

Network Attributes

- Zones
- Signals

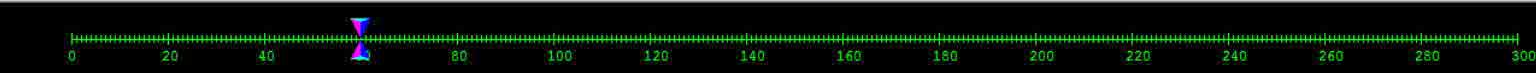
Traffic Attributes

- Density
- Speed
- Queue Length
- Vehicles
- Travel Time Contour
- No Selection

Name	Value
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Name	Value
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CONCLUDING COMMENTS

- We have seen advances in state-of-art in integrating user responses to dynamic pricing, congestion and unreliability in network modeling procedures.
- New methodologies are software independent and can be applied with any simulation-based DTA tool (caveats...)
- Application to very large New York regional network first successful application to network of this size of equilibrium DTA with heterogeneous users.
- Integration process could be improved with additional choice dimensions, and eventually fully-configured activity-based model.

KEY ISSUES and OPPORTUNITIES

- Theoretical constructs:
 - Notions of consistency in stochastic dynamic context
 - ➔ convergence measures?
 - Path dependence in dynamic simulation forecasts
 - Consistency of attribute valuation throughout activity submodels– e.g. should travel time be valued similarly in route vs mode vs departure time choices?
- Methodological issues: multi-scale modeling, path finding, activity scheduling combinatorics, cooperation and competition in multi-agent system
- Application issues: **Planning and Operations Decision Support System**
 - Different applications/problems call for different capabilities: plug-and-play built on basic platform
- Major opportunity: more active tie in with trajectory data from probes and sensor information– responsive, calibrated, relevant platform for decision support

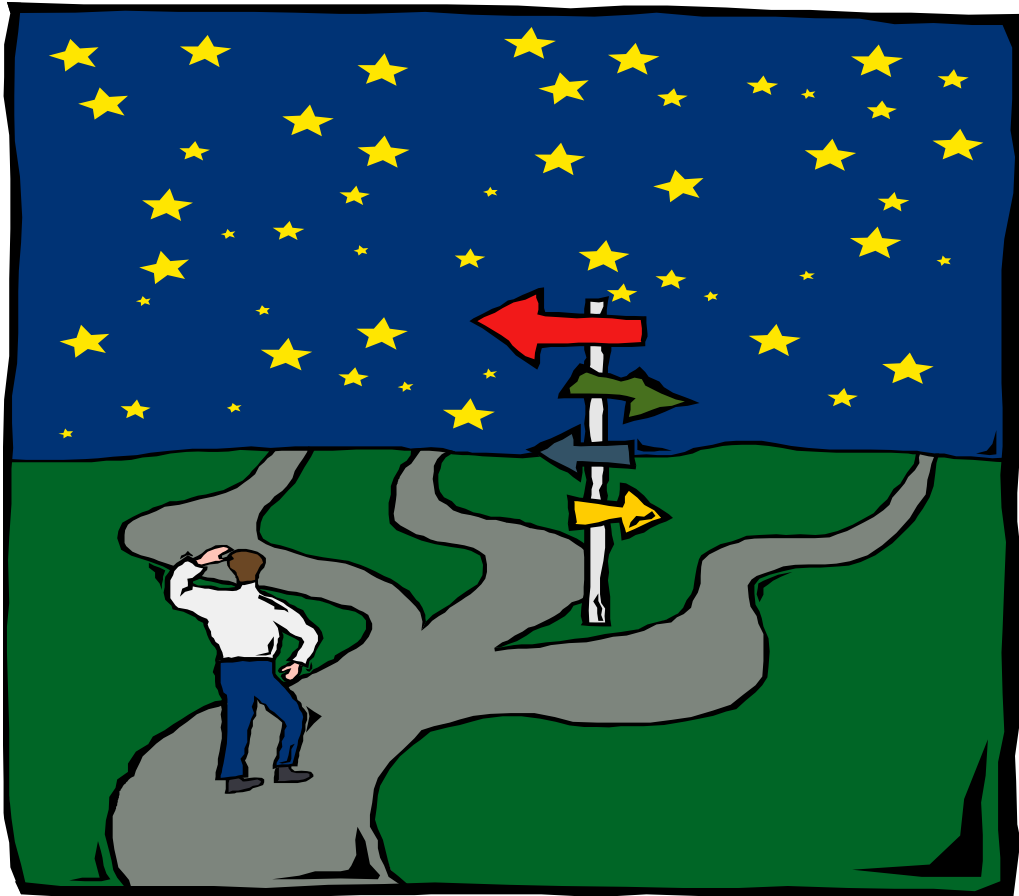
Integrated
activity-based
demand &
network
microsimulation

Process models of
cognition
and learning in
networks

Demand forecasting for planning decisions

- Transportation planning has lacked a forecasting paradigm that recognizes the complex nature of the system and the limitations of available tools
- Behavioral models more for deriving insights and understanding behavior than to serve as crystal ball
- Greater uncertainty in the input (future technology, economy, spatial patterns, lifestyles) than in the tripmaking behavior of users *given* these inputs

Towards new forecasting paradigm...



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