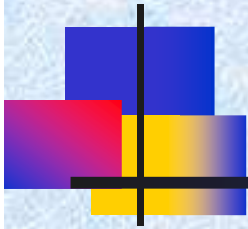


Incorporation of Travel Time Reliability in Integrated Demand and Network Simulation Models



Peter Vovsha, Parsons Brinckerhoff

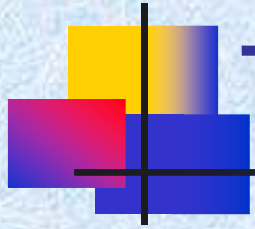


ITM, Tampa, FL, April 28, 2012



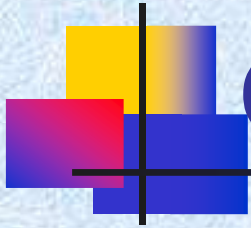
SHRP 2 Projects

- C04 “Improving Our Understanding How Highway Congestion and Pricing Affect Travel Demand”
- L04 “Incorporating Reliability Performance Measures in Operations and Planning Modeling Tools”

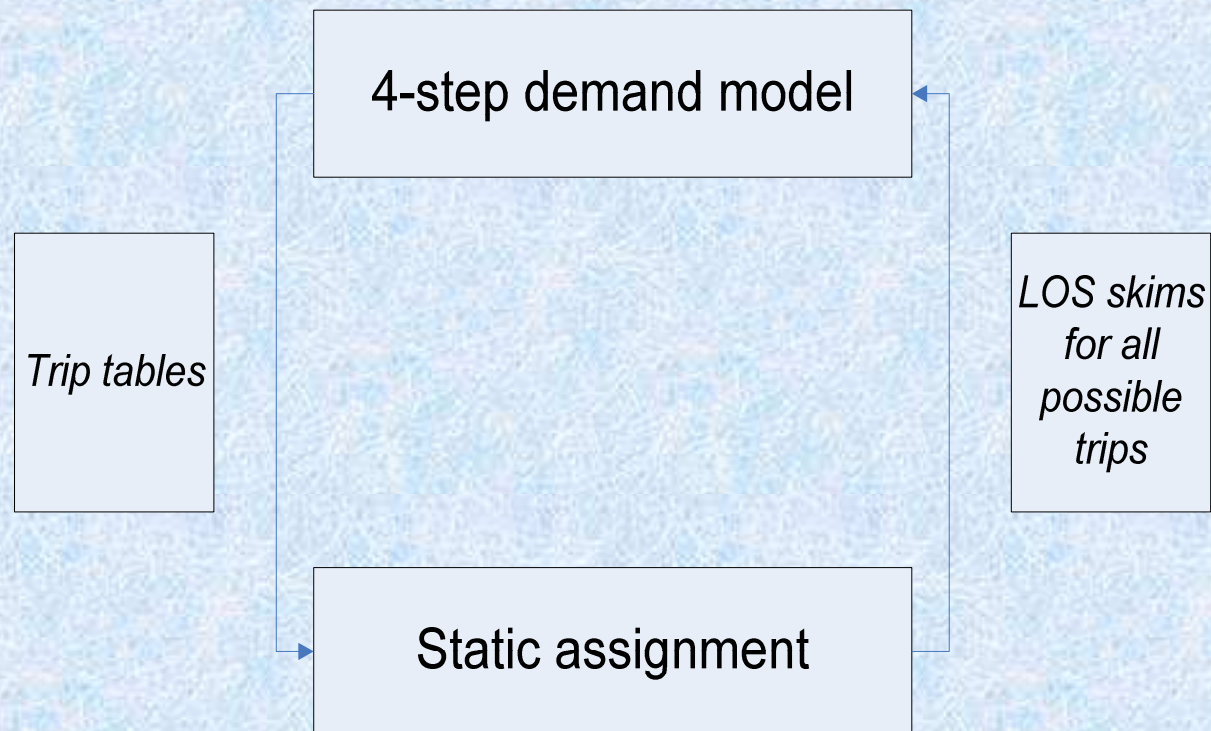


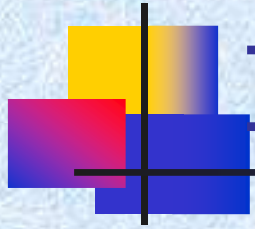
Topics to Discuss / Concepts

- ABM-DTA integration and 2-way linkage:
 - ABM-to-DTA
 - DTA-to ABM
 - Individual schedule consolidation
 - Pre-sampling
- Incorporation of travel time reliability:
 - Perceived time by congestion levels
 - Mean-variance methods
 - Schedule delay methods
 - Temporal utility profiles

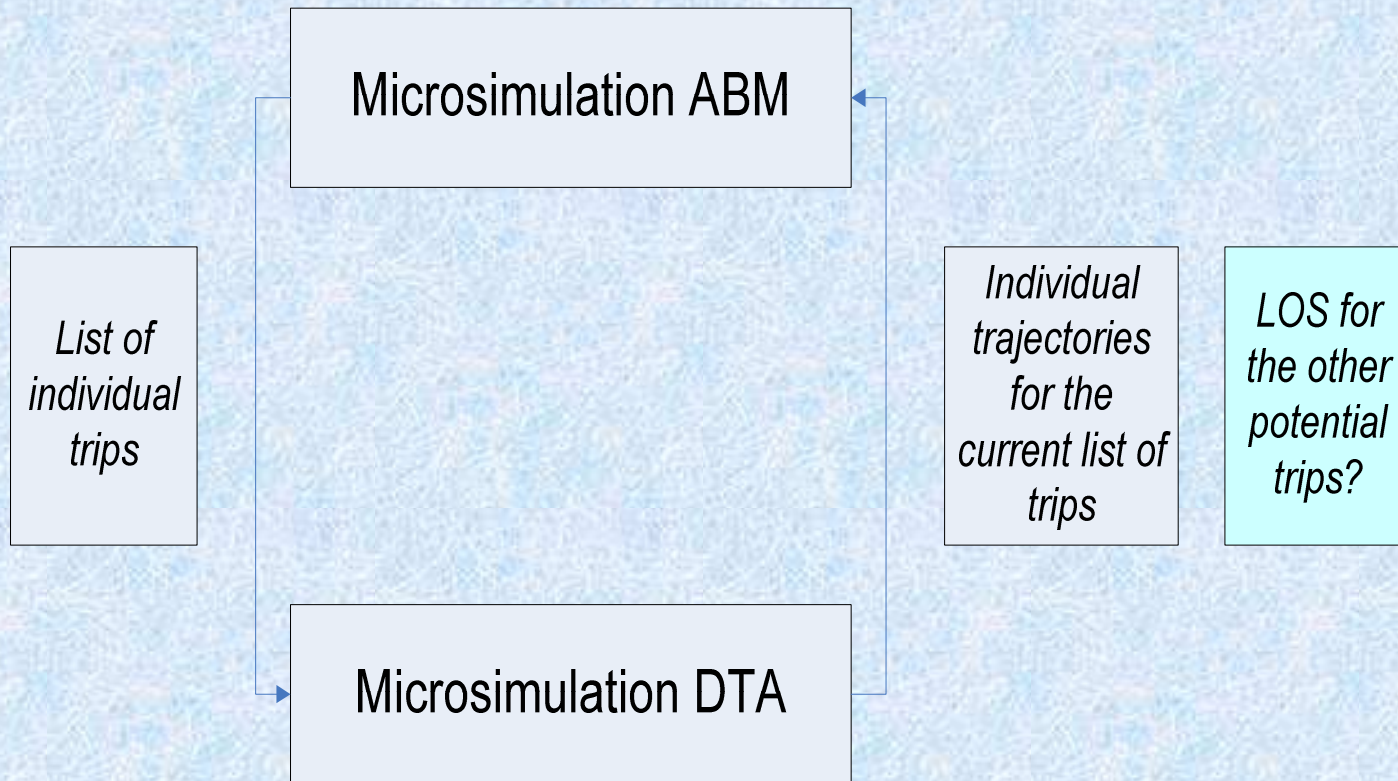


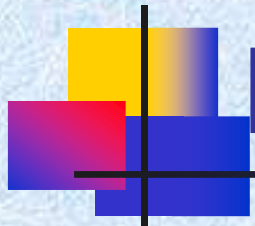
Conventional Integration Scheme



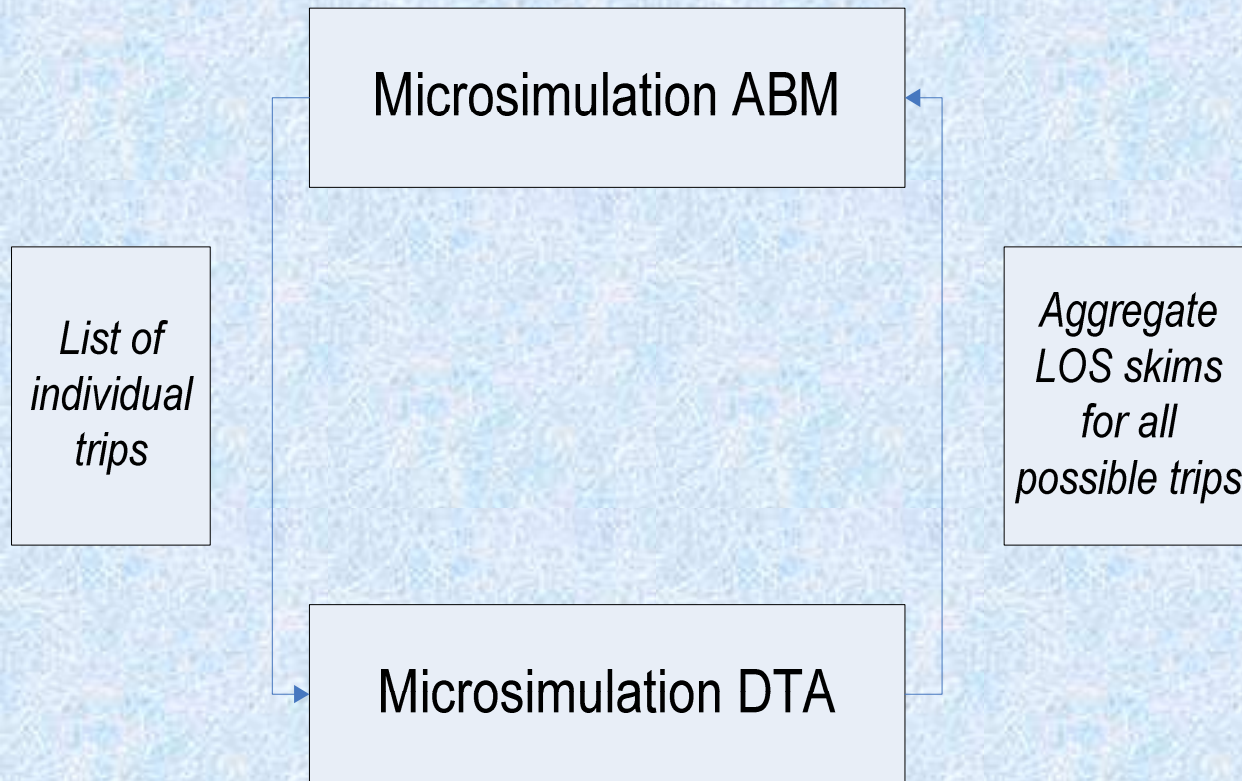


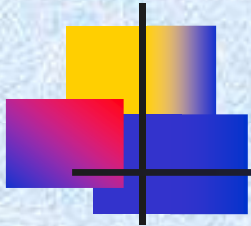
Integration Issue DTA-to-ABM



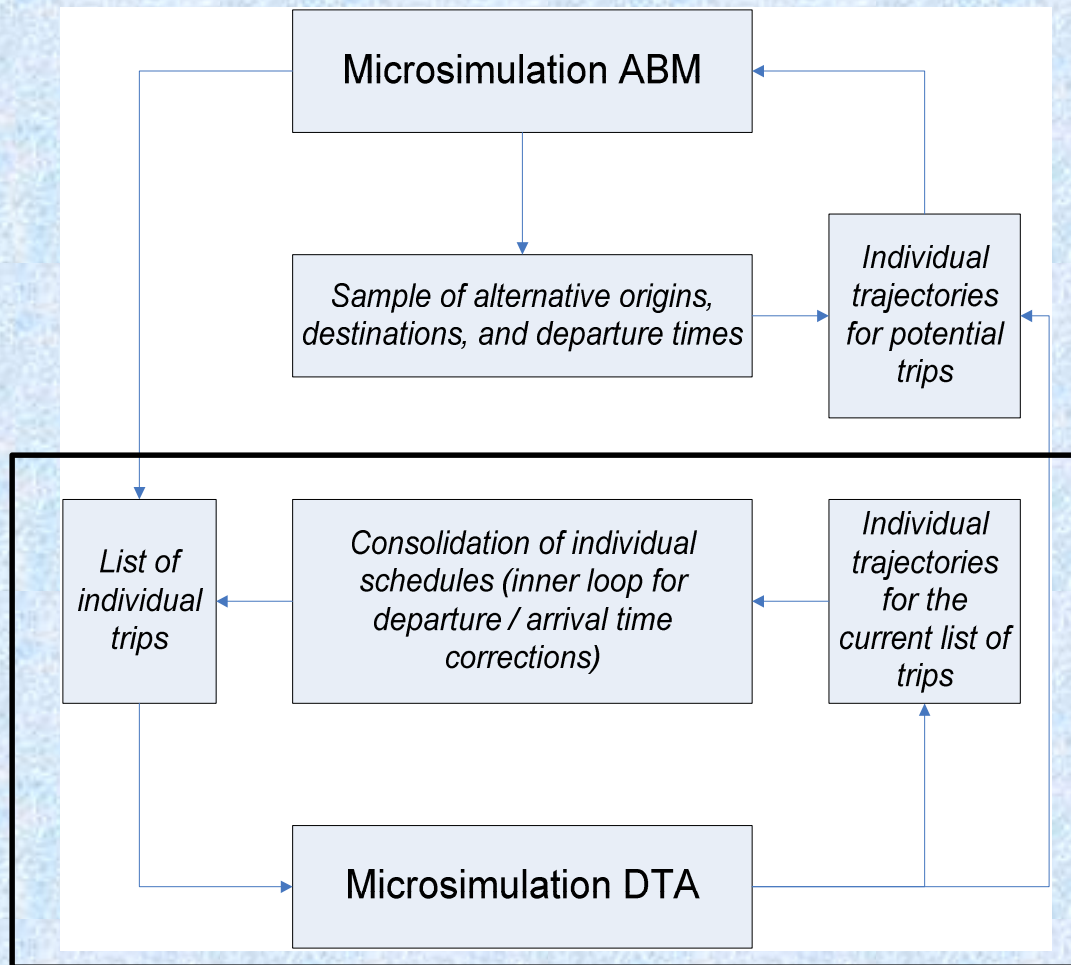


Possible Surrogate

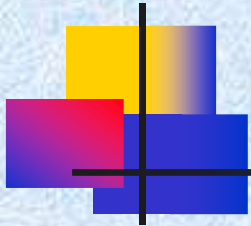




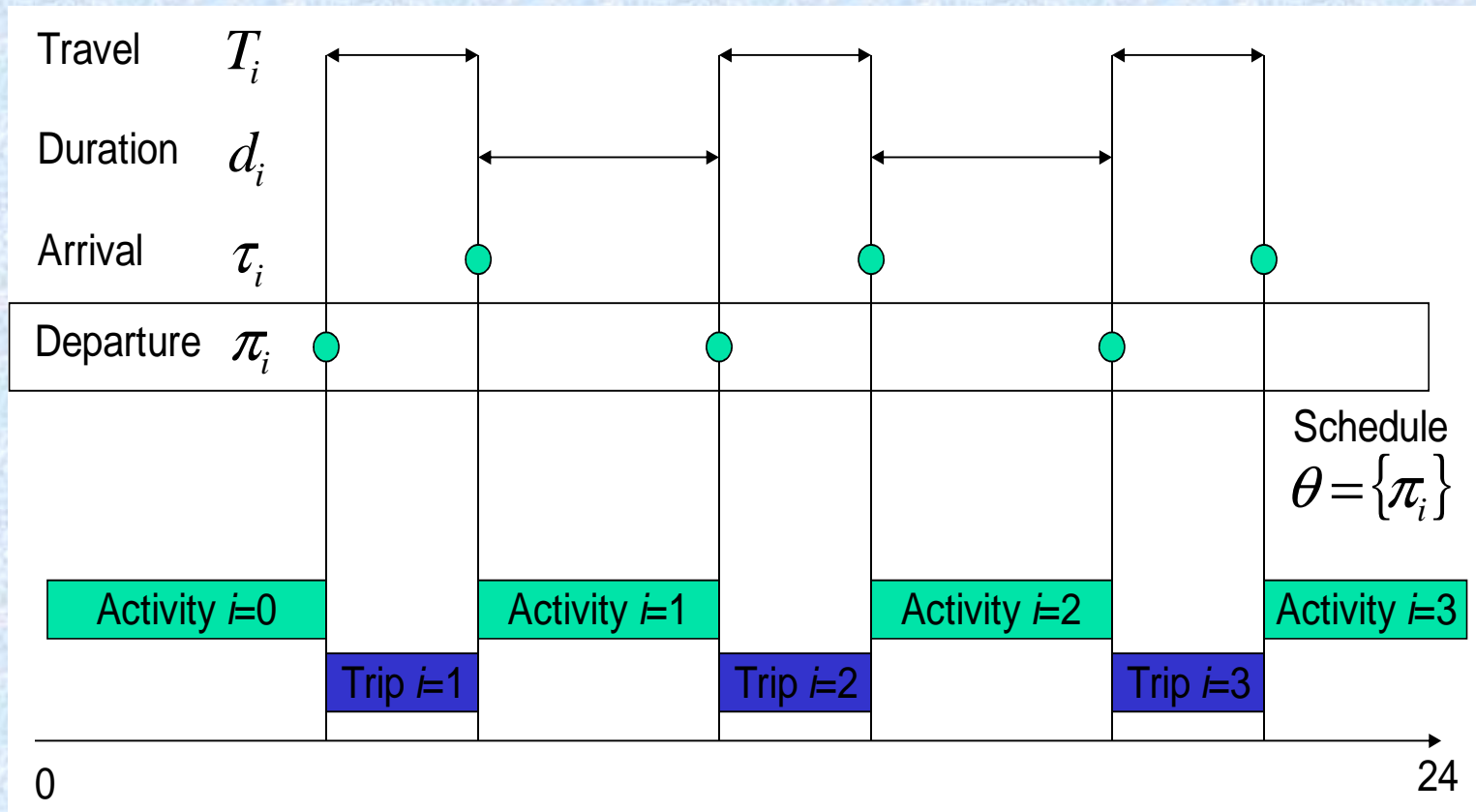
Suggested Approach

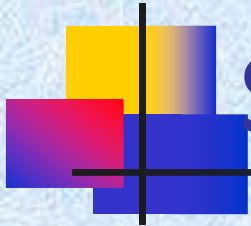


Temporal equilibrium to achieve individual schedule consistency



Individual Schedule Consistency





Schedule Adjustment Prototype

Find new schedule close to previous durations and departures

$$\min \left\{ \sum_i \left(x_i \ln \frac{x_i}{d_i} + y_i \ln \frac{y_i}{\pi_i} \right) \right\}$$

New durations New departures
Previous durations Previous departures

Daily consistency

$$\sum_i (x_i + t_i) = 24$$

Departure time

$$y_i = \sum_{j \leq i} (x_j + t_j)$$

Changed travel times

Solution

$$x_i = k \times d_i \times \prod_{j \geq i} \frac{\pi_j}{y_j}$$



Schedule Adjustment Extended

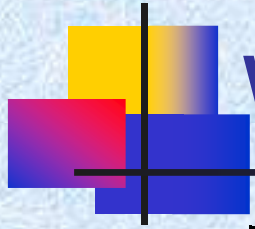
$$\min_{\{x_i\}} \left\{ \left[\sum_{i=0}^I w_i \times x_i \times \ln \left(\frac{x_i}{d_i} \right) \right] + \left[\sum_{i=1}^{I+1} u_i \times y_i \times \ln \left(\frac{y_i}{\pi_i} \right) \right] + \left[\sum_{i=0}^I v_i \times z_i \times \ln \left(\frac{z_i}{\tau_i} \right) \right] \right\}$$

$$y_i = \tau_0 + \left(\sum_{j=0}^{i-1} x_j \right) + \left(\sum_{j=0}^{i-1} t_j \right), \quad i = 1, 2, \dots, I + 1$$

$$z_i = \tau_0 + \left(\sum_{j=0}^{i-1} x_j \right) + \left(\sum_{j=0}^i t_j \right), \quad i = 1, 2, \dots, I$$

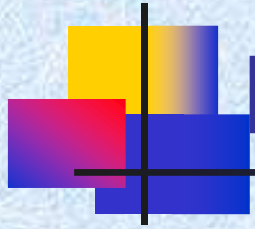
$$x_i > 0, \quad i = 0, 1, 2, \dots, I$$

$$x_i = d_i \times \left\{ \prod_{j>i} \left[\left(\frac{\pi_j}{y_j} \right)^{u_j} \times \left(\frac{\tau_j}{z_j} \right)^{v_j} \right] \right\}^{\frac{1}{w_i}}$$



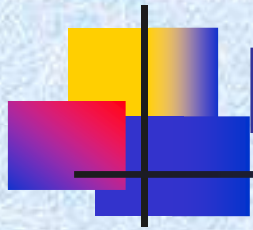
Weights for Schedule Adjustment

Activity type	Duration	Trip departure (to activity)	Trip arrival (at activity location)
Work (low income)	5	1	20
Work (high income)	5	1	5
School	20	1	20
Last trip to activity at home	1	1	3
Trip after work to NHB activity	1	5	1
Trip after work to NHB activity	1	10	1
NHB activity on at-work sub-tour	1	5	5
Medical	5	1	20
Escorting	1	1	20
Joint discretionary, visiting, eating out	5	5	10
Joint shopping	3	3	5
Any first activity of the day	1	5	1
Other activities	1	1	1



Pre-Sampling of Trip Destinations

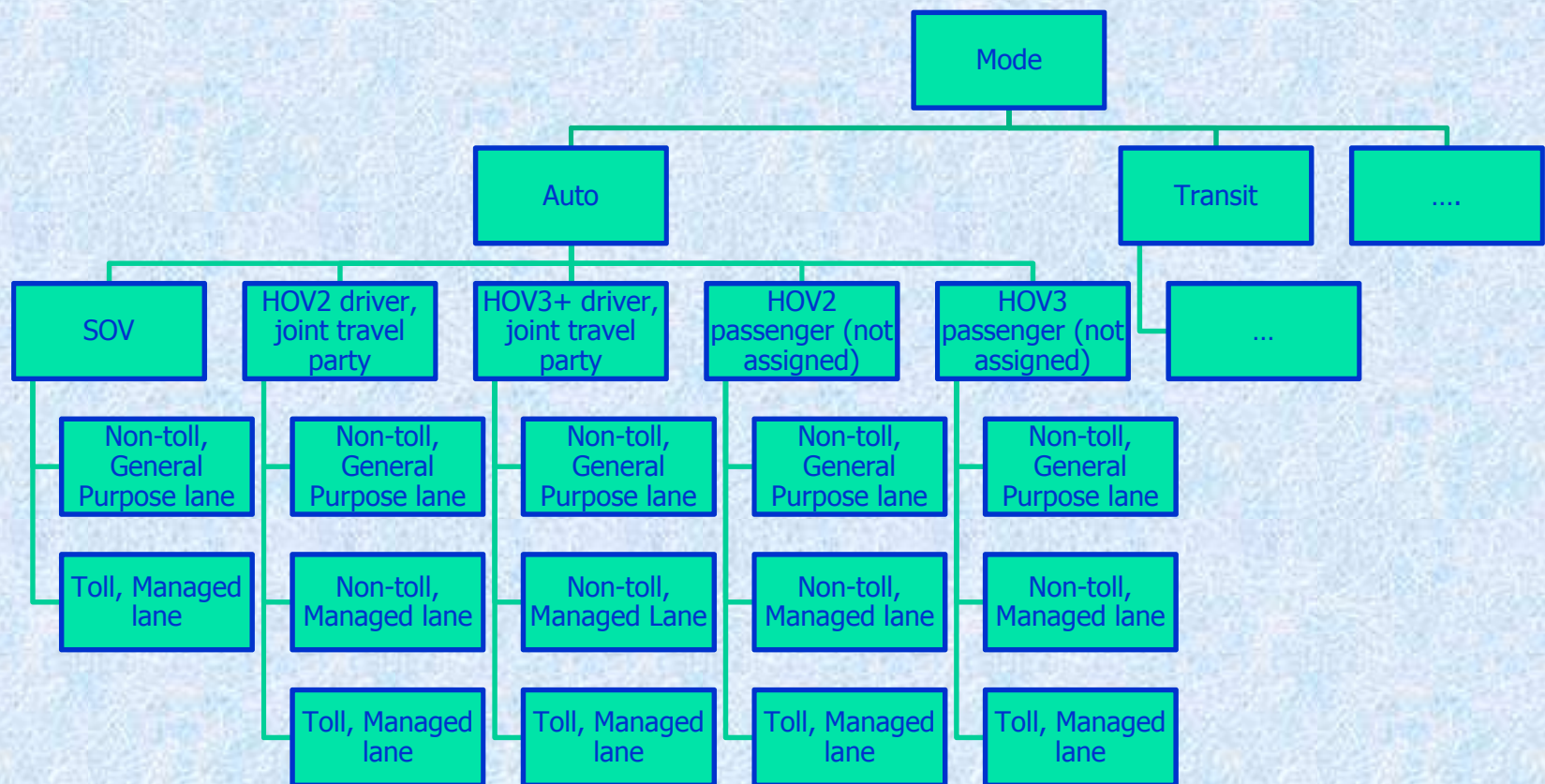
- Primary destinations are pre-sampled:
 - 300 out of 30,000 for each origin and travel segment,
 - 30 out of 300 for each individual and travel segment
- Stop locations are pre-sampled:
 - 300 out of 30,000 for each OD pair and travel segment
 - 30 out of 300 for each individual and travel segment
- Importance sampling w/o replacement from expanded set of destinations $300 \times 30,000$ and 30×300 to ensure uniform unbiased samples
- Efficient accumulation of individual trajectories in microsimulation process

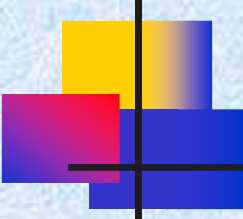


LOS Skims for Outer Loop

- Individual trajectories by departure time period for the same driver (personal learning experience), if not:
 - Individual trajectories by departure time period across individuals (what driver can hear from other people through social networks), if not:
 - Aggregate OD skims by departure time period (Advice from navigation device)

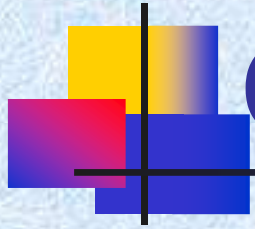
Mode Choice Refinement: Driver vs. Passenger for HOV





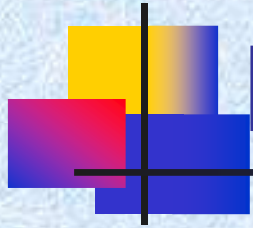
Trip Departure Time Choice Refinement (5 min resolution)

- Tour TOD choice model:
 - bi-directional and has 841 departure-arrival alternatives with 30 min resolution
 - Number of alternatives will quadruple with 15 min resolution
- Trip departure time choice model:
 - One-directional
 - 5 min resolution is feasible and results in under 100 ordered alternatives
 - Multiple Discrete-Continuous approach is being tested for Phoenix ABM (ASU)



Quantification of (Un)reliability

- Systematic variation of travel time is not unreliability:
 - Season
 - Day of week (weekdays vs. weekends)
 - Hour
- Random unpredictable variation on top of it is unreliability:
 - Day-to-day
 - Special events
 - Accidents
 - Weather, etc



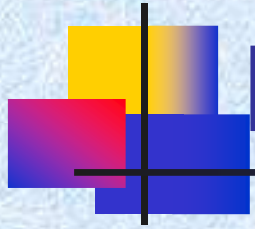
Four Methods

- Perceived highway time by congestion levels
- Time-distribution-based measures (Mean-Variance)
- Schedule delay cost
- Temporal profiles for activity participation



Time-Distribution-Based Measures

- (**Mean-Variance**) Standard Deviation (symmetric)
- (**Buffer time**) Difference between 80-90-95th and 50th percentile (asymmetric)
- (**Risk measure**) Probability of delay of certain length (asymmetric)
- (**Lateness measure**) Average delay (asymmetric)



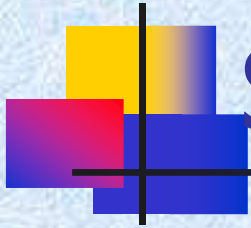
Reliability Ratio (ρ)

- **$U = \alpha \times \text{Time} + \beta \times \text{Cost} + \gamma \times \text{Reliability}$**
 - $\text{VOT} = \alpha / \beta$
 - $\text{VOR} = \gamma / \beta$
 - $\rho = \gamma / \alpha = \text{VOR} / \text{VOT}$
- It is more complicated with non-linear models:
 - VOT, VOR, and ρ becomes functions of time, cost, or distance
 - These variables must be fixed at certain values to calculate VOT, VOR, and ρ

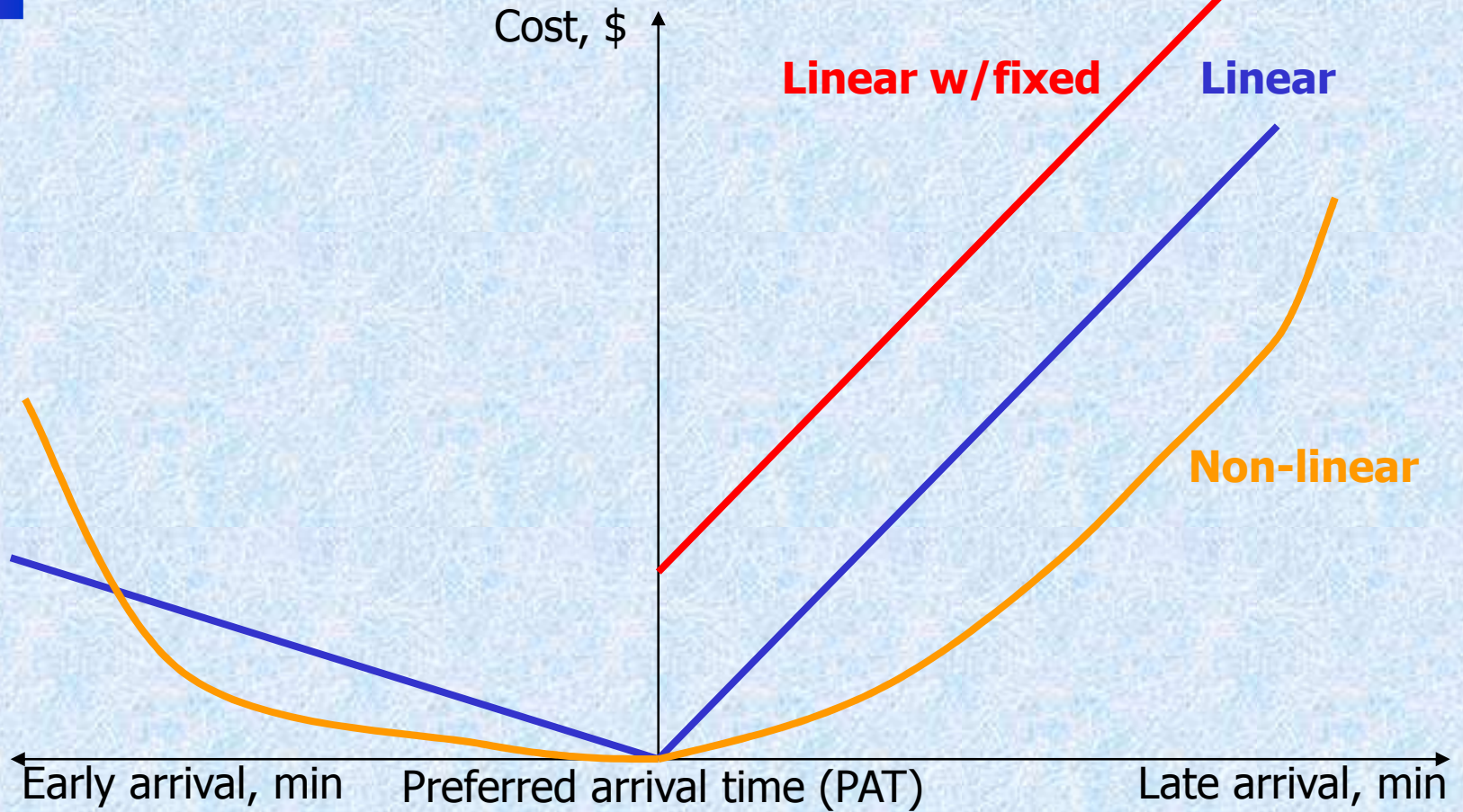


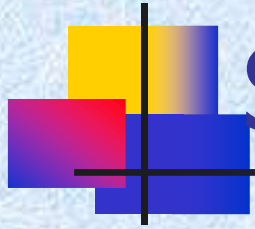
Recommended Weights for Perceived Time

Travel time conditions	Weight	LOS	V/C
Free Flow	1.00	A, B	Under 0.5
Busy	1.05	C	0.5-0.7
Light Congestion	1.10	D	0.7-0.8
Heavy Congestion	1.20	E	0.8-1.0
Stop Start	1.40	F	1.0-1.2
Gridlock	1.80	F	1.2+



Schedule Delay Cost





Schedule Delay Cost

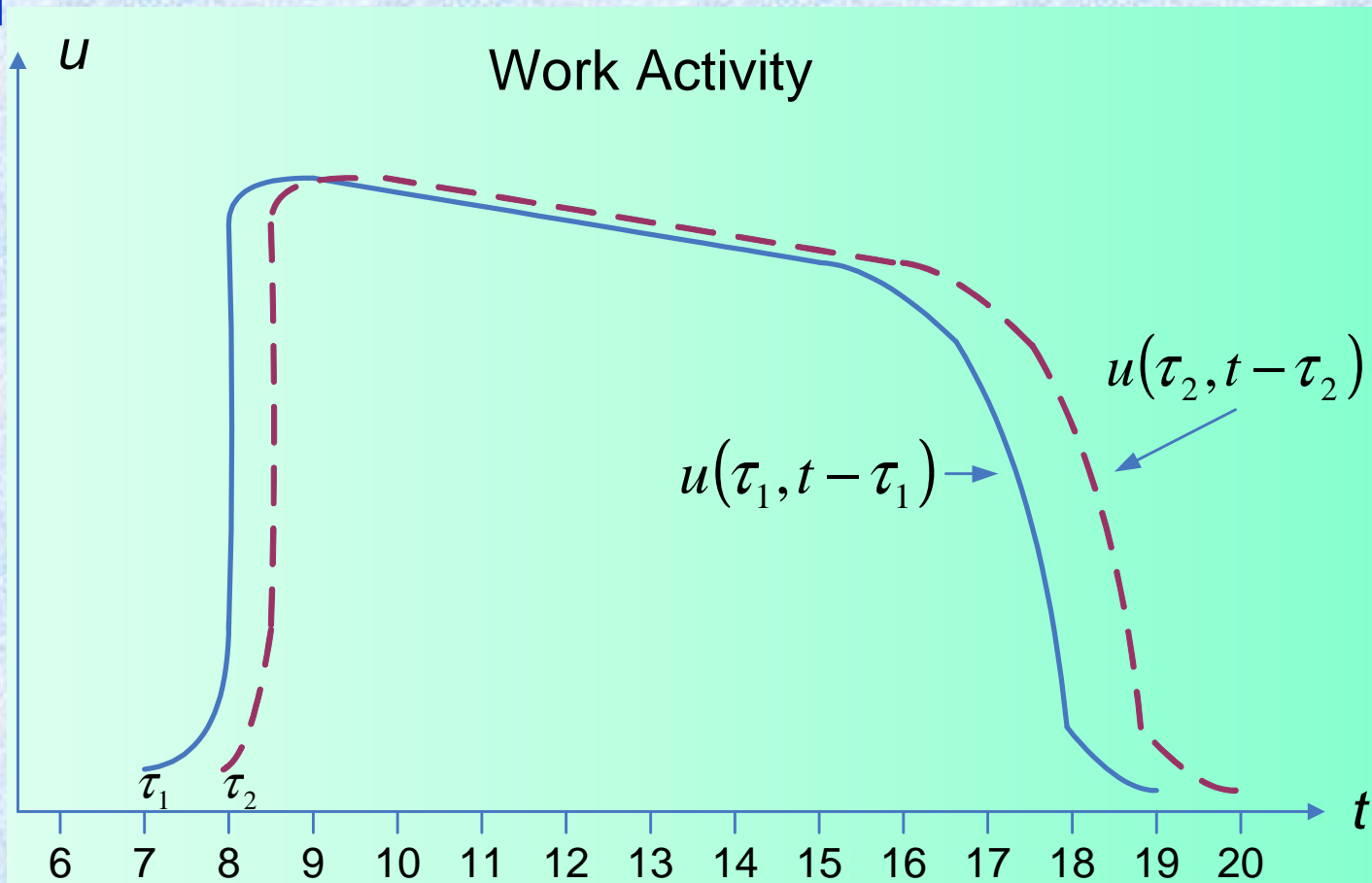
- **$U = \alpha \times T + \beta \times SDE + \gamma \times SDL + \delta \times L$**
- In presence of random travel times:
 - $f(T)$ – travel time distribution
 - $E(U)$ – expected utility dependent on $f(T)$ and departure time/ PAT
 - Improvement of reliability in terms of $f(T)$ can be evaluated in terms of $E(U)$
- Considerable body of literature:
 - SP estimates: $\gamma \geq \alpha$



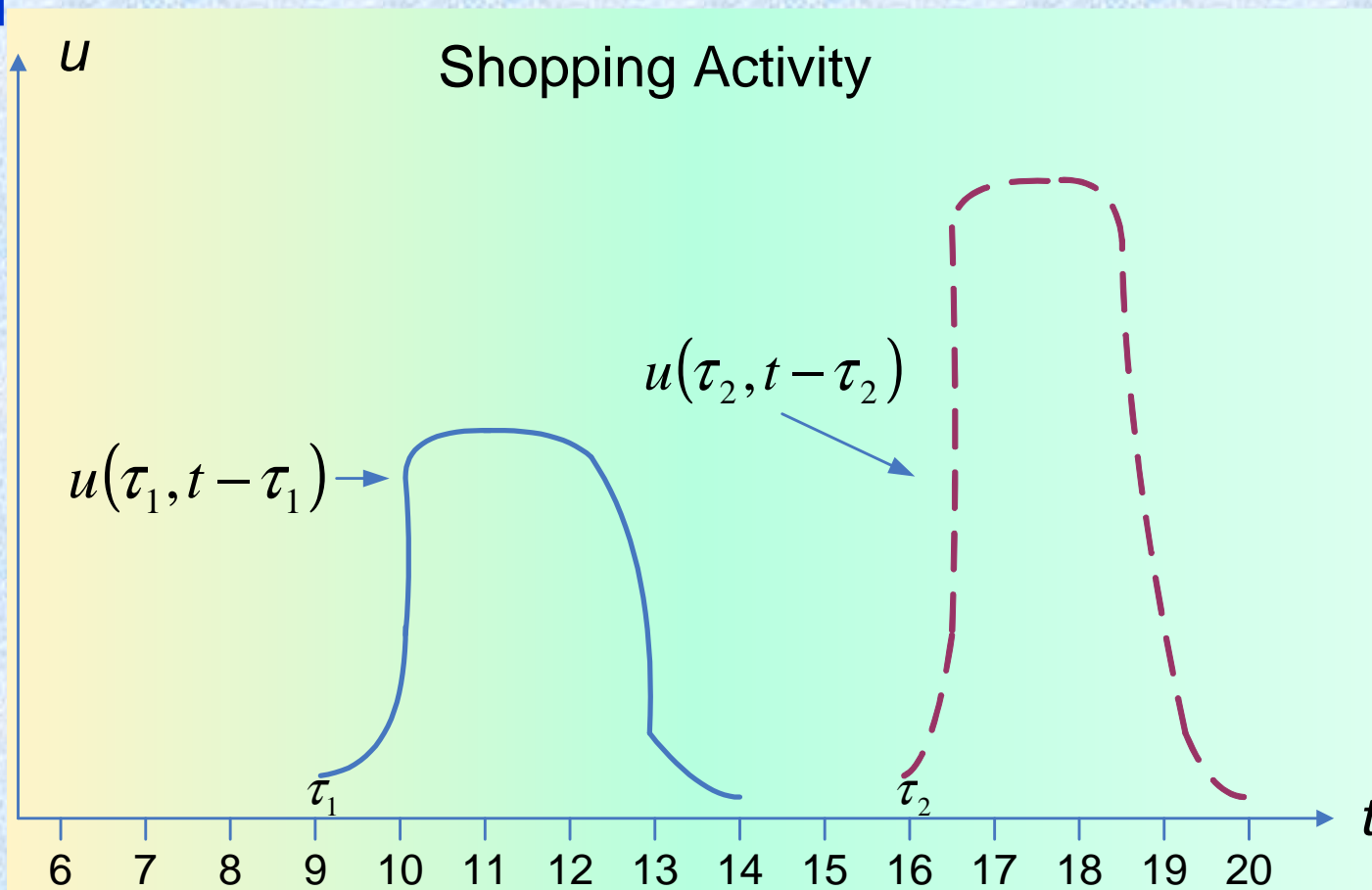
Summary of Defaults for ρ

Population segment	Travel segment	Perceived congested time vs. free-flow	STD vs. mean time	Buffer 90 th -50 th vs. median time	Lateness against PAT vs. mean time
High income (60K+)	To work	2.0	0.8	1.0	3.0
	From work	1.5	0.6	0.7	2.0
	Non-work	1.2	0.4	0.5	1.8
Low income (U60K)	To work	2.5	1.0	1.2	6.0
	From work	1.2	0.3	0.4	1.7
	Non-work	1.1	0.2	0.2	1.5

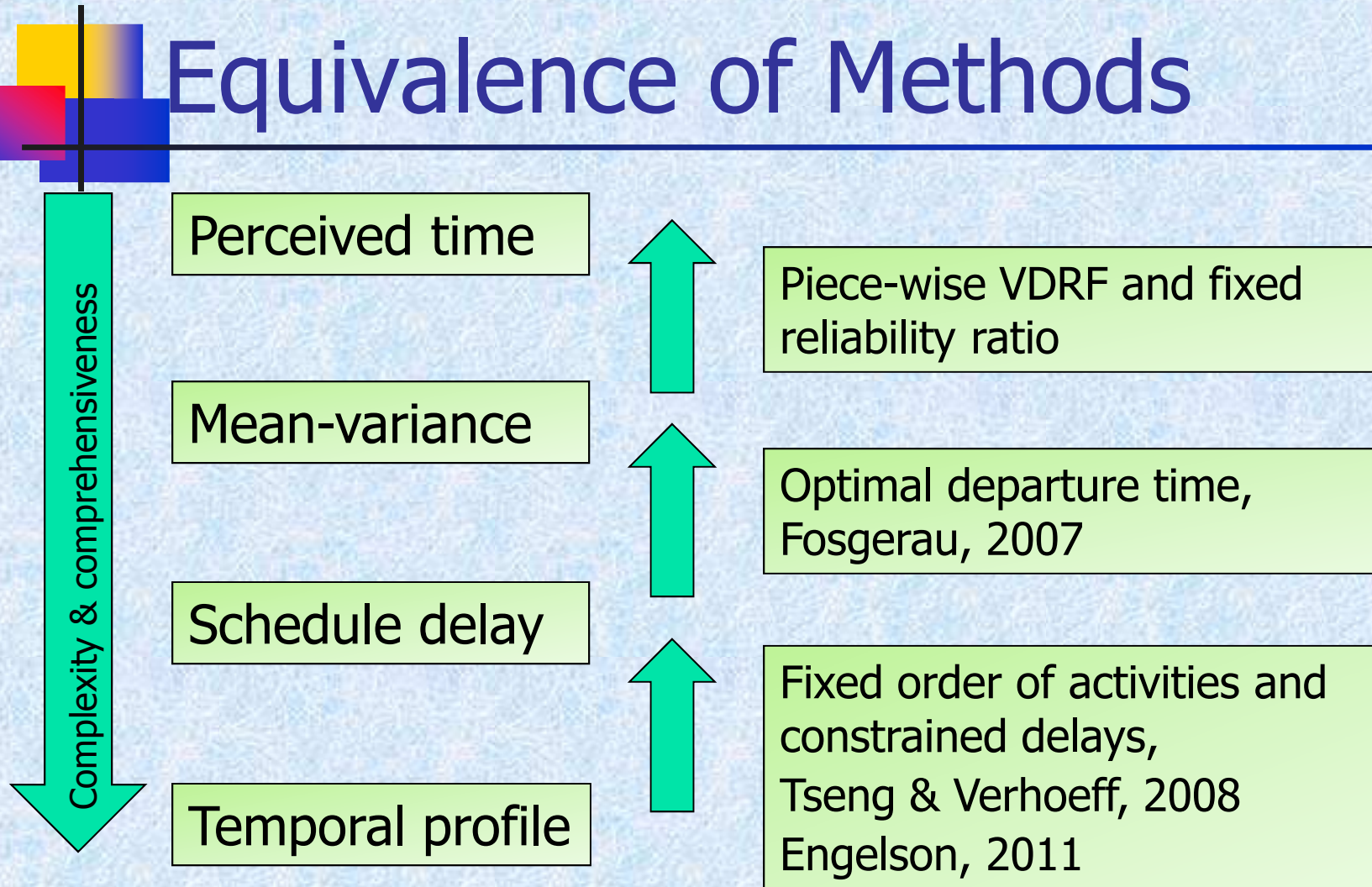
Temporal Utility Profile for Activity Participation

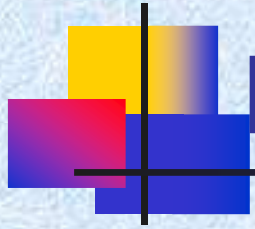


Temporal Utility Profile for Activity Participation



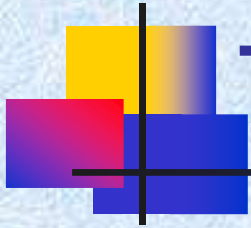
Equivalence of Methods





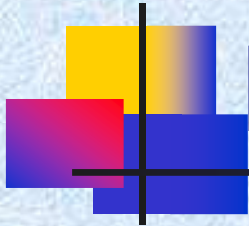
Reliability in Network Simulations

- Challenges:
 - Incorporate reliability in route choice
 - Generate OD reliability measures (skims)
- Methods:
 - Analytical (single run)
 - Simulation (multiple runs)



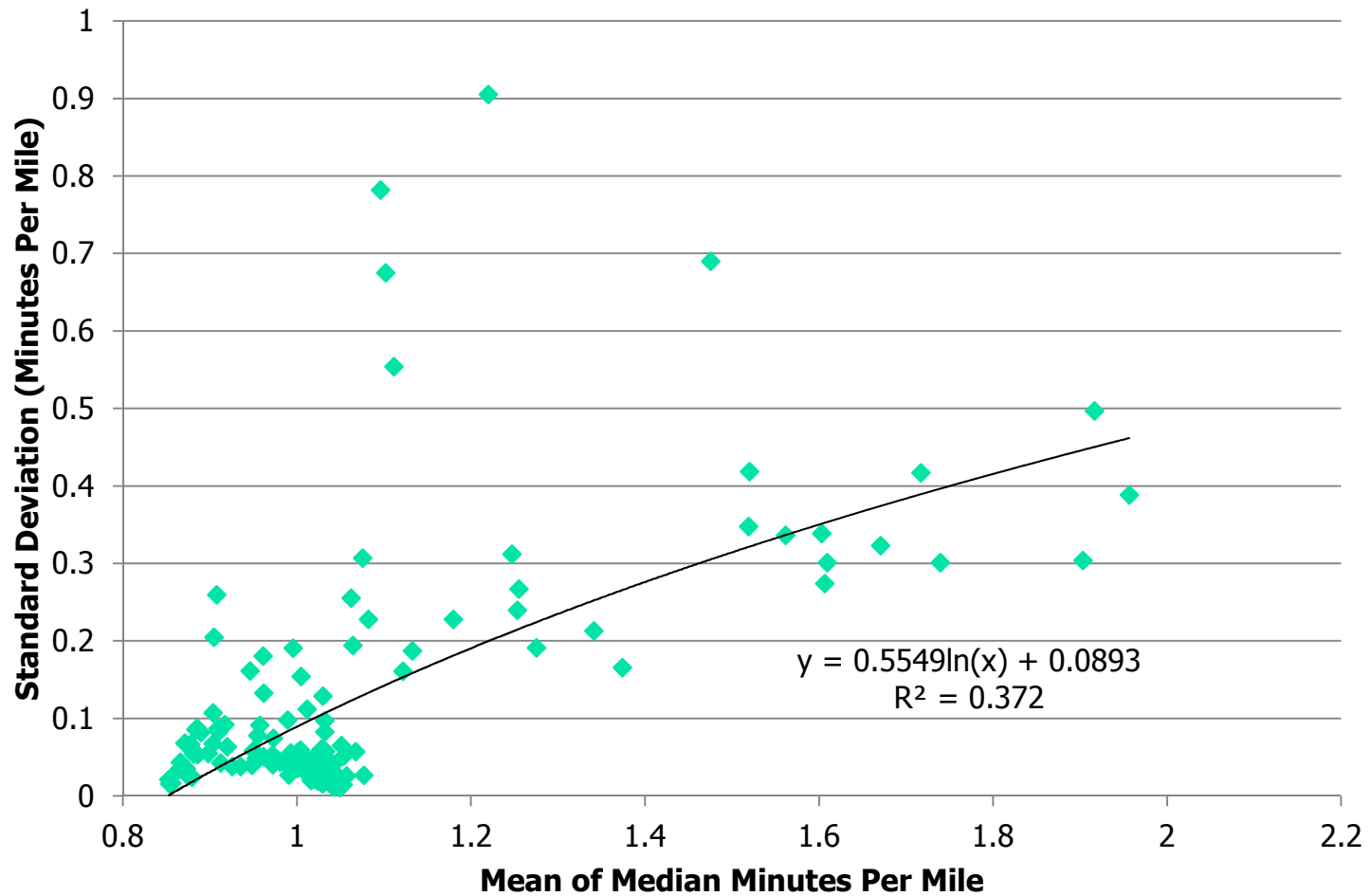
Traffic Physics at Link Level

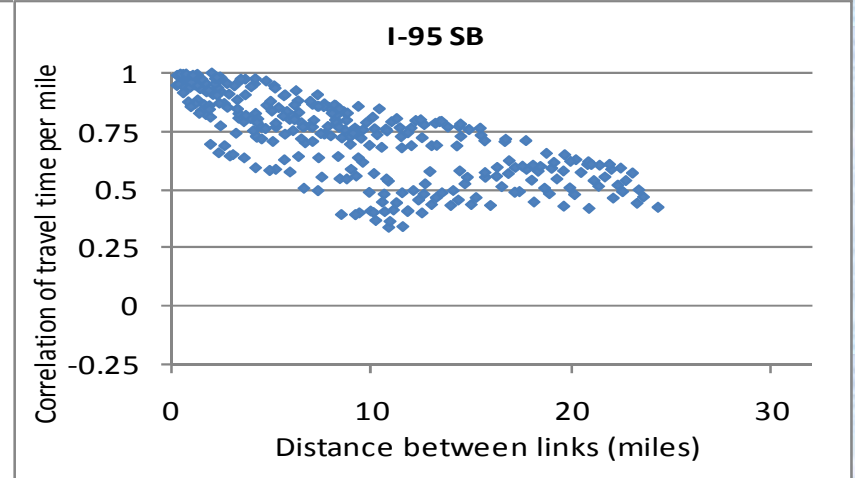
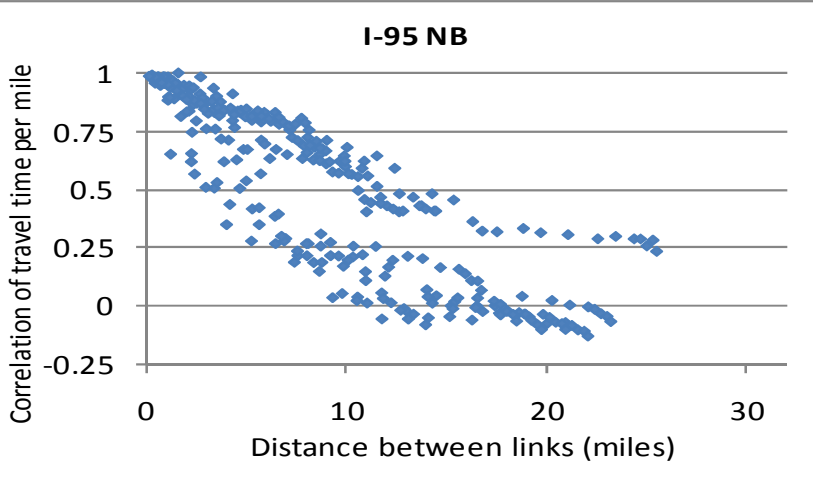
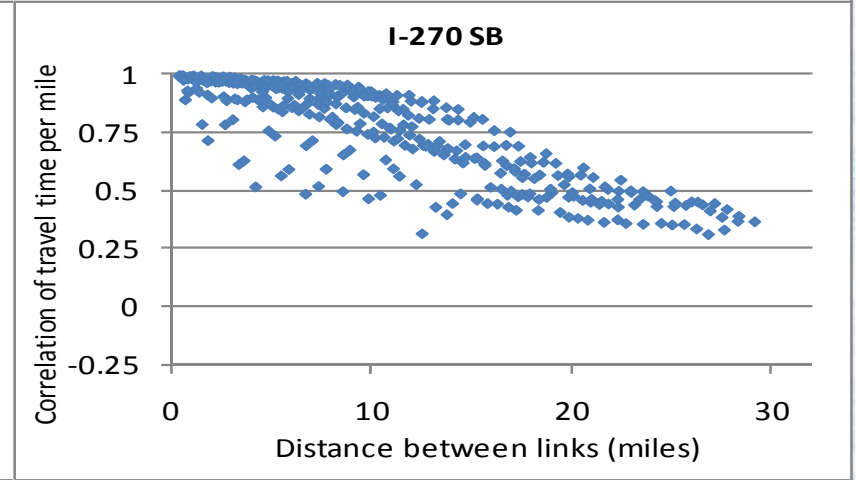
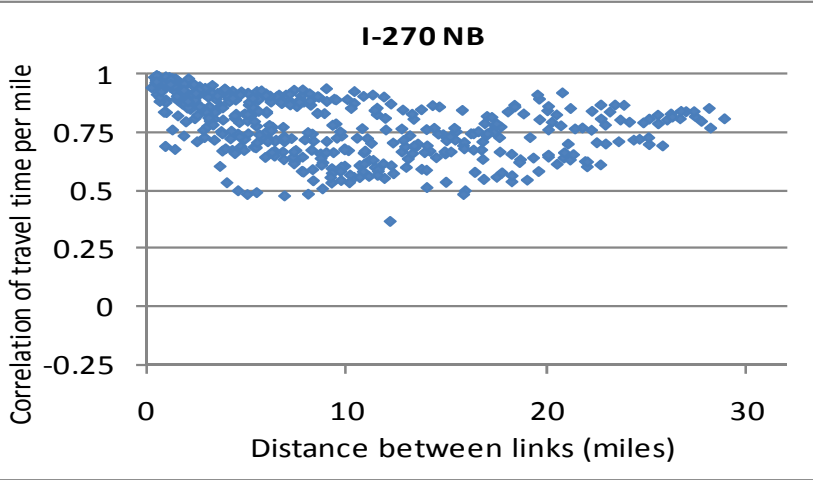
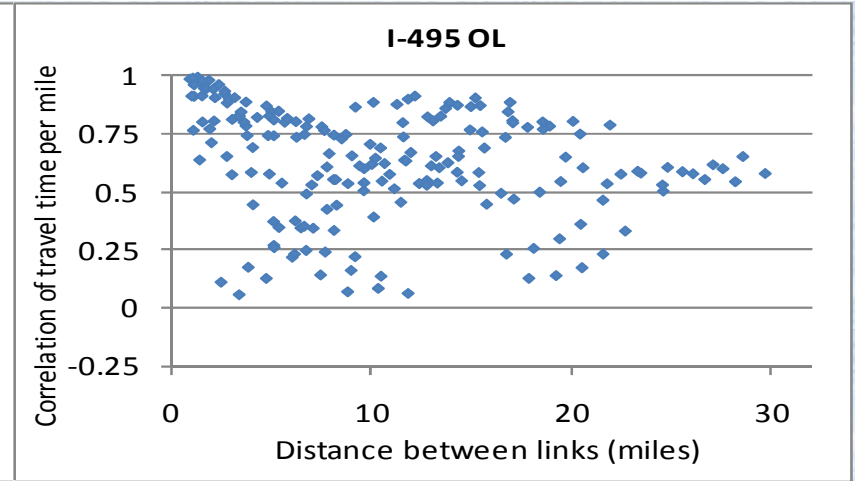
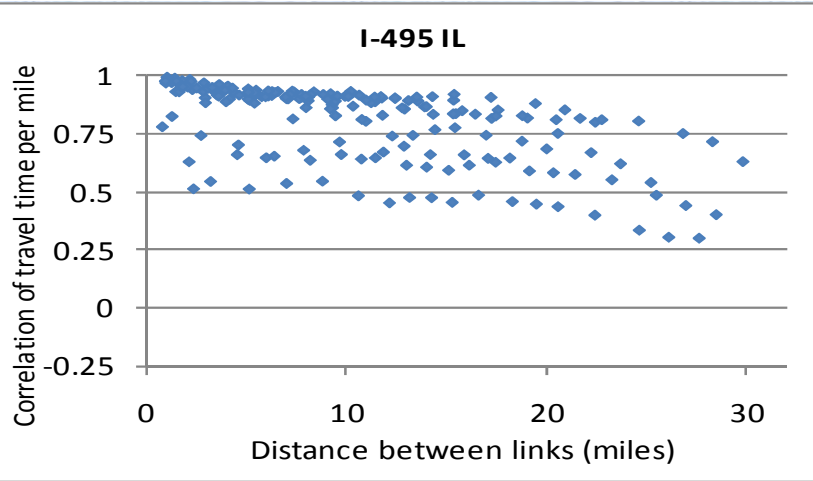
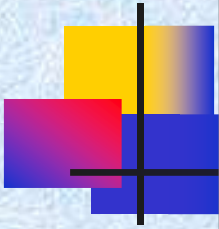
- Volume-Delay-Reliability Function (VDRF):
 - Average time $\mathbf{t_a = f(v_a)}$
 - STD (or other Reliability measure):
 $\mathbf{\sigma_a = g(t_a) = g[f(v_a)]}$ or $\mathbf{\sigma_a = h(v_a)}$
- Growing number of VDRF estimated:
 - $\mathbf{\sigma_a = g(t_a)}$ – linear, slightly non-linear
 - $\mathbf{\sigma_a = h(v_a)}$ – highly non-linear (convex)



Link-Level Functions (L03)

ALA-580 EB, I-680 to I-205, 20.25 miles

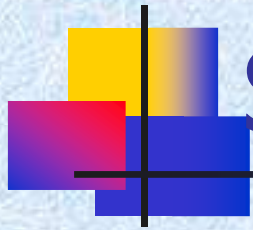






2 Implementation Frameworks for Mean-Variance Method

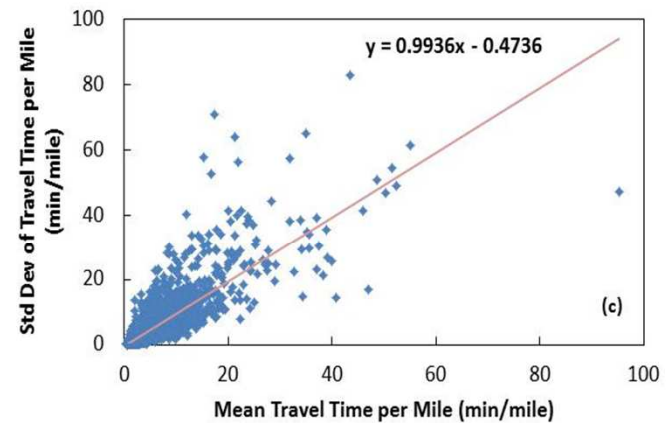
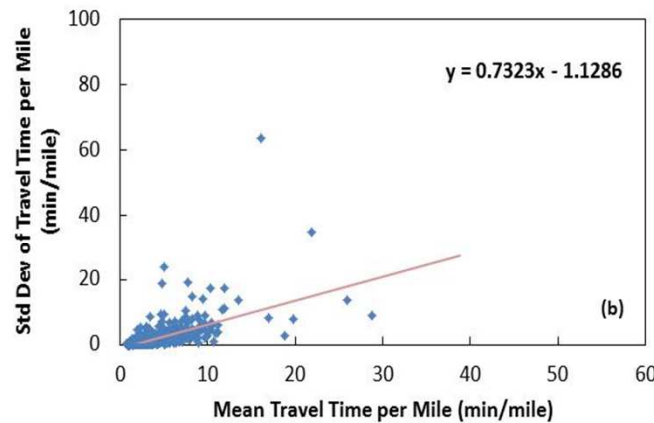
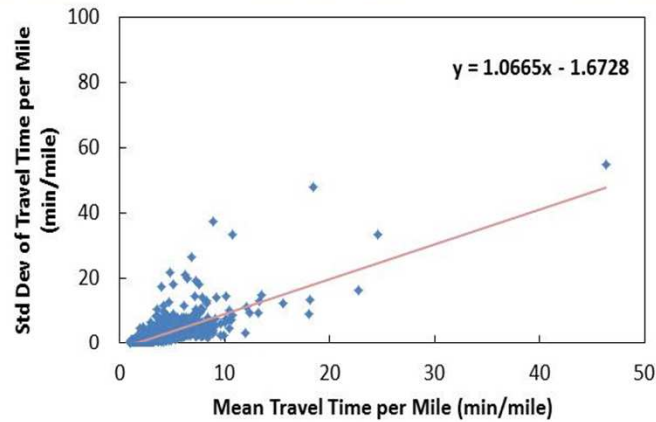
- **Single-run framework:**
 - One demand scenario
 - One network simulation
 - Travel time variation derived from a single equilibrium state (implicitly)
- **Multiple-run framework:**
 - One or several demand scenarios
 - Several network simulations
 - Travel time variation modeled explicitly



Single-Run Framework

- Demand model (C04):
 - Adding variance or standard deviation as LOS variable along with mean travel time and cost to mode choice and other travel choices
- Network Simulation Model (L04):
 - Adding variance or STD to route generalized cost along with mean travel time and cost
 - Generation of route variance or standard deviation skims for demand model

STD of Travel Time / Mile as Function of Mean Travel Time / Mile (Seattle, GPS Traffic Choices Study, 2008)



(a) O-D Level; (b) Path Level; (c) Link Level

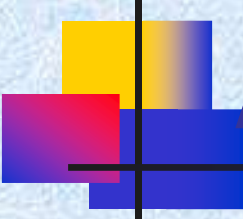
Construction of OD Trip Reliability Measures

- Link-level function does not solve the problem:
 - STD and buffer time measures are not additive
 - Variance is additive if link travel times are independent (not in general case)
- Route-level and OD Reliability Measures:
 - Robust statistical relationships between mean travel time and STD (path-based assignment)
 - Scaling procedures for link-level STD (link-based assignment)



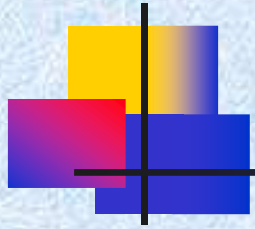
Example of Scaling Procedures to Construct Route STD

- For elemental unit (mile):
 - $\sigma = k \times t$
 - k = coefficient of variation
- For entire OD route:
 - $\sigma = k \times t \times (d)^{-\mu}$
 - d = distance
 - (independence) $-0.5 \leq -\mu \leq 0$ (perfect correlation)



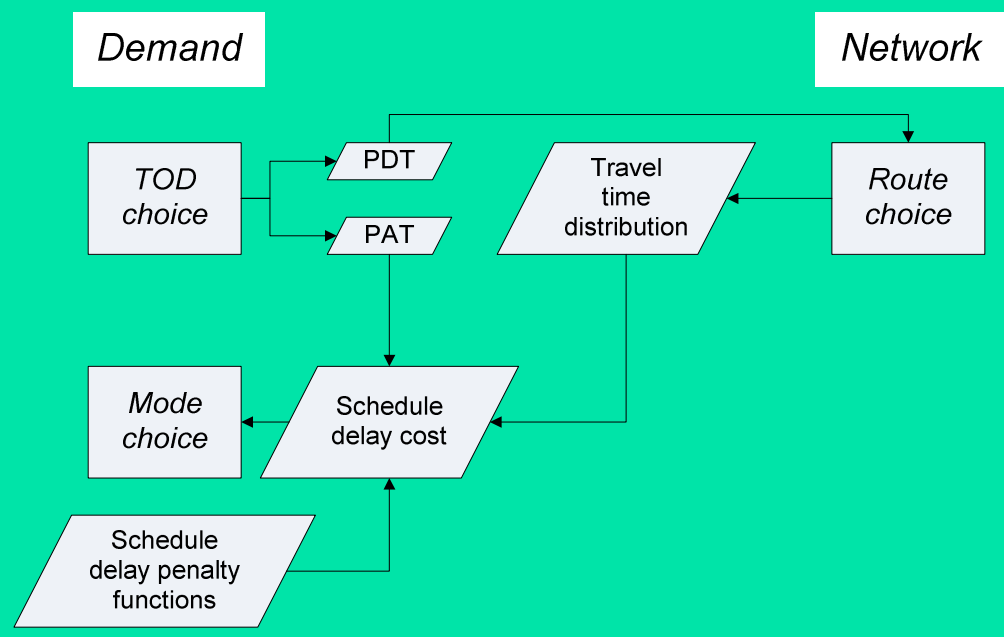
Self-Calibration of μ in Link-Based Assignment

- For each OD pair based on the previous iterations:
 - $(d_{OD})^{-\mu(OD)} = \sigma_{OD} / (\sum_a \sigma_a) = \eta_{OD}$
- Assume link generalized cost function:
 - $c_a = t_a(v_a) + \rho \times \sigma_a [t_a(v_a)]$
- Scale reliability ratio for next iteration:
 - $\rho_{OD} = \rho \times \eta_{OD}$

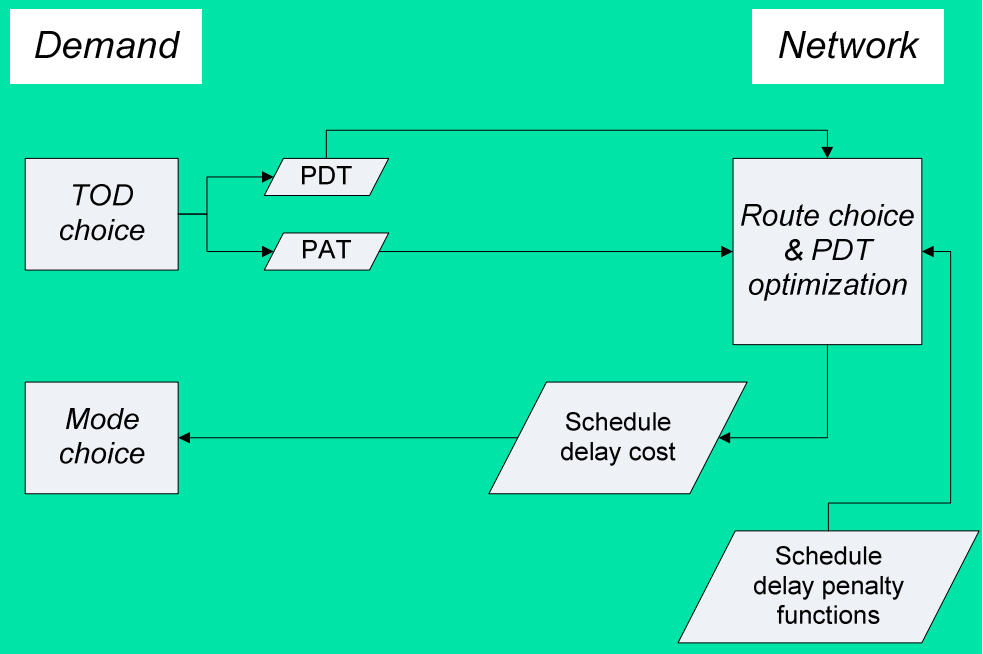


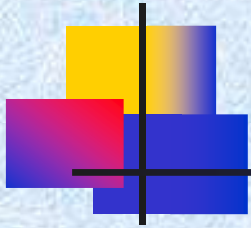
Incorporation of Schedule Delay Cost

1st approach: schedule delay cost calculation in demand model

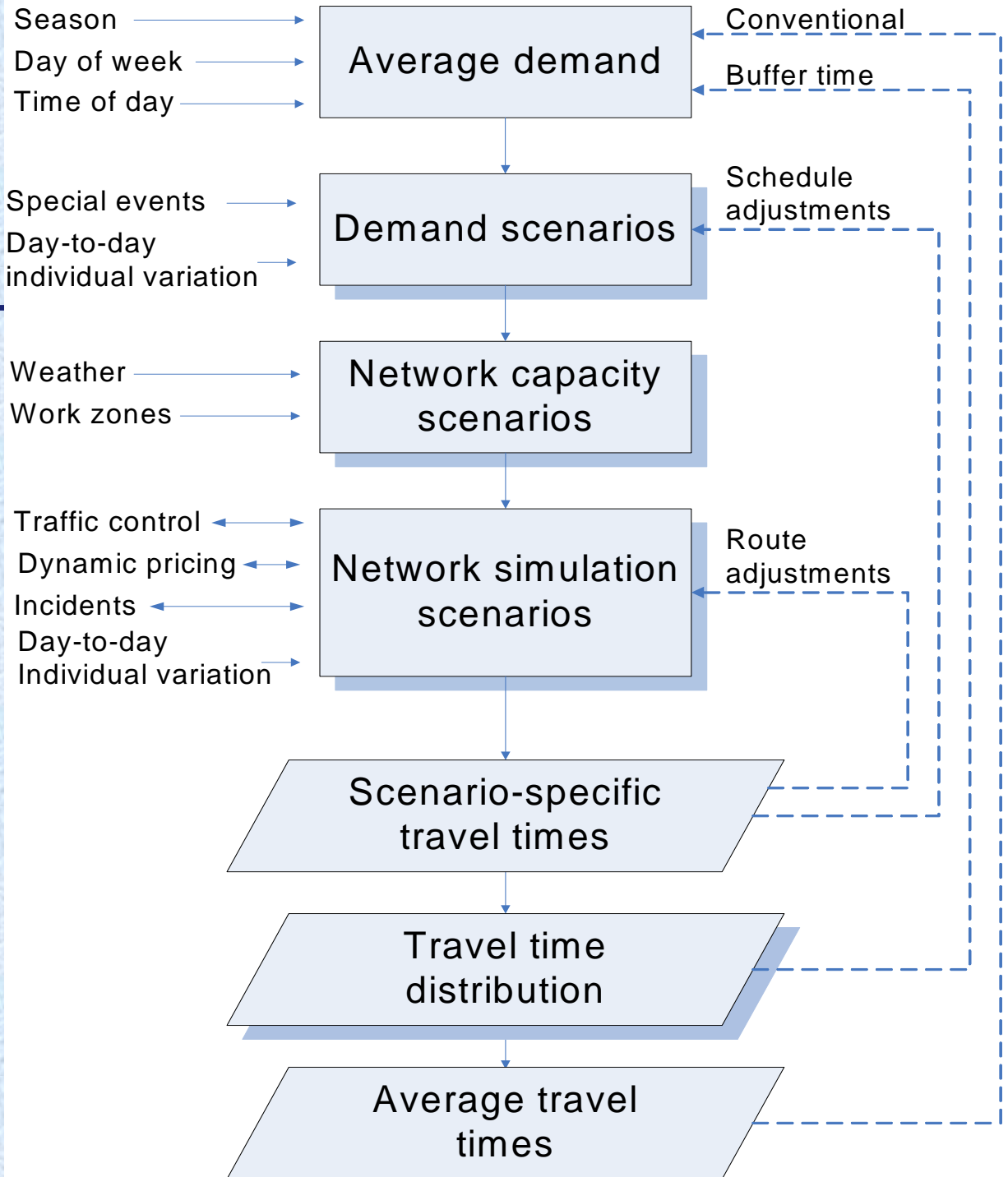


2nd approach: schedule delay cost calculation in network model





Multiple-Run Framework

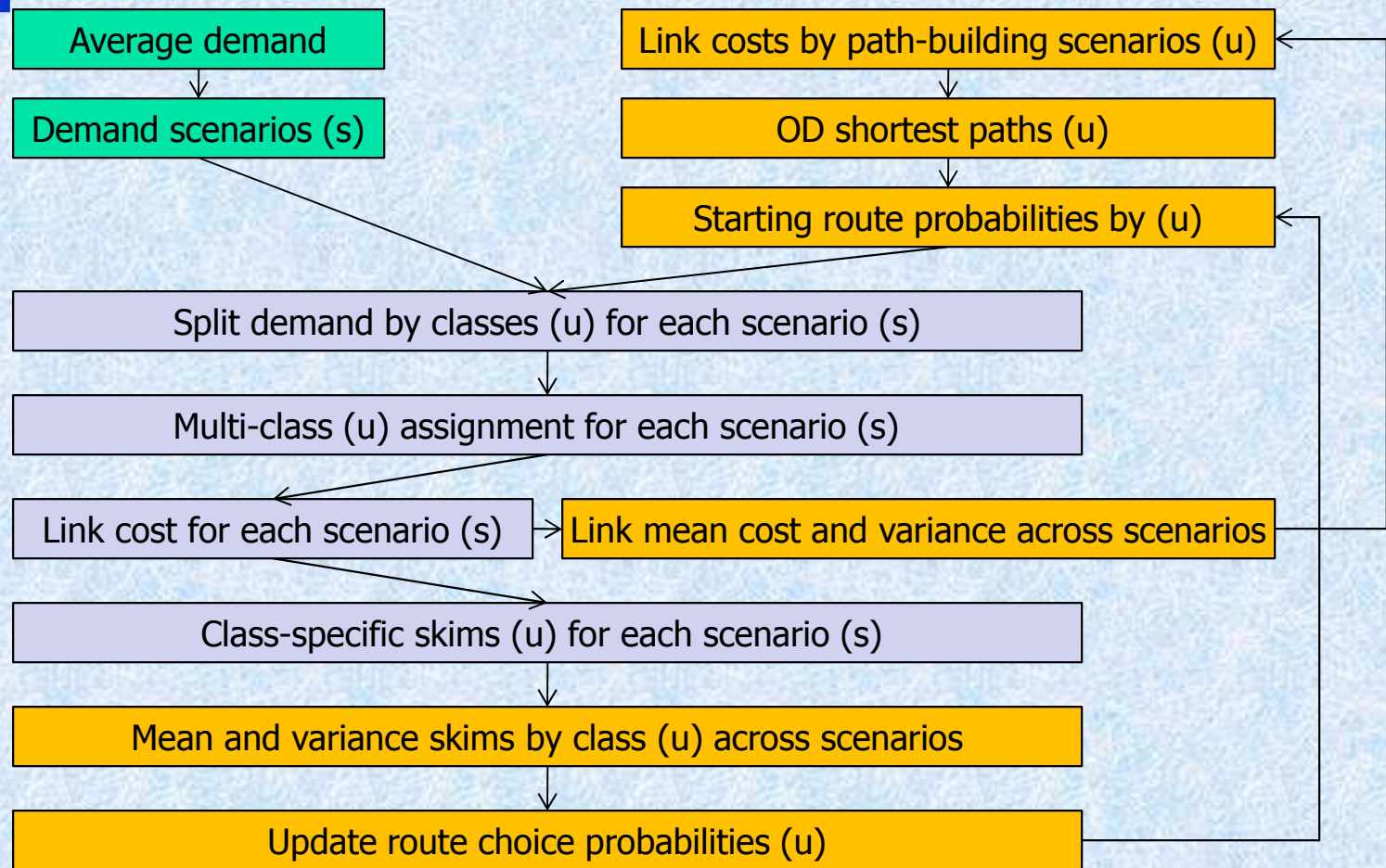




Equilibrium Assignment with Random Demand and Reliability

- Source of travel time variation is variable demand by scenarios $D(s)$
- Link travel time on given day is deterministic function $c(v)$
- Travelers do not know demand and link travel times on given day; they only know link and route mean and variance
- Travelers chose routes based on the mean-variance generalized cost function; probabilities are the same across days

Equilibrium Assignment with Random Demand and Reliability





Conclusions

- Methods to integrate microsimulation demand and network models:
 - Intermediate (temporal) equilibration for individual schedule consistency
 - Pre-sampling of locations to accumulate individual trajectories
- Methods to incorporate travel time reliability:
 - Perceived highway time by congestion levels – easy but just a surrogate
 - Mean-variance – main method substantiated in C04 and L04
 - Schedule delay cost & temporal activity profiles – more advanced methods that need further research and improved data
- Operational models / single-run framework:
 - Demand models include STD in generalized cost
 - Construction of STD measures at OD-route level to feed into demand model (robust stats or scaling)
 - Incorporation of reliability in route choice in (efficient) traffic assignment equilibrium (path-based or link-based)
- Operational models / multiple-run framework:
 - More promising and holistic way but more complicated
 - Ongoing L04 research (Scenario Manager)