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## Consistent Estimation of Route Choice Models for Dynamic Transit Assignment

Jeff Hood

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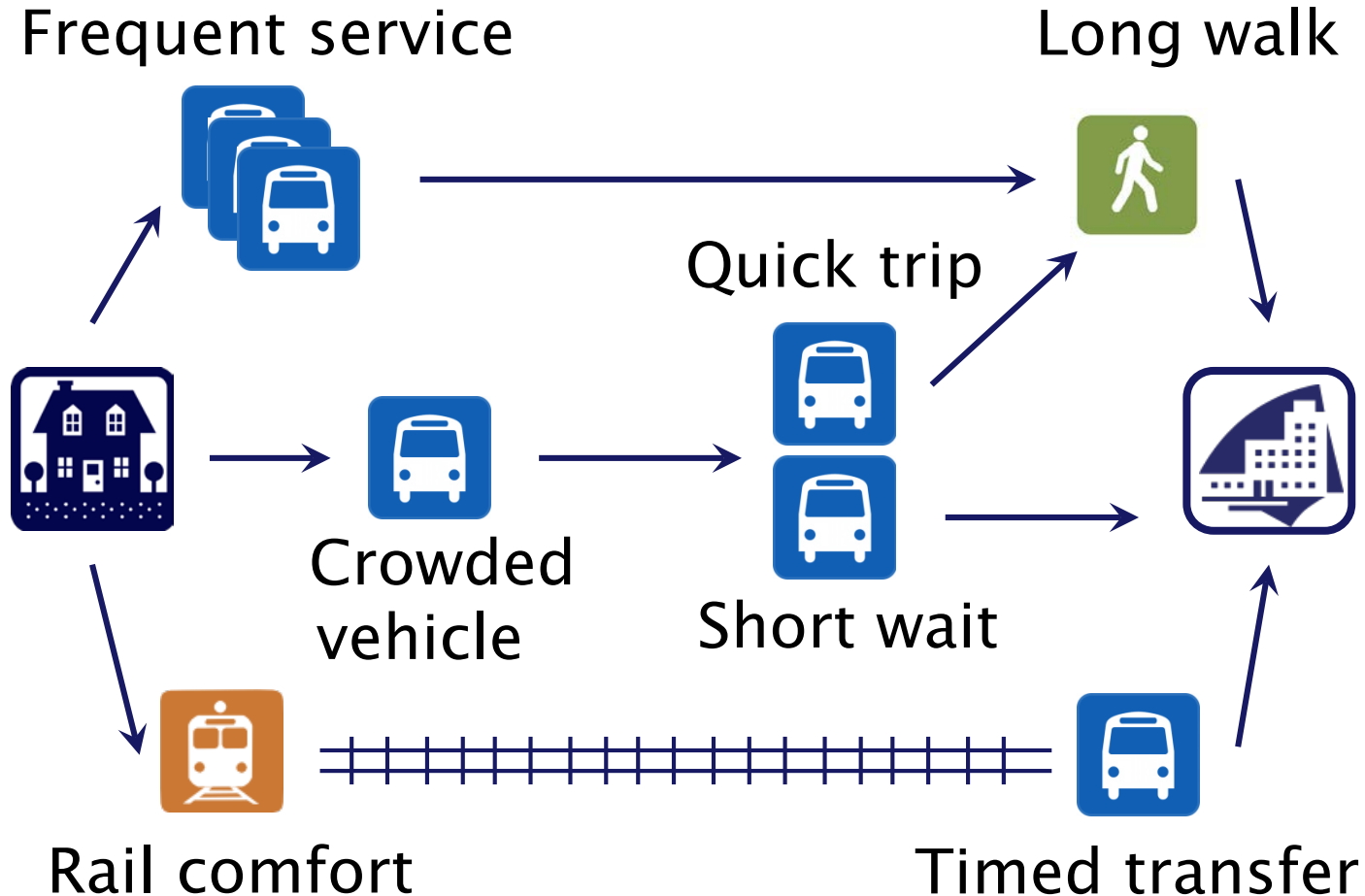
# Consistent Estimation of Route Choice Models for Dynamic Transit Assignment

Jeff Hood



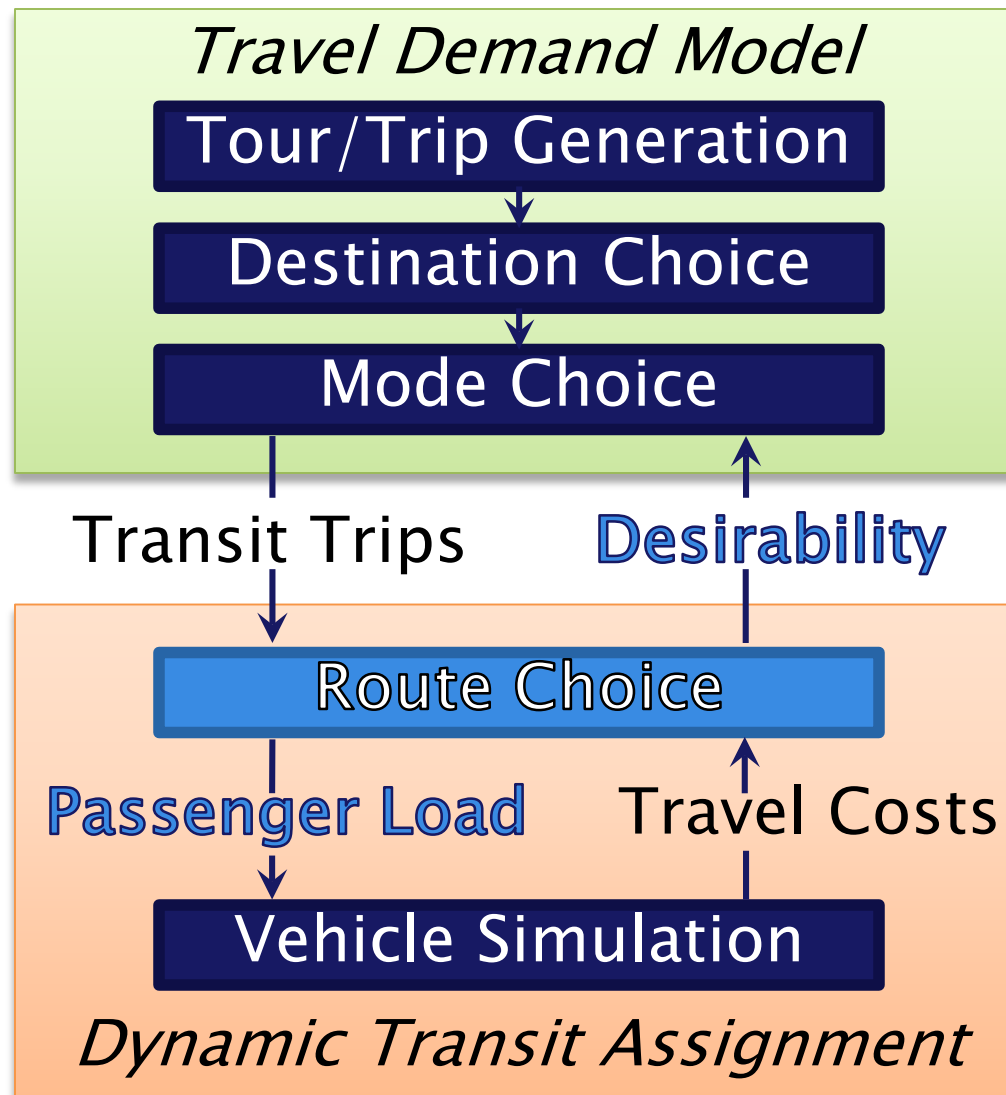
Portland State University Transportation Seminar  
Portland, Oregon  
February 16, 2016

# How do transit passengers choose a route?

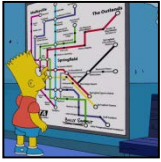




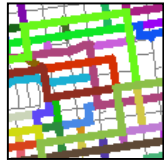
# Predict passenger load and measure desirability in forecasting system



# Outline



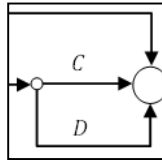
Introduction (you are here!)



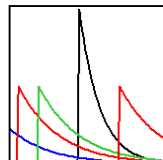
Limitations of path-based methods

$$P(a|k)$$

The recursive logit model



New correction for route overlap



Reliability and stochastic arrivals



## Limitations of Path-Based Methods

# Path-based models require choice set generation





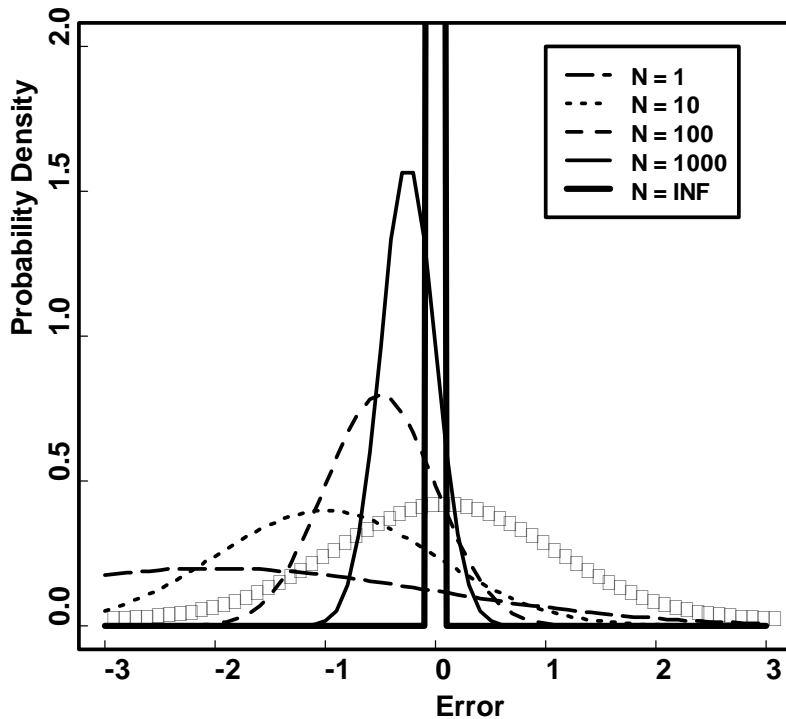
# Example: stochastic sampling with unknown distribution



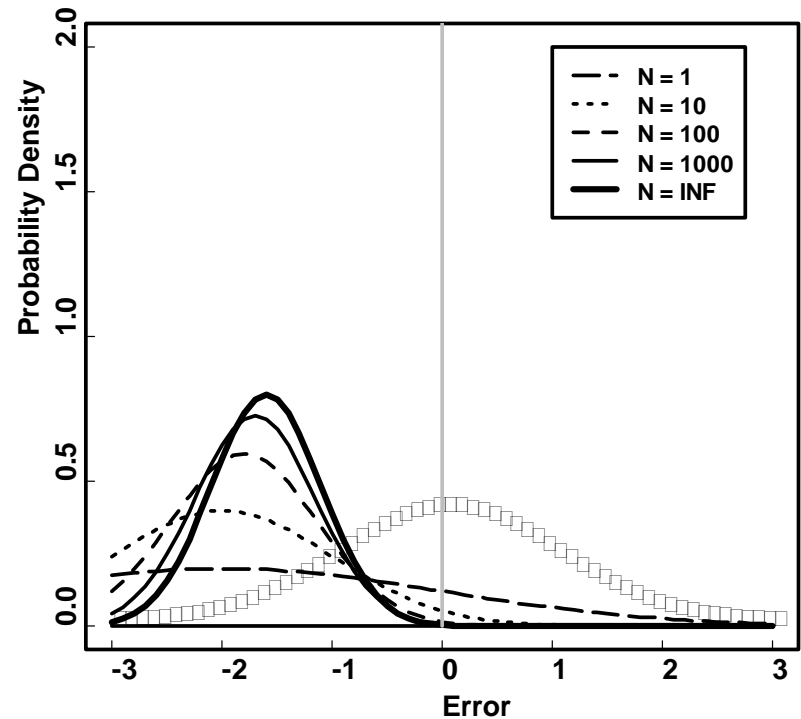
Hood et al. (2011). *Transport. Letters* 3,63–75.

# For most choice set generation schemes, estimators are inconsistent

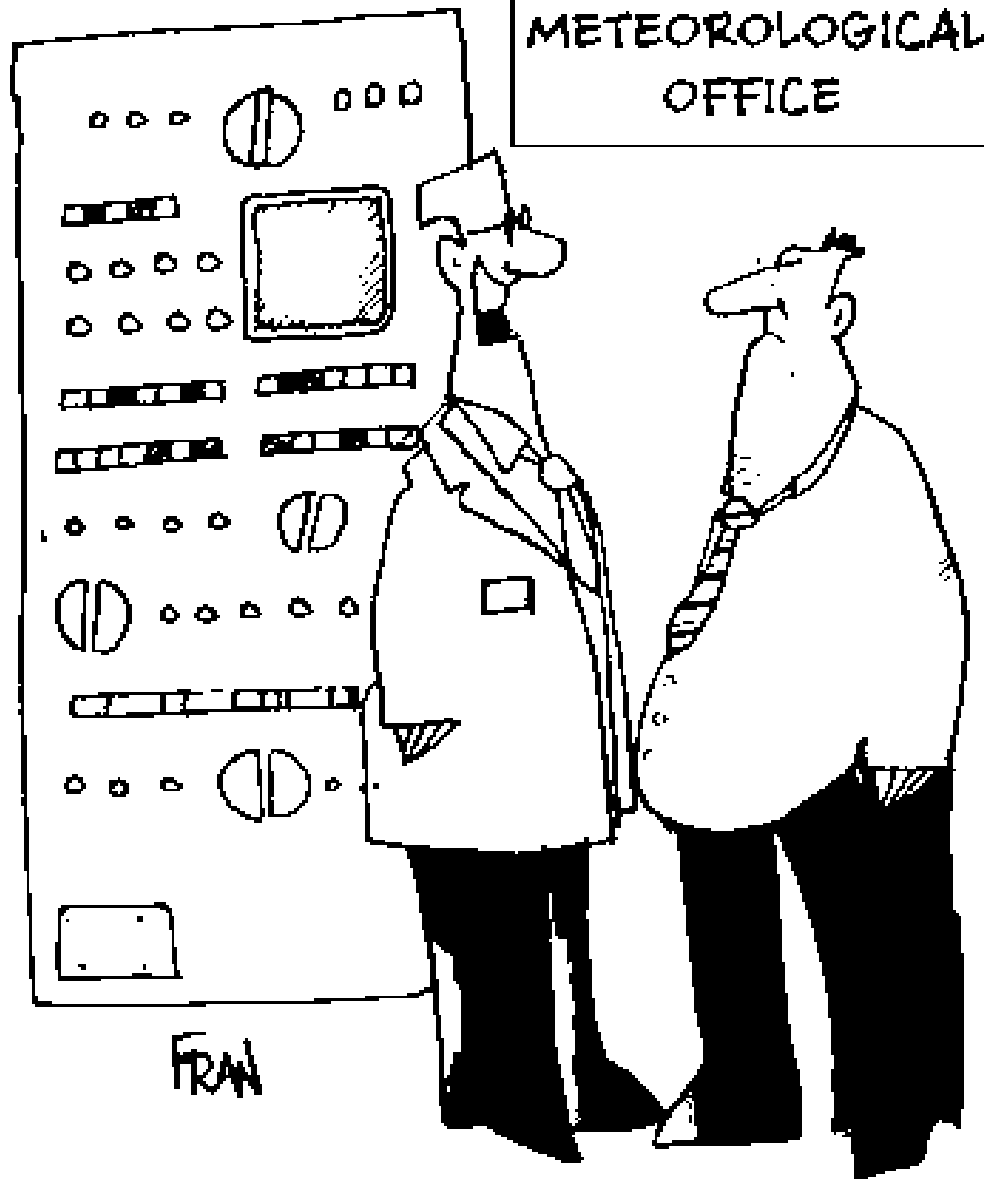
## Consistent Estimator



## Inconsistent Estimator



METEOROLOGICAL  
OFFICE



FRAN

With our new superfast computers we  
can get the forecast wrong TWICE as  
fast as we used to!

# Consistent methods are too slow



Base

Build

Stochastic Sampling



Base

Build

Path Size Link Penalty

Nassir et al. (2014). *Transport. Res. Rec.* 2430, 170–181.

Run time is quadratic in zones.

1000 zones on 4 processors requires a week!



GOOD LUCK

45  
M.P.H.





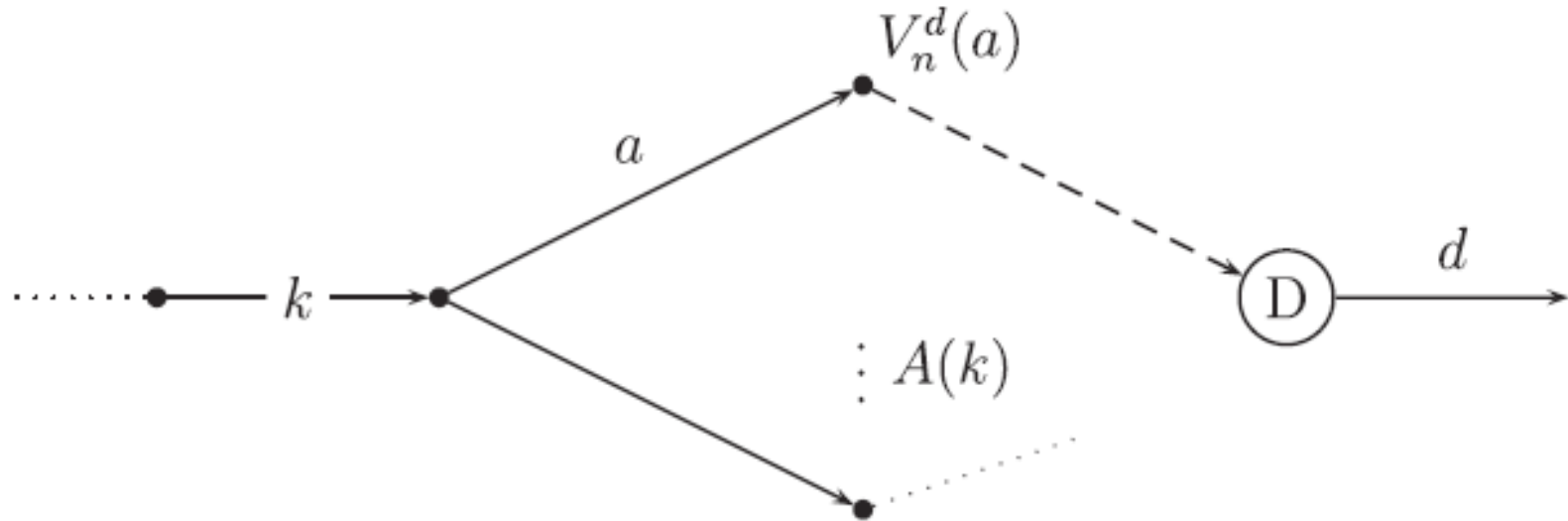
TURN BACK



$$P(a|k)$$

## The Recursive Logit Model

# Dynamic programming solution to Markov decision process



Fosgerau et al. (2014) *Transport. Res. B*: 56, 70–80

$k$ : current link

$a$ : possible movement from  $k$

$d$ : destination

$V(a)$ : expected max. utility of all paths from  $a$  to  $d$

$A(k)$ : set of all successors of  $k$

$v(a|k)$ : “instantaneous” utility of moving from  $a$  to  $k$



# Traveler maximizes sum of instantaneous utility and expected utility to destination

Recursive value equation:

$$V(k) = E \left[ \max_{a \in A(k)} (v(a|k) + V(a) + \mu \varepsilon(a)) \right]$$

$\varepsilon(a)$  i.i.d. extreme value type I implies...

Logit transition probabilities:

$$P(a|k) = \frac{\exp\frac{1}{\mu}[v(a|k)+V(a)]}{\sum_{a' \in A(k)} \exp\frac{1}{\mu}[v(a'|k)+V(a')]}$$

Value equation is logsum:

$$V(k) = \mu \log \sum_{a \in A(k)} \exp\frac{1}{\mu} [v(a|k) + V(a)]$$

# Recursive equations solved efficiently with sparse linear system

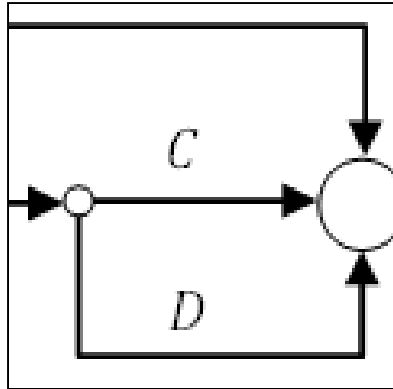
If

- $\mathbf{M}$  = matrix of exponentiated utilities  $\exp[v(a|k)/\mu]$
- $\mathbf{b}$  = indicator vector for the destination
- $\mathbf{z}$  = desired vector of values  $\exp[V(k)/\mu]$

Then the Bellman value equation is

$$\mathbf{z} = \mathbf{Mz} + \mathbf{b}$$

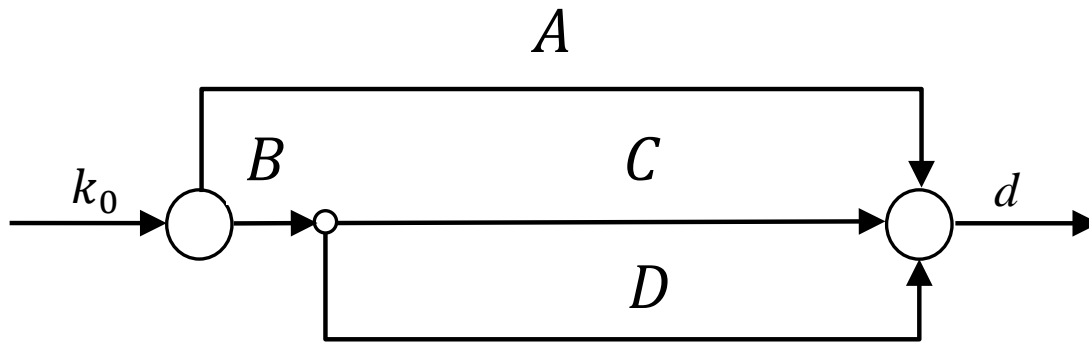
Result is equivalent to link-additive path-based model with unrestricted choice set



## New Correction for Route Overlap

# In reality, random errors are correlated due to overlapping routes

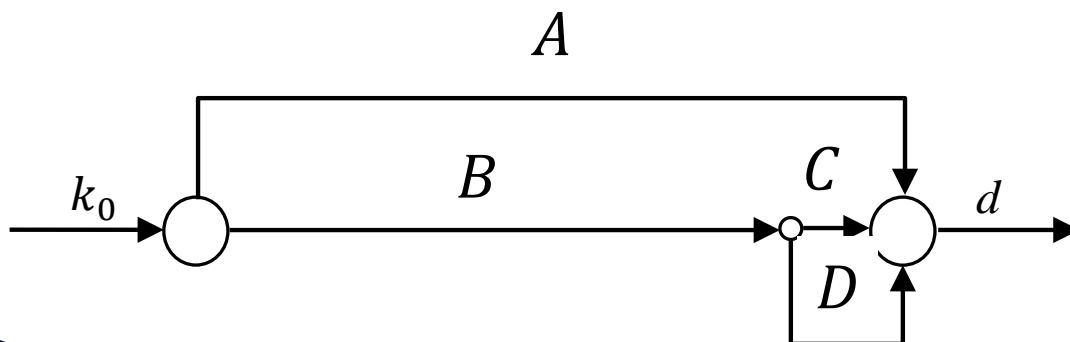
## Uncorrelated Errors



## Path-Based "Size" Correction

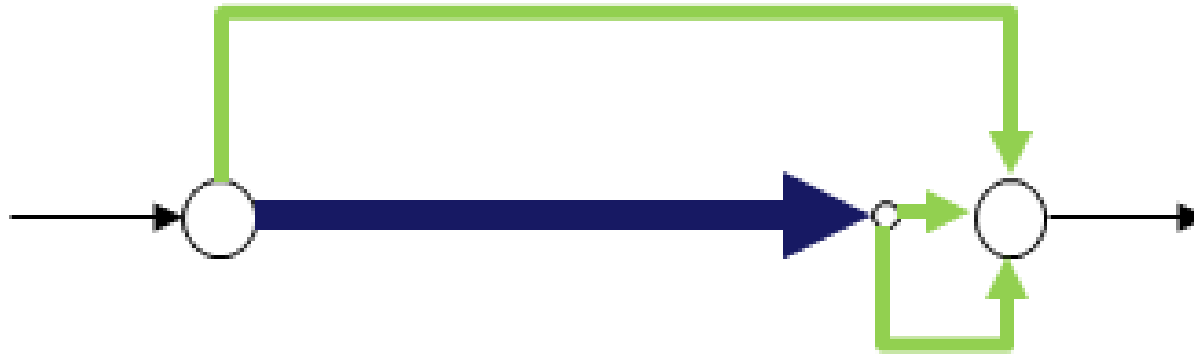
Path	Size	Prob.
A	1	0.33
BC	1	0.33
BD	1	0.33

## Correlated Errors



Path	Size	Prob.
A	1	0.50
BC	0.5	0.25
BD	0.5	0.25

For recursive logit, Fosgerau et al. recommend link flow proxy



### Performs Poorly

- Not sensitive to extent of overlapping links
- Conflates route overlap and route utility
- Requires scaling parameter
- Not topologically-invariant

# New link size variables extend recursive approach to counting of paths

# downstream path segments

$$N^d(k) = \sum_{a \in A(k)} N^d(a)$$

# upstream path segments

$$N^u(k) = \sum_{a \in A(k)} N^u(a)$$

# paths containing  $k$

$$N^d(k) \times N^u(k)$$

Equations have no solution in cyclic networks!

# Path counts should be scaled by probability anyway

## Probability-scaled downstream path segments

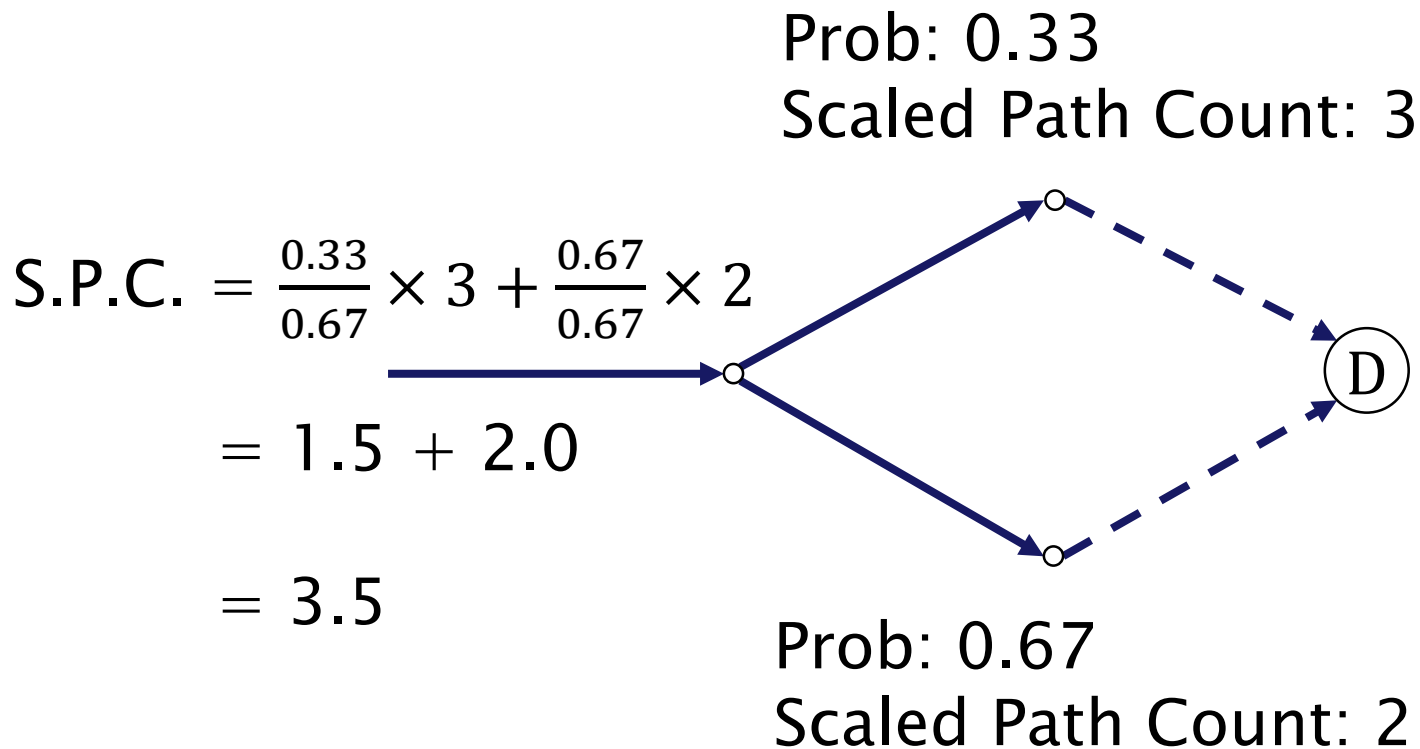
$$\tilde{N}^d(k) = \sum_{a \in A} \frac{P(a|k)}{\max_{a' \in A} P(a'|k)} \tilde{N}^d(a)$$

## Probability-scaled upstream path segments

$$\tilde{N}^u(k) = \sum_{a \in A} \frac{F(a)P(k|a)}{\max_{a' \in A} F(a')P(a')} \tilde{N}^u(a)$$

( $F(a)$  is uncorrected link flow)

# Example: scaled path count recursion





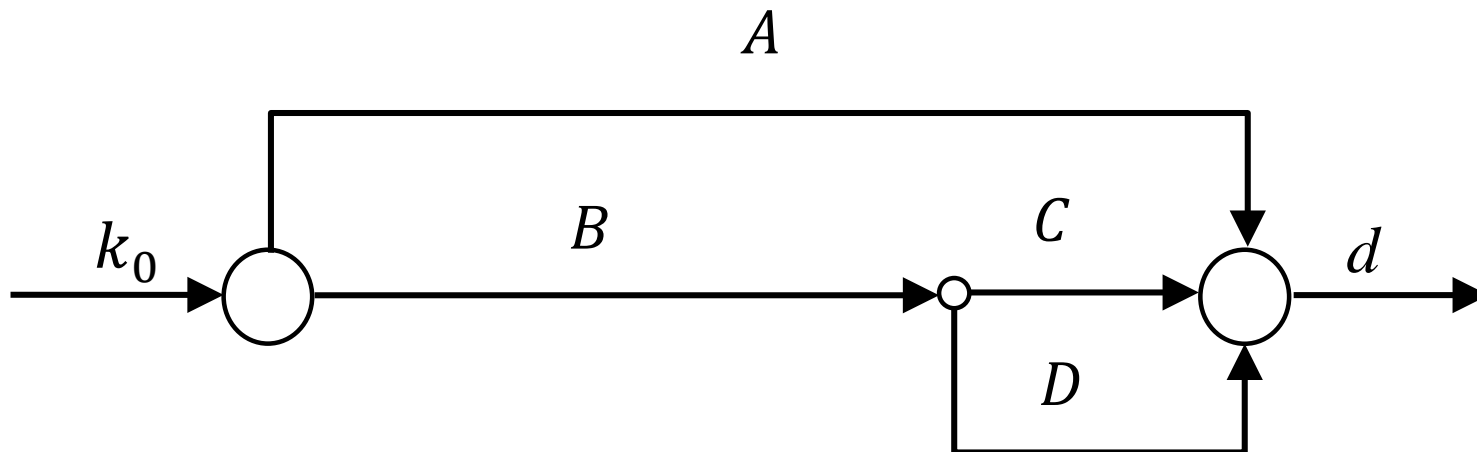
# Link size variable follows established form from path-based methods

$$LS(k) = \frac{m(k)}{\hat{M}(k)} \log \tilde{N}^d(k) \tilde{N}^u(k)$$

$m(k)$ : measure of link extent

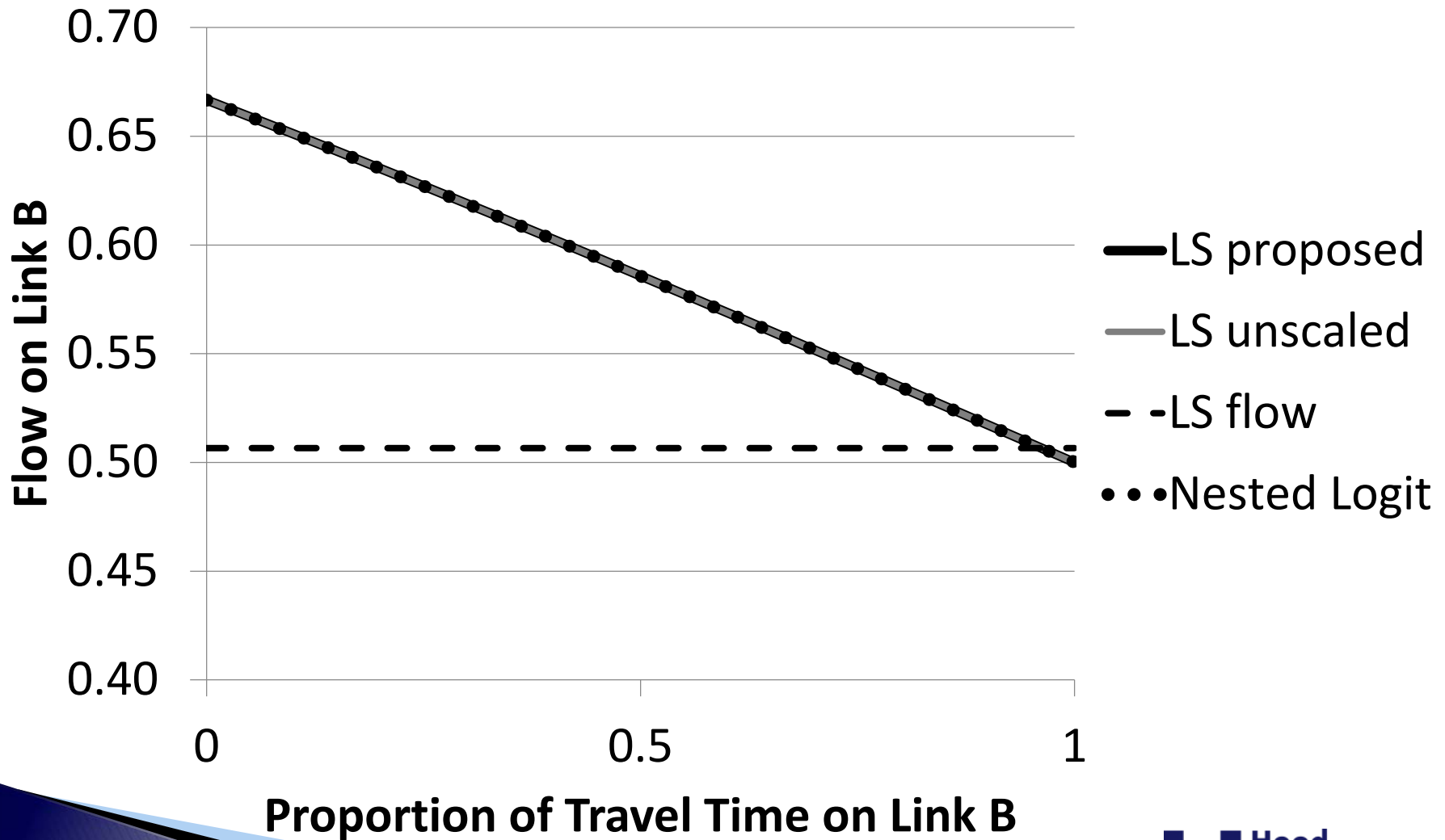
$\hat{M}(k)$ : expected measure of all paths containing  $k$

# Example: collapsing downstream alternatives

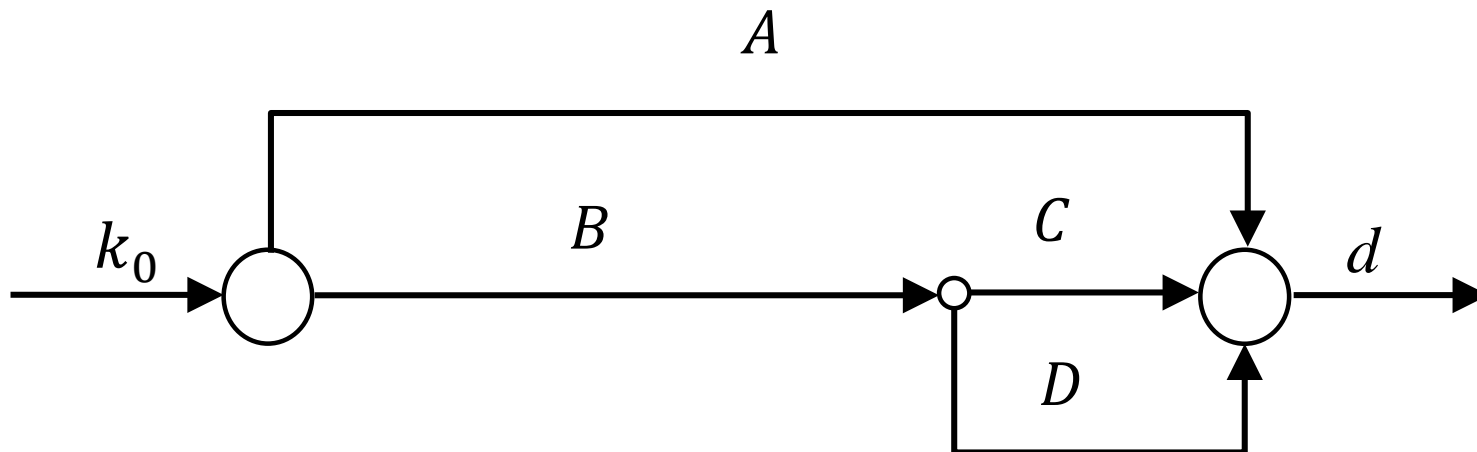


Link	Measure $m(a)$	Travel Time
A	1	1
B	$t$	$t$
C	$1 - t$	$1 - t$
D	$1 - t$	$1 - t$

# Example: collapsing downstream alternatives

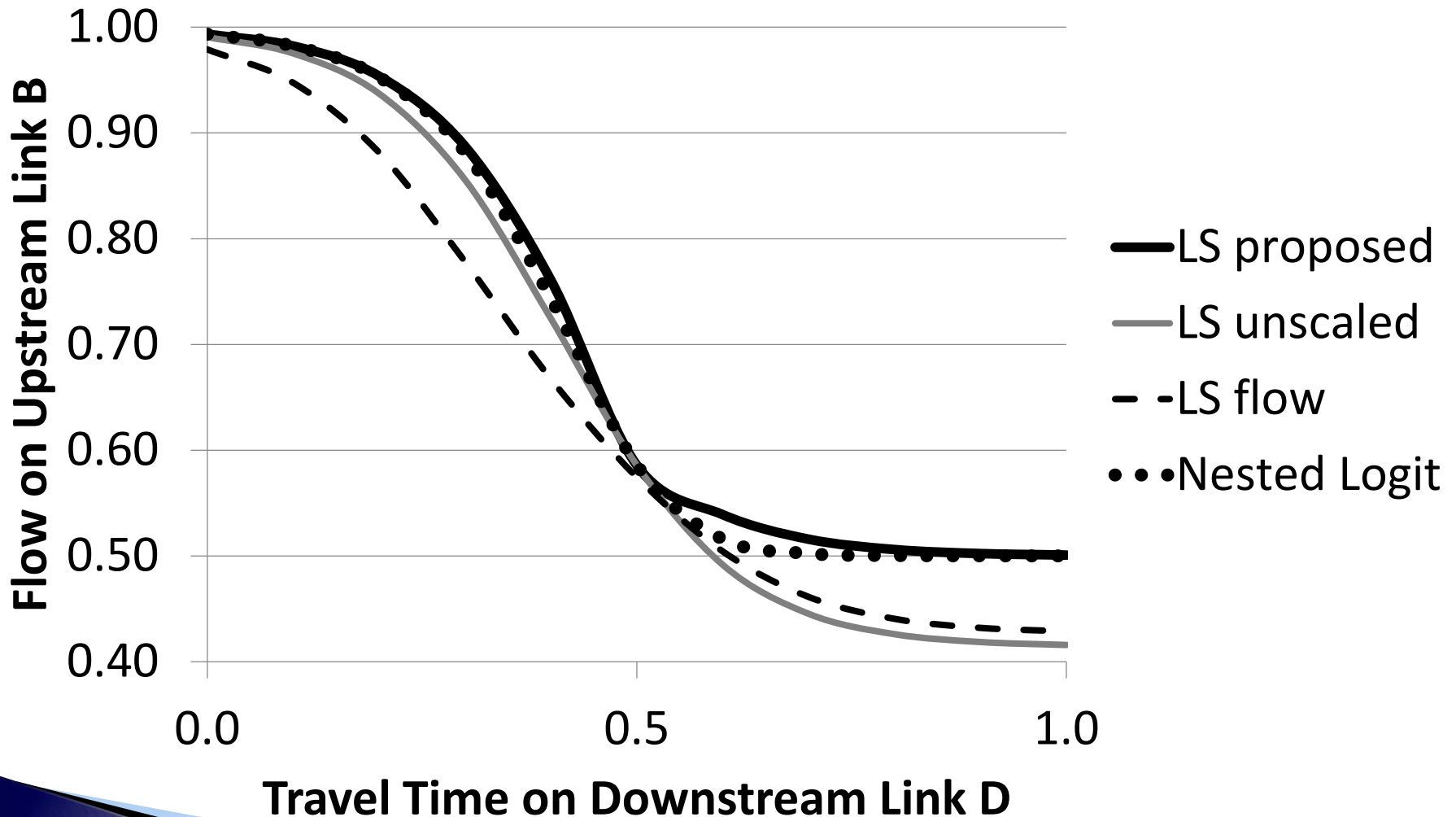


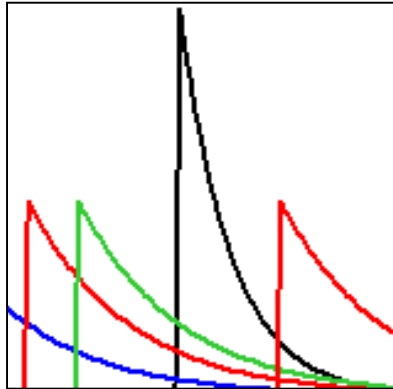
# Example: varying downstream utilities



Link	Measure $m(a)$	Travel Time
A	1	1
B	0.5	0.5
C	0.5	0.5
D	$t$	$t$

# Example: varying downstream utilities





# Reliability and Stochastic Arrivals

# Traveler responses to reliability are more complex for transit

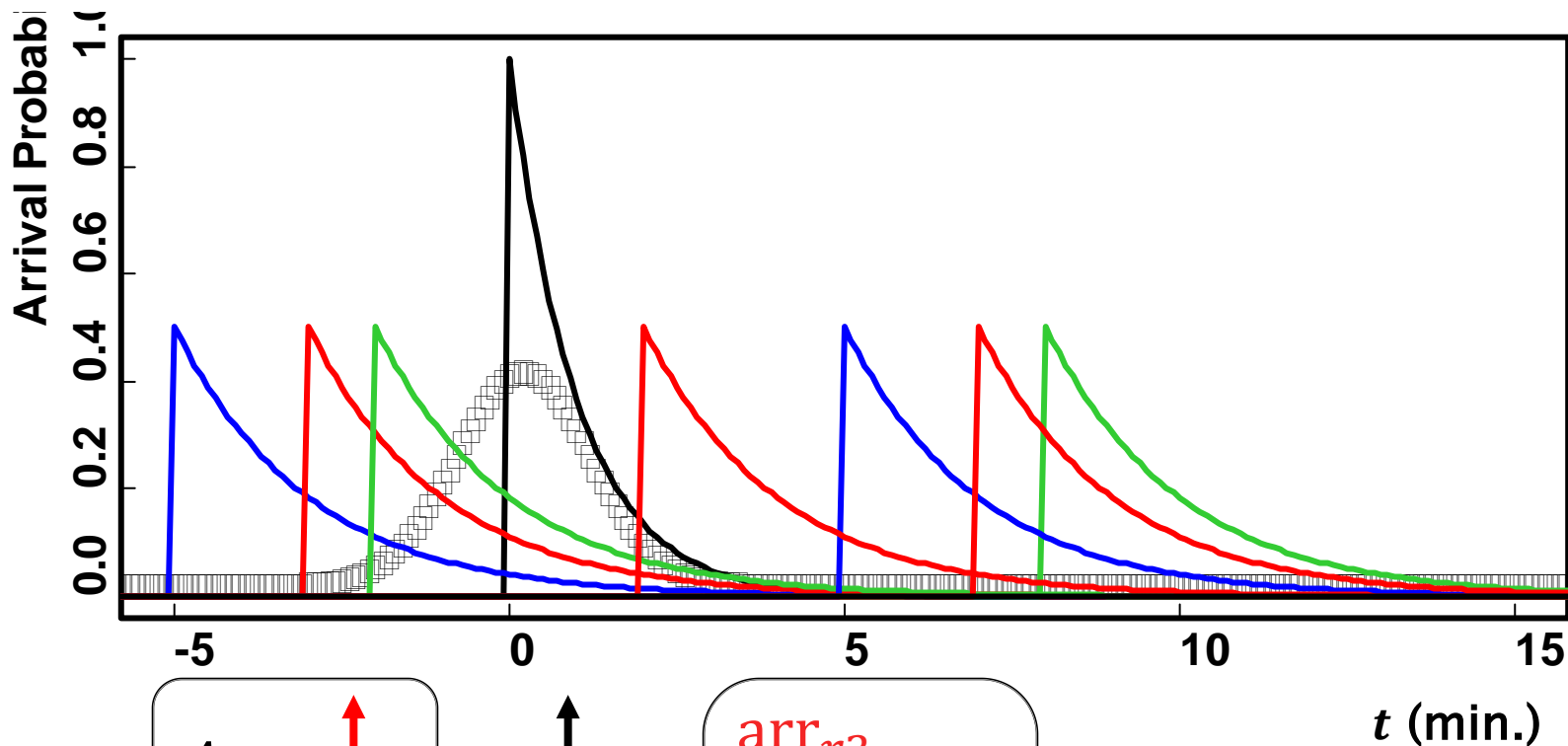
Poor schedule adherence reduces boarding probability for multiple reasons

- Direct disutility of excess wait times
- Missed connections
- Lateness in arrival sequence

“Reliability” term in utility function will not work

Problem requires dynamic choice probabilities

# Boarding is conditional on which vehicles have departed, are dwelling, or yet-to-arrive



$\underline{A}$   $\uparrow$   
 $dep_{r1}$

$\uparrow$   
 $arr_k$

$dwe_{b1}$   $\text{—————}$   
 $A^*$   $dwe_{g1}$   $\text{—————}$

$arr_{r2}$   
 $arr_{b2}$  = ?  
 $arr_{r3}$   
 $arr_{g2}$   $\overline{A}$



# Probability of boarding a dwelling vehicle is

$$P(a|k, \bar{A}, a \in A^*) \\ = \frac{e^{\frac{1}{\mu}[\beta_{dwe}E(dwe_a)+v(a|k)+V(a)]}}{\sum_{a' \in A^*(k)} e^{\frac{1}{\mu}[\beta_{dwe}E(dwe_{a'})+v(a'|k)+V(a')]} + e^{\frac{1}{\mu}E(U_{wait}|k, \bar{A})}}$$

Depends on expected utility of waiting  $E(U_{wait}|k, \bar{A})$

# What is the expected utility of waiting?

## Recursive formula depending on

- Conditional distribution of  $\text{arr}_k$  given  $\bar{A}$

$$\Phi_k(t|\bar{A}) = P(\text{arr}_k < t|\bar{A})$$

- Conditional probability that next arrival is  $a_i$

$$P\left(\min_{a_j \in \bar{A}} \text{arr}_{a_j} = \text{arr}_{a_i}\right)$$

- Expected utility of waiting for  $a_i$

$$\begin{aligned} & E(w(a_i|k, \text{arr}_k)) \\ &= \int_0^{\infty} w(a_i|k, t^+) d\tilde{\Phi}_{a^+}(t|A^+) \end{aligned}$$

# Probability of boarding a later vehicle given the decision to wait is

$$\begin{aligned} & P(a|\text{wait}(\bar{A})) \\ &= \sum_{a_i \in \bar{A}} P\left(\min_{a_j \in \bar{A}} \text{arr}_{a_j} = \text{arr}_{a_i}\right) \\ & \times \left( \begin{array}{l} \delta(a = a_i)P(\text{board}(\bar{A} \setminus \{a_i\})) \\ + \delta(a \neq a_i)P(\text{wait}(\bar{A} \setminus \{a_i\}))P(a|\text{wait}(\bar{A} \setminus \{a_i\})) \end{array} \right) \end{aligned}$$

## Marginal probability of boarding vehicle $a$ is

$$P(a|k) = \sum_{i=0}^{|A(k)|} \sum_{j=0}^i \sum_{\underline{A} \in C(A(k), i)} \sum_{\bar{A} \in C(A(k) \setminus \underline{A}, j)} P(\underline{A}, \bar{A}, A^*)$$

$$\times \left( \begin{array}{l} \delta(a \in A^*) \\ \times P(a|k, \bar{A}, a \in A^*) \\ + \delta(a \in \bar{A}) \left( 1 - \sum_{a' \in A^*} P(a|k, \bar{A}, a' \in A^*) \right) \\ \times P(a|\text{wait}(\bar{A})) \end{array} \right)$$

If delays are independent exponentials, there is an (excruciating) closed-form solution



Kathryn Coffel  
 David Ory  
 Lisa Zorn  
 Shimon Israel

Suzanne Childress  
 Stefan Coe  
 Joe Castiglione  
 Drew Cooper

Elizabeth Sall  
 Alireza Khani  
 Emma Frejinger  
 Tien Mai

**Thank you!**

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