



NETWORK EQUILIBRIUM UNDER CUMULATIVE PROSPECT THEORY WITH ENDOGENOUS STOCHASTIC DEMAND & SUPPLY

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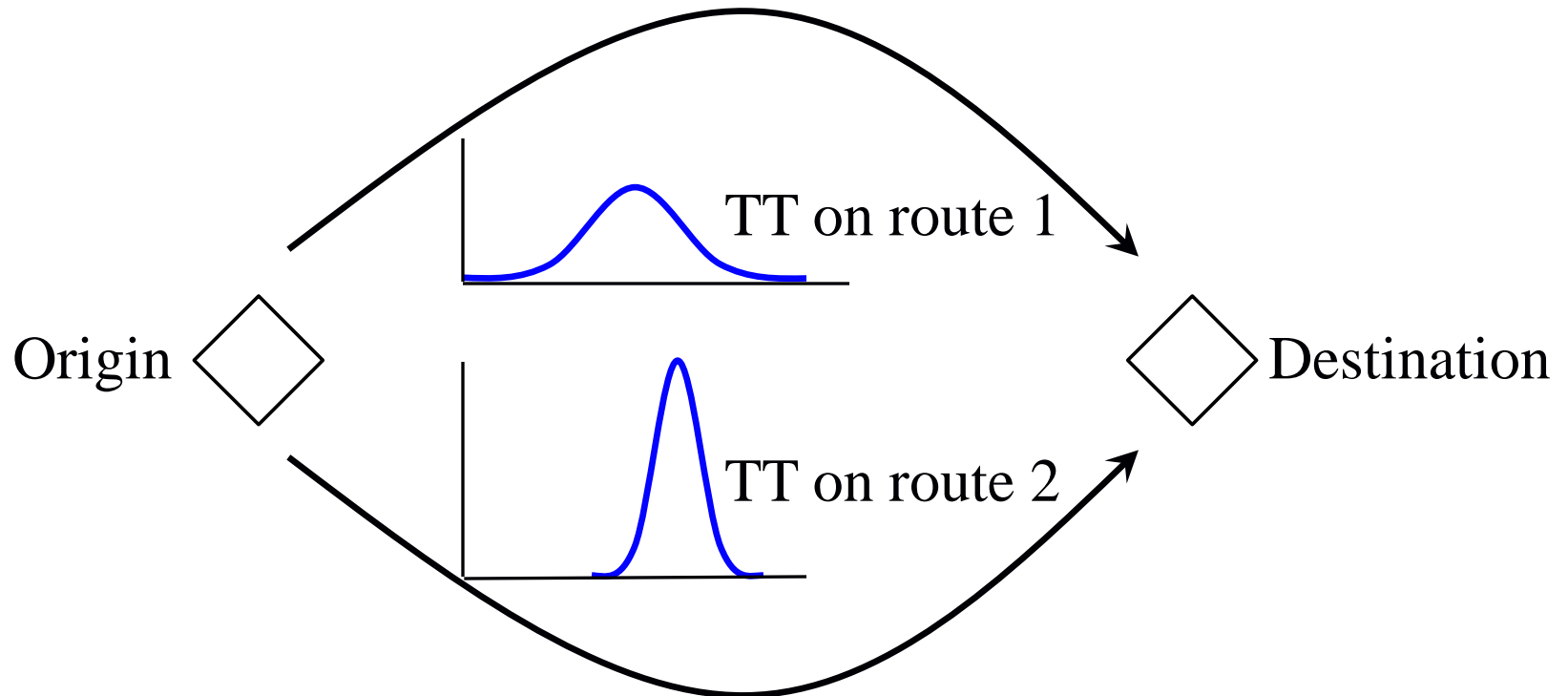
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Based on the paper Sumalee et al (2009) presented at ISTTT

Two-link Network

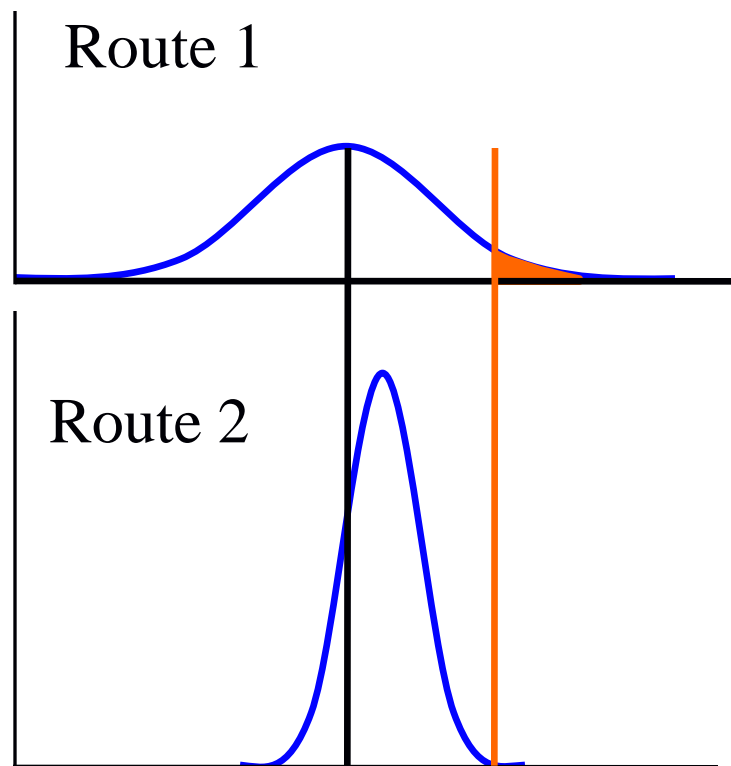


How do drivers choose between routes with uncertain travel times?





Approaches



Mean Travel Time \Rightarrow Route 1

Prob Late Arrival \Rightarrow Route 2

Mean + stdDev

Mean + Variance

95th Percentile of TT distribution

...



Empirical Evidence leads to PT

- Choice modellers have shown that *decisions under uncertainty* do not conform to maximising the expected benefit.
- The empirical evidence suggests two major modifications to ‘Expected Utility Maximisation’:
 - the carriers of value are gains/losses relative to a reference point
 - the value of each outcome is multiplied by a decision weight, not by an additive probability.

This generalisation of EUmax is Prospect Theory



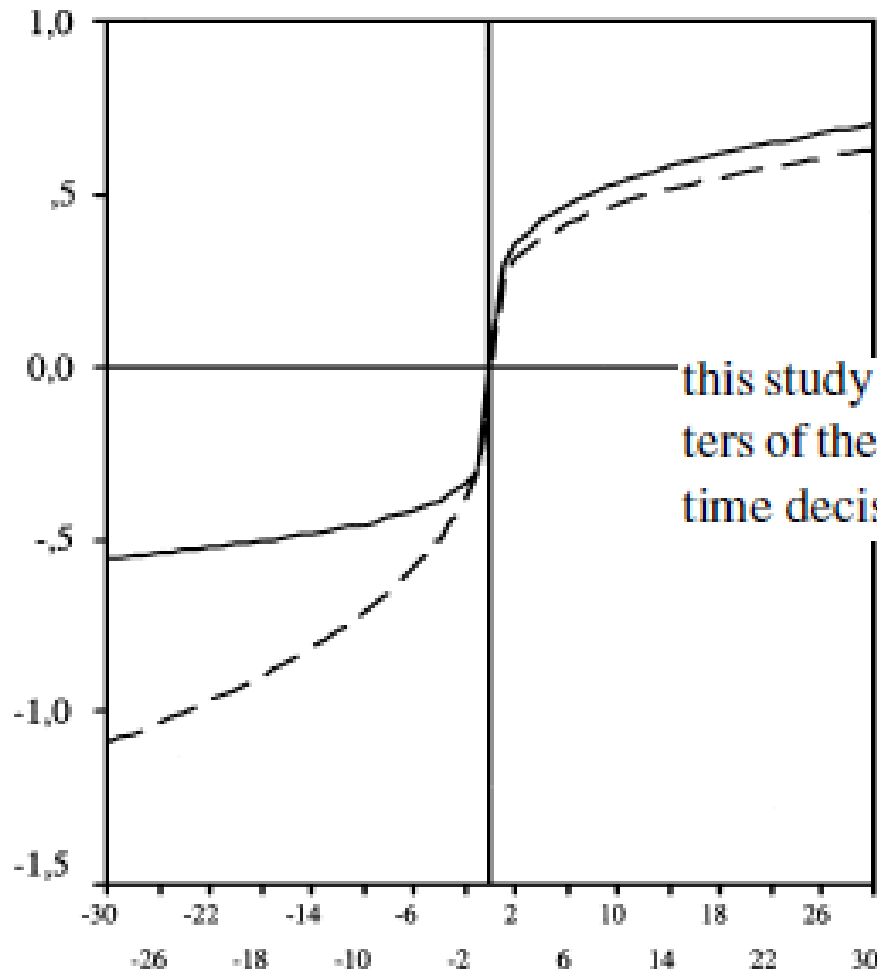
Context

- A traffic network has many sources of variability in demand and supply: results in **Travel Time Variability**.
- Drivers are aware that travel times are *uncertain* and include this TTV in their decision making.
- Senbil, M., & **Kitamura, R.**, (2004). Reference points in commuter departure time choice: a prospect theoretic test of alternative decision frames. *Journal of Intelligent Transportation Systems* 8, 19–31
- Jou, R.C., **Kitamura, R.**, Weng, M.C., Chen, C.C., (2008). Dynamic commuter departure time choice under uncertainty. *Transportation Research Part A* 42(5), 774-783.

assumed to reduce the likelihood of choosing that departure time. The empirical results indicate that around 20% of commuters are likely to switch their departure times and routes and most of commuters experience gains, and that preferred arrival times of commuters tend to be near their work starting times. Most importantly, it is shown that, consistent with prospect theory, commuters react asymmetrically to gains and losses.



Empirical evidence



this study the *quasi-gain region*. Estimation of the parameters of the value functions has also indicated that departure time decision is consistent with prospect theory.

— —
Value (First Approach)
——
Value (Second Approach)

Senbil and Kitamura (2004)

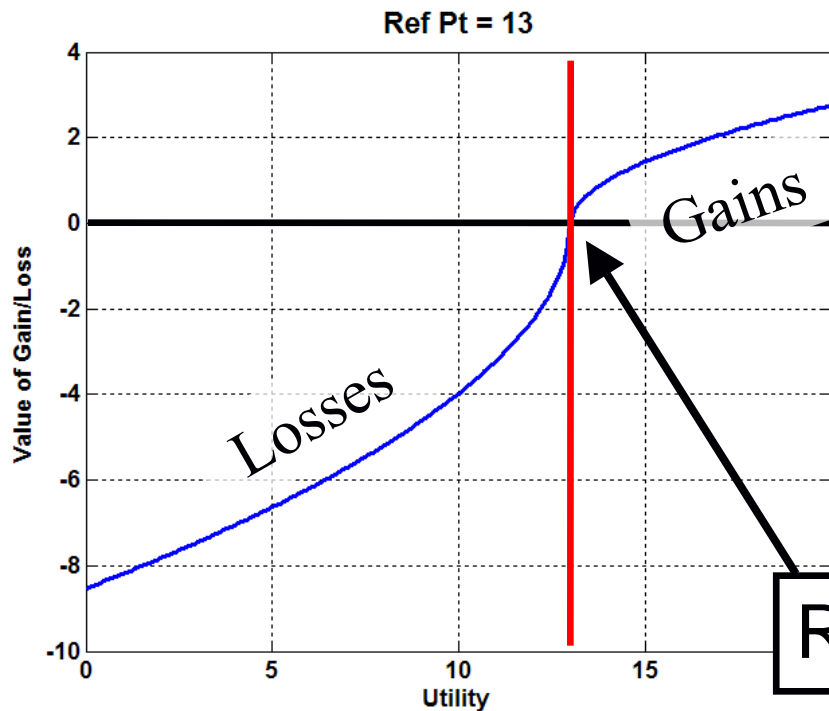
Schedule Delay



Prospect Theory Transformations

The fundamental embodiment (and impact) of PT is through the following two transformations:

Value Function $g(.)$



Probability Weighting Function $w(.)$



Reference Point



Reference Dependence

Empirical studies show that people's preferences over final outcomes depend on the reference point from which they are judged:

Kahneman & Tversky (1979)	for decision under risk
Kahneman <i>et al</i> (1990)	for choice among commodity bundles
Loewenstein & Prelec (1992)	for inter-temporal choice
Bateman <i>et al</i> (1997)	for contingent valuation
Dolan & Robinson (2001)	for welfare theory
Bleichrodt & Pinto (2002)	for multi-attribute utility
Senbil&Kitamura (2004)	for departure time choice
Jou, Kitamura et al (2008)	for departure time choice

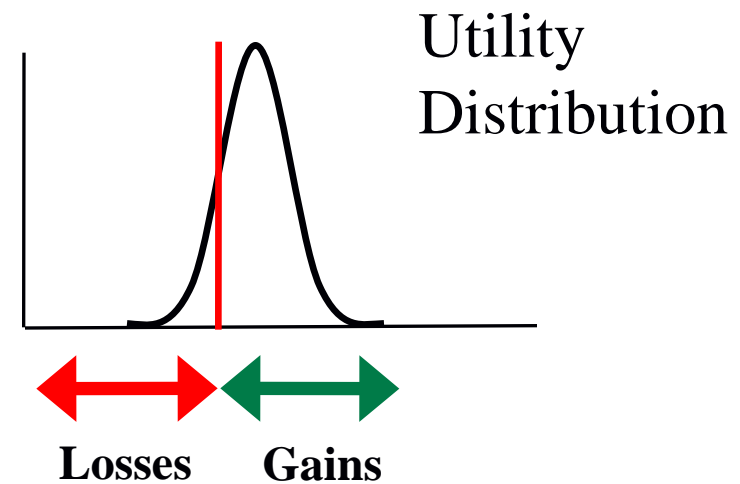
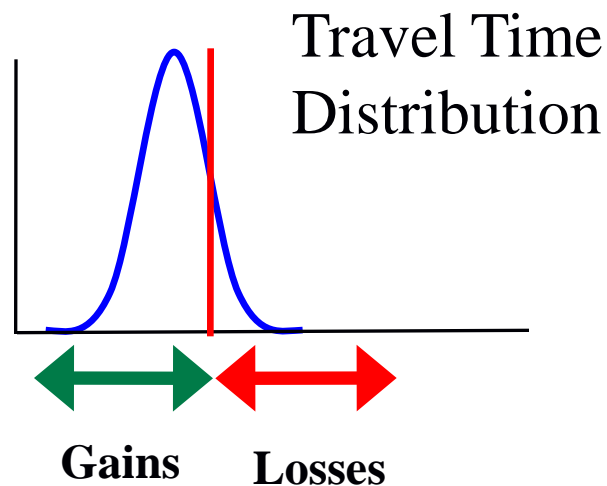


Travel Time \rightarrow Utility

If we have a distribution of path costs $C_k \sim N(c_k(\mathbf{x}), \sigma_k)$

It is convenient to consider the utility

$$U_k = U_{dest} - C_k = U_{dest} - c_k(\mathbf{x}) - \varepsilon_k$$





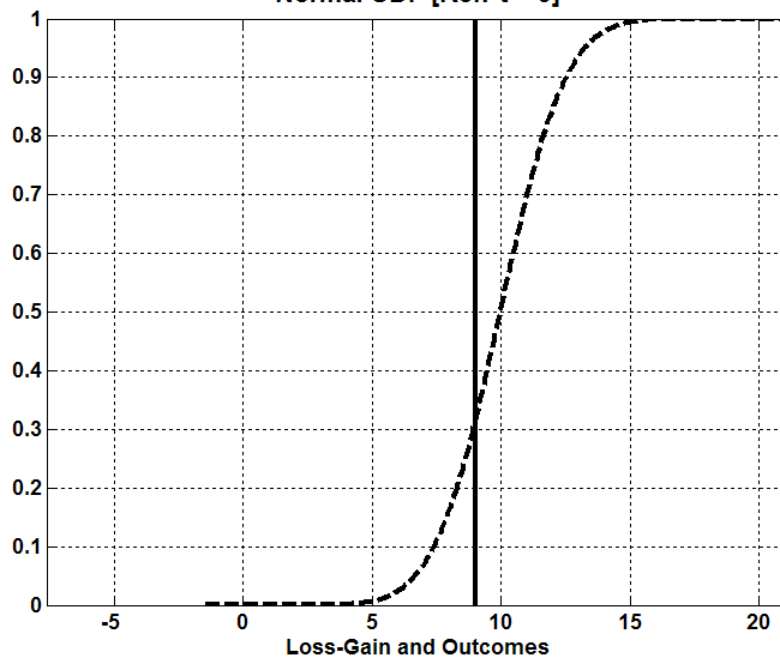
Applying the CPT Transforms



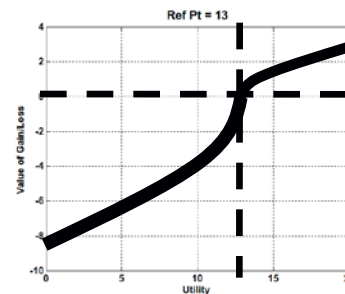
Weighting Function
Maps Probabilities



Normal CDF [RefPt = 9]



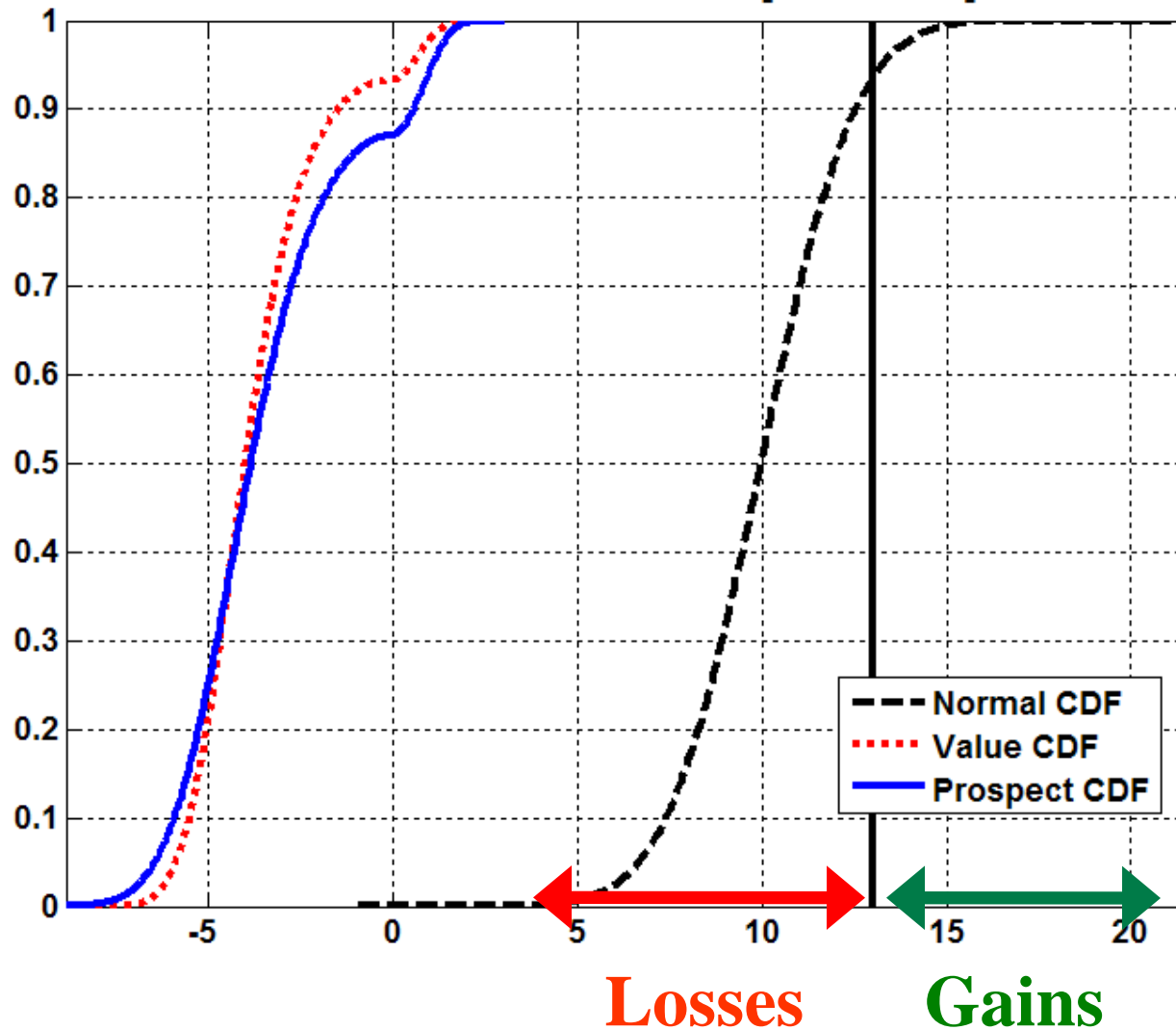
Value Function
Maps Outcome
Utilities





CPT Transformations

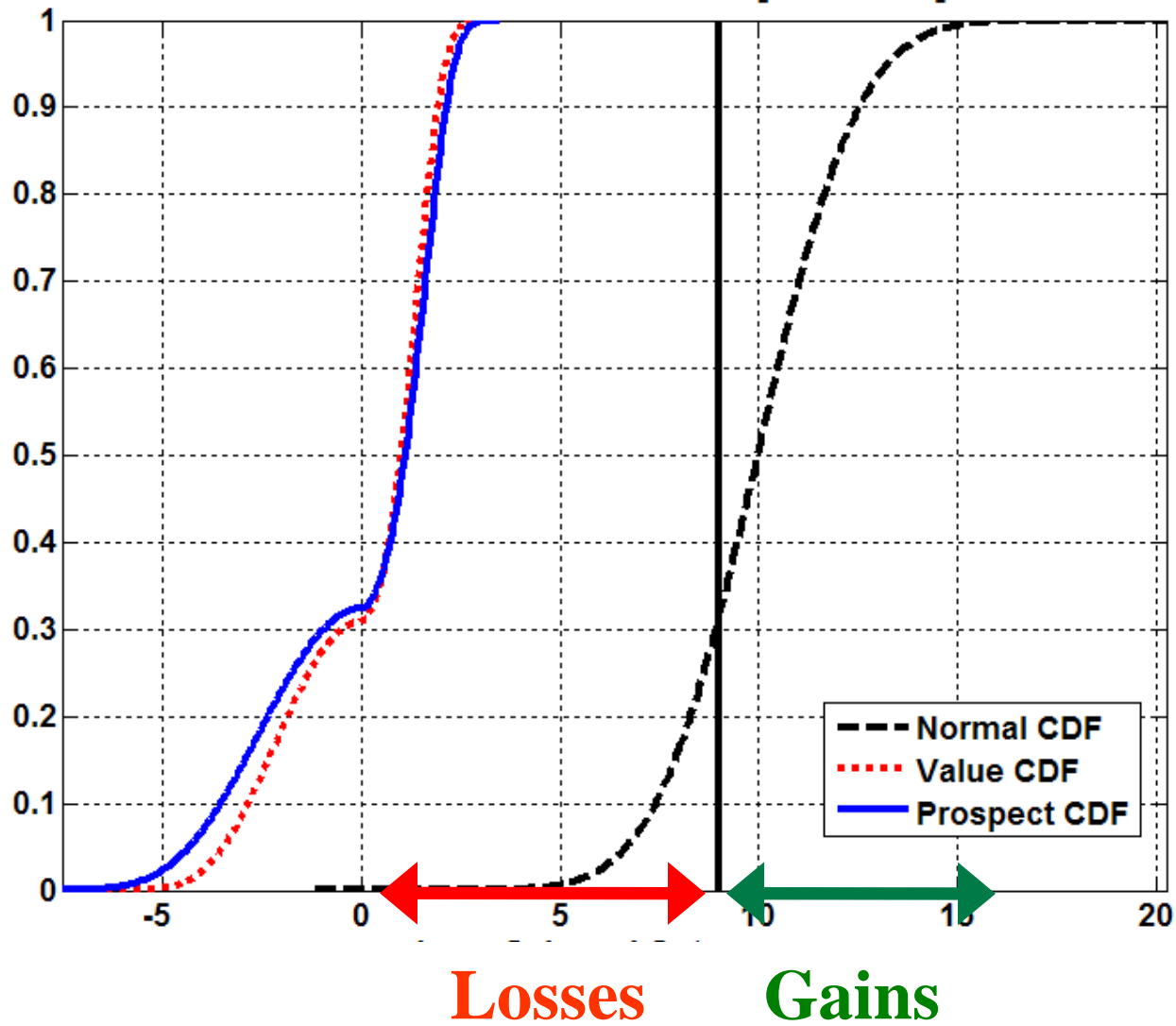
Transformed Normal CDF [RefPt = 13]



**Utility:
Cumulative
Distribution**



Transformed Normal CDF [RefPt = 9]





Cumulative Prospect Value

CPV for continuous distribution of outcomes

$$cpv = \int_{u_0}^{\infty} -\frac{dw(1 - F_U(u))}{du} \cdot g(u) du + \int_{-\infty}^{u_0} \frac{dw(F_U(u))}{du} \cdot g(u) du$$

$F_U(u)$ = Path Utility CDF (the outcome distribution)

$g(\cdot)$ = Value Function

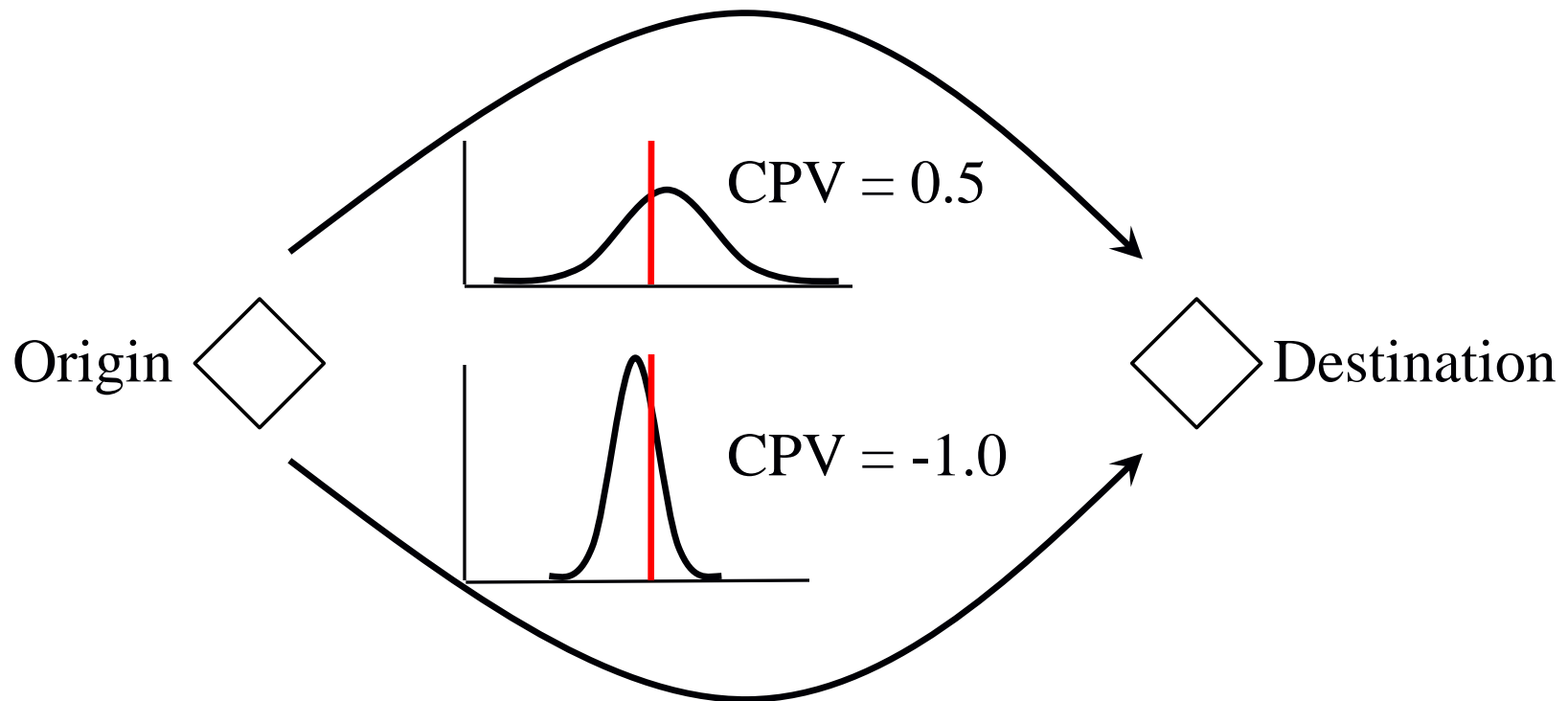
$w(\cdot)$ = Weight Function

u_0 = Reference Point

Two-link Network



How do drivers choose between routes with uncertain travel times?



Choose between alternatives according to CPV



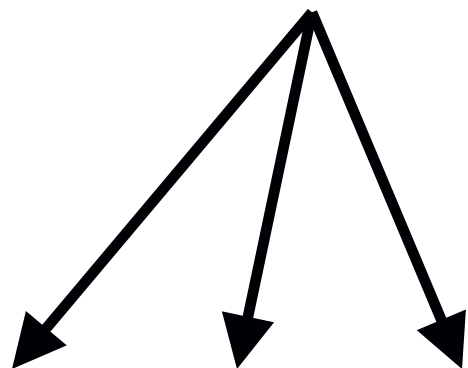
Making Variability Endogenous

- Part B Paper
 - Stochastic link travel times with exogenously defined distribution
 - Reference point (and other CPT parameters) assumed
 - Compute equilibrium: show dependence on RefPt and other parameters
- Extension here – to endogenize variability
 - **Stochastic OD demand...**
 - ...split into path flows via ratio of means f/q hence stochastic path flows
 - Gives rise to stochastic link flows (with correlations)
 - Independently **stochastic link capacities**
 - Can compute resulting stochastic travel times (with correlations)
 - CPT applied to route choice



Stochastic Demand

Lognormal Demand $Q \sim LN(\mu, \sigma)$ with mean q



Split Q into path flows $F_k = (f_k/q) \cdot Q$

Where f_k is mean flow on path k

Path Flows Lognormal Distributed $F_k \sim LN(\mu_k, \sigma_k)$

Conservation condition on means: $q = \sum_k f_k$

Gives stochastic, correlated link flows.



Stochastic Supply

Link capacities are assumed to be independent random variables

$$T_1 = t_a^0 + b_a \left(\frac{X_a}{C_a} \right)^{n_a}$$

With $C_a \sim LN(\mu_{ca}, \sigma_{ca})$ and independent of X_a

We therefore have non-trivial covariance matrices for link flows and link travel times.

Path costs arise from standard link-additive model.

BUT CPV is not link-additive!



Equilibrium Condition

With $cpv_k(x)$ the flow dependent CPV on route k we have that (demand feasible) \mathbf{f}^* is a CPV-UE if and only if:

$$\mathbf{cpv}(\mathbf{f}^*)^T (\mathbf{f} - \mathbf{f}^*) \leq 0 \quad \forall \mathbf{f} \in F$$

$$\min_{\mathbf{f} \in \Omega} \{ G(\mathbf{f}) \}$$

$$G(\mathbf{f}) = \max_{\mathbf{g} \in \Omega} \left(\mathbf{cpv}(\mathbf{f})^T \cdot \mathbf{g} - \mathbf{cpv}(\mathbf{f})^T \cdot \mathbf{f} \right)$$



Equilibrium assignment algorithm

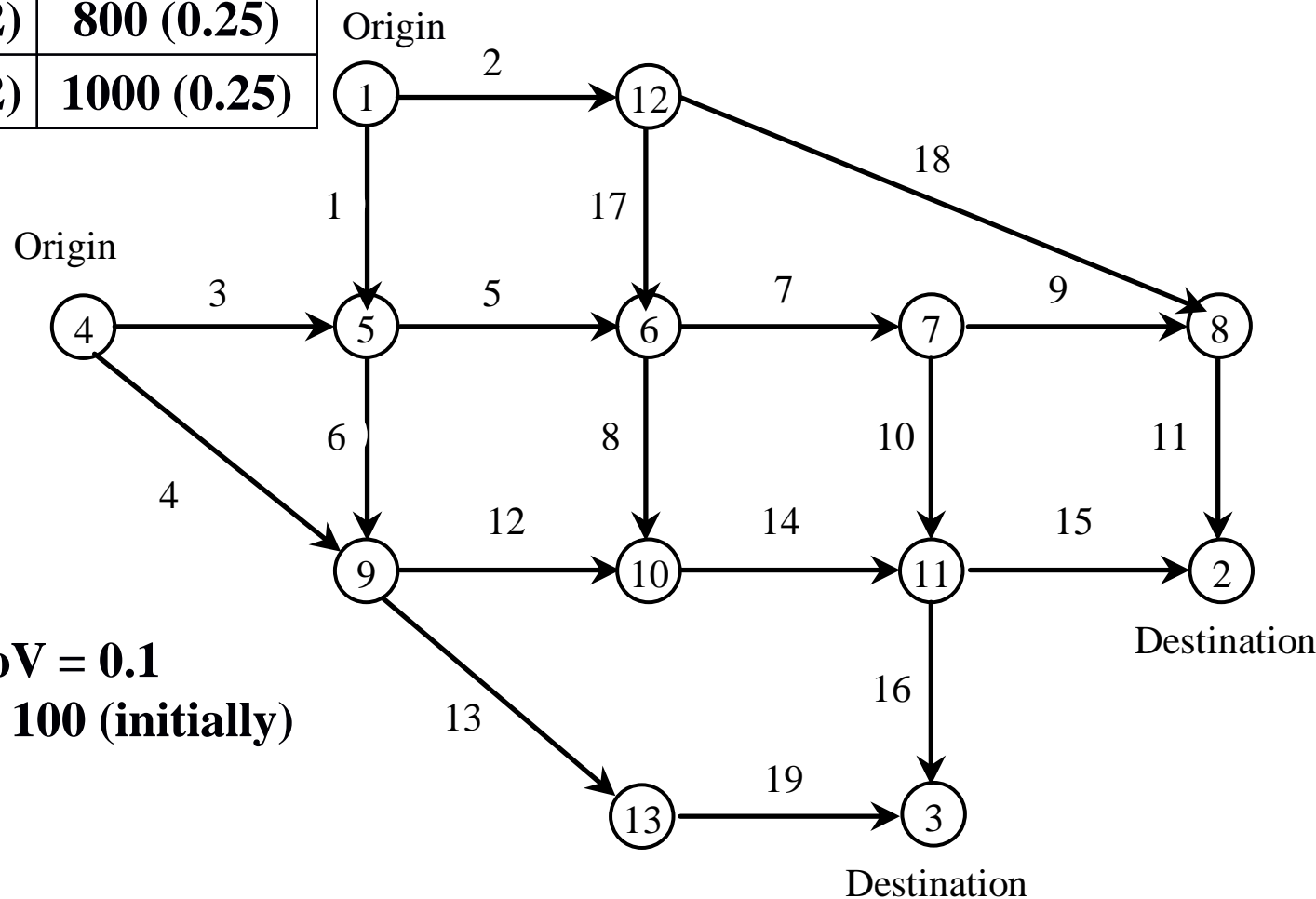
- Generate path set. Find initial feasible mean path flow \mathbf{f}
- Compute travel time distribution for each path.
- Compute prospect value for each path & find ‘best’ path
- Assign mean OD flow to best path; gives auxiliary flow \mathbf{g}
- Step from \mathbf{f} toward \mathbf{g} (as in MSA)
- Test convergence



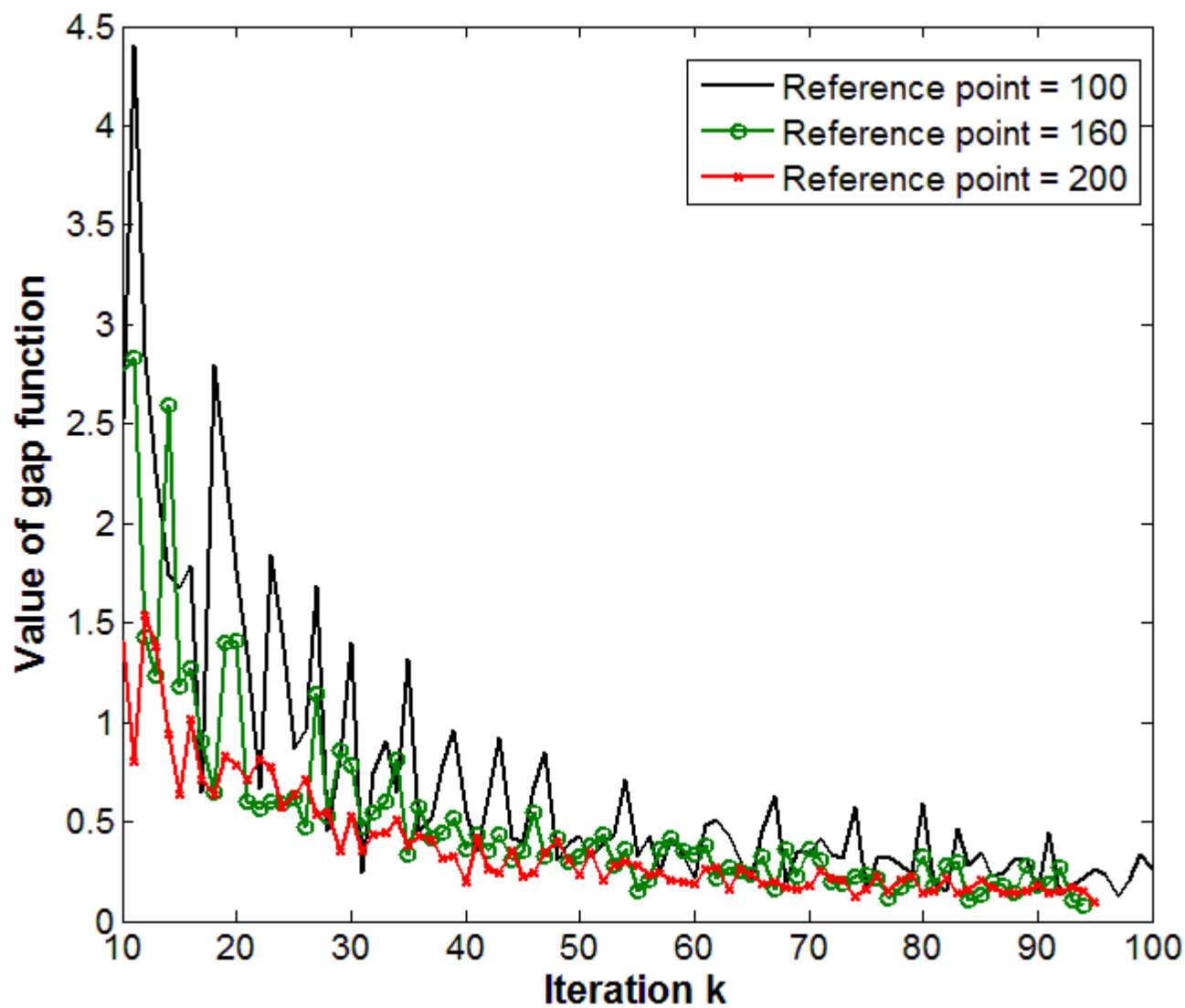
Nguyen and Dupuis Network

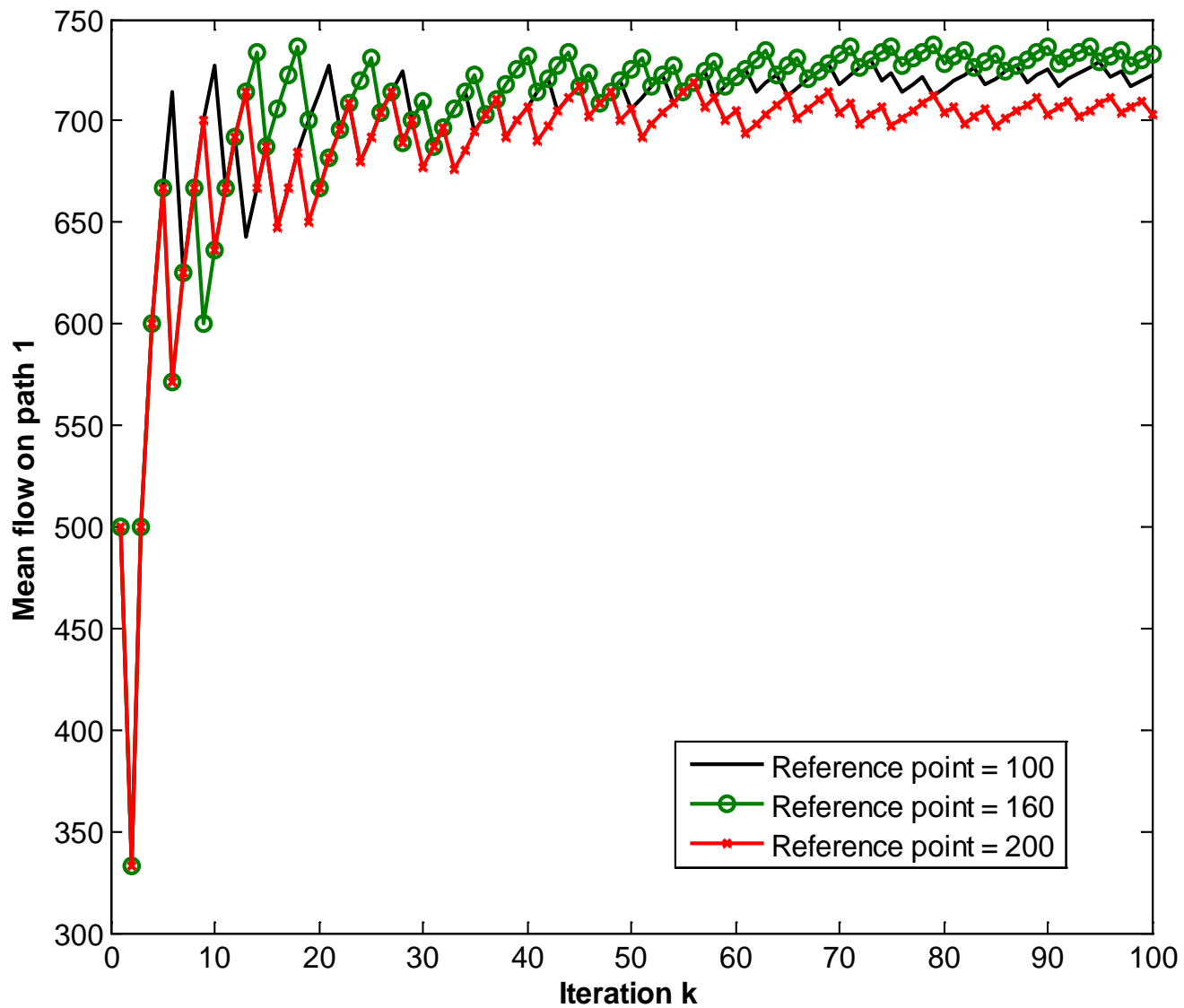
OD demand (CoV)

Orig\Dest	2	3
1	1000 (0.2)	800 (0.25)
4	1500 (0.2)	1000 (0.25)



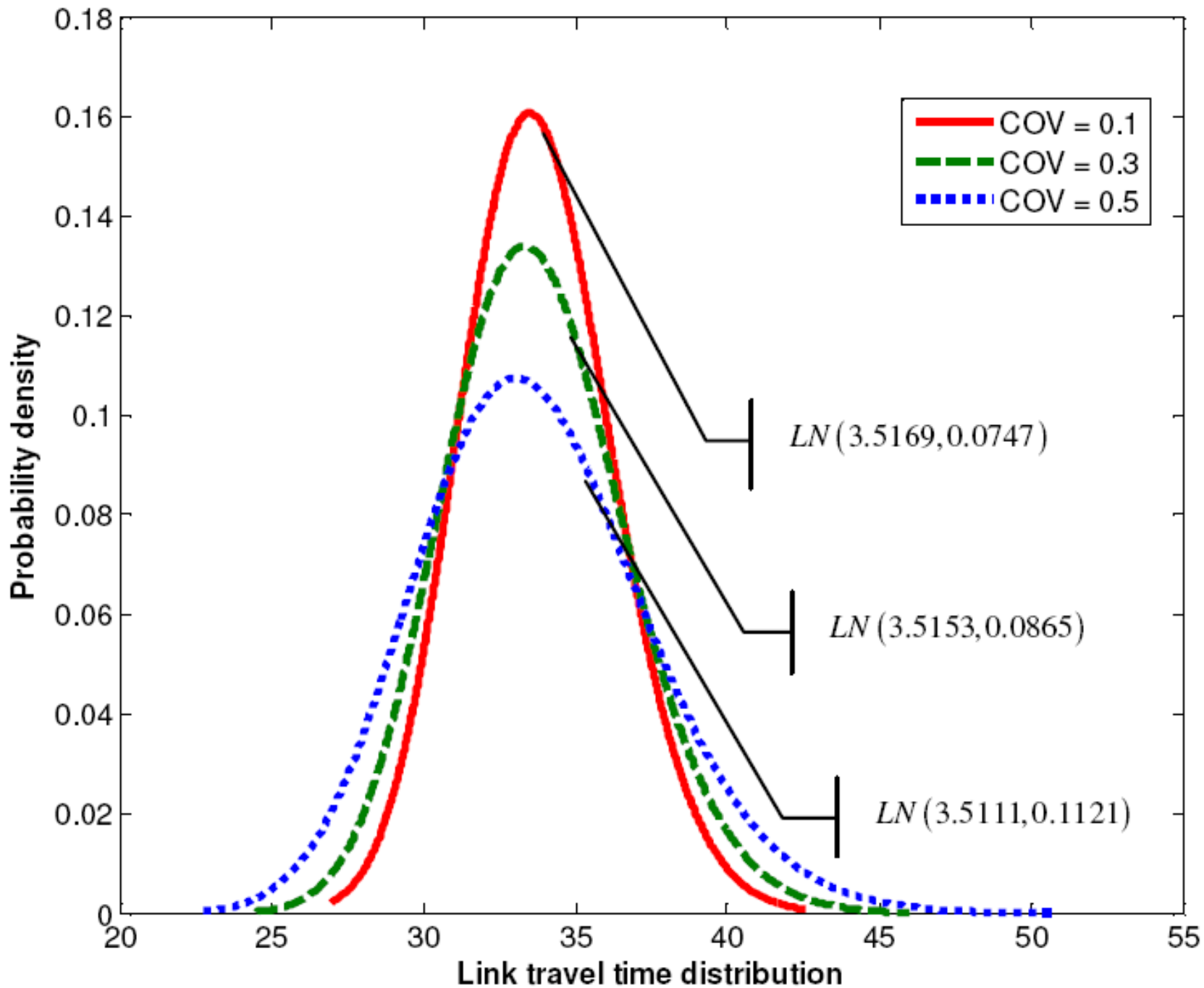
Link Capacities CoV = 0.1
Reference Points = 100 (initially)





OD movement	OD pair	Path	Link sequence	PV	Flow		Travel time	
					Expected	SD	Expected	SD
1	(1,2)	1	2-18-11	49.188	685	137	61.578	5.090
		2	1-5-7-9-11	49.144	150	30	63.680	4.784
		3	1-5-7-10-15	-	-	-	-	-
		4	1-5-8-14-15	-	-	-	-	-
		5	1-6-12-14-15	-	-	-	-	-
		6	2-17-7-9-11	-	-	-	-	-
		7	2-17-7-10-15	49.123	165	33	64.000	4.729
		8	2-17-8-14-15	-	-	-	-	-
2	(4,2)	9	4-12-14-15	40.976	310	62	67.106	6.154
		10	3-5-7-9-11	40.967	411	82	68.130	6.086
		11	3-5-7-10-15	40.928	294	59	68.621	6.057
		12	3-5-8-14-15	40.970	169	34	68.398	6.067
		13	3-6-12-14-15	40.965	316	63	68.675	6.048
3	(1,3)	14	1-6-13-19	23.643	351	88	60.576	4.516
		15	1-5-7-10-16	23.626	10	0	61.846	4.062
		16	1-5-8-14-16	23.660	95	24	61.623	4.136
		17	1-6-12-14-16	-	-	-	-	-
		18	2-17-7-10-16	23.649	243	61	61.675	4.121
		19	2-17-8-14-16	23.659	101	25	61.452	4.206

Link 18

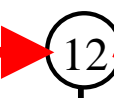




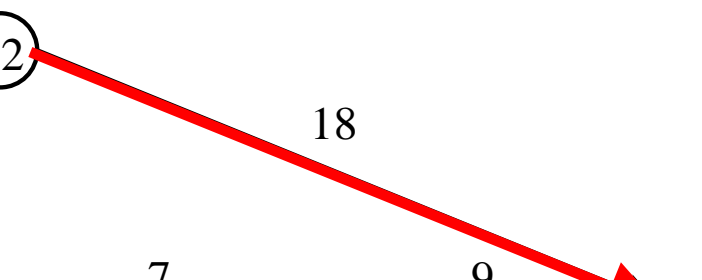
Origin



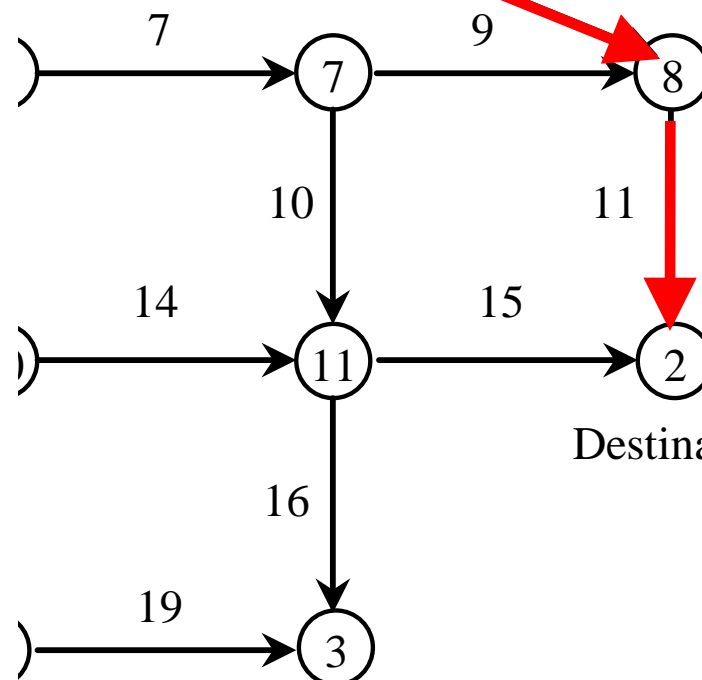
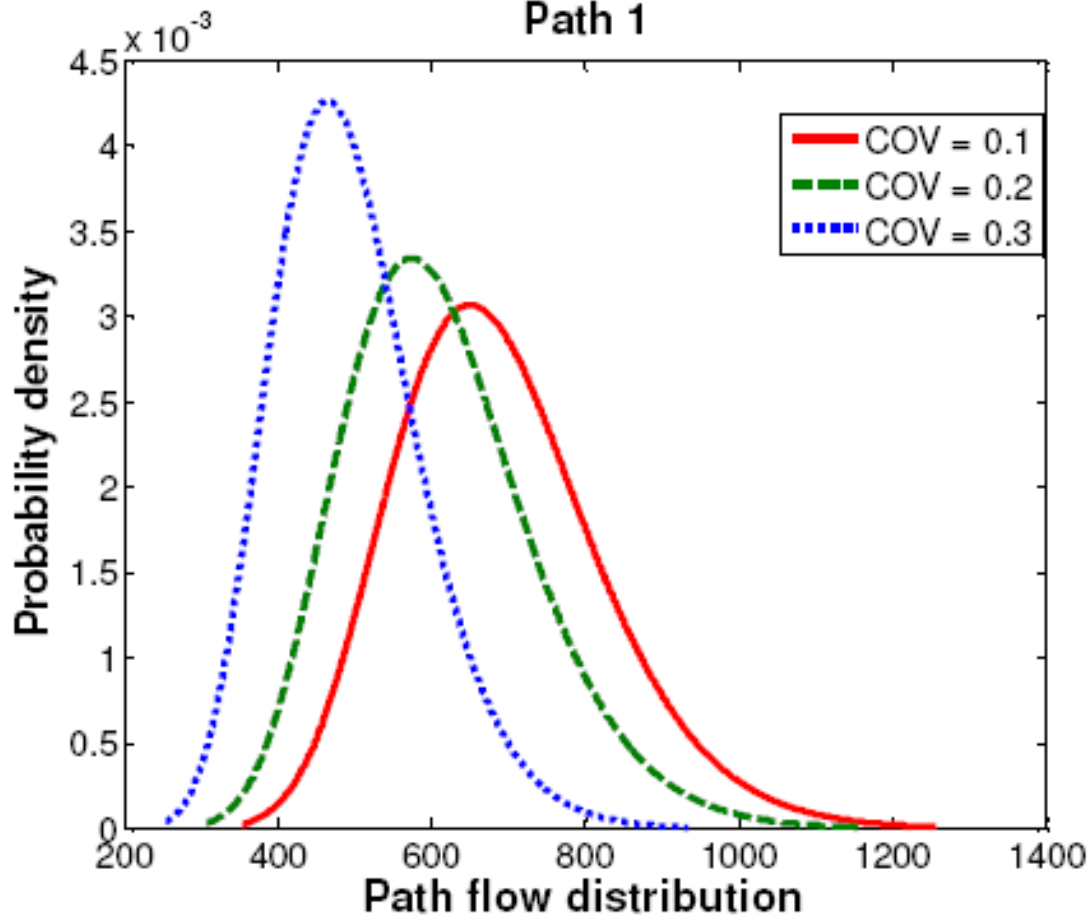
2



18



Path 1

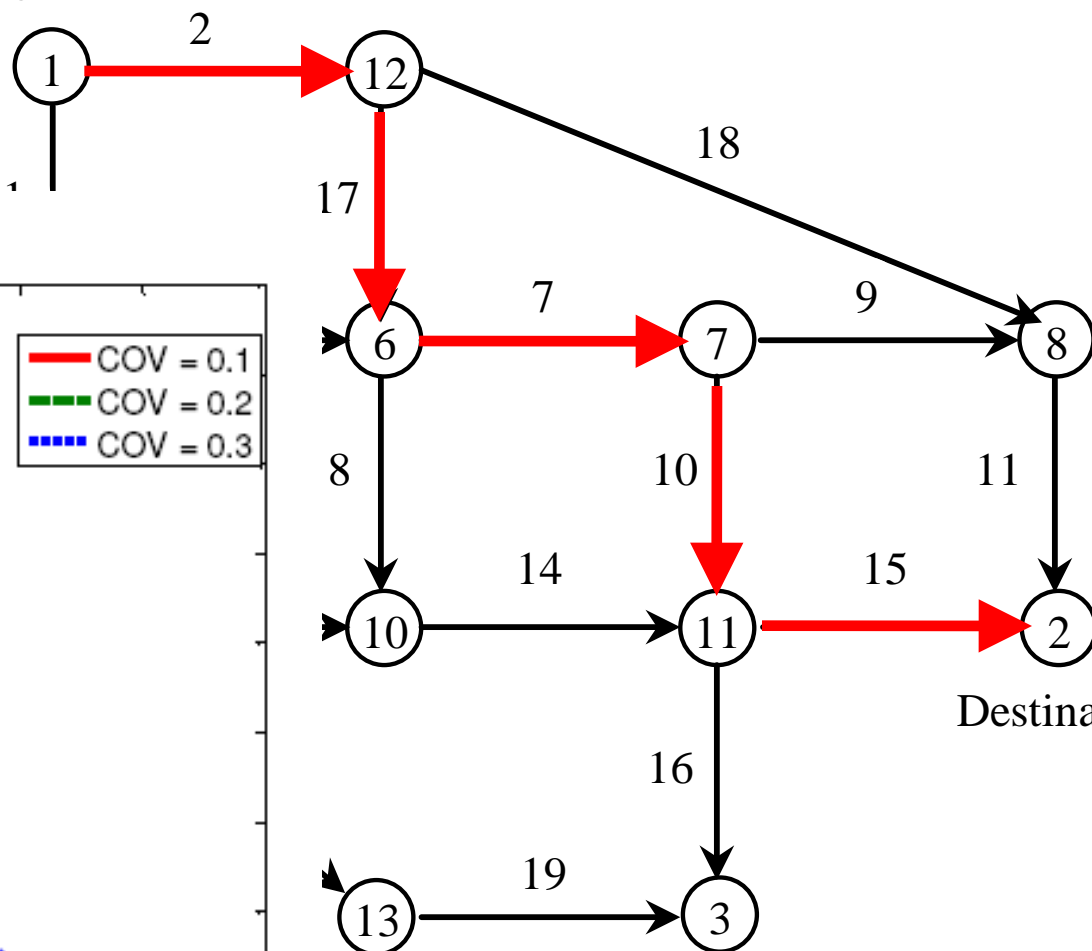


Destination

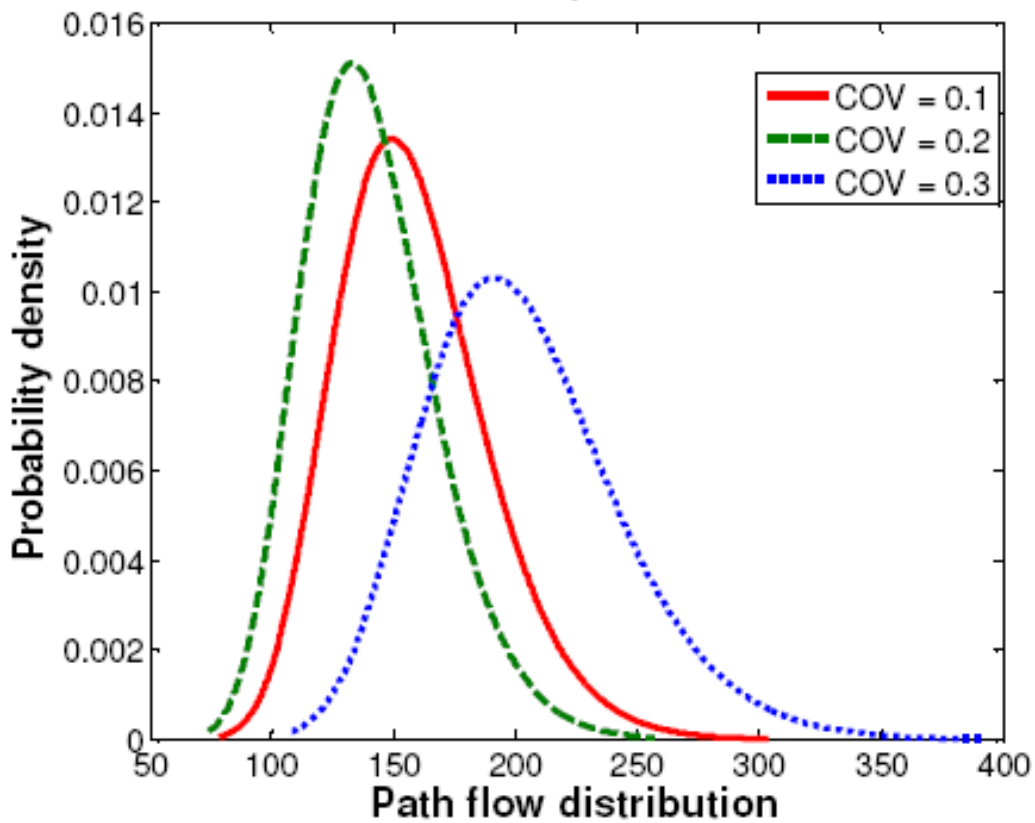
Destination



Origin



Path 7



Destination

Destination

Summary



Cumulative Prospect Theory as paradigm for route choice with stochastic travel times.

Continuous formulation of cumulative prospect theory presented.

Equilibrium condition defined.

Potential future studies:

- Solve for equilibrium efficiently and consider larger networks
 - **Path based** criterion so needs thought
- Network inference (estimation) from the observed link/path data
- Investigate impact of assuming CPV over $E[U_{\max}]$ or ‘reliability’ terms
 - How is UE recovered from CPV in the limit?
- Define reference point via a day-to-day learning process?
 - Initial reference point updated by experience, new alternatives added to choice set as reference point changes?
 - Does this converge? To a sensible solution? Is this process stable?