

Tomography-based Overlay Network Monitoring

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Motivation

- Infrastructure ossification led to thrust of overlay and P2P applications
- Such applications flexible on paths and targets, thus can benefit from E2E distance monitoring
 - Overlay routing/location
 - VPN management/provisioning
 - Service redirection/placement ...
- Requirements for E2E monitoring system
 - **Scalable & efficient**: small amount of probing traffic
 - **Accurate**: capture congestion/failures
 - Incrementally deployable
 - Easy to use

Existing Work

- General Metrics: RON (n^2 measurement)
- Latency Estimation
 - Clustering-based: IDMaps, Internet Isobar, etc.
 - Coordinate-based: GNP, ICS, Virtual Landmarks
- Network tomography
 - Focusing on inferring the characteristics of physical links rather than E2E paths
 - Limited measurements -> under-constrained system, unidentifiable links

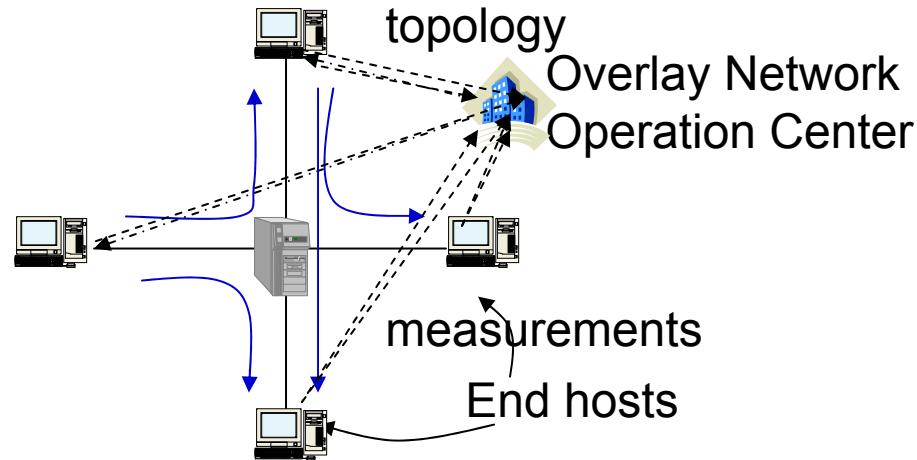
Problem Formulation

Given an overlay of n end hosts and $O(n^2)$ paths, how to select a minimal subset of paths to monitor so that the loss rates/latency of all other paths can be inferred.

Assumptions:

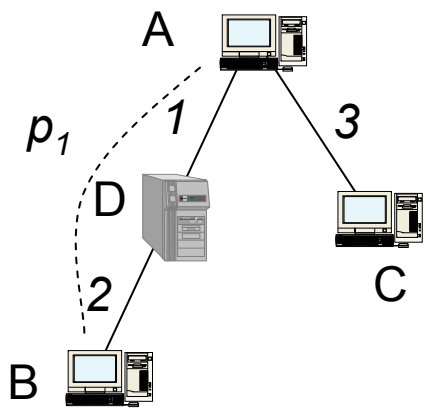
- Topology measurable
- Can only measure the E2E path, not the link

Our Approach



Select a basis set of k paths that fully describe $O(n^2)$ paths ($k \ll O(n^2)$)

- Monitor the loss rates of k paths, and infer the loss rates of all other paths
- Applicable for any additive metrics, like latency

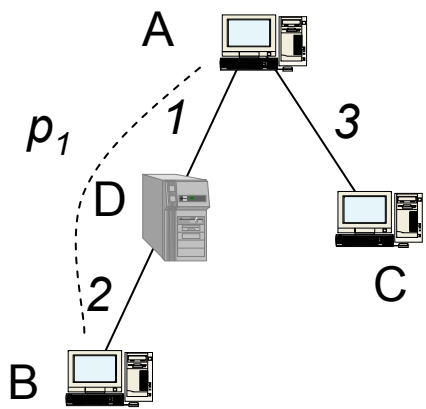


Modeling of Path Space

Path loss rate p , link loss rate / $1 - p_1 = (1 - l_1)(1 - l_2)$

$$\log(1 - p_1) = \log(1 - l_1) + \log(1 - l_2) = \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} \log(1 - l_1) \\ \log(1 - l_2) \\ \log(1 - l_3) \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = b_1$$

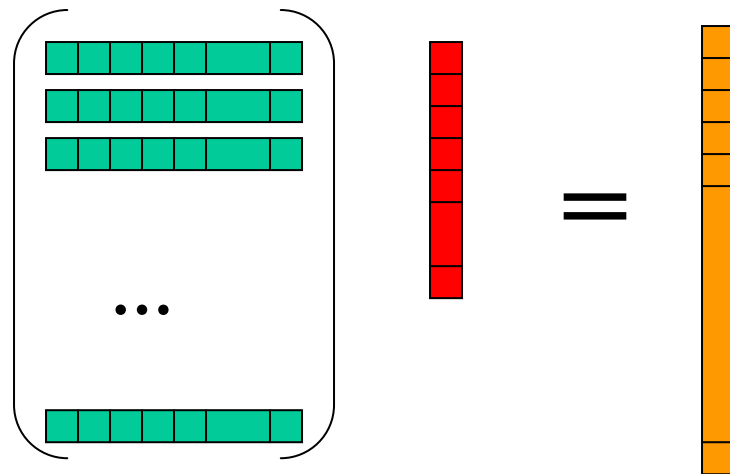


Putting All Paths Together

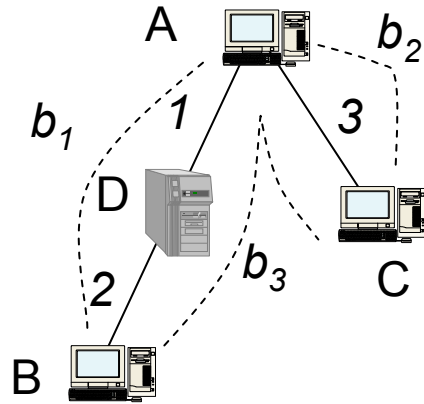
Totally $r = O(n^2)$ paths, s links, $s \ll r$

$$Gx = b, \text{ where path matrix } G \in \{0 | 1\}^{r \times s}$$

link loss rate vector $x \in \mathfrak{R}^{s \times 1}$, path loss rate vector $b \in \mathfrak{R}^{r \times 1}$

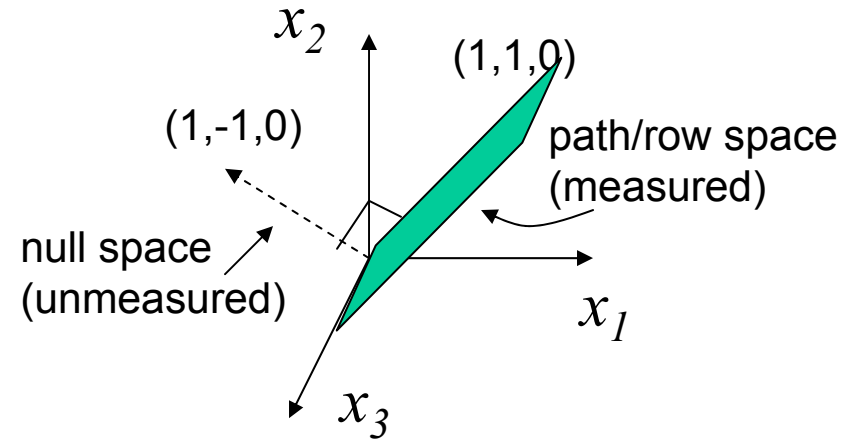


Sample Path Matrix



$$G = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

$$G \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$



- $x_1 - x_2$ unknown \Rightarrow cannot compute x_1, x_2
- Set of vectors $\alpha[1 \ -1 \ 0]^T$ form null space
- To separate identifiable vs. unidentifiable components:
 $x = x_G + x_N$

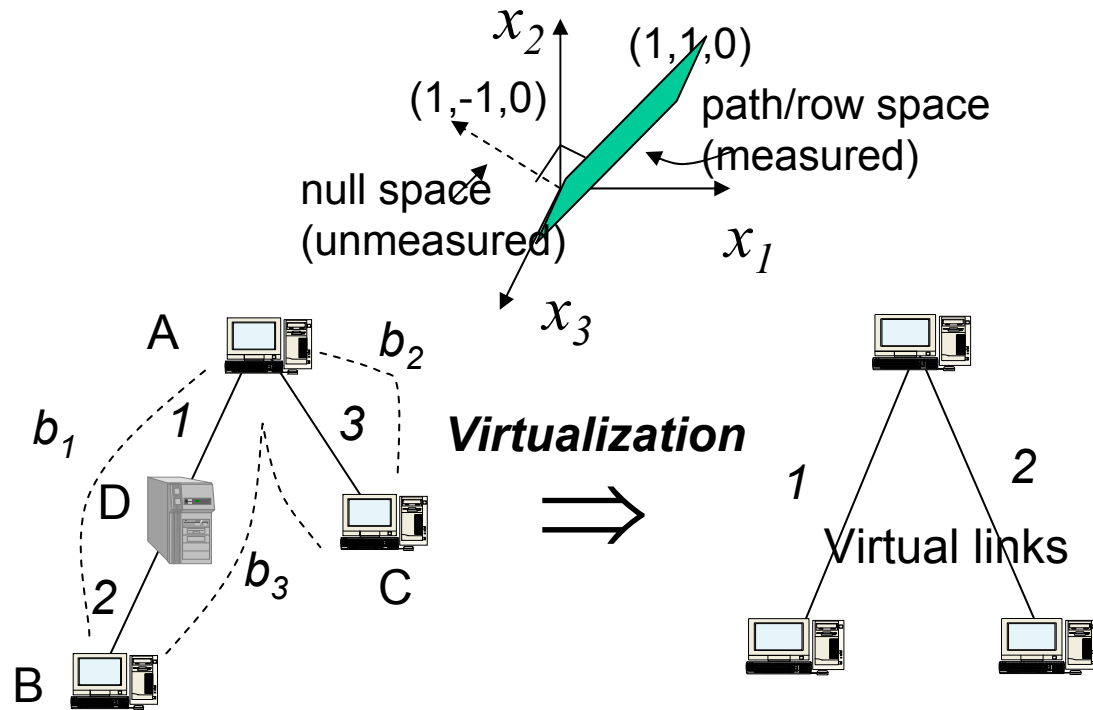
$$x_G = \frac{(x_1 + x_2)}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} b_1 / 2 \\ b_1 / 2 \\ b_2 \end{bmatrix}$$

$$x_N = \frac{(x_1 - x_2)}{2} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

Intuition through Topology Virtualization

Virtual links:

- Minimal path segments whose loss rates uniquely identified
- Can fully describe all paths
- x_G is composed of virtual links



$$x_G = \frac{(x_1 + x_2)}{2} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} b_1 / 2 \\ b_1 / 2 \\ b_2 \end{bmatrix}$$

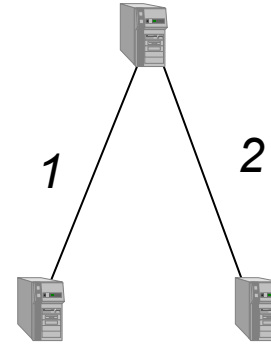
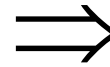
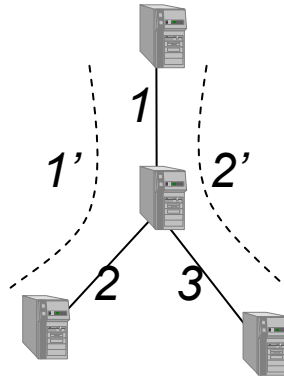
All E2E paths are in path space, i.e., $Gx_N = 0$

$$b = Gx = Gx_G + Gx_N = Gx_G$$

More Examples

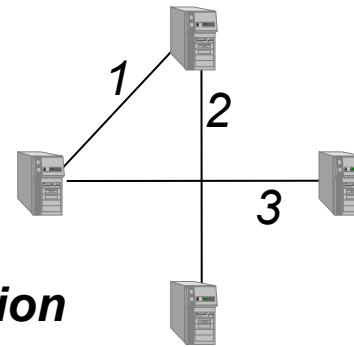
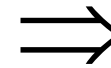
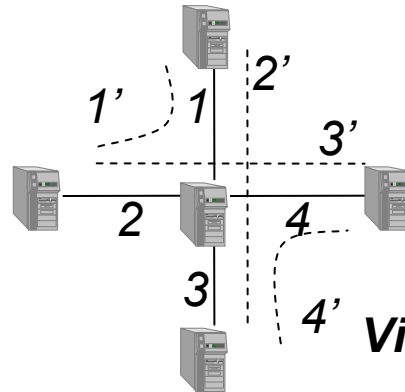
$$G = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

$$\text{Rank}(G)=2$$



$$G = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

$$\text{Rank}(G)=3$$



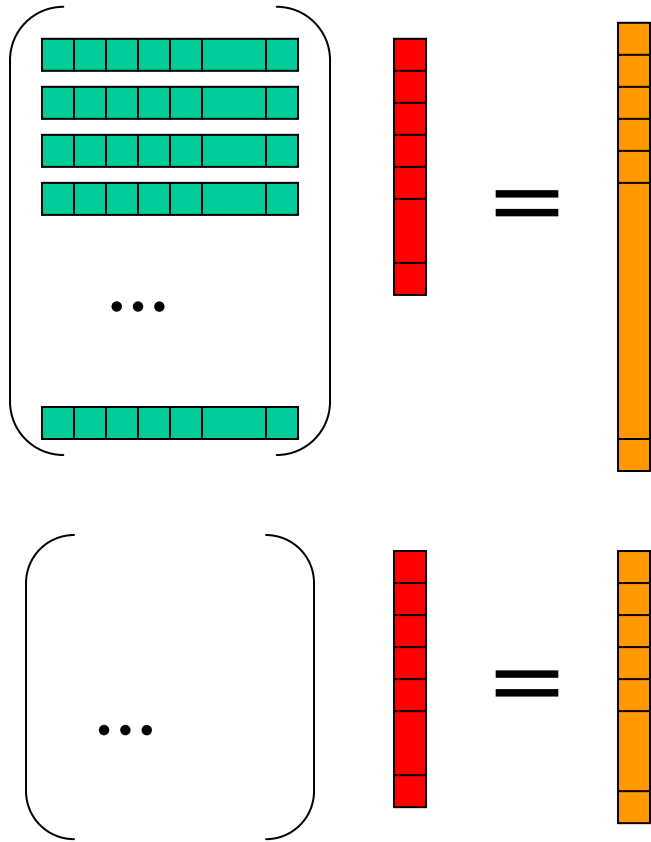
Virtualization

Real links (solid) and all of the overlay paths (dotted) traversing them

Virtual links

Algorithms

$$Gx_G = b$$



- Select $k = \text{rank}(G)$ linearly independent paths to monitor
 - Use QR decomposition
 - Leverage sparse matrix: time $O(rk^2)$ and memory $O(k^2)$
 - E.g., 10 minutes for $n = 350$ ($r = 61075$) and $k = 2958$
- Compute the loss rates of other paths
 - Time $O(k^2)$ and memory $O(k^2)$

How many measurements saved ?

$$k \ll O(n^2) ?$$

For a power-law Internet topology

- When the majority of end hosts are on the overlay

$$k = O(n) \text{ (with proof)}$$

- When a small portion of end hosts are on overlay
 - If Internet a pure hierarchical structure (tree): $k = O(n)$
 - If Internet no hierarchy at all (worst case, clique):
 $k = O(n^2)$
 - Internet has moderate hierarchical structure [TGJ+02]

For reasonably large n , (e.g., 100), $k = O(n \log n)$

(extensive linear regression tests on both synthetic and real topologies)

Practical Issues

- Topology measurement errors tolerance
- Measurement load balancing on end hosts
 - Randomized algorithm
- Adaptive to topology changes
 - Add/remove end hosts and routing changes
 - Efficient algorithms for incrementally update of selected paths

Evaluation

- Extensive Simulations
- Experiments on PlanetLab
 - 51 hosts, each from different organizations
 - $51 \times 50 = 2,550$ paths
 - On average $k = 872$
- Results Highlight
 - Avg real loss rate: 0.023
 - Absolute error mean: 0.0027
90% < 0.014
 - Relative error mean: 1.1
90% < 2.0
 - On average 248 out of 2550 paths have no or incomplete routing information
 - No router aliases resolved

Areas and Domains		# of hosts	
US (40)	.edu	33	
	.org	3	
	.net	2	
	.gov	1	
	.us	1	
International (11)	Europe (6)	France	1
		Sweden	1
		Denmark	1
		Germany	1
		UK	2
	Asia (2)	Taiwan	1
		Hong Kong	1
	Canada		2
	Australia		1

Conclusions

- A tomography-based overlay network monitoring system
 - Given n end hosts, characterize $O(n^2)$ paths with a basis set of $O(n \log n)$ paths
 - Selectively monitor the basis set for their loss rates, then infer the loss rates of all other paths
- Both simulation and PlanetLab experiments show promising results