

Network Loss Inference with Second Order Statistics of End-to-End Flows

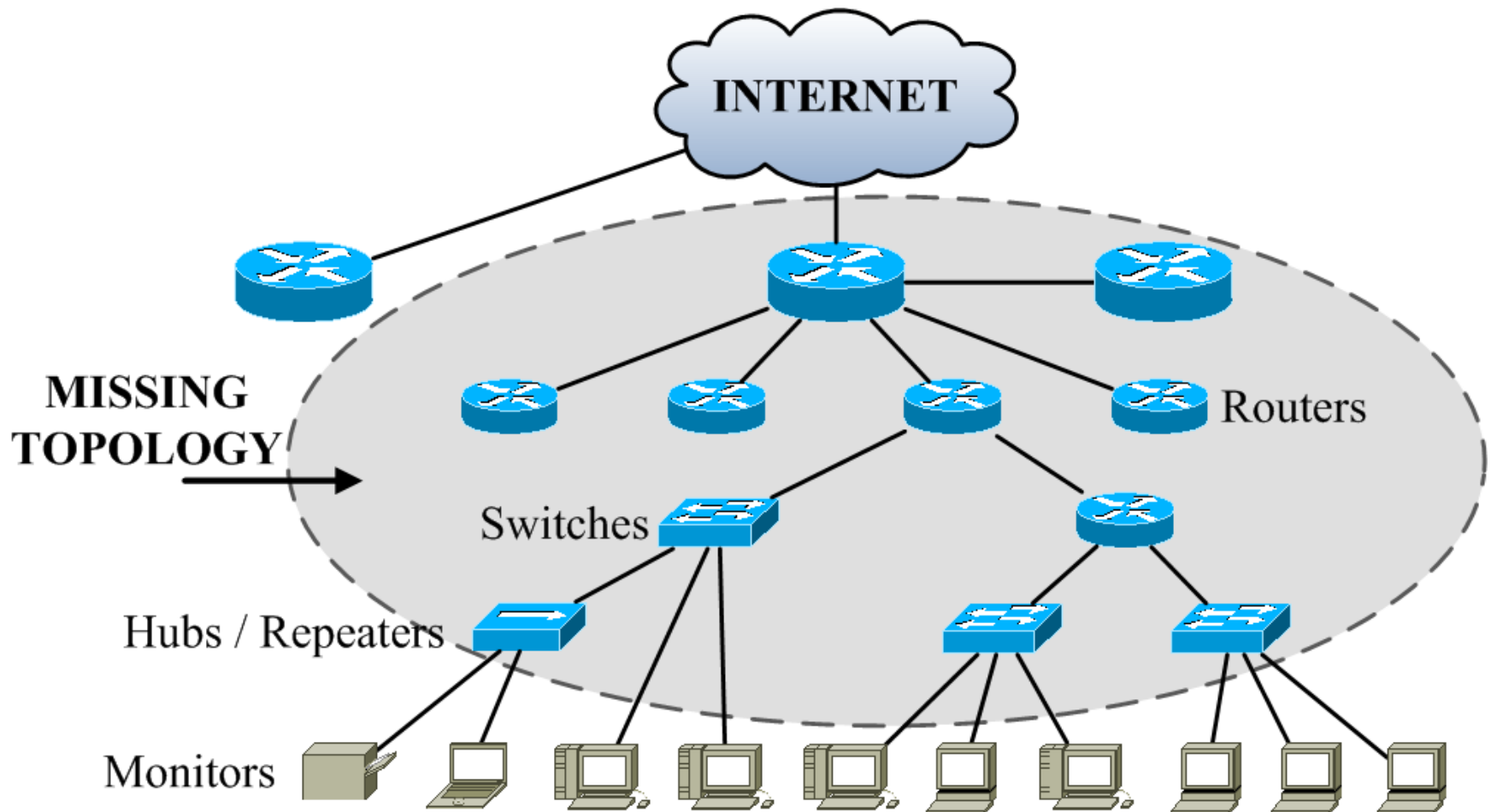
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My INFOCOM 2011 submission



My INFOCOM 2011 submission

- Problem solved
 - Cast the Tree Topology Inference problem → BICA
 - Derive seqBICA and incBICA
 - Better rate of convergence (compare to BLTP)
- Current problems
 - High computation complexity
 - Multicast is limited
 - Missing real-life experiments

Unicast probes

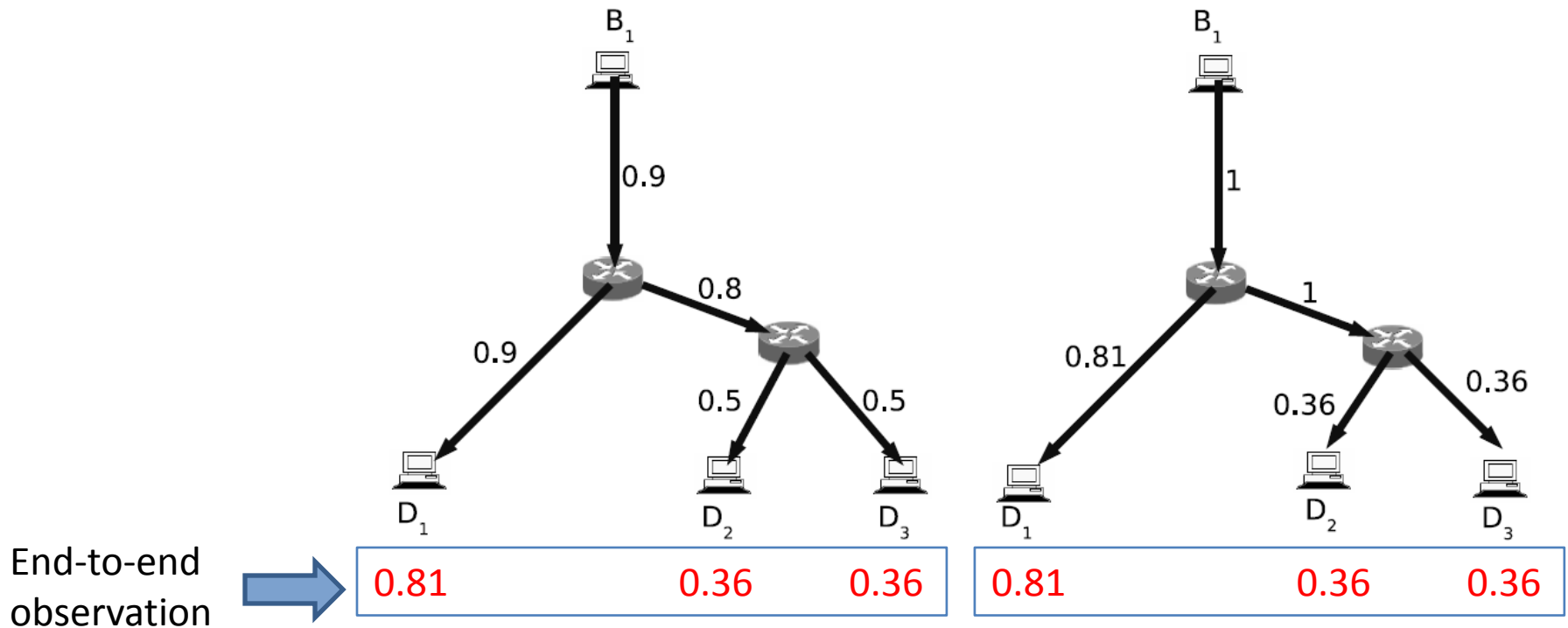
PlanetLab exp

Outline

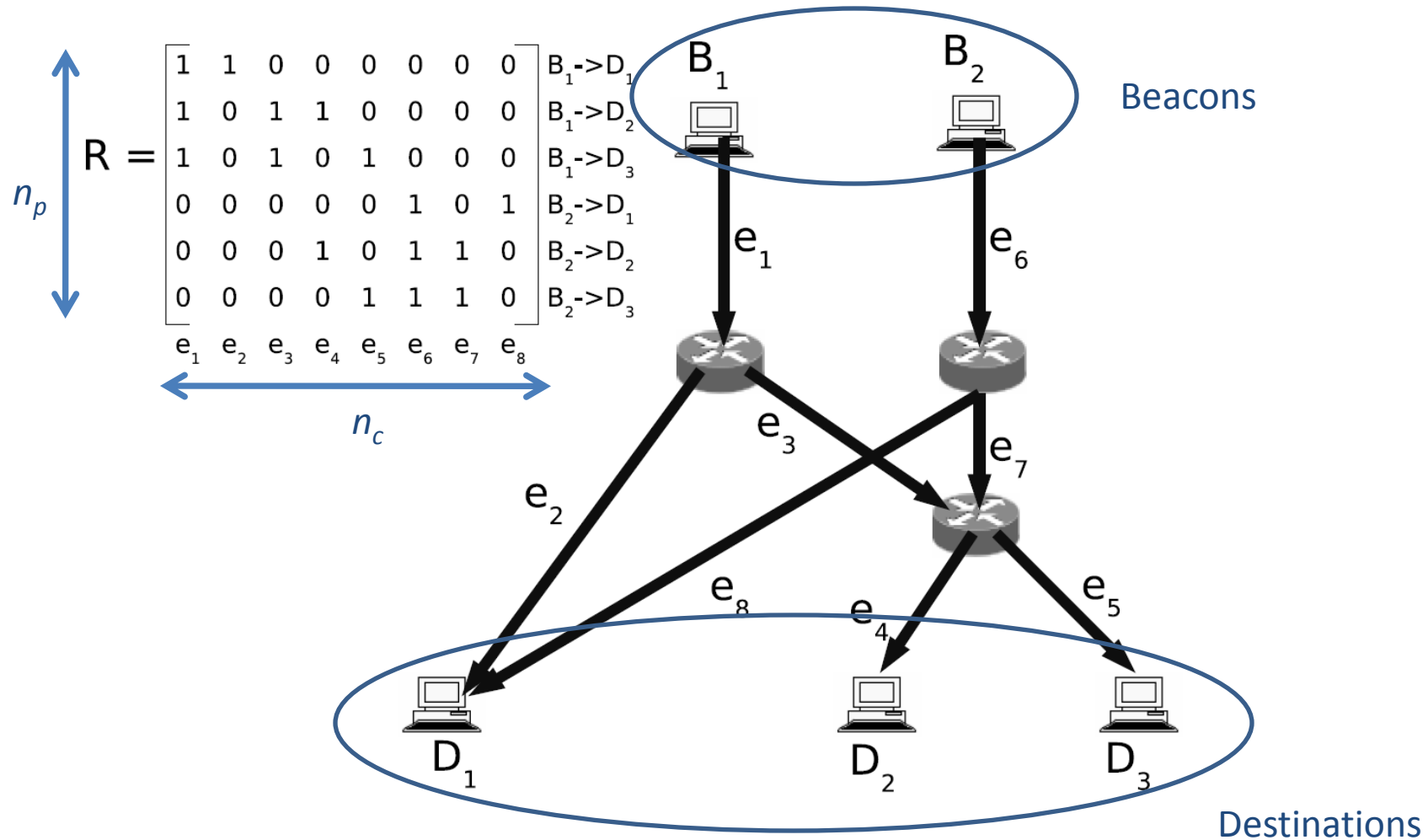
1. Problem definition
2. Network model and assumptions
3. The algorithm
4. Simulation and Experiments
5. Conclusion

1. Problem definition

- Given the Network Topology, compute the link loss rates from end-to-end measurements
- Under-determined problem



2. Network model

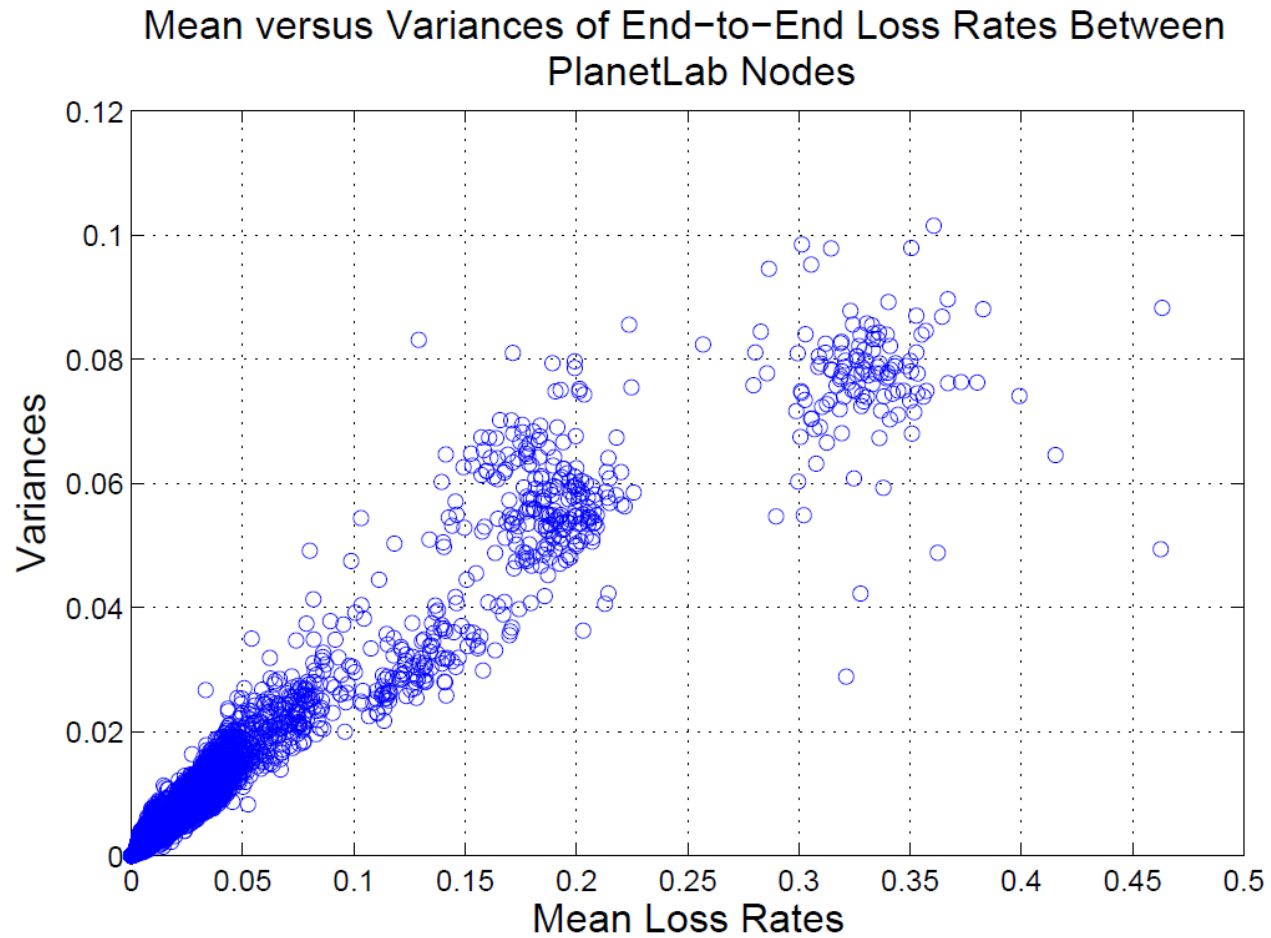


2. Assumptions

- $\hat{\phi}_{e_k}$: transmission rate on link e_k
- $\hat{\phi}_{i,e_k}$: transmission rate of path i on link e_k
- $\hat{\phi}_i$: transmission rate of path i
- $Y_i = \log \hat{\phi}_i$ and $X_k = \log \hat{\phi}_{e_k}$

Network topology	End-to-end probes
<ul style="list-style-type: none">• T.1 Time-invariant routing: R remains unchanged throughout the measurement period.• T.2 No route fluttering: There is no pair of paths P_i and $P_{i'}$ that share two links e_j and $e_{j'}$ without also sharing all the links located in between e_j and $e_{j'}$. That is, the two paths never meet at one link, diverge, and meet again further away at another link.	<ul style="list-style-type: none">• S.1 Identical sampled rates: $\hat{\phi}_{i,e_k} = \hat{\phi}_{e_k}$ almost surely (a.s.) for all paths P_i that traverse e_k.• S.2 Link independence: The random variables X_k are independent.• S.3 Monotonicity of variance: The variance v_k of X_k is a non-decreasing function of $1 - \phi_{e_k}$.

2. Assumptions



2. Problem formulation

- Let

$$\mathbf{Y} = [Y_1 \ Y_2 \ \dots \ Y_{n_p}]^T \text{ and } \mathbf{X} = [X_1 \ X_2 \ \dots \ X_{n_c}]^T$$

- Now we need to solve

$$\mathbf{Y} = R\mathbf{X}.$$

- Unfortunately, R is (always) rank deficient

3. The algorithm

- Define the covariance matrices of \mathbf{X} and \mathbf{Y}

$$\Gamma_{\mathbf{X}} = \text{diag}(\mathbf{v}) = \text{diag}([v_1 \ v_2 \ \dots \ v_{n_c}])$$

$$\Sigma = \begin{bmatrix} \sigma_{Y_1}^2 & \text{COV}[Y_1, Y_2] & \dots & \text{COV}[Y_1, Y_{n_p}] \\ \text{COV}[Y_2, Y_1] & \sigma_{Y_2}^2 & \dots & \text{COV}[Y_2, Y_{n_p}] \\ \vdots & \dots & \ddots & \vdots \\ \text{COV}[Y_{n_p}, Y_1] & \text{COV}[Y_{n_p}, Y_2] & \dots & \sigma_{Y_{n_p}}^2 \end{bmatrix}$$

- From $\mathbf{Y} = R\mathbf{X}$. we have $\Sigma = R\Gamma_{\mathbf{X}}R^T = R\text{diag}(\mathbf{v})R^T$
- Define augmented matrix \mathbf{A}

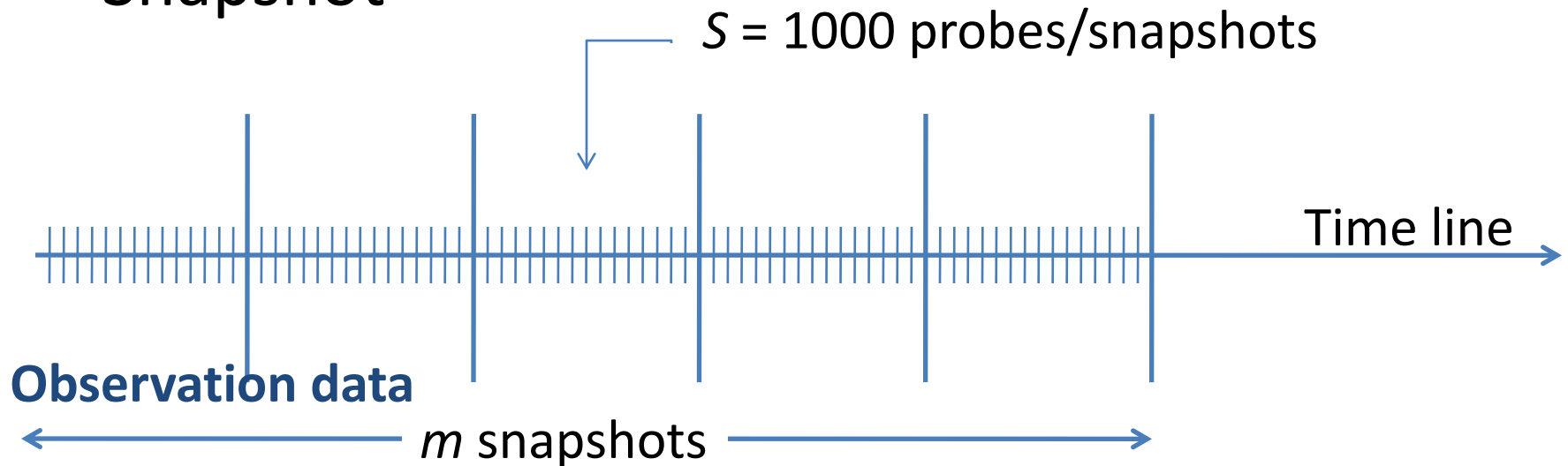
DEFINITION 1. Let \mathbf{A} be the augmented matrix of dimension $n_p(n_p + 1)/2 \times n_c$ whose rows consist of the rows of R and the component wise products of each pair of different rows from R . The rows of \mathbf{A} are arranged as follows: $\mathbf{A}_{((i-1) \times n_p + (j-i)+1)*} = \mathbf{R}_{i*} \otimes \mathbf{R}_{j*}$ for all $1 \leq i \leq j \leq n_p$.

$$R = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \end{bmatrix}, \quad \mathbf{A} = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 \end{bmatrix}$$

3. The algorithm

- LEMMA 1. *The equations $\Sigma = R \text{diag}(\mathbf{v}) R^T$ are equivalent to the equations $\Sigma^* = A \mathbf{v}$, where Σ^* is a vector of length $n_p(n_p + 1)/2$ and $\Sigma_{(i-1)n_p+j-i+1}^* = \Sigma_{i,j}$ for all $1 \leq i \leq j \leq n_p$.*

- Snapshot



3. The algorithm

- From m previous snapshots, calculate

$$\hat{\Sigma}_{ii'} = \frac{1}{m-1} \sum_{l=1}^m Y_i^{(l)} Y_{i'}^{(l)} - \bar{Y}_i^{(l)} \bar{Y}_{i'}^{(l)}, 1 \leq i \leq i' \leq n_p$$

- Then we can calculate $\hat{\Sigma}^*$ and solve

$$\hat{\Sigma}^* = A\mathbf{v}$$

- Eliminate good links:

- Sort links by their variances $v_1 \leq v_2 \leq \dots \leq v_{n_c}$
- Remove columns in R and entries in \mathbf{X} until R is full rank, we now have R^* and \mathbf{X}^*

- Finally solve $\mathbf{Y} = R^* \mathbf{X}^*$

3. The algorithm

Input: The reduced routing matrix R and $m + 1$ snapshots: $\mathcal{Y} = \{\mathbf{y}^1, \mathbf{y}^2, \dots, \mathbf{y}^m, \mathbf{y}^{m+1}\}$.

Phase 1 (Learning the link variances):

Solve (8) with the first m snapshots $\{\mathbf{y}^1, \mathbf{y}^2, \dots, \mathbf{y}^m\}$ to find v_k for all links $e_k \in \mathcal{E}_c$.

Phase 2 (Inferring link loss rates):

Step 1. Sort v_k in increasing order.

Step 2. Initialize $R^* = R$.

(**Loop**) While R^* is not of full column rank
remove R_{1*}^* from R^* .

Step 3. Solve (9) for snapshot $(m + 1)$ th.

Approximate $\phi_{e_k} \approx 1$ for all links e_k
whose columns were removed from R .

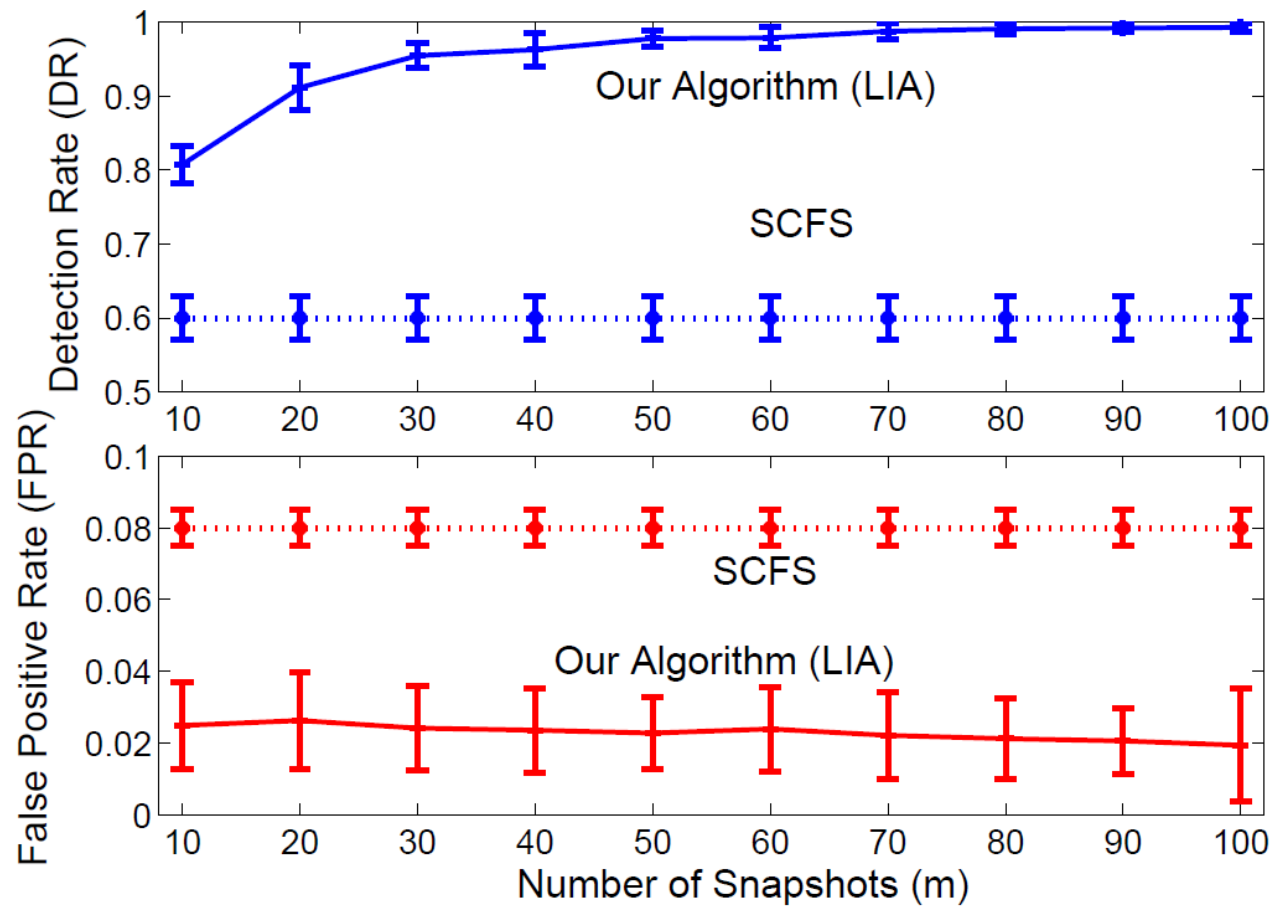
Output: Link transmission rates $\phi = [\phi_1 \ \phi_2 \ \dots \ \phi_{n_c}]^T$ of the $(m + 1)$ th snapshot.

Loss Inference Algorithm (LIA)

4. Simulation

- Matlab implementation
- Congested links = 10%
- Loss rate
 - Good links: $[0, 0.002]$
 - Congested links: $[0.002, 0.2]$
- Bernoulli/Gilbert packet loss model
- $S = 1000$ probes for each snapshot
- Performance metrics
 - Detection rate
 - False positive rate

4. Simulation

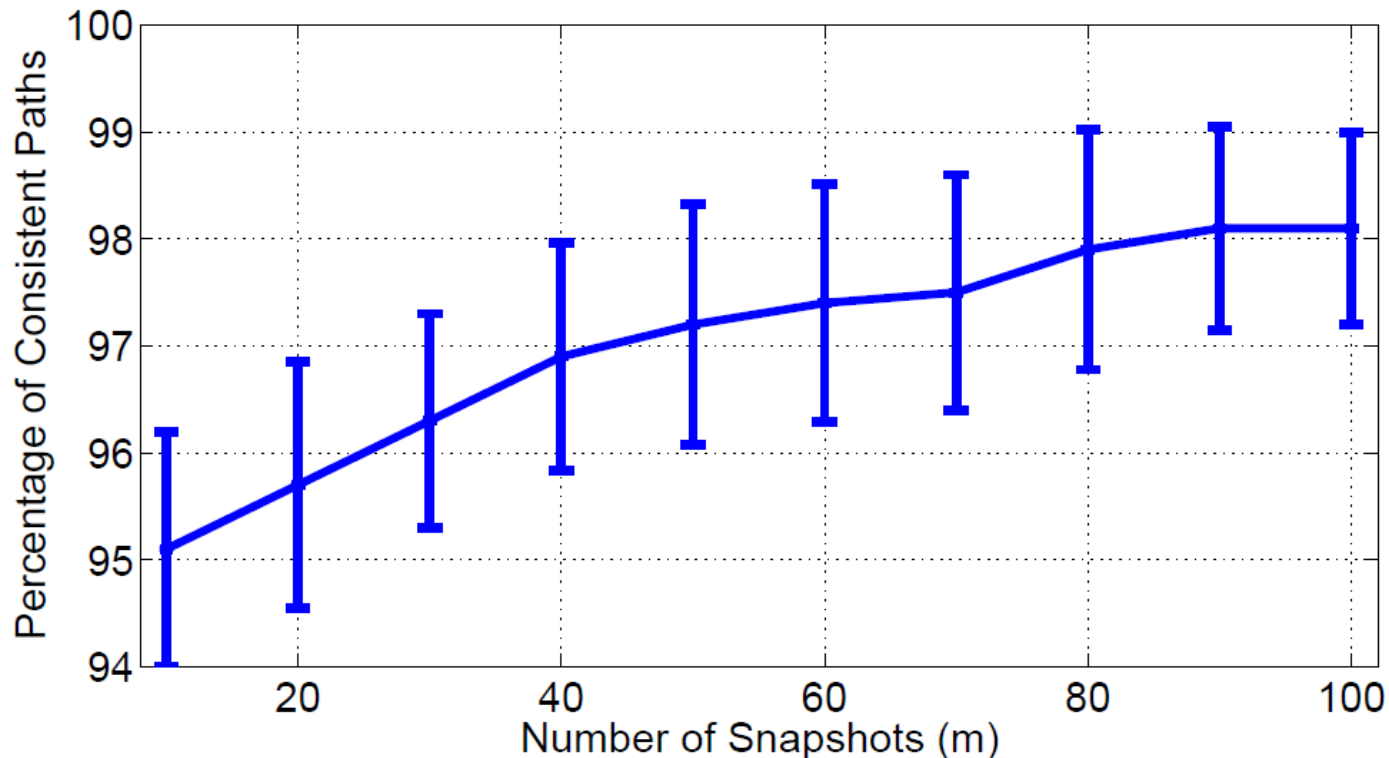


4. Internet experiment

- PlanetLab network with 716 hosts (only 381 hosts cooperate) on April 20, 2007
- Use Traceroute to measure network topology (potential noise in R)
- Remove fluttering paths from R
- Probe = UDP packet (40 bytes)
- Avoid creating probe congestion: probing rate = 100KB/sec
- Cross validation method

4. Internet experiment

- A path is consistent if: $\left| \hat{\phi}_i - \prod_{e_k \in P_i \cap \mathcal{E}_{\text{inf}}} \phi_{e_k} \right| \leq \epsilon, \quad \epsilon = 0.005.$



5. Conclusion

- Link Lost Rate Inference by using second-order statistics
- What can be applied to our research?
 - Link delay inference
 - PlanetLab experiments methodology
 - Covariance exploitation