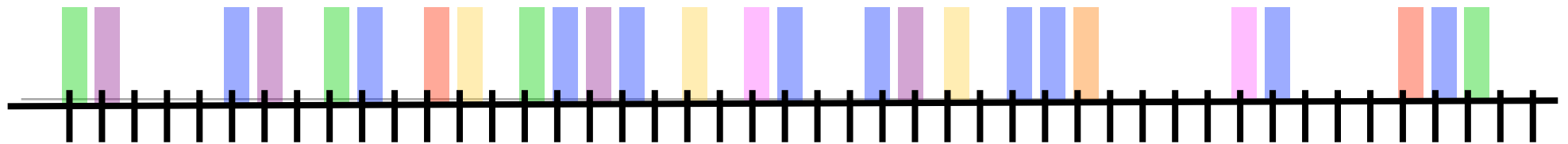


# Accurate and Efficient SLA Compliance Monitoring



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# Motivation

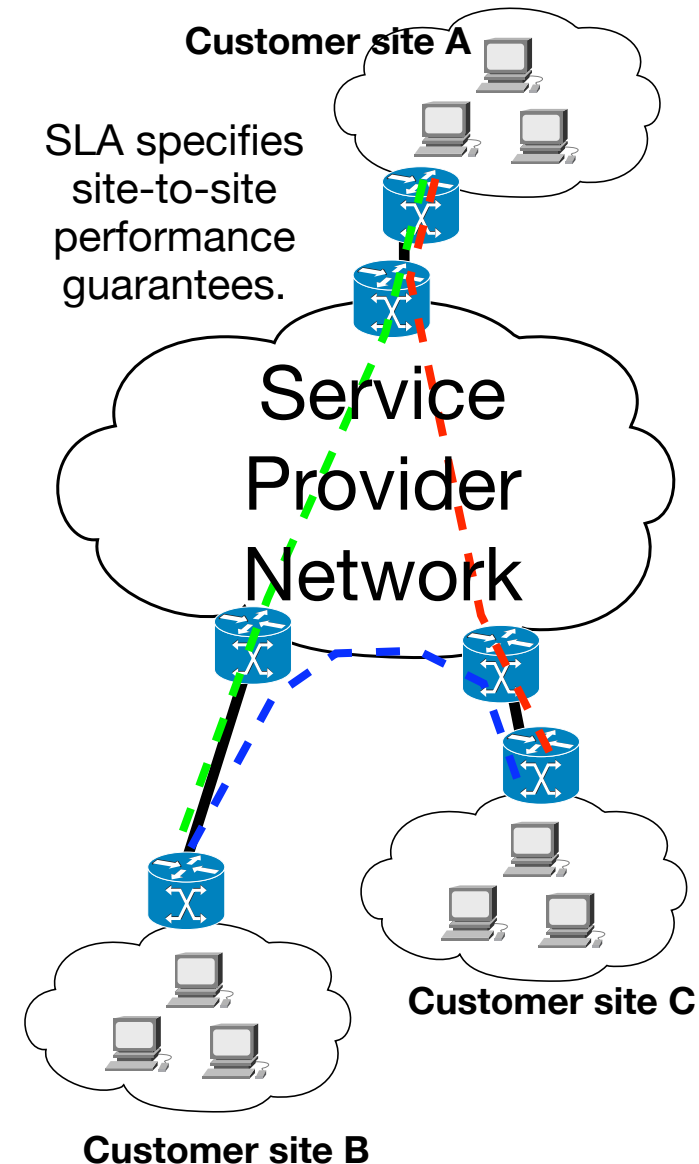
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- Service level agreements (SLAs) specify performance guarantees made by Internet service providers.
  - Example metrics: packet loss, delay, delay variation.
- Accurate and robust SLA compliance monitoring is important for service providers and their customers.
  - Lightweight, effective monitoring is a key challenge.
    - Measurement on a single path.
    - Network-wide monitoring.
  - Non-compliance can have serious consequences!



# Service Level Agreements

- Performance guarantees made by providers to customers.
  - *E.g.*, customer buys VPN service, wants guarantee of good service.
  - Metrics: packet loss, delay, delay variation, network availability.
  - Specific to origin-destination sites (*e.g.*, delay between A and C versus B and C).
  - Different statistics used, *e.g.*, mean, 95<sup>th</sup> percentile, maximum.
  - Metrics typically averaged over long time periods.





# SLA Monitoring Challenges

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- Overhead of simultaneous active measurement of multiple metrics is problematic.
- In-network characteristics are difficult to accurately measure with packet probes.
- Coordination and overhead of network-wide measurements.
- Data management.
  - Collection, processing, storing and archiving, coping with measurement errors, filtering outliers, ...



# Approach

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- **Multi-objective probing**: simultaneous measurement of multiple performance objectives.
  - Reduce overhead, simplify the measurement process.
- **New and more accurate/robust active methodologies** for measuring delay, loss, and delay variation
  - A new methodology for estimating **mean end-to-end delay**.
    - Based on Simpson's method for numerical integration.
  - A new methodology for estimating **quantiles of the delay distribution**.
    - No assumptions made about nature of the underlying distribution.



# Approach (2)

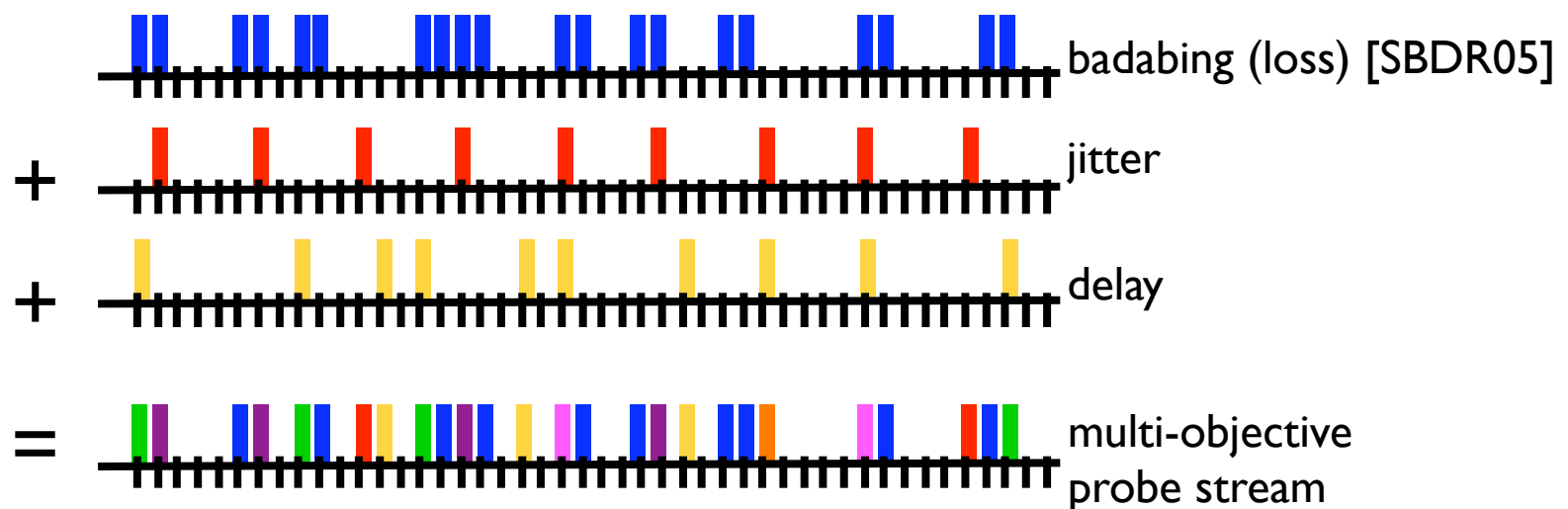
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- A new methodology for **inferring an upper bound on the distribution of delay** for an *unmeasured* path, given measurements for other, related paths in the network.
  - Extends algebraic approaches of prior work to *distributions*.
- A new heuristic for measurement of **packet loss rate** based on the badabing probe process [SBDR05].
  - Badabing originally designed to measure aspects of congestion episodes, not loss rate.
- A new methodology for more **robust measurement of delay variation (jitter)** on an end-to-end path.
  - A qualitative assessment of congestion, analogous to RTP.



# Multi-objective probing

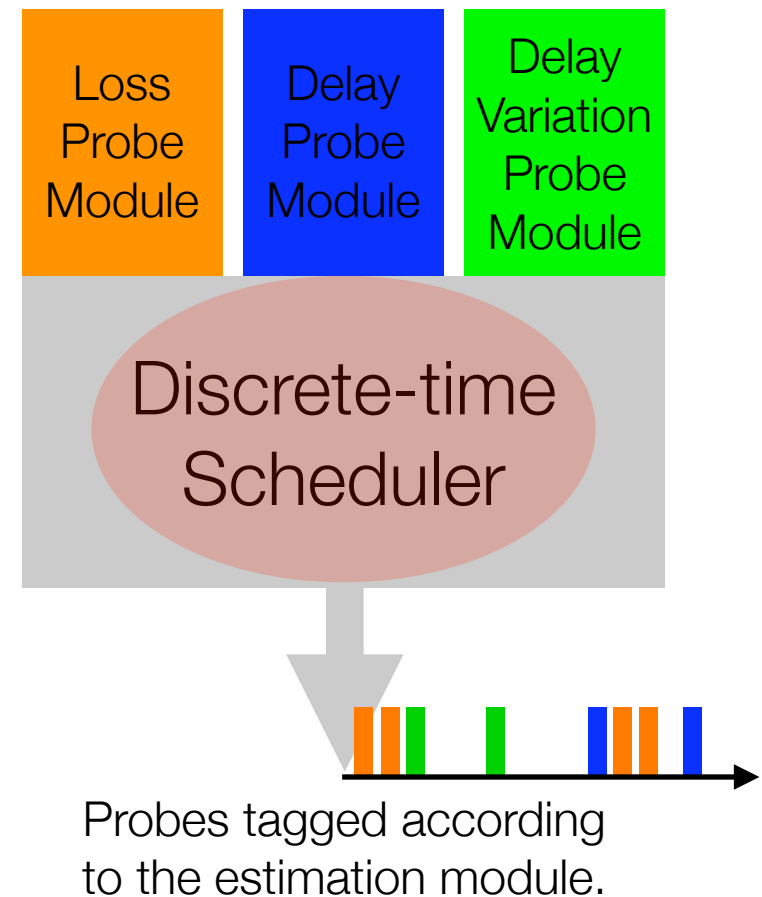
- Assume estimation algorithms operate in discrete time.
  - Probes may be scheduled to be sent at same time slot.
  - Tag probes according to the estimator module to which they apply.





# Implementation

- Discrete-time scheduler core.
  - Modular probe algorithms register with scheduler.
  - Module contains all logic to implement specific measurement algorithm.
  - Modules receive callbacks from scheduler, send probes through scheduler.







# Methodology: Mean Delay

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- Model delay as a continuous function  $f(t)$ .
- Simpson's method for numerical integration is a natural approach for estimating the mean of  $f(t)$ .
  - $a, b$  are the endpoints, and  $c$  is the midpoint of interval  $l_j$ .
$$\frac{1}{6}(f(a_j) + f(b_j) + 4f(c_j)) + e_j$$
- At time slot  $i$ , choose value  $k$  from geometric distribution with parameter  $p$ .
  - Send probes at time slot  $i, i+(k+1), i+2(k+1)$ .
  - Apply Simpson's method to measured probe delays.



# Methodology: Delay Quantiles

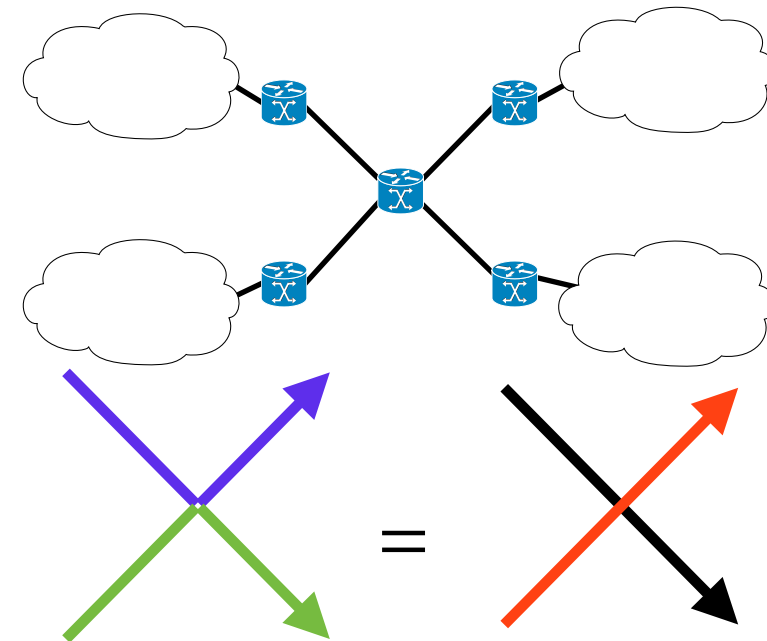
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- Estimate quantiles using delay samples from probes.
- Let  $\{x_i : 1, \dots, n\}$  be  $n$  samples drawn from distribution  $F$ .
  - Let  $Q_p$  denote the  $p^{\text{th}}$  quantile, the solution to  $F(Q_p) = p$ .
- $X_k \leq x$  is the event that at least  $k$  samples are less than or equal to  $x$ ;  $\Pr[X_k \leq Q_p] = G(n, p, k)$ .
- Want: level  $X^+(n, p, \epsilon)$  such that  $Q_p$  is guaranteed to exceed it with some small probability  $\epsilon$ .
  - Use  $K^+(n, p, \epsilon)$ , the  $1 - \epsilon^{\text{th}}$  quantile of the binomial  $B_{n, p}$  distribution.
  - Similar formulation for the lower bound  $K^-(n, p, \epsilon)$ .
  - Bounds can be calculated exactly using binomial distribution.



# Methodology: Distribution Inference

- Consider scalar additive metrics (e.g., delay, log transmission probability)
- Given a subset of performance measures across intersecting paths, is it possible to infer the whole set of measures?
  - Chen *et al.* (SIGCOMM 2004) and Chua *et al.* (INFOCOM 2005) examined problem for *scalar* measures.
- What about inferring a distribution of performance measures from a subset?

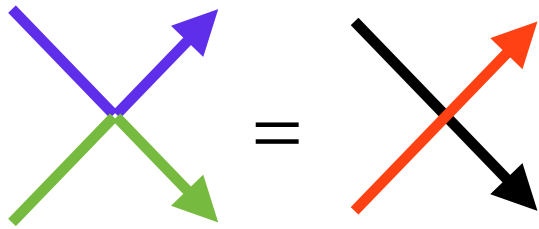


$$y_1 + y_4 = y_2 + y_3$$

(e.g.,  $y_i$ 's are measured mean delays.)



# Methodology: Distribution Inference



$Y_1 + Y_4 = Y_2 + Y_3$   
(e.g.,  $Y_i$ 's are distributions  
of one-way delays.)

- R is the set of routes forming routing matrix A.
- There is a minimal set of paths  $S \subsetneq R$  s.t. every row of A can be expressed as a linear combination of S.
  - Partition S in  $S^-$  and  $S^+$  based on sign of coefficient in the linear combination:  $Y_1 = Y_2 + Y_3 - Y_4$ ;  $S_1^+ = \{2,3\}$ ;  $S_1^- = \{4\}$ 
    - Can formulate the convolution problem in terms of these partitions.
    - The distributions are discretized prior to convolution.
      - Our results provide a lower bound on the quantiles (upper bound on CDF).



# Methodology: Loss Rate

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- Start with badabing loss probe stream for measuring *frequency* and *duration* characteristics of loss episodes.
  - Probe pairs, sent according to a geometric distribution.
    - Each probe consists of three packets, sent back-to-back.
- Heuristic: loss rate measured by badabing during a loss episode is related to what a typical TCP flow might measure.

$$\hat{L} = \hat{F}\hat{l}$$



# Methodology: Delay Variation

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- Consider a stream of probes of length  $k$ .
  - $s_{i,j}$  denotes difference in **send** time between probes  $i,j$
  - $r_{i,j}$  denotes difference in **receive** time between probes  $i,j$
- Construct a matrix  $M$  where element  $i,j$  contains the ratio  $s_{i,j}/r_{i,j}$ :
  - $s_{i,j}/r_{i,j} = 1$  if spacing does not change.
  - $s_{i,j}/r_{i,j} > 1$  if spacing increases.
  - $s_{i,j}/r_{i,j} < 1$  if spacing decreases.
  - $s_{i,j}/r_{i,j} = 0$  if either probe is lost.



# Methodology: Delay Variation

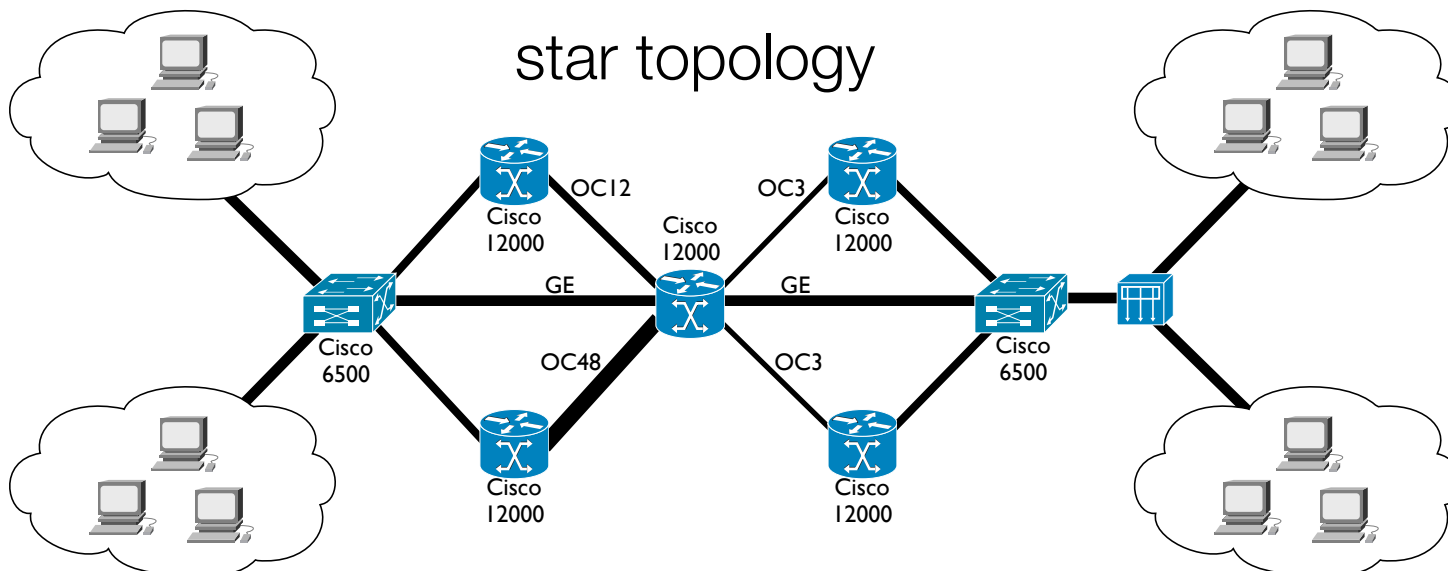
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- Compute eigenvalues of matrix  $M$ .
  - Results in vector  $e$  of eigenvalues, sorted large to small.
  - If all ratios are 1, largest eigenvalue is  $k$  (stream length).
    - Denote this “expected” vector of eigenvalues as  $e'$ .
  - Subtract  $e'$  from  $e$ , taking the  $L_1$  norm of the resulting vector.
$$\sum_{i=1}^k |e_i - e'_i|.$$
- Result is called **DV matrix metric**.
  - A qualitative assessment of the amount of distortion from what we expect.



# Experiments

- Created tool called SLAm (SLA monitor).
- Evaluated in controlled laboratory environment.
  - Two topologies: dumbbell and star.
- Compare SLAm with RFC standard probe streams at same bitrate.







# Results: Bandwidth Savings

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- Three probe algorithms operating simultaneously.
  - 5 millisecond discrete time interval.
  - Loss probe:  $p_{loss} = 0.3$ , 600 byte packets.
  - Delay probe:  $p_{delay} = 0.048$ , 100 byte packets.
  - Delay variation periodic probe: 30 millisecond interval, 48 byte packets.
- Savings is parameter dependent, and can be big.

Loss	Delay	Delay Variation	Sum	SLAm	Savings
489 Kb/s	20 Kb/s	60 Kb/s	569 Kb/s	470 Kb/s	99 Kb/s (17%)



# Results: Delay

- Results for SLAm are closer to true value than standard Poisson-based stream (RFC 2679).
- Fast convergence to true mean delay (in paper).

mean delay comparison	SLAm		RFC 2679	
	true	estimate	true	estimate
dumbbell (60%)	0.006	0.006	0.007	0.009
dumbbell (75%)	0.014	0.014	0.006	0.013
star: route 1	0.007	0.006	0.007	0.005
star: route 2	0.009	0.008	0.009	0.006
star: route 3	0.005	0.005	0.005	0.004
star: route 4	0.007	0.006	0.007	0.004

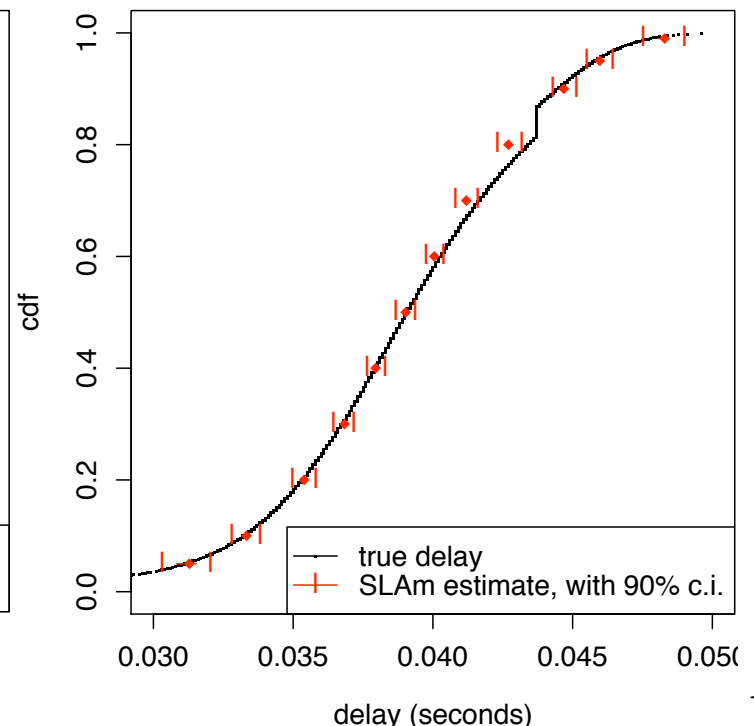
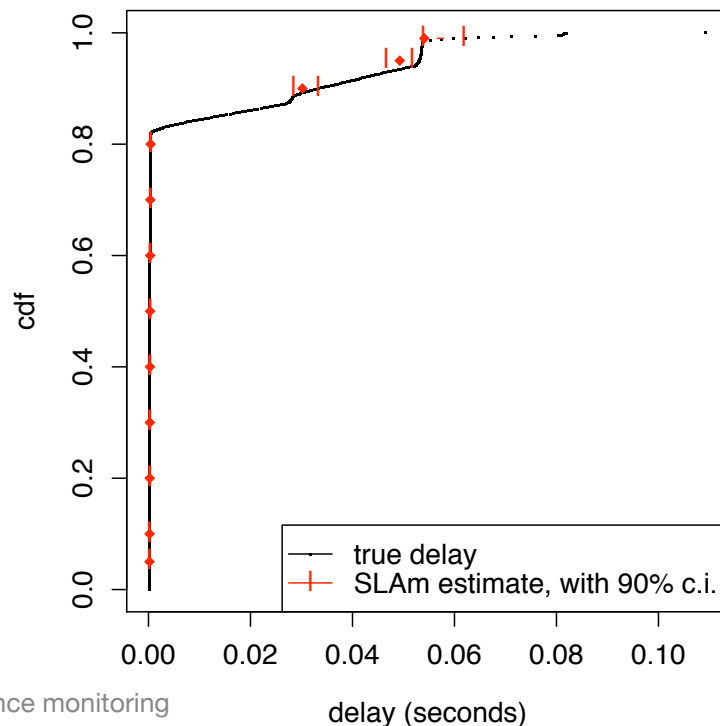
Results for self-similar background traffic generated using Harpoon.



# Results: Delay Quantiles

- Calculated quantiles with 90% confidence interval.
- Intervals generally include true quantile, with few exceptions.
  - For all traffic scenarios used, in both dumbbell and star topologies.

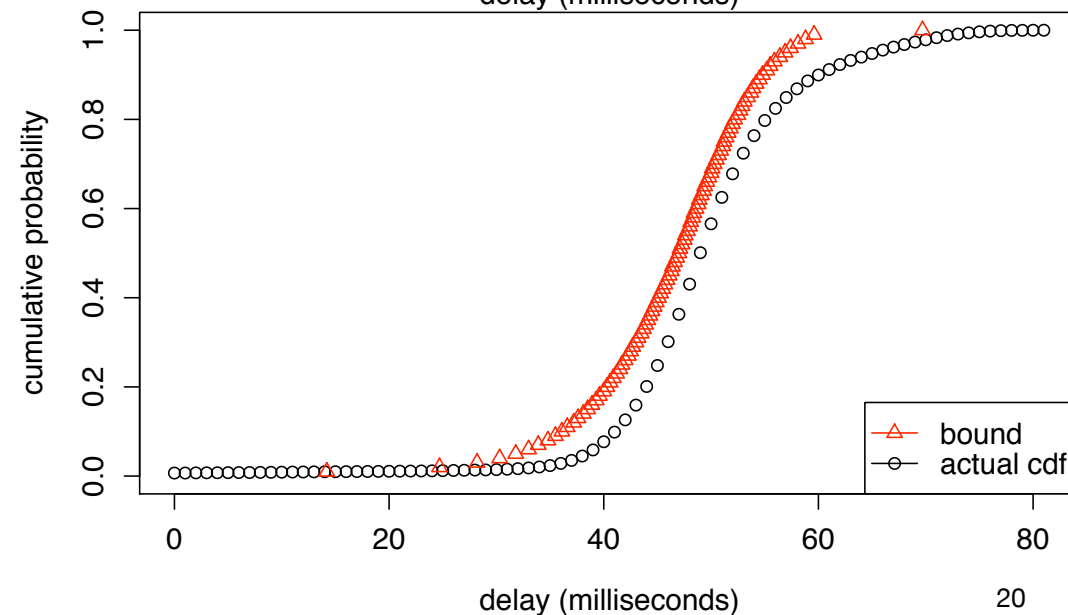
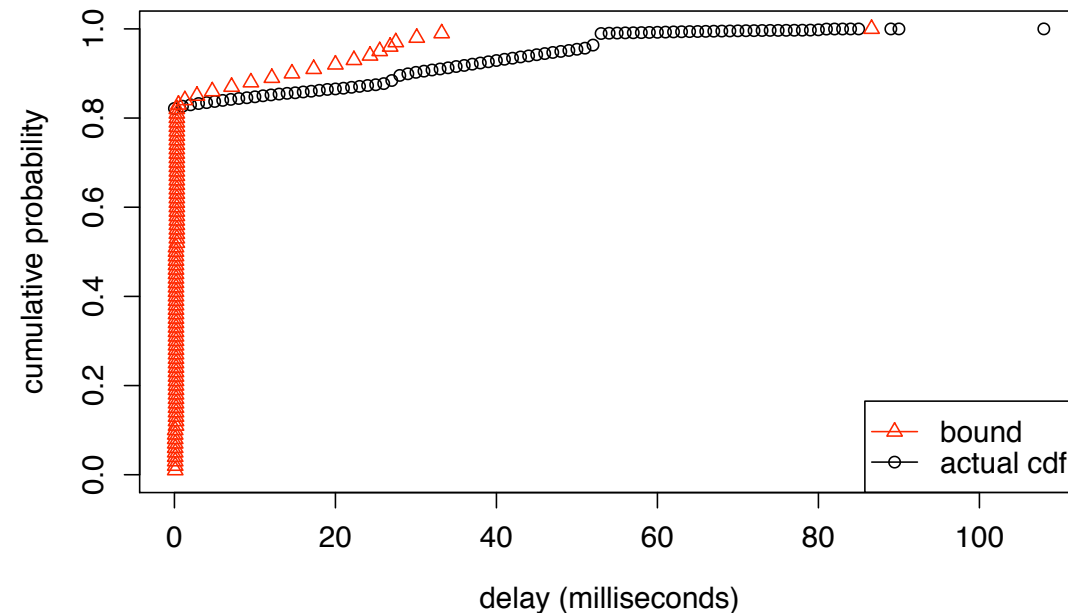
Results for CBR in star topology (left) and long-lived TCP in dumbbell topology (right).





# Results: Delay Distribution Inference

- Inferred distributions are close to the true ones.
  - Discretization of 100 microseconds for convolution.
- Results shown for UDP CBR traffic scenario (top) and self-similar traffic scenario (bottom).





# Results: Loss Rate

- Loss rate estimates are much more accurate than standard Poisson-based stream.
- Fast convergence to true loss rate (in paper).

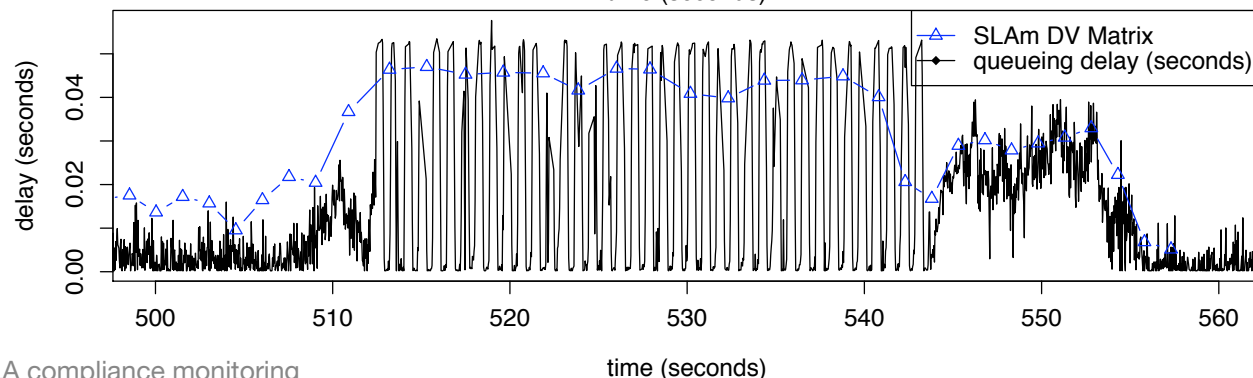
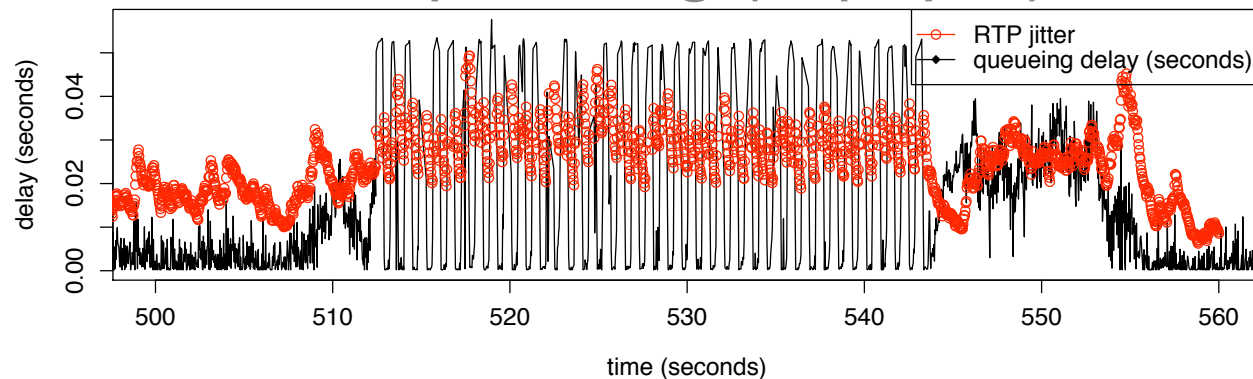
loss rate comparison	SLAm		RFC 2680	
	true	estimate	true	estimate
dumbbell (60%)	0.0008	0.0007	0.0017	0
dumbbell (75%)	0.0049	0.0050	0.0055	0
star: route 1	0.0170	0.0205	0.0289	0.0058
star: route 2	0.0008	0.0006	0.0069	0.0000
star route 3	0.0192	0.0178	0.0219	0.0036
star: route 4	0.0005	0.0006	0.0002	0.0000

Results for self-similar background traffic generated using Harpoon.



# Results: Delay Variation

- SLAm DV matrix metric is more robust than RTP.
  - More accurately tracks congested and turbulent conditions.
  - Also robust in two-hop setting (in paper).





# Summary

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- A set of new methodologies for accurate, lightweight SLA compliance monitoring.
  - Multi-objective probing: reduces overhead.
  - Delay: accurate estimates of mean and quantiles; inferred distributions are close to true distributions.
  - Loss rate: accurate heuristic based on badabing probes.
  - Delay variation: robust qualitative estimate of congestion.
- Methodologies implemented in a tool called SLAm.
  - Laboratory tests with one- and two-hop topologies.
  - Source code will be released soon.



# The end

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- Ongoing and future work:
  - Probe stream coordination in the network-wide setting based on knowledge of topology.
  - How to optimize for accuracy given a daily (or hourly, etc.) probe budget?
  - SLA compliance monitoring does not require perfect accuracy; what appropriate tradeoffs be made between “good enough” accuracy and overhead?