

# Multi-Hop Probing Asymptotics In Available Bandwidth Estimation: Stochastic Analysis

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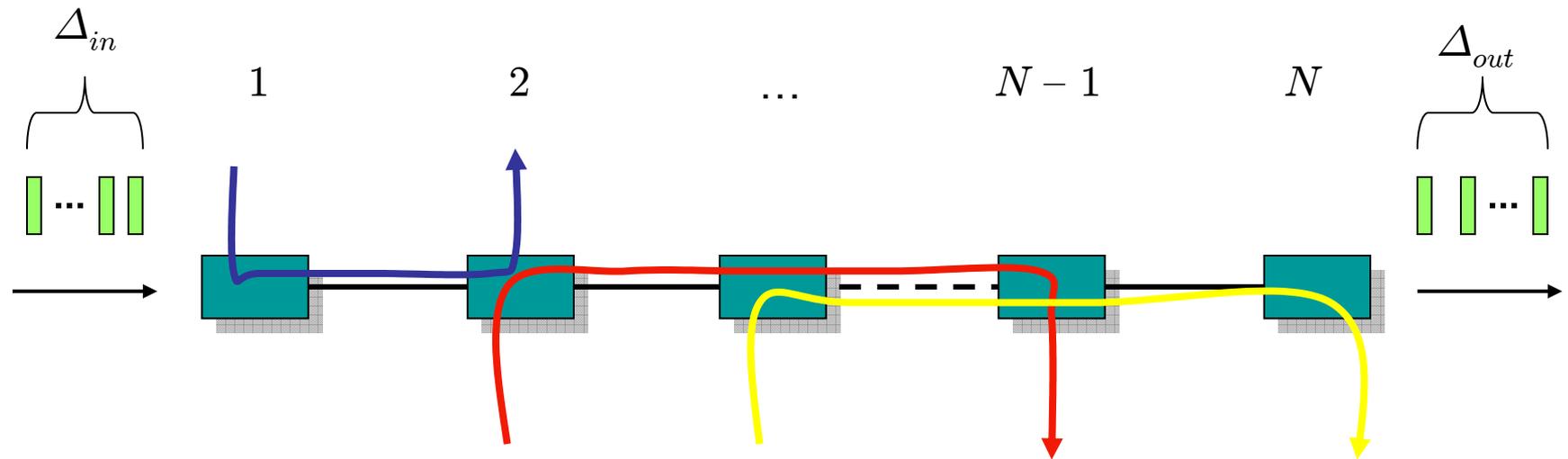
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# Problem Statement

- Bandwidth estimation using **first-order** statistics of packet-train output dispersions
  - Assume an  $N$ -hop path probed by packet trains of length  $n$



- **Goal:** derive the relationship between the statistical mean of  $\Delta_{out}$  and  $\Delta_{in}$  under arbitrary cross-traffic
  - We call this the **probing response curve**

# Outline

## Related work and background:

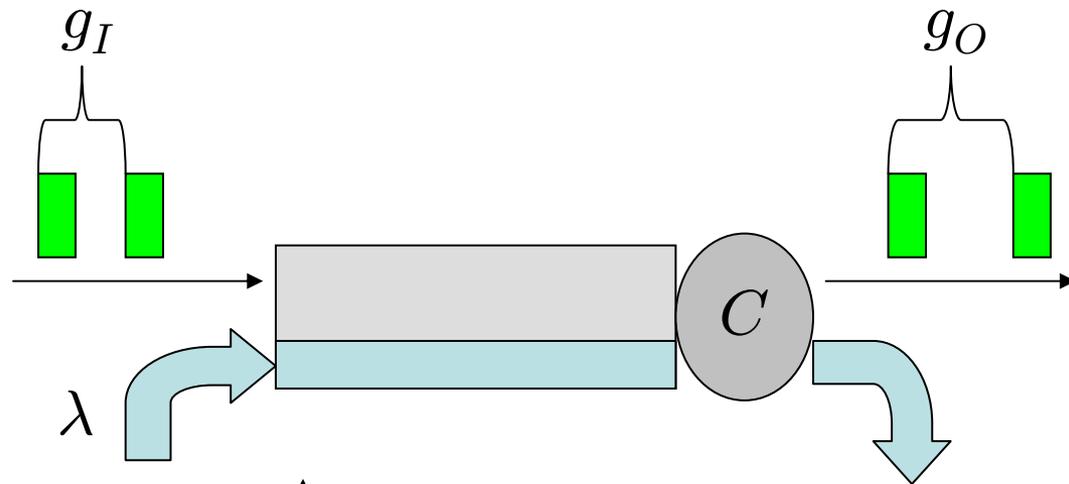
- Single-hop fluid curve
  - E.g., Melander (2001), Dovrolis (2001)
- Multi-hop fluid curve (one-hop persistent)
  - E.g., Dovrolis (2001)

## Our contribution:

- Multi-hop **fluid** response curve
  - Arbitrary cross-traffic routing
- Multi-hop **stochastic** response curve
  - Packet-level model of cross-traffic
- Experimental verification
- Implications on existing techniques

# Related Work

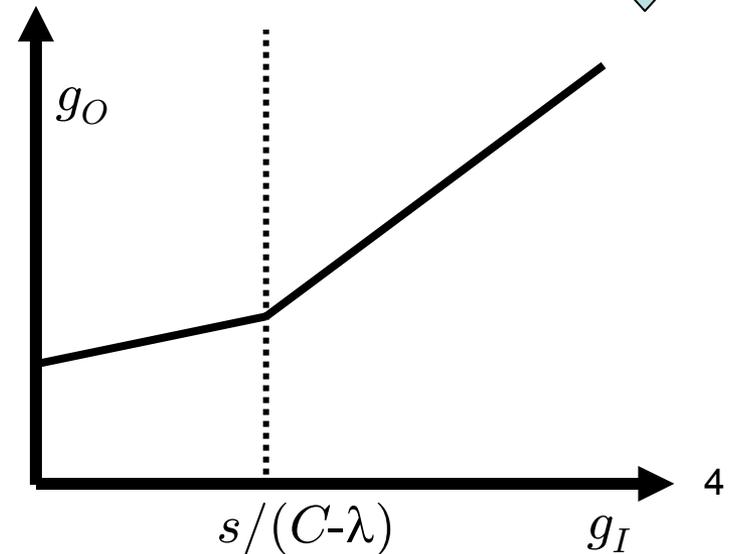
- Single-hop fluid setting:



- Response curve:

$$g_O = \begin{cases} \frac{s}{C} + \frac{g_I \lambda}{C} & g_I \leq \frac{s}{C-\lambda} \\ g_I & g_I \geq \frac{s}{C-\lambda} \end{cases}$$

$$= \max \left( g_I, \frac{s + g_I \lambda}{C} \right).$$



## Related Work (cont'd)

Let  $r_I = \frac{s}{g_I}$  and  $r_O = \frac{s}{g_O}$ , the fluid rate curve is

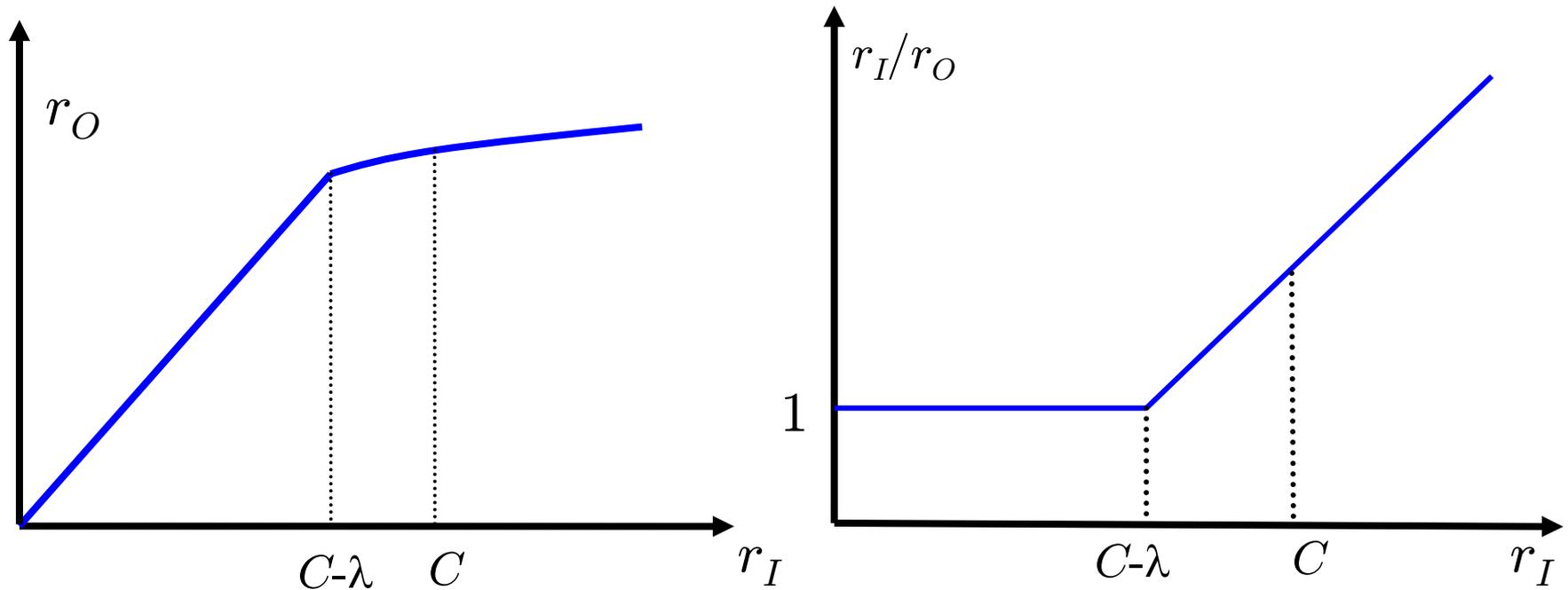
$$r_O = \begin{cases} r_I & r_I \leq C - \lambda \\ \frac{r_I}{r_I + \lambda} C & r_I \geq C - \lambda \end{cases} \quad (1)$$

A transformed version of rate curve depicts the relation between  $\frac{r_I}{r_O}$  and  $r_I$

$$\frac{r_I}{r_O} = \begin{cases} 1 & r_I \leq C - \lambda \\ \frac{r_I + \lambda}{C} & r_I \geq C - \lambda \end{cases} \quad (2)$$

# Related Work (cont'd)

- Rate-response single-hop fluid curves:



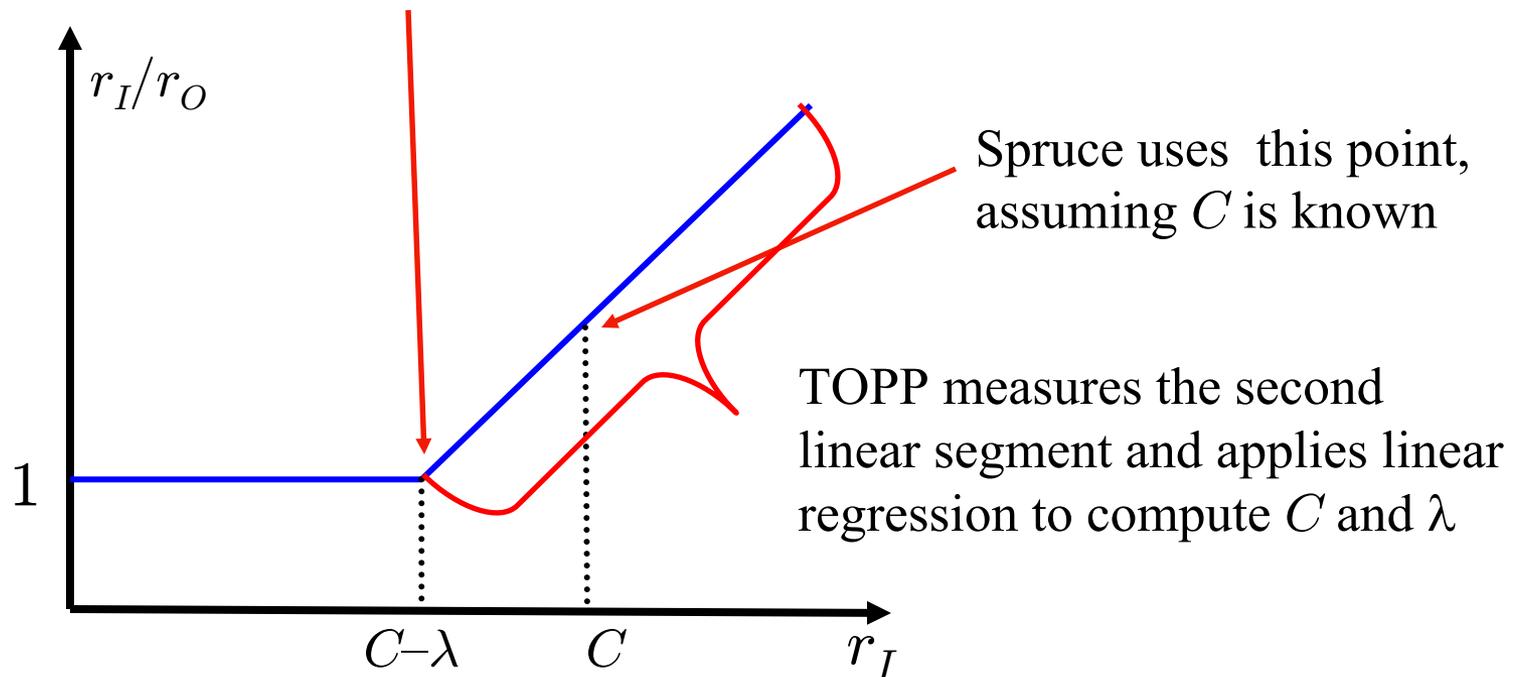
$$r_O = \min \left( r_I, \frac{r_I}{r_I + \lambda} C \right).$$

$$\frac{r_I}{r_O} = \max \left( 1, \frac{r_I + \lambda}{C} \right).$$

# Related Work (cont'd)

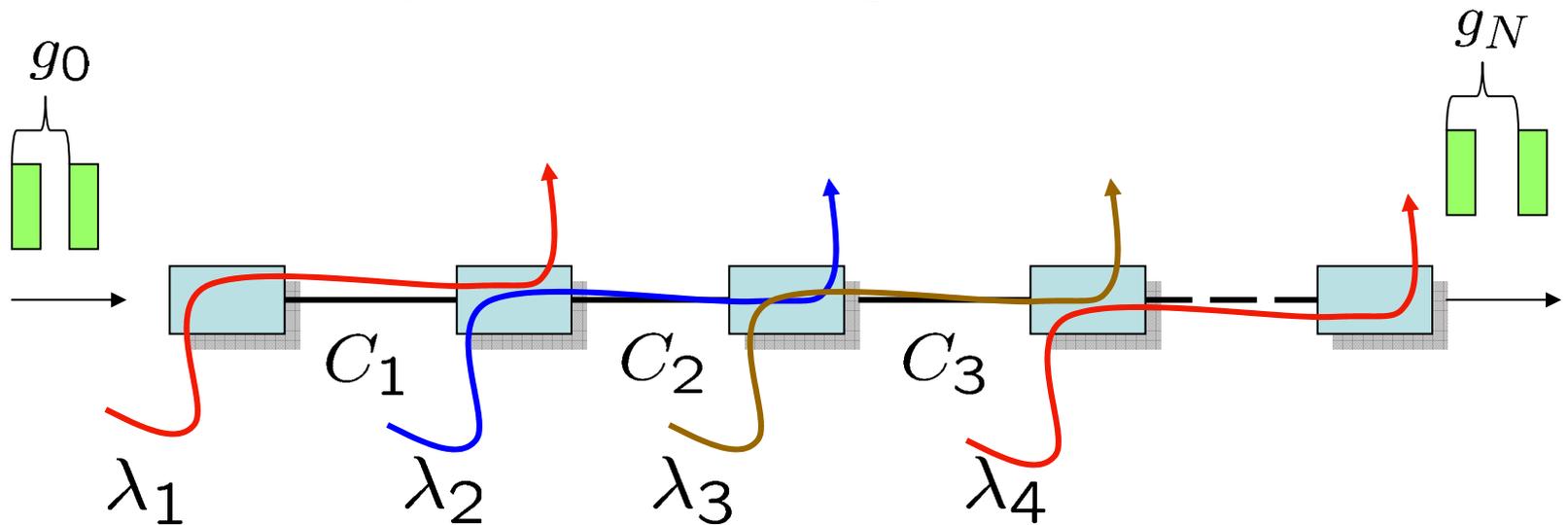
- Question: how do existing techniques relate to single-hop fluid curves?

PTR searches for the turning point.



# Related Work (cont'd)

- Previous multi-hop models
  - Analytical results are only available for fluid cross-traffic with **one-hop persistent** routing



- Mathematically:

$$g_i = \max \left( g_{i-1}, \frac{s + \lambda_i g_{i-1}}{C_i} \right)$$

# Outline

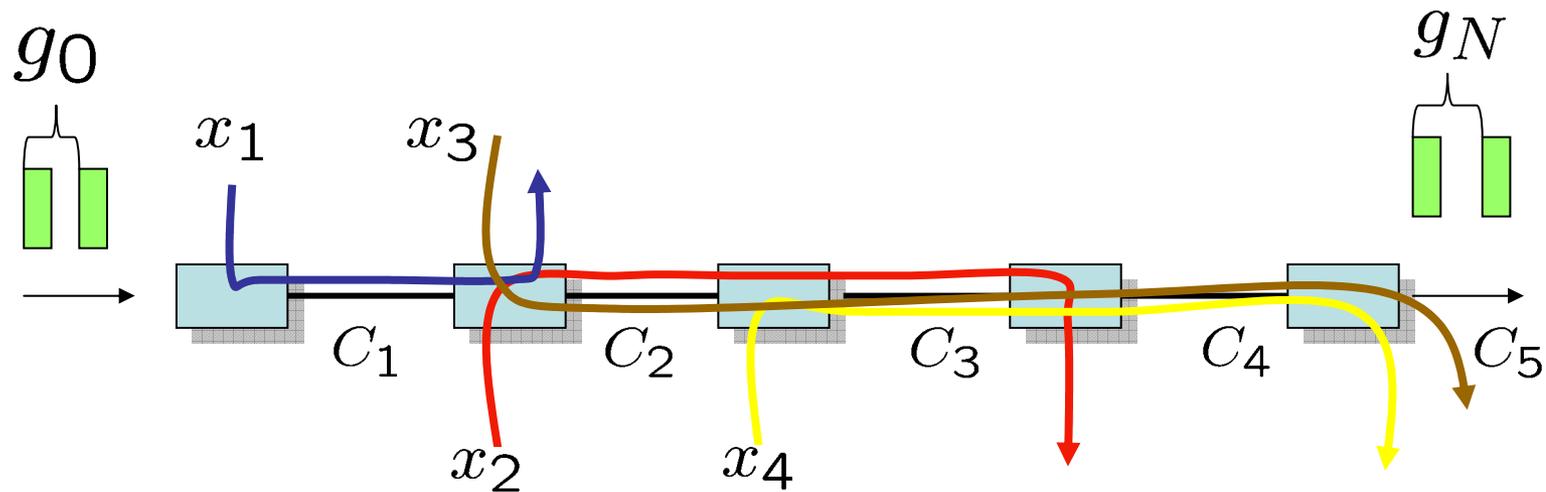
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- Implications on existing techniques

# Multi-hop Fluid Response Curve



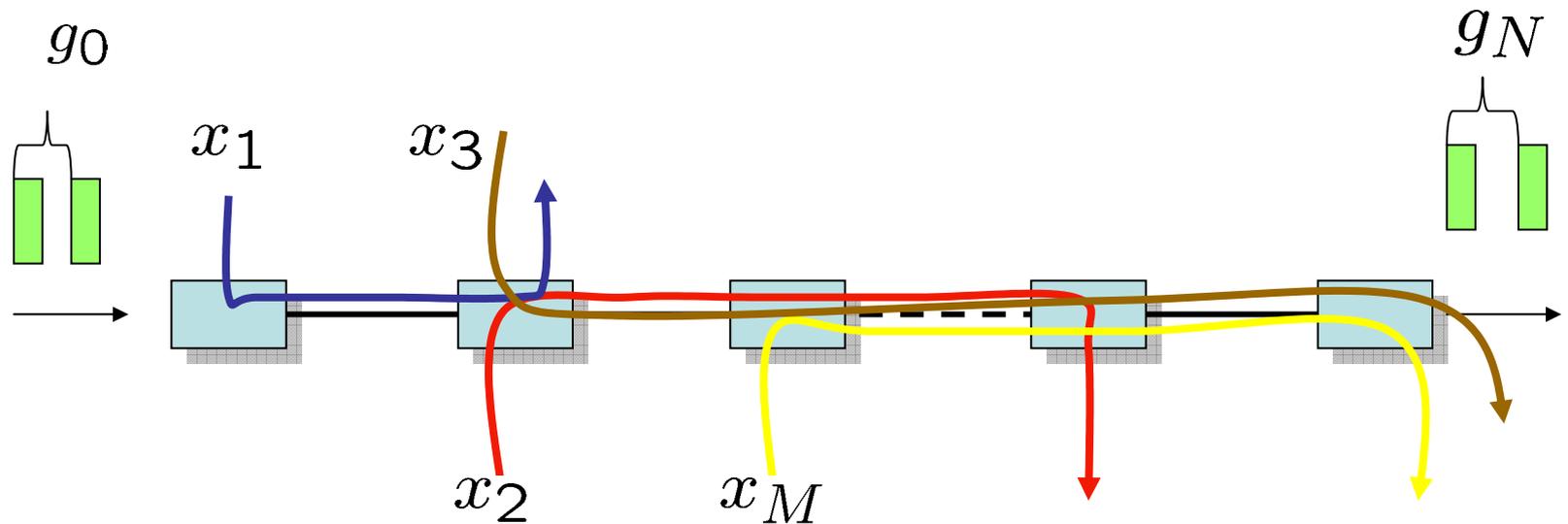
**Notation 1:** flow rate vector  $\mathbf{x}$ :  $(x_1, x_2, \dots, x_M)$

**Notation 2:** flow aggregation  $\Gamma_{k,i}$ : A binary vector indicating all flows that enter the path at link  $k$  and traverse link  $i$

**Example:**  $\Gamma_{2,3} = (0, 1, 1, 0)^T$

The arrival rate of  $\Gamma_{k,i}$  is  $\mathbf{x}\Gamma_{k,i}$ .

# Multi-hop Fluid Response Curve (cont'd)

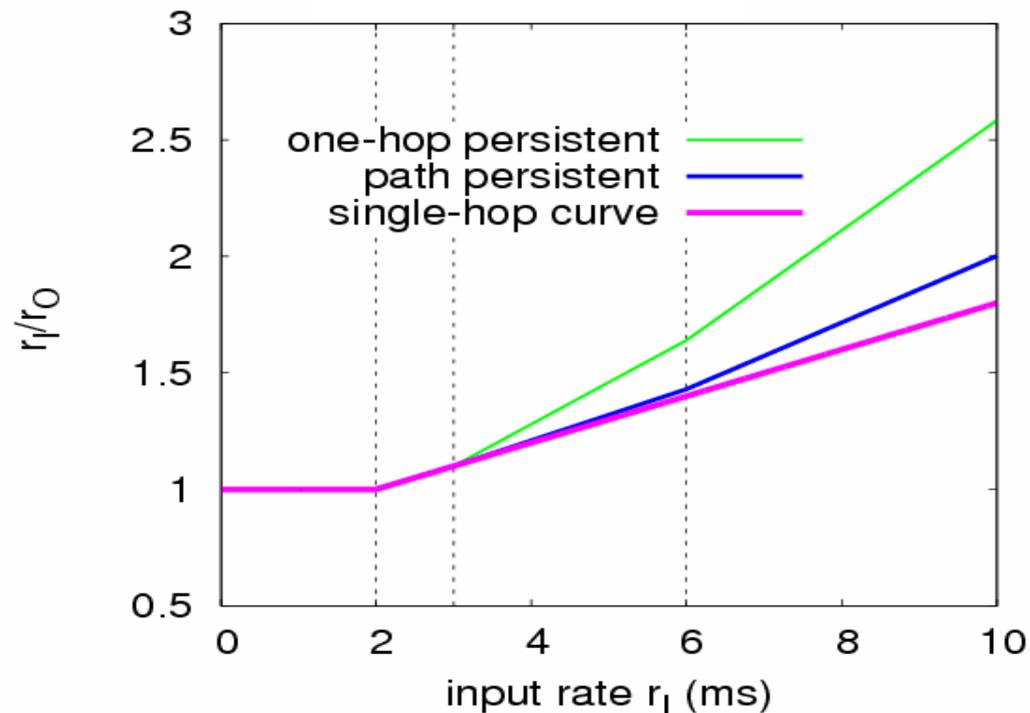


## Main Result 1:

$$g_i = \max \left( g_{i-1}, \frac{s + \sum_{k=1}^i g_{k-1} x \Gamma_{k,i}}{C_i} \right)$$

# Multi-hop Fluid Response Curve (cont'd)

- Implications of this result
  - One-hop persistent curve is the **upper** bound
  - Single-hop curve is the **lower** bound



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# Multi-Hop Stochastic Response Curve

**Assumption:** The output dispersion R.V.  $G_N$  for different probings have identical distributions.

**Main Result 2:**

$$\begin{aligned} E[G_i] &= E[G_{i-1}] + \frac{E[R_i]}{n-1} \\ &= \frac{\sum_{k=1}^i (\mathbf{x}\Gamma_{k,i} E[G_{k-1}]) + s}{C_i} + \frac{E[\tilde{I}_i]}{n-1}. \end{aligned}$$

# Multi-Hop Stochastic Response Curve (cont'd)

- Stochastic curve:

$$\begin{aligned}
 E[G_i] &= E[G_{i-1}] + \frac{E[R_i]}{n-1} \\
 &= \frac{\sum_{k=1}^i (\mathbf{x}\Gamma_{k,i} E[G_{k-1}]) + s}{C_i} + \frac{E[\tilde{I}_i]}{n-1}.
 \end{aligned}$$

- Fluid curve:

$$g_i = \max \left( g_{i-1}, \frac{\sum_{k=1}^i \mathbf{x}\Gamma_{k,i} g_{k-1} + s}{C_i} \right).$$

These terms do not have fluid counterparts.

# Multi-Hop Stochastic Response Curve (cont'd)

The presence of the two additional terms causes the multi-hop gap response curve to positively deviate from its fluid counterpart.

The response deviation  $\beta_N = E[G_N] - g_N$  can be recursively expressed as follows

$$\beta_i = \begin{cases} \beta_{i-1} + \frac{E[R_i]}{n-1} & g_i = g_{i-1} \\ \frac{1}{C_i} \sum_{k=1}^i (\mathbf{x}\Gamma_{k,i}\beta_{k-1}) + \frac{E[\tilde{I}_i]}{n-1} & g_i > g_{i-1} \end{cases}$$

# Multi-Hop Stochastic Response Curve (cont'd)

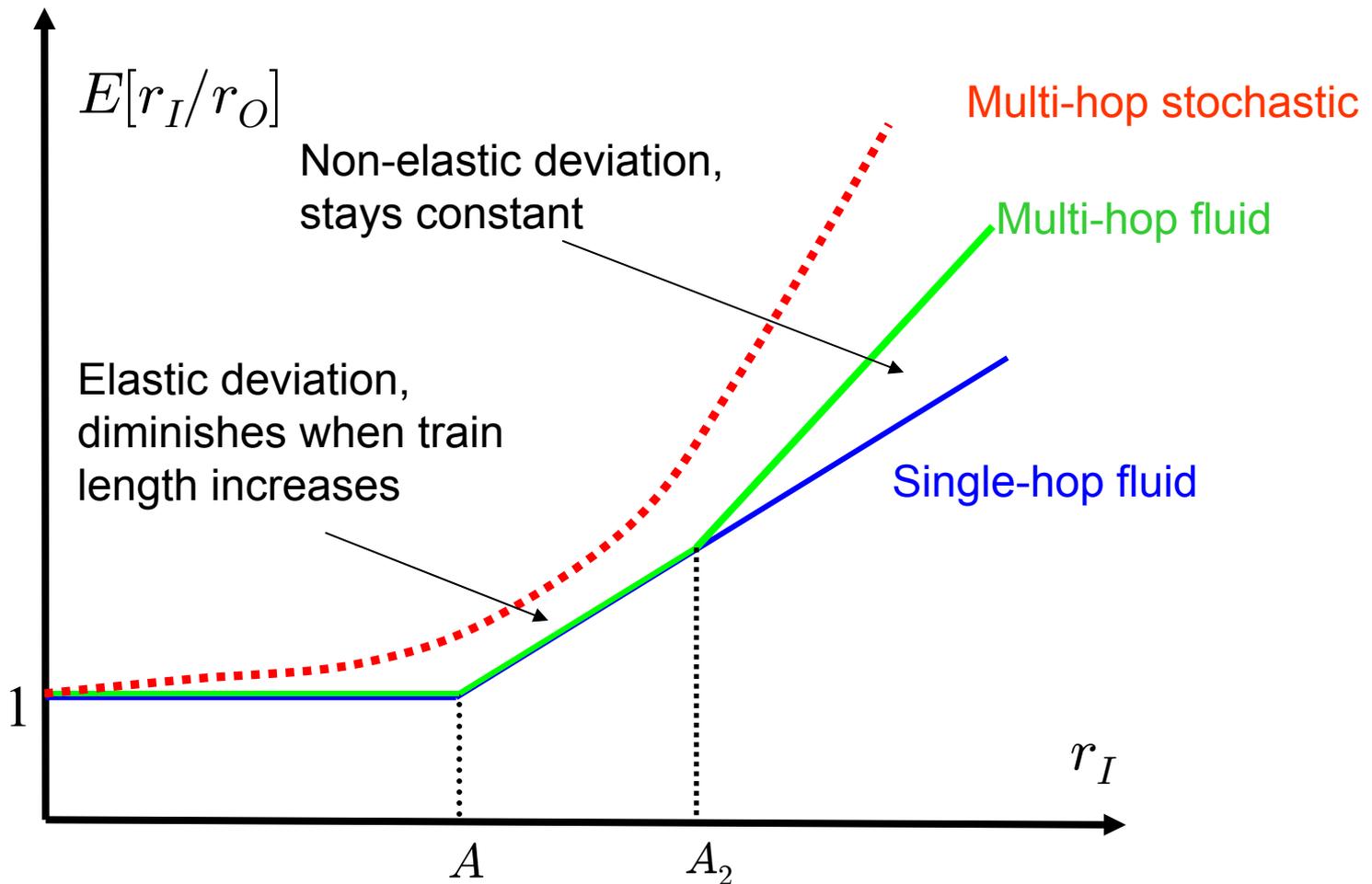
- Impact of Packet-train Parameters

Under certain additional assumptions, we get the following two results in the paper:

**Main Result 3:** For any given input rate, as probing packet size  $s$  increases,  $G_N$  converges to  $g_N$  in the mean.

**Main Result 4:** For any given input rate, as packet-train length  $n$  increases,  $G_N$  converges to  $g_N$  in the mean-square sense.

# Full Picture of the 3 Response Curves



# Outline

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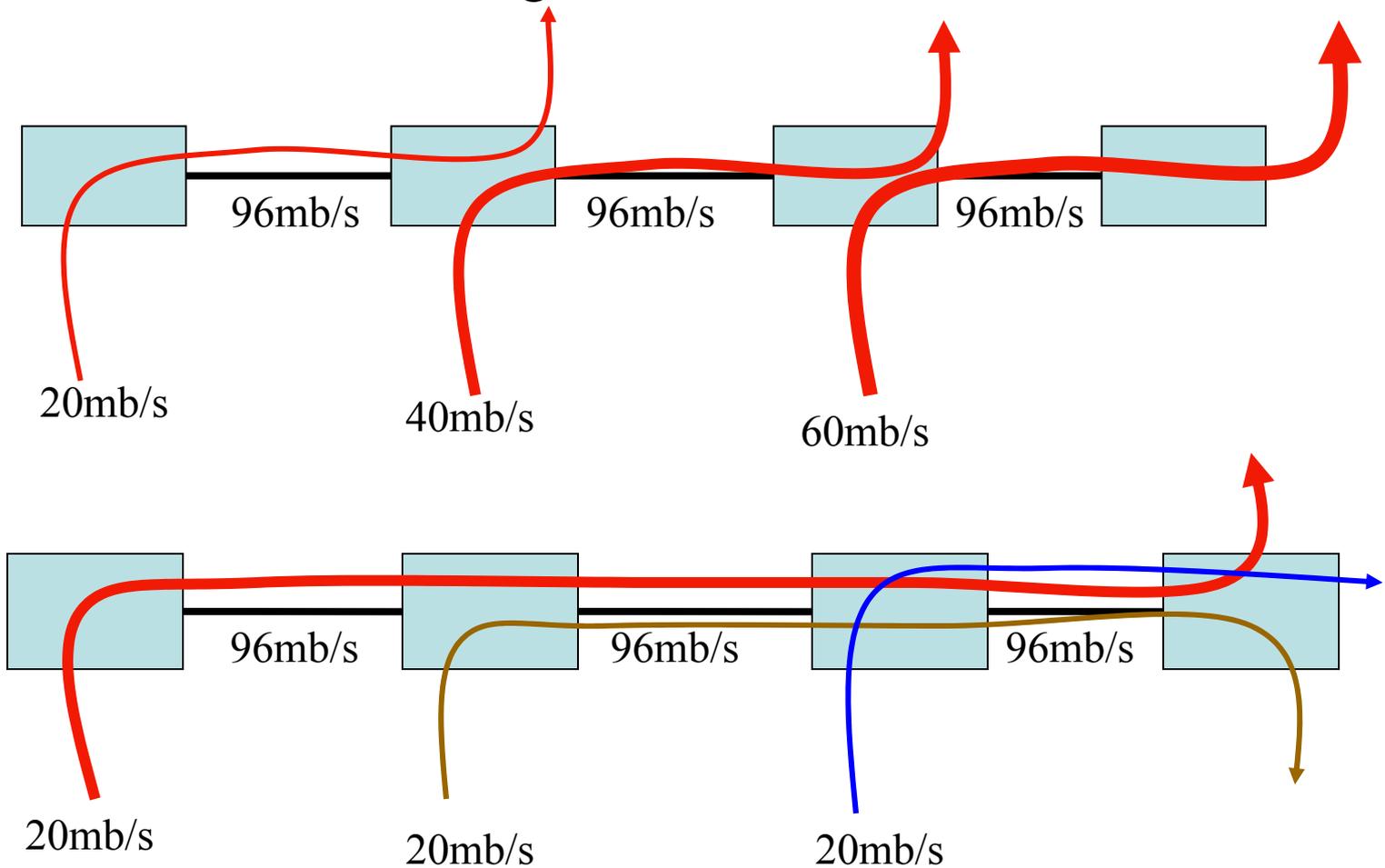
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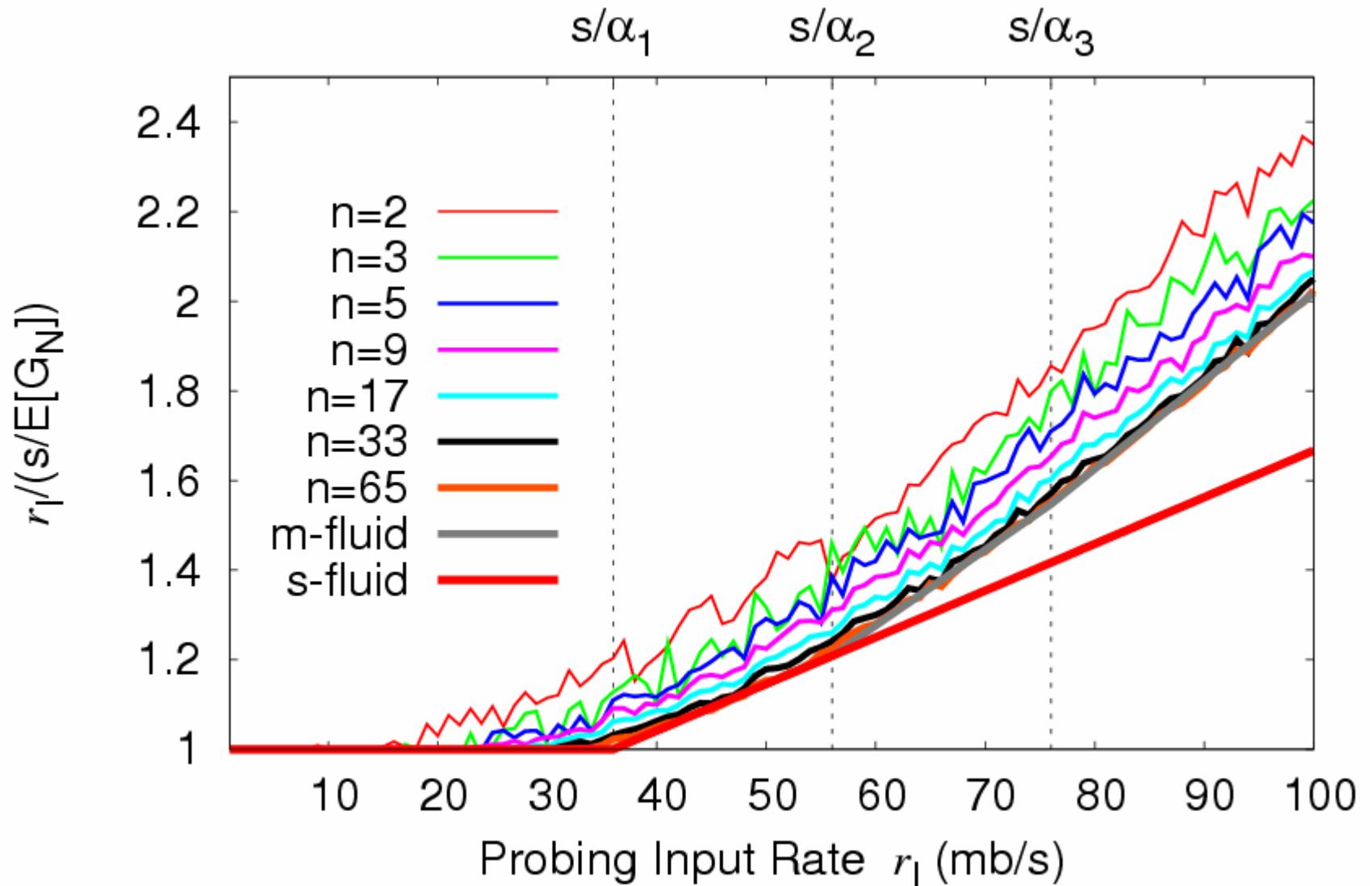
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- **Experimental verification**
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# Experimental Verifications

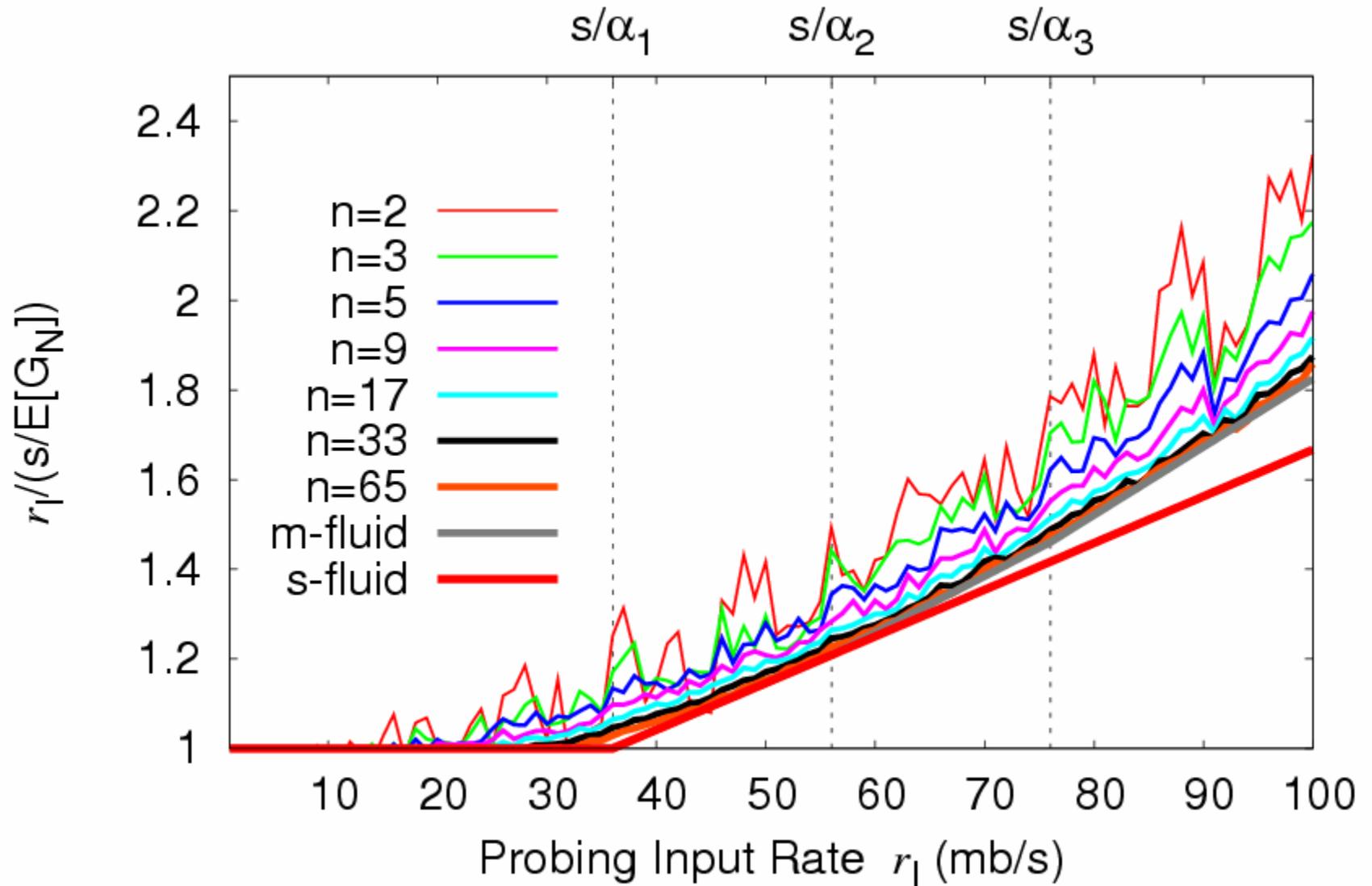
- Emulab Testbed Settings



# One-Hop Persistent Cross Traffic Routing



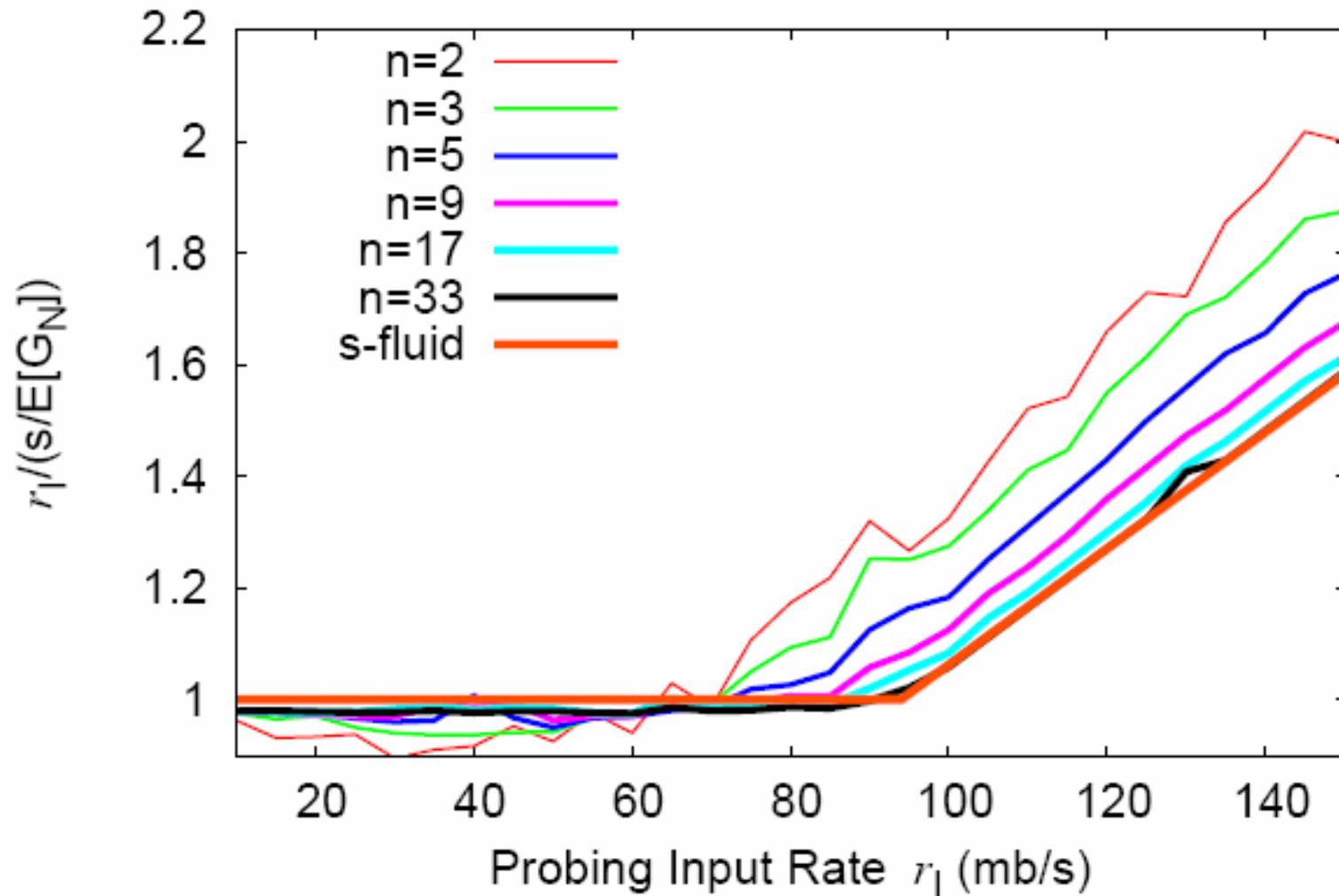
# Path Persistent Routing



# Real Internet Experiments

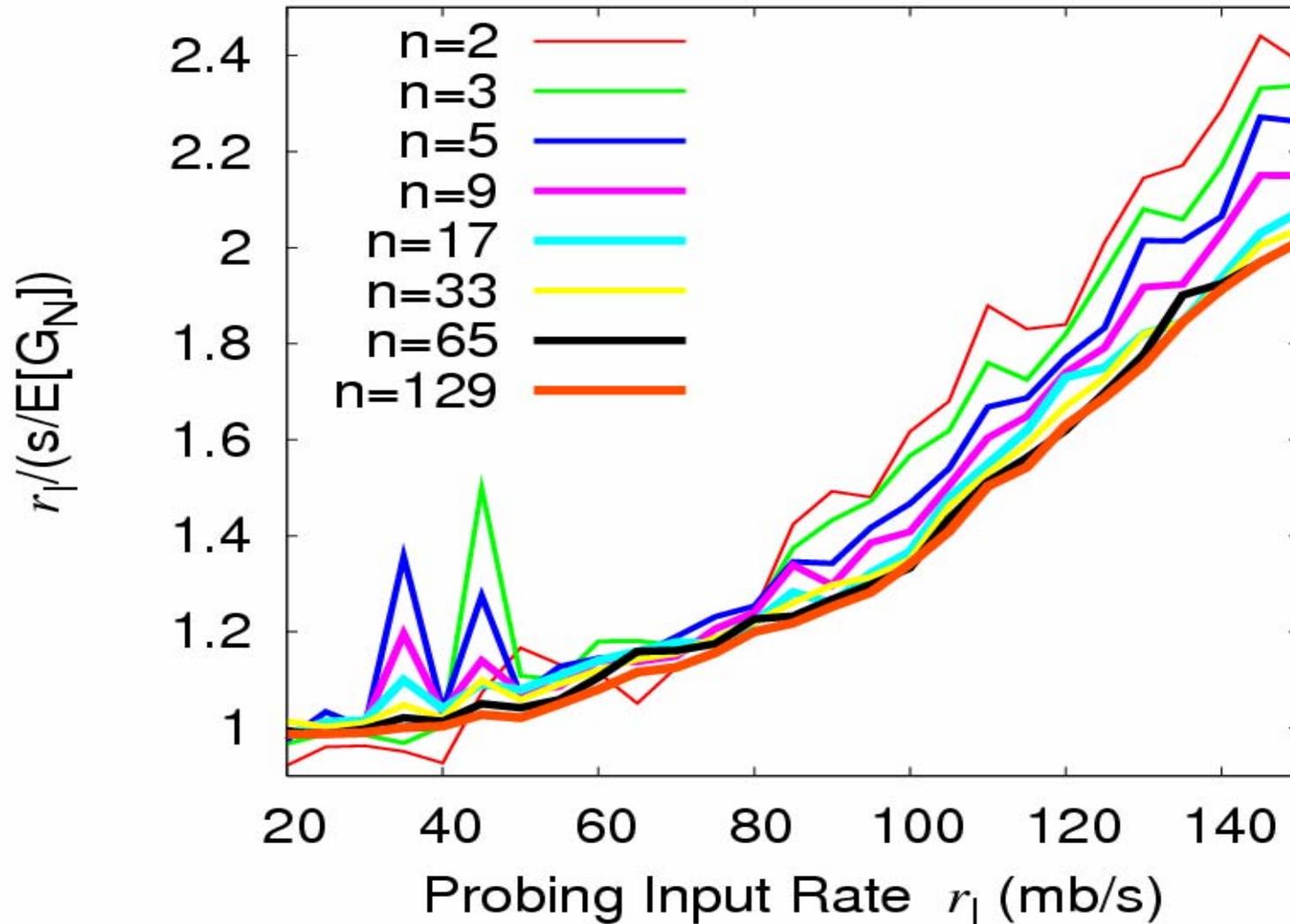
- We measure the rate response curves for more than 270 Internet paths over the RON testbed.
- Parameters:
  - Input rates: from 10 to 150 mb/s with step 5 mb/s
  - Packet-train length: from 33 to 129 packets
  - Packet-size: 1500 bytes
  - For each rate, we use 200 trains to estimate  $E[G_N]$
- Experiment durations are 20-100 minutes

# Lulea → CMU (1/16/2005)



(a) lulea → CMU

# Ana1-gblx → Cornell ( 4/29/2005)



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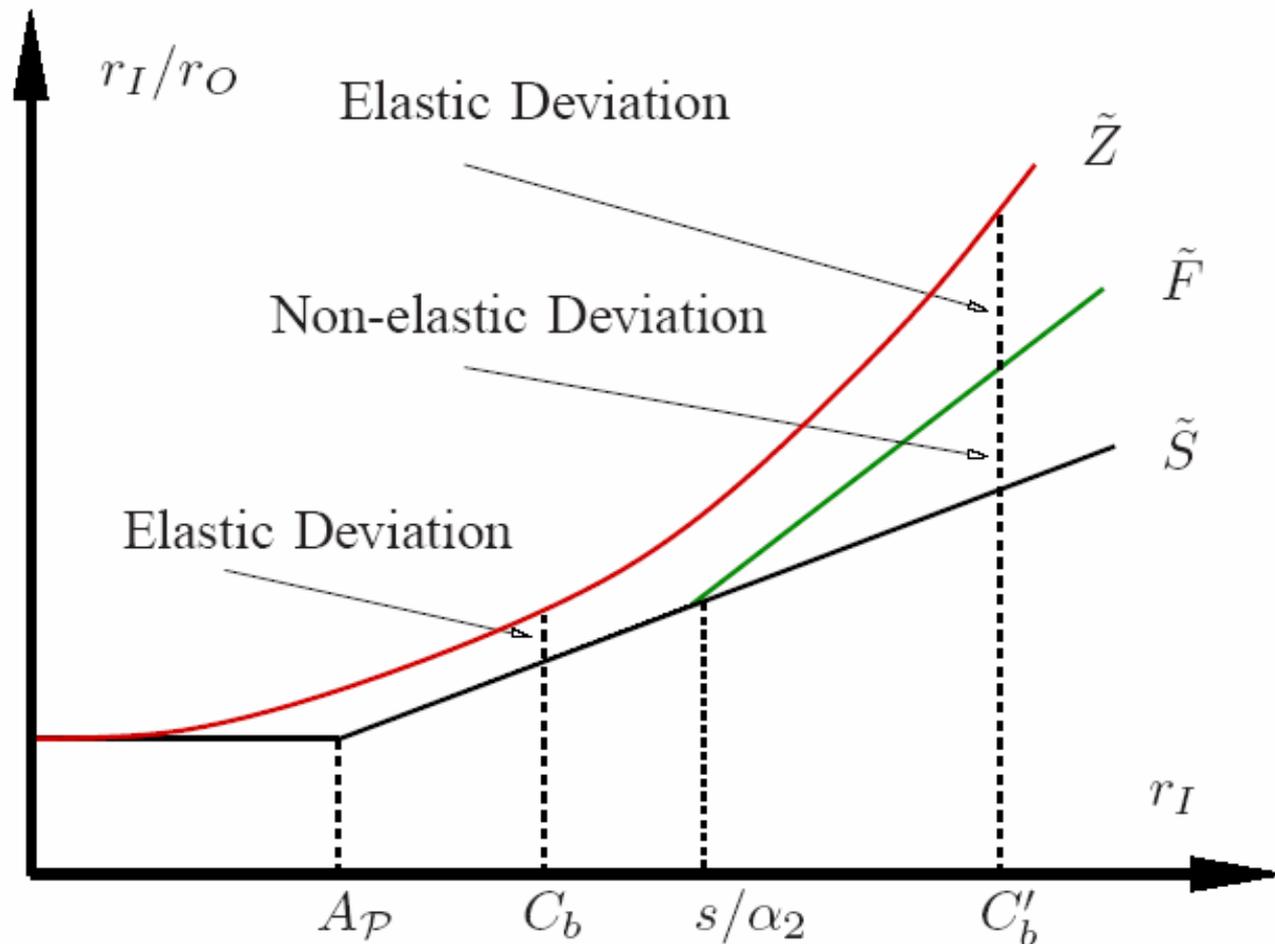
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# Implications on Existing Techniques

- TOPP uses **packet-pairs** to measure the stochastic response curve and implicitly assumes that it is the same as the fluid curve
  - To avoid bias, TOPP must use trains of sufficient length
- Pathload and PTR are related to searching for the turning point in the single-hop fluid response curve
  - Since they are using long trains, they are often immune to measurement bias

# Implications on Existing Techniques (cont'd)

- Spruce measurement bias



# Implications on Existing Techniques (cont'd)

- Spruce measurement biases

Experiment	Elastic Bias	Non-Elastic Bias	Total bias	Real avail- bw	Spruce Measurement
Emulab-1	53.76	30.24	84	36	0
Emulab-2	26.88	12	38.88	36	0
Lulea-cmu	24	0	24	94	70

# Conclusions

- We derived the multi-hop fluid response curve with arbitrary cross-traffic routing
- Also derived the multi-hop response curve using packet-based cross-traffic models and showed its convergence to its fluid counterpart when packet-train length increases
- Our results provided a stochastic justification of the existing techniques using long-trains
- Uncovered the sources of measurement biases for the techniques using short trains
- Leads to new techniques for measuring the tight link capacity (implementation in progress)