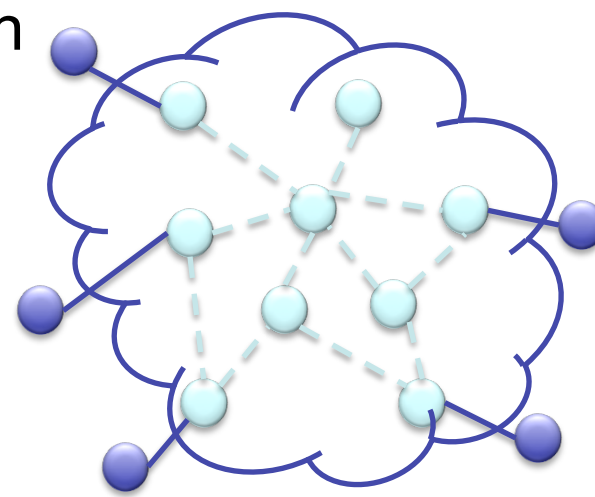


Network tomography has been proved to be a valuable tool for network monitoring and management in large scale networks. However, previous work in network tomography do not consider the occurrence of failures, while designing the network measurement framework. In this work, we study the problem of selecting paths to improve the performance of network tomography applications in the presence of network element failures.

## Network Tomography

- Goals:
  - Determine the state of internal elements using e2e measurements
  - Estimate the complete set of e2e measurements from an incomplete set
- Additive metrics (e.g. delay, log of loss rate)
- Solve a linear system:  $Ax = y \rightarrow$  e2e meas.
  - Link metrics
  - Path matrix
- Reduce overhead by probing a subset (basis) of the available paths
- Previous approaches **do not consider failures** of network elements



Under the assumption that the failure distribution is known, we propose an algorithm with guaranteed approximation ratio and feasible run-time; it is called ROBust MEasurements (RoMe). We also consider the case in which the failure distribution is not known, and propose a reinforcement learning algorithm to solve our optimization problem, using RoMe as a subroutine.

## Known failure distribution

- RoMe
  - Approximation factor  $(1-1/\sqrt{e})$
  - Exponential complexity in computing ER value
- Approximating ER value to achieve polynomial complexity
 
$$E(Z_{\hat{R}_q}) \leq \sum_{w \in R_{ind}} E(X_w) + \sum_{q \in R_{dep}} E(D_q)$$
- Optimal with additional constraint of linear independence in paths

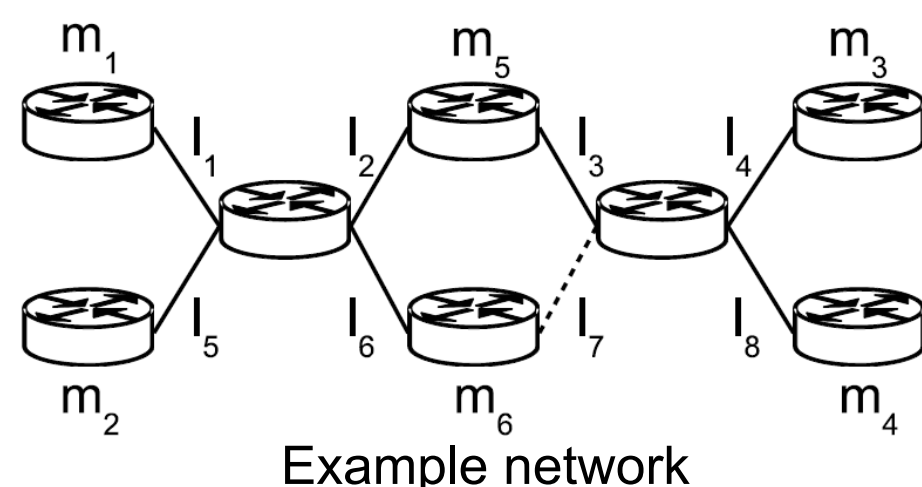
```

Algorithm 1: the RoMe algorithm
Input: Set  $R_M$ , budget  $B$ , cost function  $PC: 2^{R_M} \rightarrow \mathbb{R}^+$ ,
       objective function  $ER: 2^{R_M} \rightarrow \mathbb{R}^+$ 
Output: A subset of  $R_M$ 
1  $\bar{R} = \arg \max_{q \in R_M} \{ER(\{q\}) : PC(\{q\}) \leq B\}$ ;
2  $R_{out} = \emptyset$ ;
3  $R = R_M$ ;
4  $w_q = \frac{ER(\{q\}) - ER(\emptyset)}{PC(\{q\})}, \forall q \in R_M$ ;
5 while  $R \neq \emptyset$  do
6    $q_{max} = \arg \max_{q \in R} w_q$ ;
7   if  $PC(R_{out}) + PC(\{q_{max}\}) \leq B$  then
8      $R_{out} \leftarrow R_{out} \cup \{q_{max}\}$ ;
9    $R = R \setminus \{q_{max}\}$ ;
10  // Weight update
11  for all the  $q \in R$  do
12     $w_q = \frac{ER(R_{out} \cup \{q\}) - ER(R_{out})}{PC(\{q\})}$ ;
13 if  $ER(R_{out}) > ER(\bar{R})$  then
14   return  $R_{out}$ 
15 else
16   return  $\bar{R}$ 
    
```

## Impact of failures

### Motivational example:

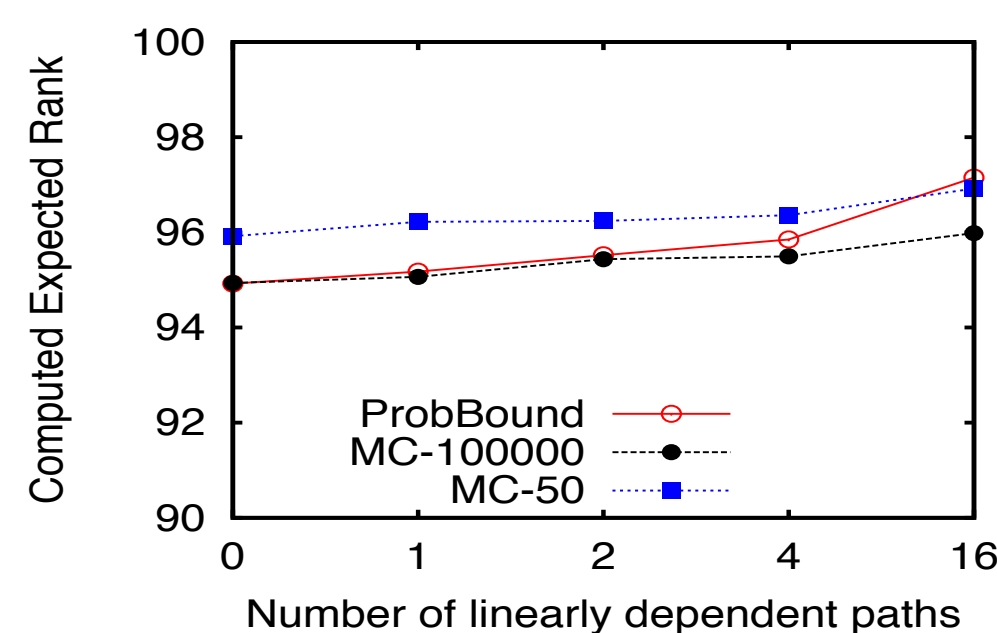
- Monitors:  $m_1, m_2, \dots, m_6$
- Considered bases
  - $R_1 = (q_1, q_2, q_4, q_{11}, q_{15}, q_5, q_6, q_7)$
  - $R_2 = (q_5, q_6, q_7, q_8, q_9, q_{10}, q_{11}, q_{12})$



- Probing  $R_1$  when  $l_7$  fails:
  - Provides rank 3
  - Does not cover links  $l_3$  and  $l_6$
  - Results in a linear system with infinite solutions
- Probing  $R_2$  when  $l_7$  fails:
  - Provides rank 6
  - Enables to determine that  $l_7$  has failed
  - Results in a linear system that has a solution for each link

### Motivational experiments:

- Realistic ISP topologies from the Rocketfuel Project [1]



## Problem formulation

- Problem setting:
  - Heterogeneous probing cost
  - Limited budget
  - Set of candidate paths  $R_M$
- Objective function: Expected Rank (ER)

$$ER(R) = \sum_{v \in \{0,1\}^{|E|}} r(R_v) P(v)$$

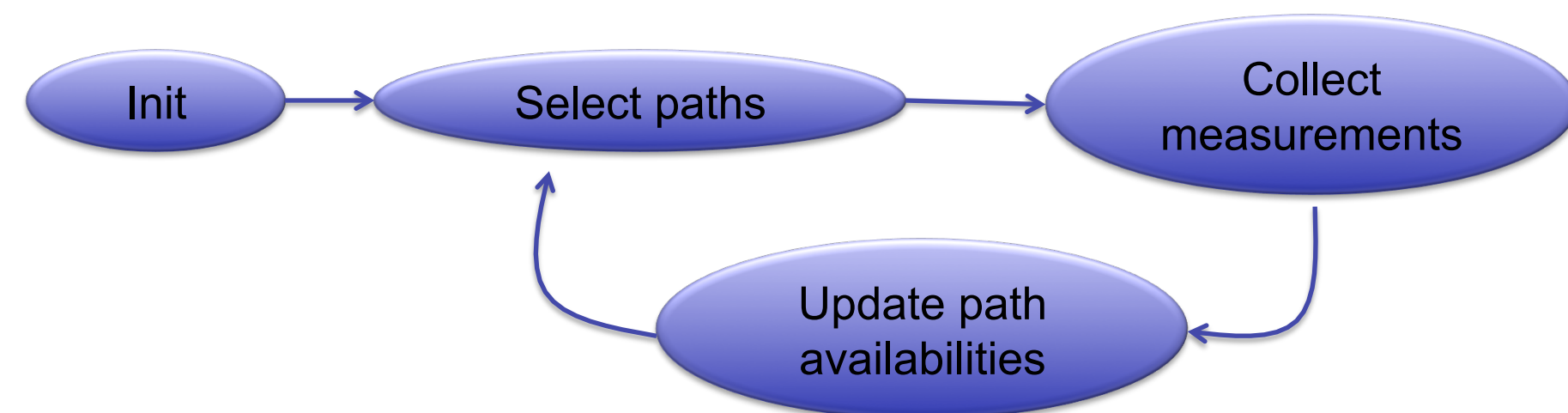
- Optimization problem:

$$\begin{aligned}
 &\text{Find } R^* \subseteq R_M \\
 &R^* = \operatorname{argmax}_{R^* \subseteq R_M} ER(R) \\
 &\text{s. t. } PC(R) \leq B
 \end{aligned}$$

Probing cost  $\rightarrow$  Budget

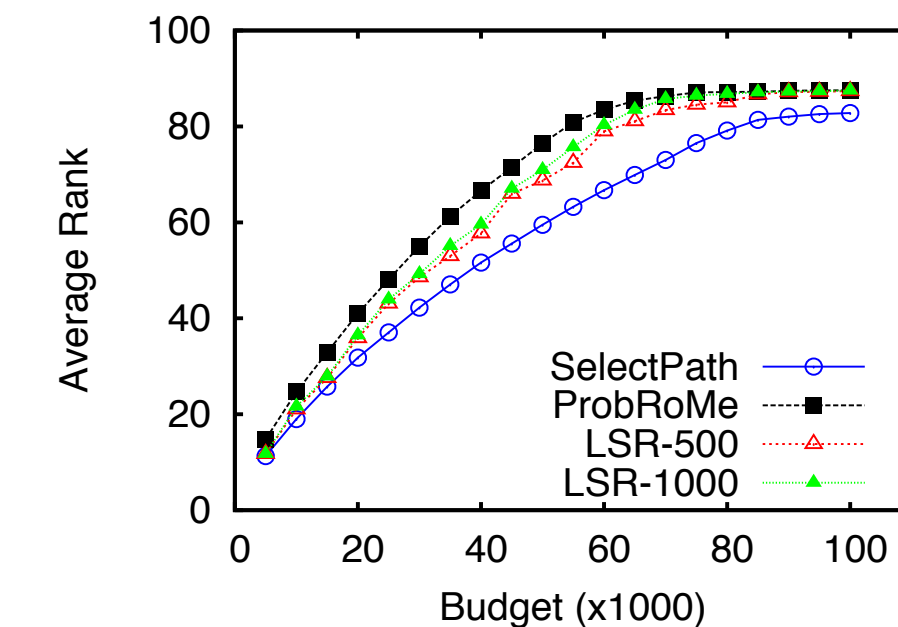
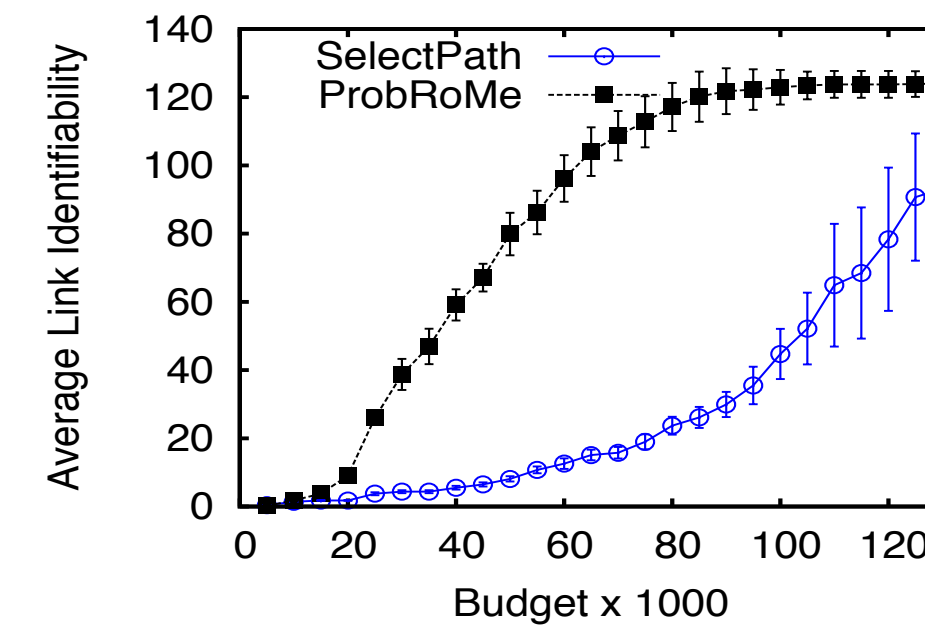
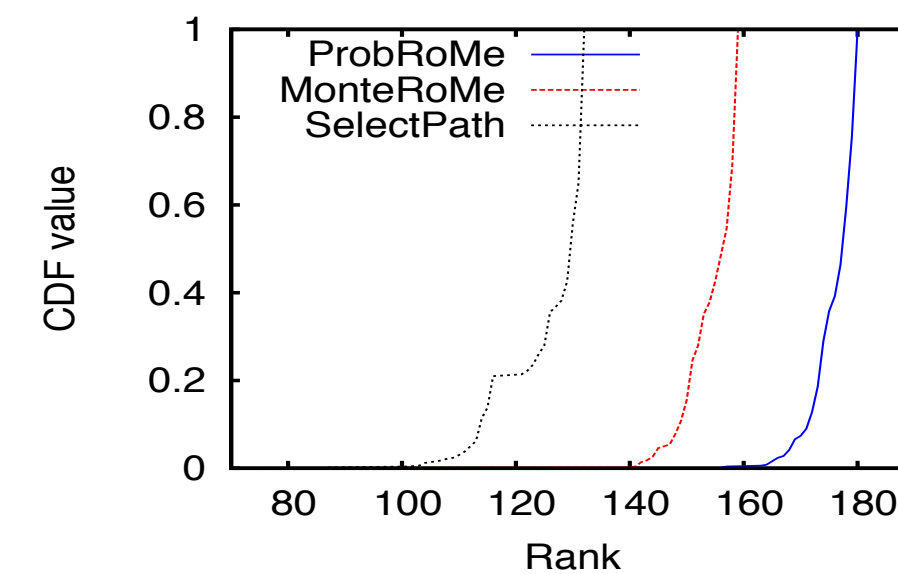
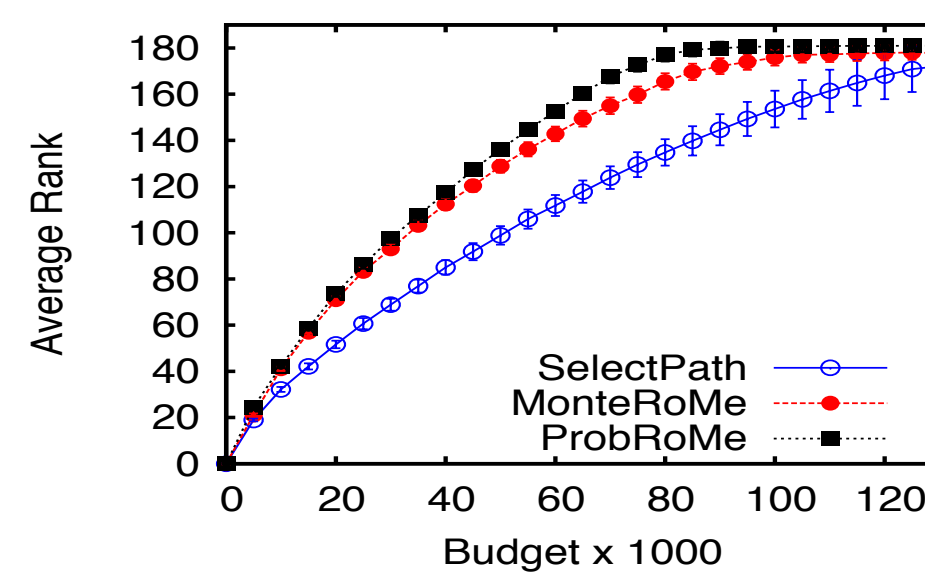
## Unknown failure distribution

- Reinforcement learning approach
  - Learning with Submodular Rewards (LSR) algorithm
  - We learn path availabilities through probing
- LSR has performance guarantees under certain conditions



## Results

- Realistic ISP topologies from the Rocketfuel Project [1]
- Realistic failure model [2]
- We compared to SelectPath [3]



## References

[1] "Rocketfuel Project: Internet topologies," <http://www.cs.washington.edu/research/networking/rocketfuel/>.

[2] A. Markopoulou, G. Iannaccone, S. Bhattacharyya, C.-N. Chuah, and C. Diot, "Characterization of failures in an IP backbone," IEEE INFOCOM, 2004.

[3] Y. Chen, D. Bindel, H. Song, and R. H. Katz, "An algebraic approach to practical and scalable overlay network monitoring," ACM SIGCOMM Comp. Com. Rev., 2004.

[4] S. Tati, S. Silvestri, T. He, T. La Porta, "Robust Network Tomography in the Presence of Failures", IEEE ICDCS 2014