

# Locating network monitors: complexity, heuristics, and coverage

Kyoungwon Suh, Yang Guo, Jim  
Kurose, and Don Towsley  
kwsuh, kurose,towsley@cs.umass.edu  
Yang.Guo@mathworks.com

presented by Kyoungwon Suh

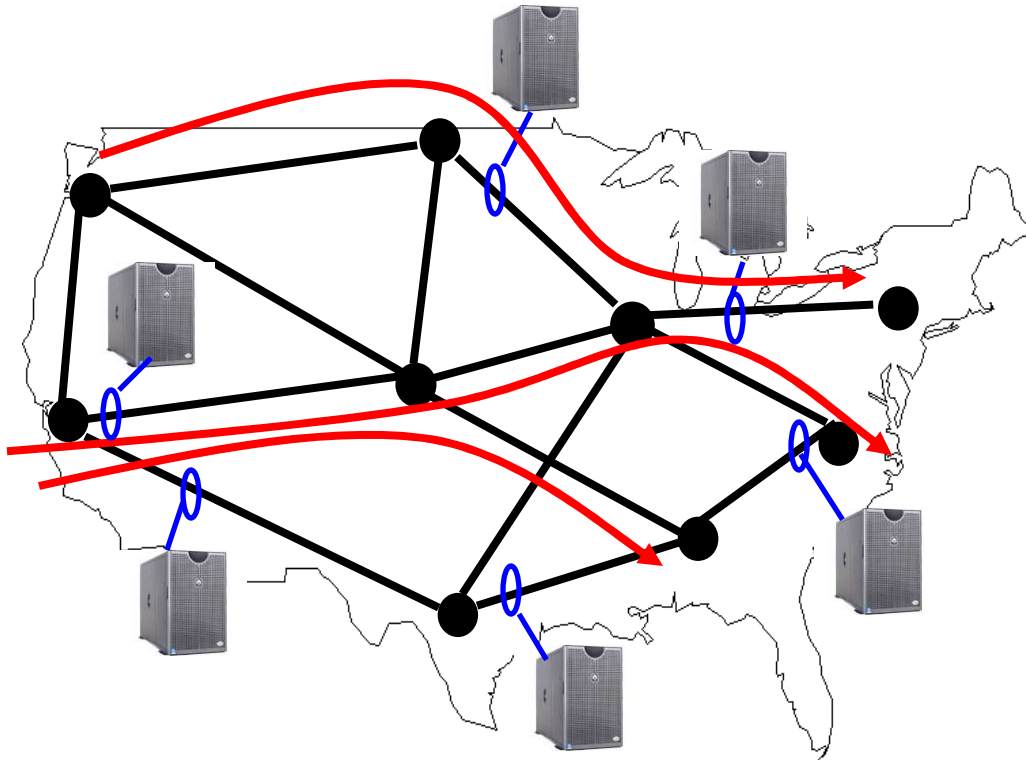


University of Massachusetts  
Department of Computer Science

# Outline

- ❑ Motivation
- ❑ Related Work
- ❑ Monitoring cost and reward models
- ❑ Passive monitoring: problems and solutions
- ❑ Evaluating our heuristics
- ❑ Summary and future work

# Challenges of distributed monitoring



- ❑ Need to observe large fraction of traffic
- ❑ Single link measurement point: NOT enough
- ❑ Monitoring cost versus monitoring reward

# Related Work on monitor placement

- ❑ Number and location of active monitors
  - ❑ NP-hardness proof and approx. algorithm
    - Jamin (INFOCOM '00), Adler (NGC '01), Horton (IMC '03), Barford (IMW '01) Bejerano (INFOCOM '03), and Li (INFOCOM '03)
- ❑ Distributed passive monitors
  - ❑ Locations were chosen in ad-hoc basis
    - Sprint IPMon Project (PAM '01)
- ❑ Adaptive sampling strategies
  - ❑ Concerned with "single" passive monitor
    - INMON's sFlow, Choi (SIGMETRICS '02) and Duffield (IMW '02)

No previous work on location and sampling strategies for passive monitors

# Monitoring cost and reward models

□ Deployment cost:

$$C_D = \sum_{i \in L} f_i y_i$$

$L$ : set of all links

$f_i$ : monitor deployment cost at link  $i$

$y_i$ : 1 if a monitor is placed at link  $i$

□ Operating cost:

$$C_O = \sum_{i \in L} y_i c_i \sum_{j \in D} \rho_j m_{ij}$$

$D$ : set of all flows

$c_i$ : unit sampling cost of monitor  $i$

$\rho_j$ : traffic demand of flow  $j$

$m_{ij}$ : fraction of flow  $j$  sampled by monitor  $i$

# Monitoring cost and reward models - Continued

## □ Monitoring reward:

- Benefit of traffic monitoring
- e.g., monitored traffic demand (i.e., coverage)

$$C_M = \sum_{j \in D} u_j(M_j)$$

$M_j$ : fraction of flow  $j$  sampled by monitors  
 $u_j(M_j)$ : benefit gained by monitoring flow  $j$ ; non-decreasing and concave utility function

Examples :

$$M_j = 1 - \prod_i (1 - m_{ij})$$

$$u_j(M_j) = \rho_j M_j ; \quad u_j(M_j) = (1 - \exp(-5.0 * M_j))$$

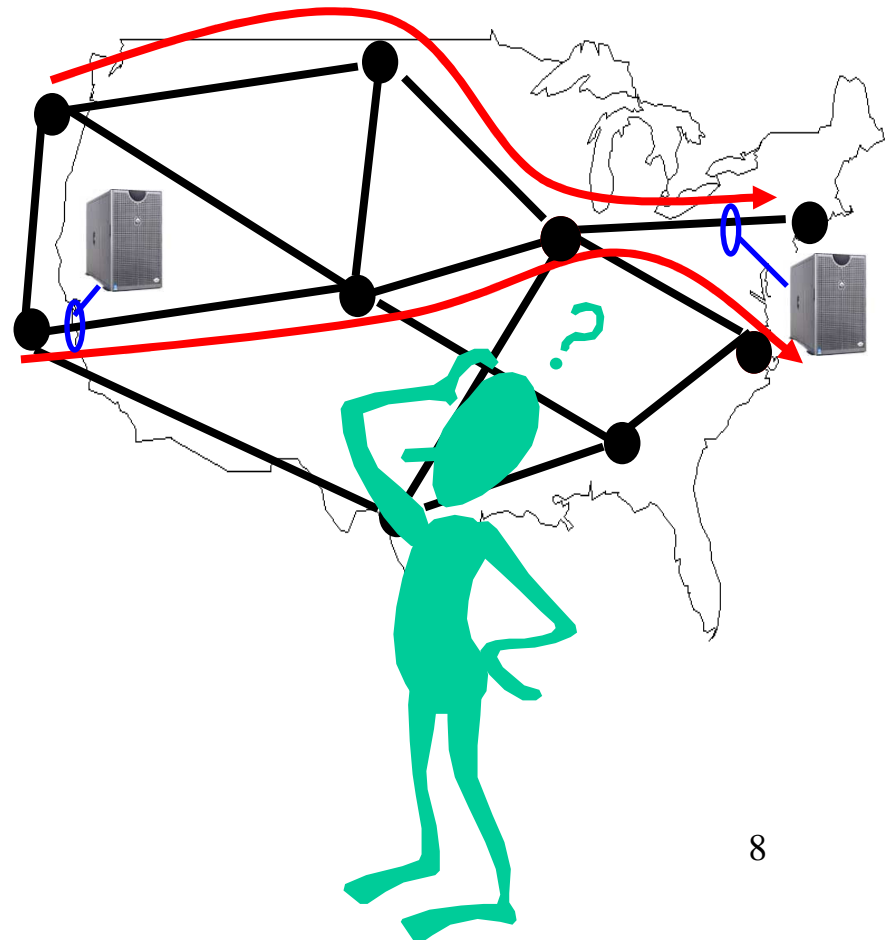
# Monitoring Problems

- ❑ Monitoring problems without sampling
  - ❑ BCMCP: Budget Constrained Maximum Coverage Problem
  - ❑ MDCP: Minimum Deployment Cost Problem
  - ❑ MDOCP: Minimum Deployment and Operating Cost Problem
- ❑ Monitoring problem with sampling
  - ❑ BCMCP-S: Budget Constrained Maximum Coverage Problem w/ Sampling

# BCMCP - Example

(Budget constrained maximum coverage problem)

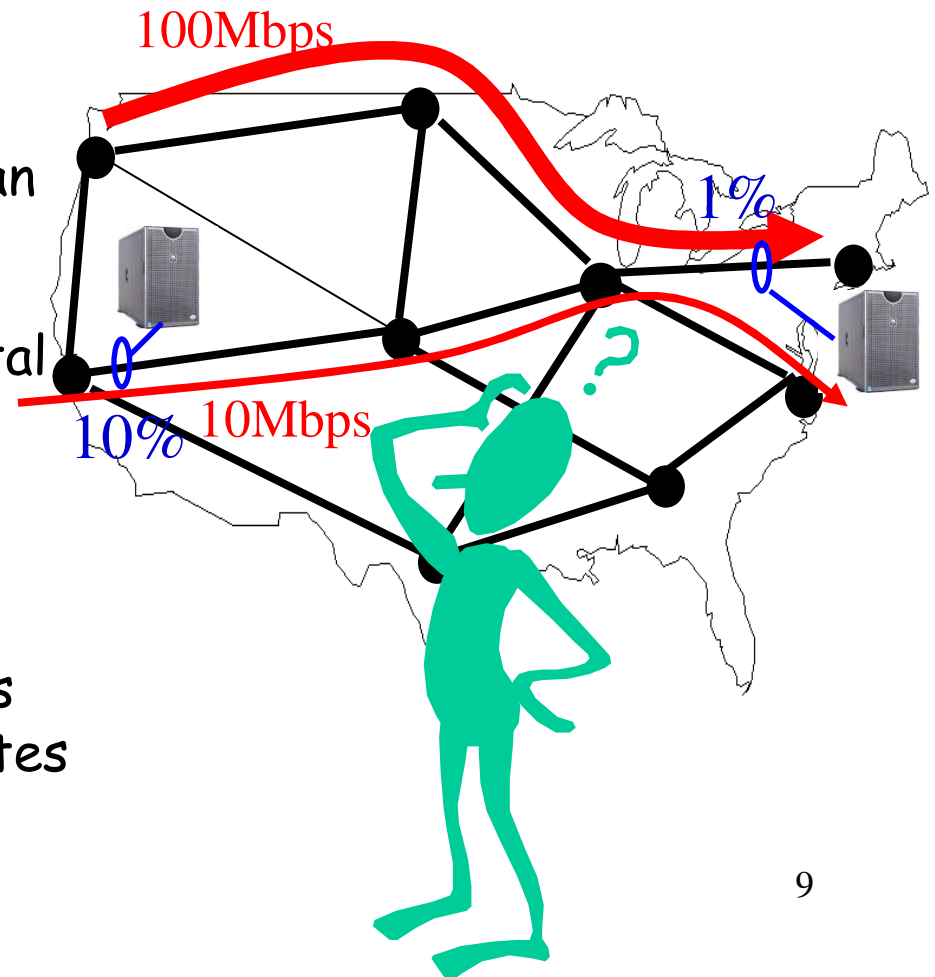
- ❑ Constraints:
  - ❑ Monitor captures all packets seen on link
  - ❑ Constraint:  $K$  monitors
- ❑ Objective:
  - ❑ Maximize total rate of monitored packets
- ❑ What to determine:
  - ❑ Optimal monitor locations



# BCMCP-S - Example

(Budget Constrained Maximum Coverage Problem with Sampling)

- ❑ **Constraints:**
  - ❑ Placed monitor *samples* packets carried by link
  - ❑ Per-flow sampling rate can be chosen
  - ❑ Constraint: K monitors
  - ❑ Constraint: R Mbps in total
- ❑ **Objective:**
  - ❑ Maximize sum of flow sampling rates
- ❑ **What to determine:**
  - ❑ Optimal monitor locations
  - ❑ Optimal flow sampling rates



# BCMCP: problem formulation and solution

- ❑ Objective
  - ❑ Maximize monitoring reward w/o violating total deployment cost constraint
- ❑ Complexity: NP-Hard
  - ❑ ILP formulation
  - ❑ Reduction from budgeted maximum coverage problem
- ❑ Proposed greedy heuristics
  - ❑ Greedy: choose links maximizing marginal gains per unit cost
  - ❑ Approximation ratio:  $1-1/e$

# BCMCP-S: problem formulation and solution

## ❑ Objective

- ❑ Maximize monitoring reward w/o violating deployment and operating cost constraints

## ❑ MINLP formulation

- ❑ Difficult to obtain optimal solution directly

## ❑ Proposed algorithm

- ❑ Greedy algorithm for location
- ❑ Gradient projection method for sampling rates

# Questions

- ❑ How close are approximation solutions to optimal solutions?
- ❑ How many monitors are required to monitor “large” fraction of traffic?

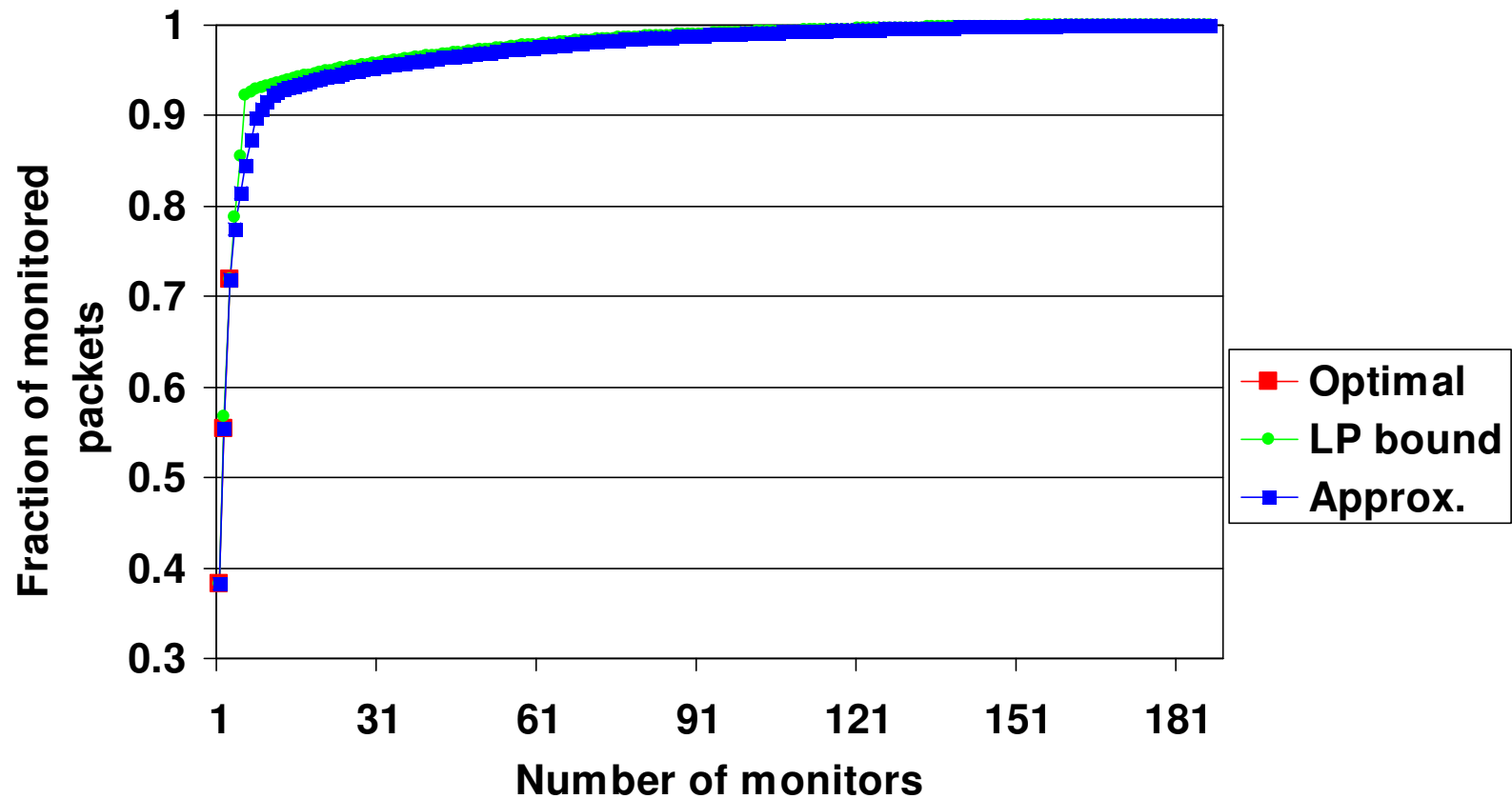
# Evaluating our heuristics

## □ Simulation settings:

- Random and transit-stub topologies generated by GT-ITM (10 and 100 nodes)
- Cable & Wireless PoP-level topology (33 nodes)
- Node: abstraction of a router or a PoP
- Utility function: linear or exponential
- Cost functions: equal deployment cost, equal operating cost
- Traffic matrices: Fortz and Thorup's traffic generation formula (INFOCOM '00)

# BCMCP

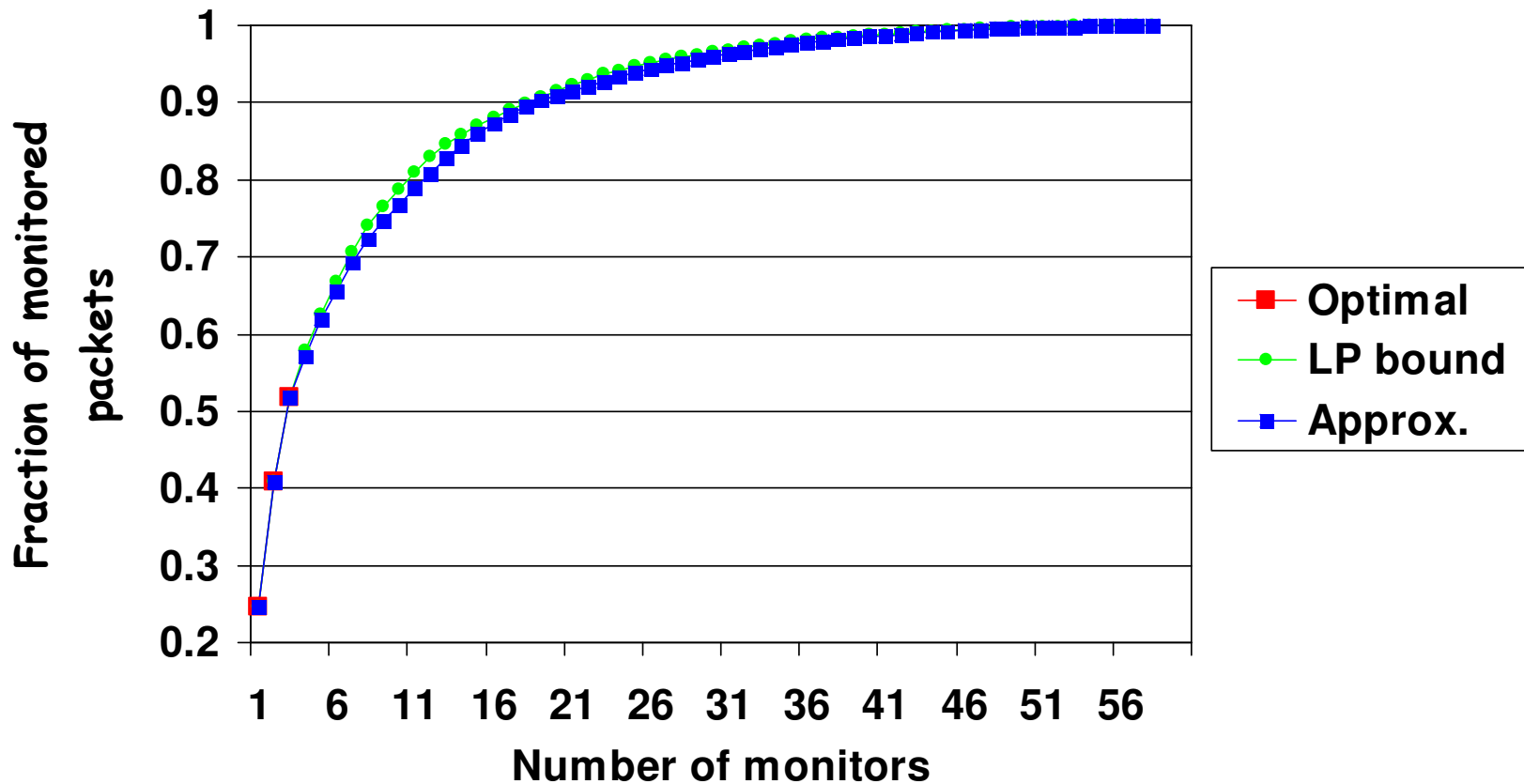
Transit Stub: 100 nodes, 187 links, 8885 flows



8 monitors (4% of possible locations) can monitor more than 90% of packets

# BCMCP

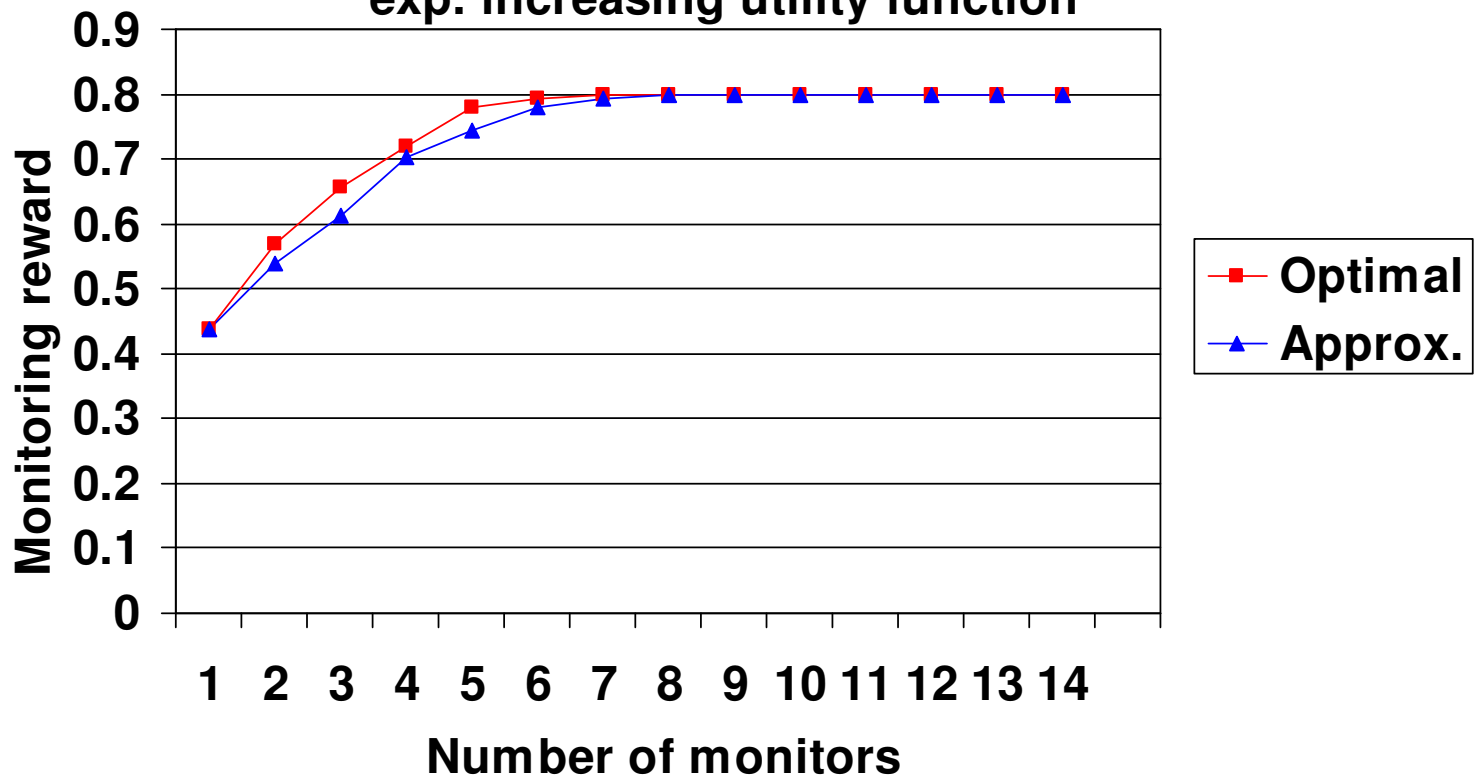
Cable & Wireless PoP-level: 33 nodes, 58 edges, 1041 flows



20 monitors (34% of possible locations)  
can monitor 90% of packets

# BCMCP-S

Random graph: 10 nodes, 14 links, 36 flows, and  
exp. increasing utility function



Marginal increase in monitoring reward is  
decreasing

# Questions/Answers

- ❑ How close are approximation solutions to optimal solutions?
  - ❑ Much better than theoretical worst-case bounds
  - ❑ Often, same as optimal solutions
  - ❑ Why: shortest-path routing, all-pair traffic demands, and network topology
- ❑ How many monitors are required to monitor "large" fraction of traffic?
  - ❑ Marginal increase in monitoring reward is decreasing
  - ❑ Only a few monitors are enough!

# Summary and Future work

- ❑ Proposed novel monitor location models
- ❑ Proposed monitor placement and operating strategies
- ❑ Experiment results show that:
  - ❑ Very good approximation ratio can be achieved
  - ❑ Only a small fraction of links is often enough
- ❑ Extension of our models
  - ❑ Routing changes caused by link/router failures
  - ❑ Stochastic traffic demand models

The End