

# Ad Hoc Distributed Simulation of Queueing Networks



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# Agenda

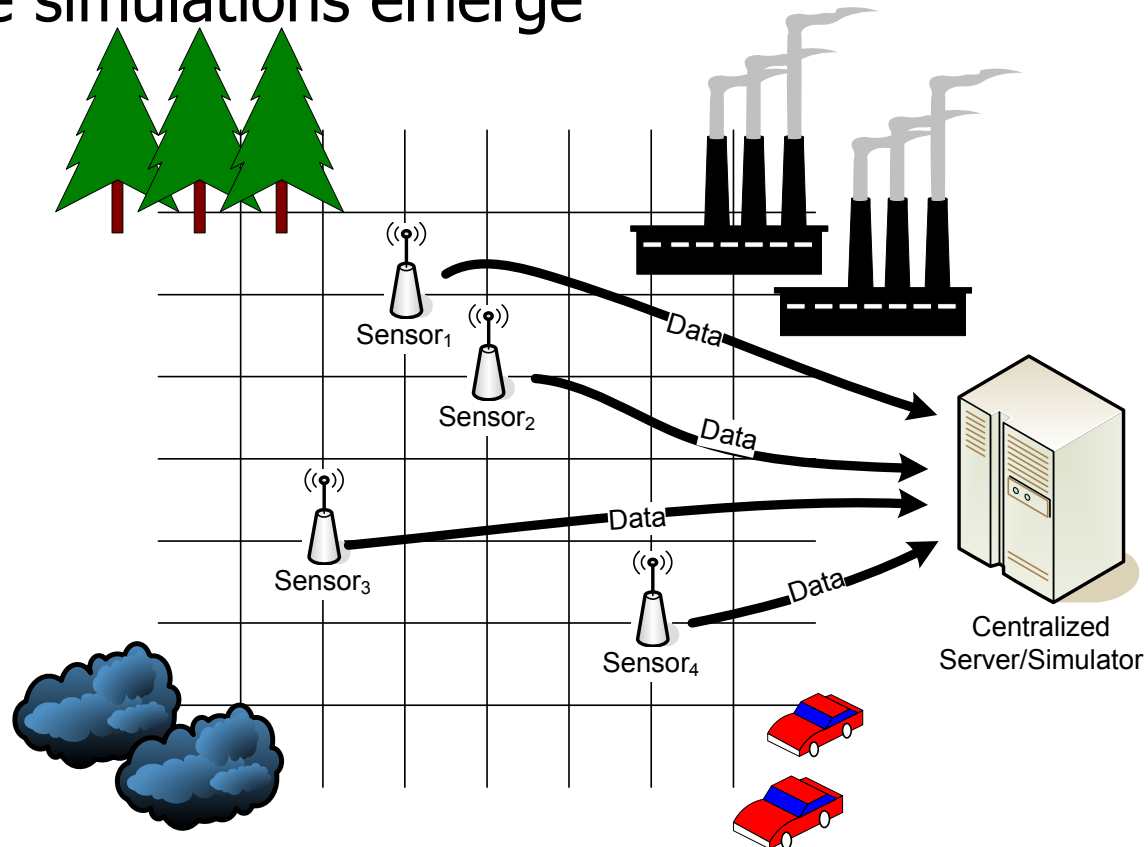


- ⌘ Motivations
- ⌘ Ad Hoc Distributed Simulations
- ⌘ Queueing Network Simulations
- ⌘ Experiments and Results
- ⌘ Conclusions and Future Works

# Motivations



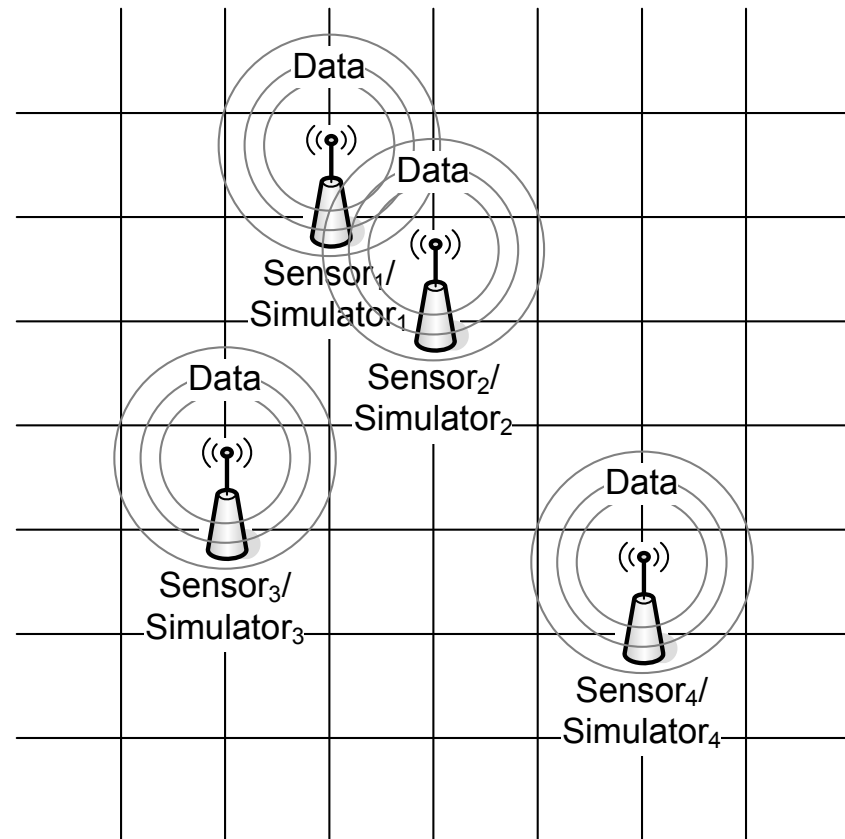
- ⌘ Sensor networks gain in importance
- ⌘ On-line simulations emerge



# Motivations (*cont.*)



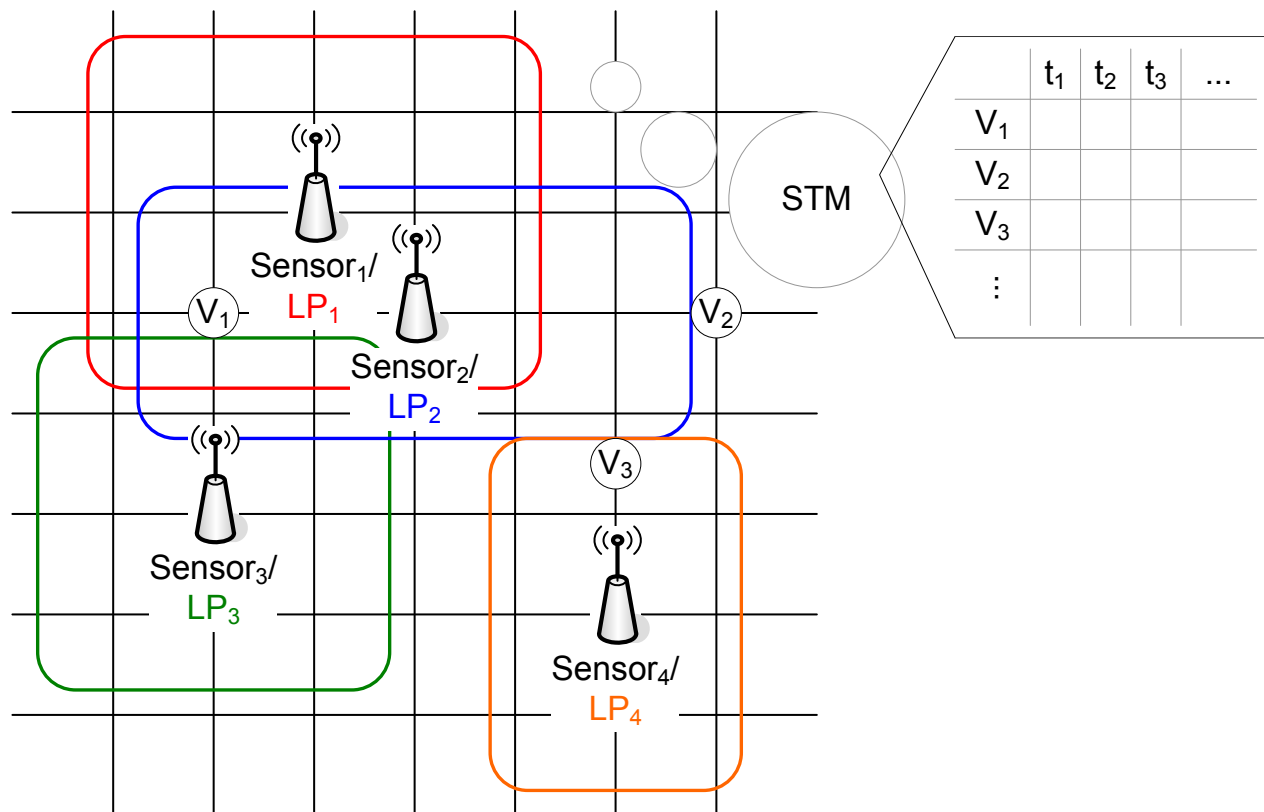
- ⌘ Operating on-line simulations within sensor networks show benefits
  - ☑ Reduced communication
  - ☑ Quick response to changes
  - ☑ Resilient to failures
- ⌘ Ad hoc distributed simulation is an approach to embedded on-line simulations in sensor networks
  - ☑ Applied to transportation management systems
  - ☑ Generalizing the approach
  - ☑ Applying to queueing systems



# Ad Hoc Distributed Simulations



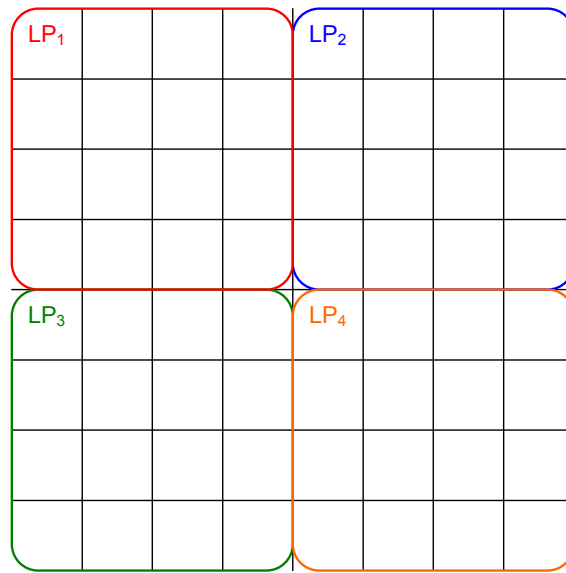
- ⌘ Ad hoc distributed simulation = {autonomous logical processes (LPs)} + space time memory (STM) + rollback mechanism



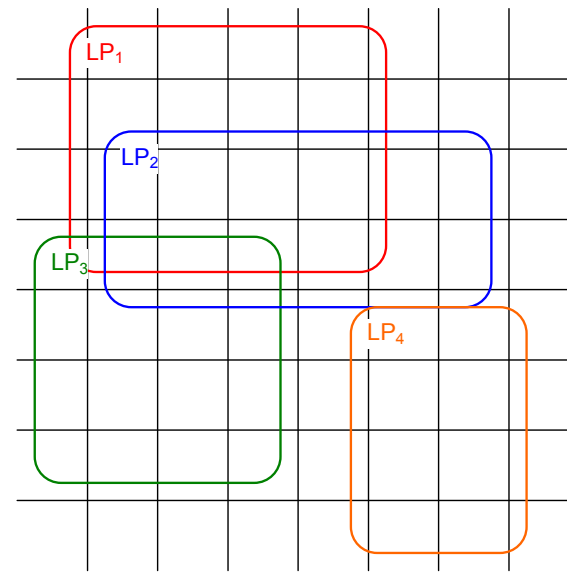
# Ad Hoc Distributed Simulations (*cont.*)



- ⌘ Each LP autonomously models a portion of the system under investigation
  - ⏏ Partitioning of the system is arbitrary (compared to conventional distributed simulations)
  - ⏏ The portion modeled by one LP may change over time



Conventional Approach

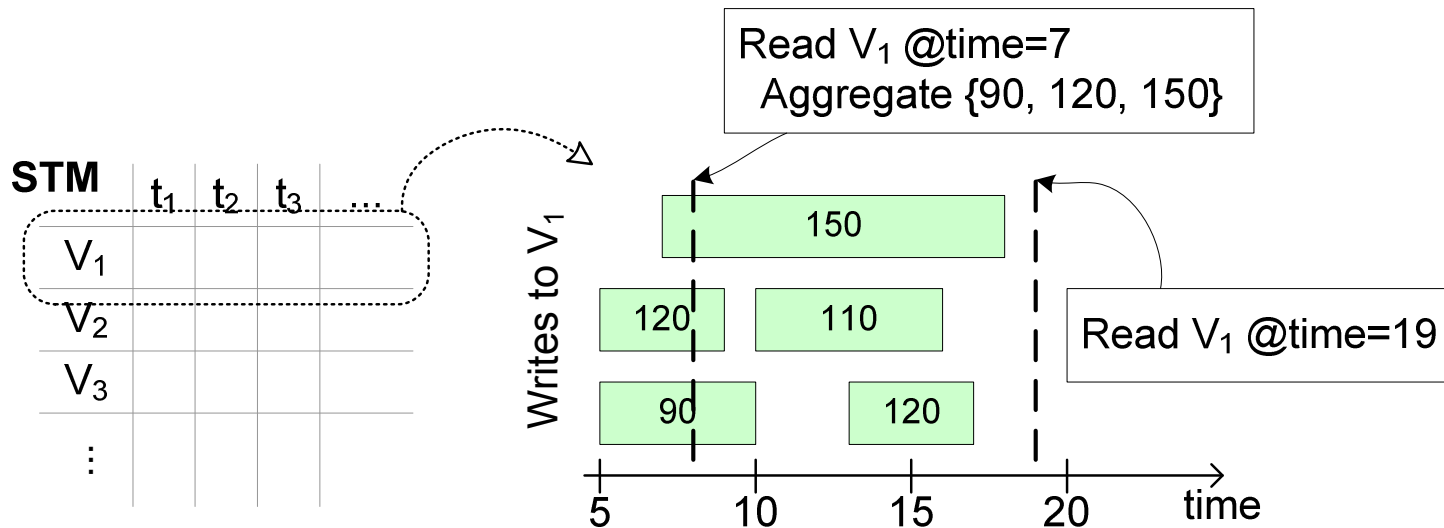


Ad Hoc Approach

# Ad Hoc Distributed Simulations (*cont.*)



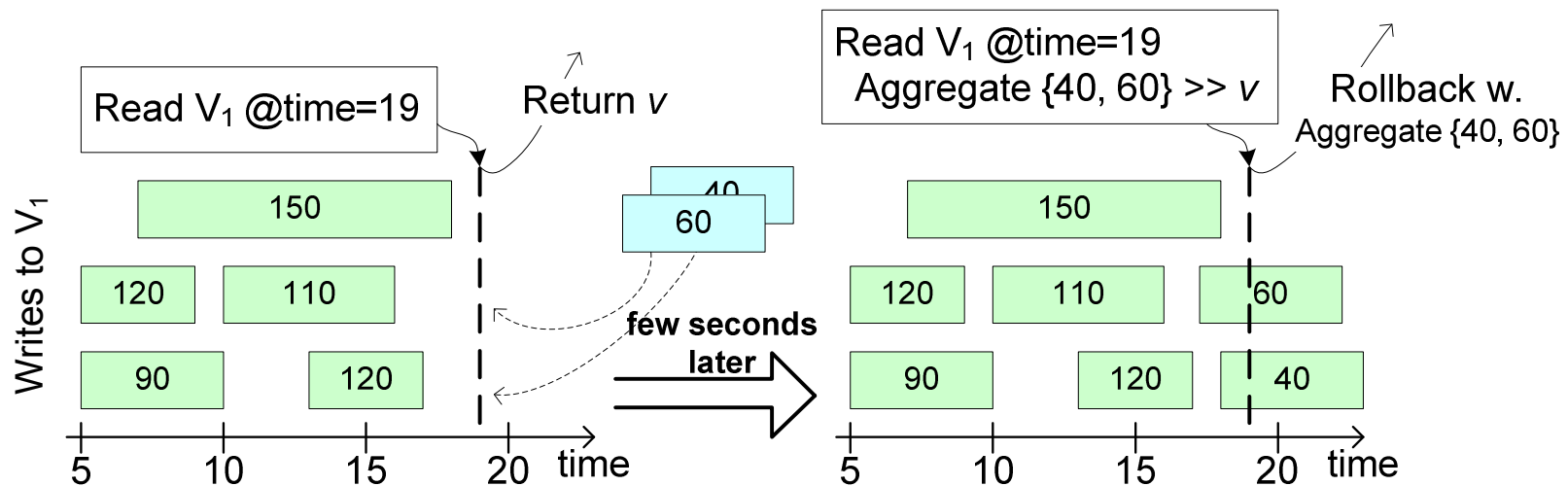
- ⌘ LPs exchange information via the STM
  - ⊞ STM holds time stamped updates of variables shared among LPs
  - ⊞ LPs update values with a time interval within which the update is valid
  - ⊞ LPs read values by specifying the desired variable and a time stamp
    - ⊞ Sufficient number of predictions
    - ⊞ Insufficient number of predictions



# Ad Hoc Distributed Simulations (*cont.*)



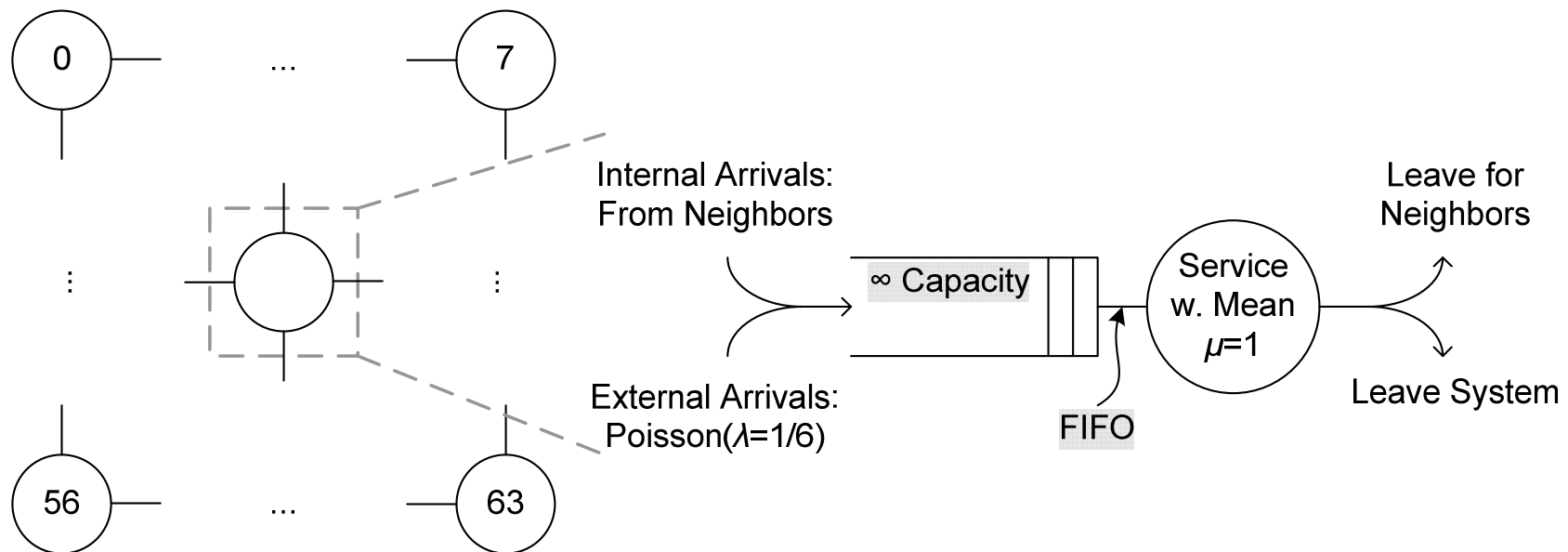
- ⌘ The rollback mechanism is to correct invalid input values used by LPs
  - ⊞ The rolled back LP rewinds its simulation time back to when the invalid value was used
  - ⊞ The LP restores its prior state
  - ⊞ The LP retracts updates that should be canceled out
  - ⊞ The LP restarts the simulation with new values



# Queueing Network Simulations



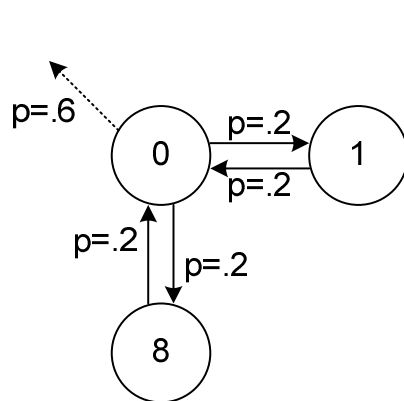
- ⌘ Queueing Networks are used to model a variety of industrial systems
- ⌘ The modeled open queueing network consists of 64 nodes arranged into an 8x8 rectangular configuration



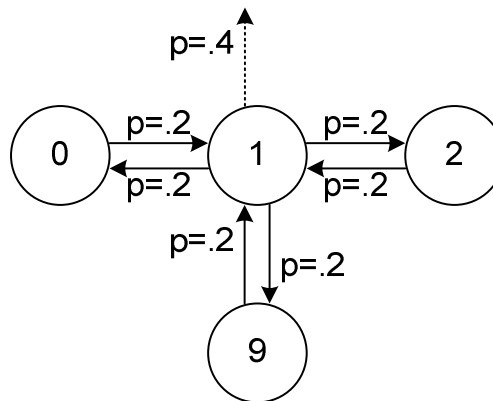
# Queueing Network Simulations (*cont.*)



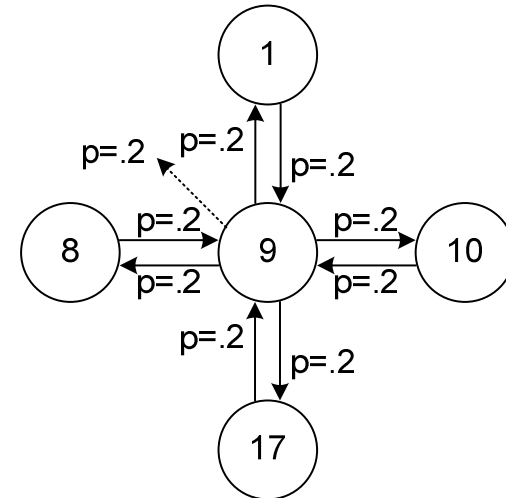
- ⌘ Served jobs either leave the system or move to any of the neighboring nodes; routing probabilities vary on node types



(a) Corner Node



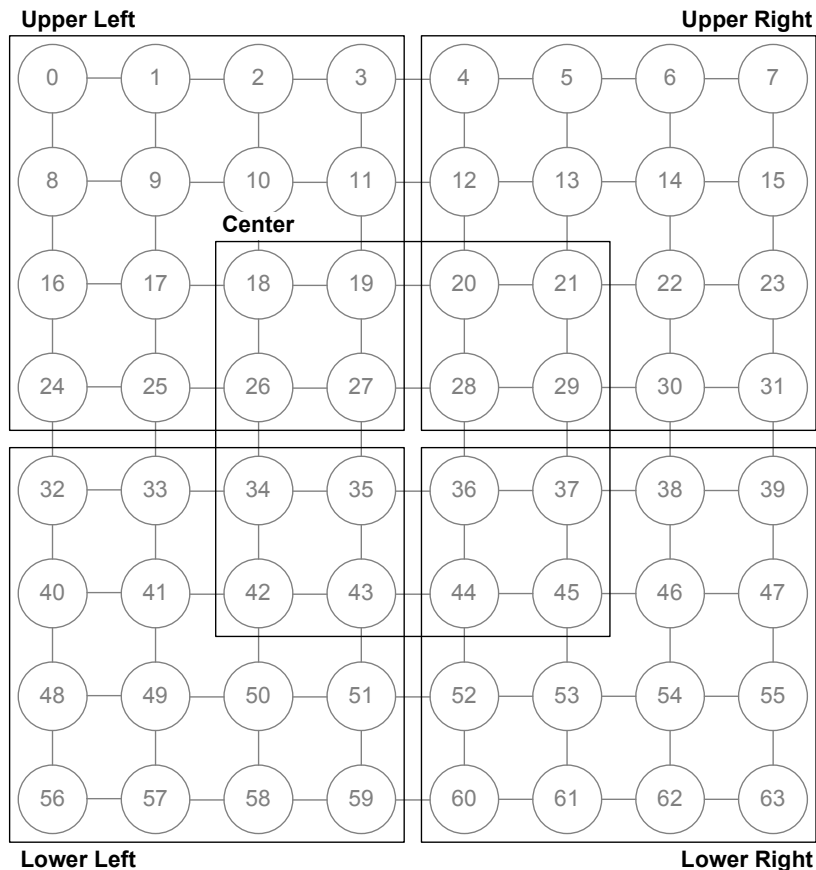
(b) Edge Node



(c) Interior Node

- ⌘ The modeled network is large with complex routings, various path cycles, and traffic intensities of the nodes range from 0.3 to 0.8

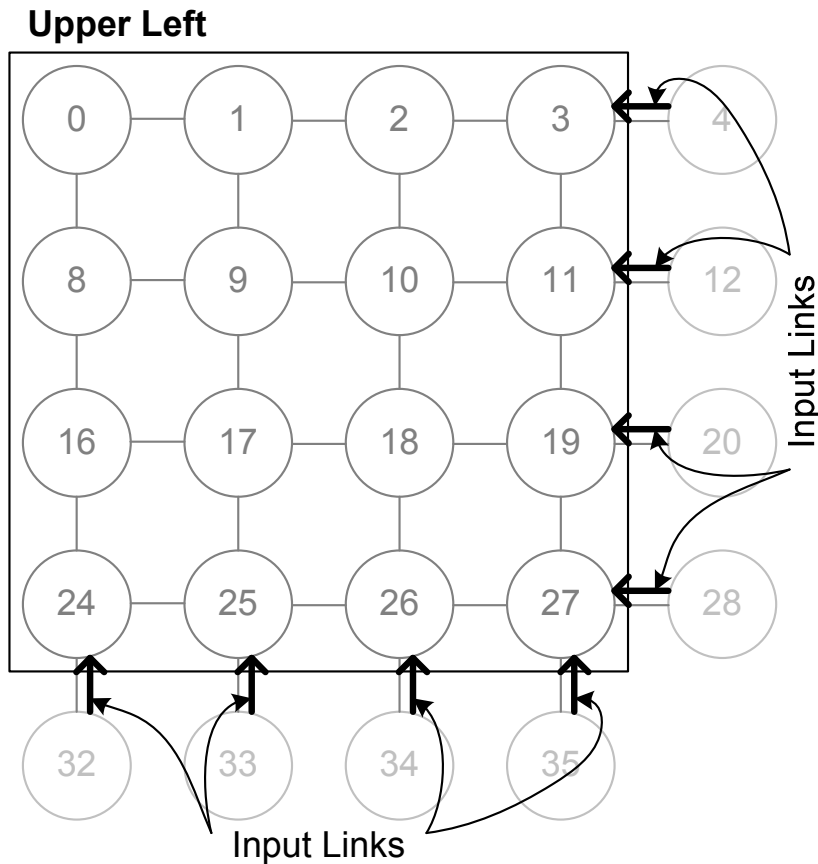
# Applying Ad Hoc Approach - Partitioning



⌘ A grid-like partitioning with significant overlapping among portions

- ⌘ Five portions, each with  $n_1$ ,  $n_2$ ,  $n_3$ ,  $n_4$ , and  $n_5$  LPs respectively
- ⌘  $\sum n_i = N$
- ⌘ Current design,  $n_1 = n_2 = n_3 = n_4 = n_5 = 8$  ( $N=40$ )

# Applying Ad Hoc Approach - Information Exchange



- ⌘ Arrivals on input links are modeled as renewal processes where distribution parameters are estimated using the information from the STM
  - ☒ No data at first
  - ☒ Set to Poisson( $\lambda=1/6$ )
- ⌘ Departure statistics of all links are written to the STM
  - ☒ Number of observed inter-departure times
  - ☒ First two moments of these inter-departure times
  - ☒ Denoted  $(n_{t,l}, m_{1,t,l}, m_{2,t,l})$

# Applying Ad Hoc Approach - Information Exchange (*cont.*)



- ⌘ Simulation starts in empty and idle state; information exchange begins after  $w$  seconds in simulation time
- ⌘ Every  $d_{\text{read}}$  seconds, each LP “reads” the statistics of the input links
  - ☒ Reuse last statistics (i.e., the one corresponding to  $d_{\text{read}}$  seconds earlier)
  - ☒ Inform the STM of the usage
  - ☒ Be rolled back if the STM finds the statistics invalid
- ⌘ Every  $d_{\text{write}}$  seconds, each LP writes statistics of all modeled links
- ⌘ The observation period is over the last  $d_{\text{observe}}$  seconds with the rolling window mechanism, preventing the statistics from being sensitive to random fluctuations
- ⌘ Current design,
  - ☒  $w=300$  (seconds)
  - ☒  $d_{\text{read}}=d_{\text{write}}=30$  (seconds)
  - ☒  $d_{\text{observe}}=300$  (seconds)

# Applying Ad Hoc Approach - Arrival Process Approximation



- ⌘ When the service times follow exponential distribution, the arrival processes on the input links are approximated as Poisson processes
- ⌘ When the service times are non-exponential, the arrival processes are approximated by renewal processes with gamma inter-arrival times<sup>1,2</sup>
- ⌘ The parameters (for either case) are estimated using the method of moments
  - ☒ Let  $m_1$  and  $m_2$  be the first two moments
  - ☒ In former case, the rate parameter  $\lambda$  follows

$$\hat{\lambda} = \frac{1}{m_1}$$

- ☒ In later case, the shape parameter  $\alpha$  and the scale parameter  $\beta$  follows

$$\hat{\alpha} = \frac{m_1^2}{m_2 - m_1^2}, \quad \hat{\beta} = \frac{m_2 - m_1^2}{m_1}$$

# Applying Ad Hoc Approach - Rollback Mechanism & Data Aggregation



- ⌘ Few predictions may bias the aggregated value; the rollback mechanism is applied if there are at least  $k$  predictions (i.e.,  $k$  triples of  $(n_{t,l}, m_{1,t,l}, m_{2,t,l})$  with the same  $t$  but different  $l$ )
- ⌘ The rollback detection function is invoked when
  - ☒ An LP informs the statistics it is using
  - ☒ Statistics are written to the STM
  - ☒ Statistics are deleted (retracted) from the STM
- ⌘ First, the rollback detection function requests an aggregated value calculated from the pooled-averaging approach

$$\hat{m}_{1,t} = \frac{\sum_l n_{t,l} \times m_{1,t,l}}{\sum_l n_{t,l}} \quad \hat{m}_{2,t} = \frac{\sum_l n_{t,l} \times m_{2,t,l}}{\sum_l n_{t,l}}$$

# Applying Ad Hoc Approach - Rollback Mechanism & Data Aggregation



- ⌘ Second, the difference between the first moments are considered; a rollback is triggered if the relative difference is greater than  $\varepsilon_{\text{relative}}$
- ⌘ Third, if a rollback is necessary, another aggregated value is requested as new input to the LP being rolled back
  - ☒ The random sampling and pooled-variance approach is adopted

$$\hat{m}_{1,t} = \text{RandUni}(\{m_{1,t,l}\}) + h\varepsilon$$

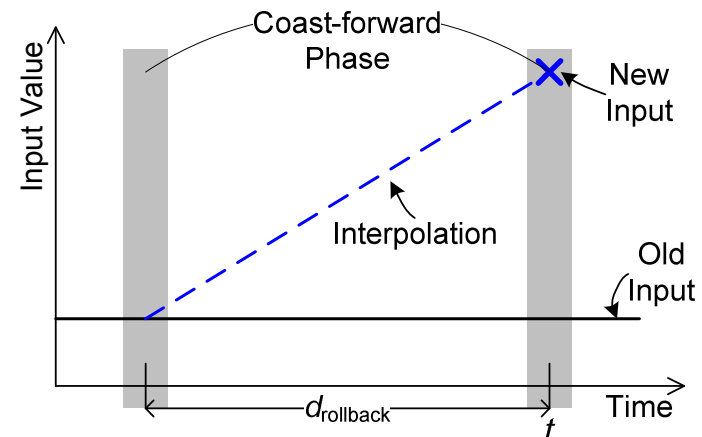
$$\hat{m}_{2,t} = \frac{\sum_l n_{t,l} - 1}{\sum_l n_{t,l}} \times \frac{\sum_l (n_{t,l} - 1) S_{t,l}^2}{\sum_l (n_{t,l} - 1)} + \hat{m}_{1,t}^2, \text{ with } S_{t,l}^2 = \frac{n_{t,l} (m_{2,t,l} - m_{1,t,l}^2)}{n_{t,l} - 1}$$

# Applying Ad Hoc Approach - Rollback Mechanism & Data Aggregation



- ⌘ When an LP is rolled back to simulation time  $t$ ,
  - ⊞ Retracts its read and write operations associated to time greater than or equal to  $t$
  - ⊞ Rewinds further back to  $t - d_{\text{rollback}}$  so that the mismatches do not result in abrupt changes in output statistics
    - ⊞ No writes to the STM during the  $d_{\text{rollback}}$ -second coast-forward phase
    - ⊞ Interpolating the input values during the coast-forward phase
      - Linear interpolation on first moments
      - Linear interpolation on coefficients of variation to calculate second moments

- ⌘ Current design,
  - ⊞  $k=8$
  - ⊞  $\epsilon_{\text{relative}} = 10\%$
  - ⊞  $d_{\text{rollback}} = 300$  (seconds)



# Experiments



⌘ We differ service time distribution to construct three scenarios

☒ Scenario 1: Exponential( $\lambda=1$ )

☒ Rate parameter:  $\lambda$

☒ Mean: 1

☒ Variance: 1

☒ Scenario 2: Gamma( $\alpha=2, \beta=0.5$ )

☒ Shape parameter:  $\alpha$

☒ Scale parameter:  $\beta$

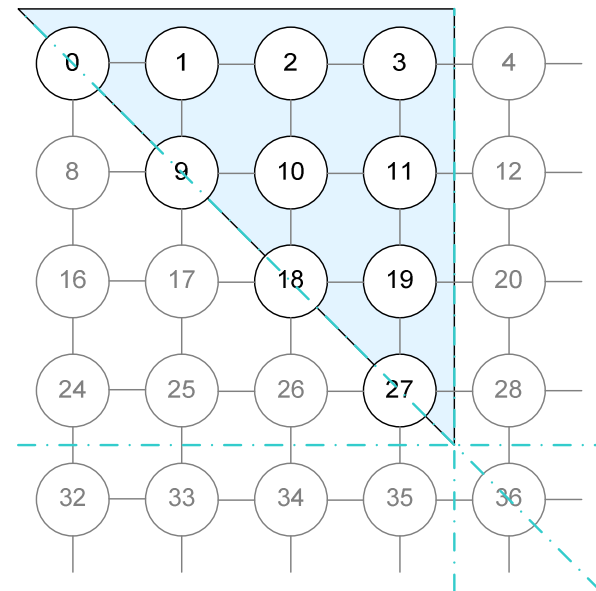
☒ Mean: 1

☒ Variance: 0.5

☒ Scenario 3: Gamma(0.25, 4)

☒ Mean: 1

☒ Variance: 4



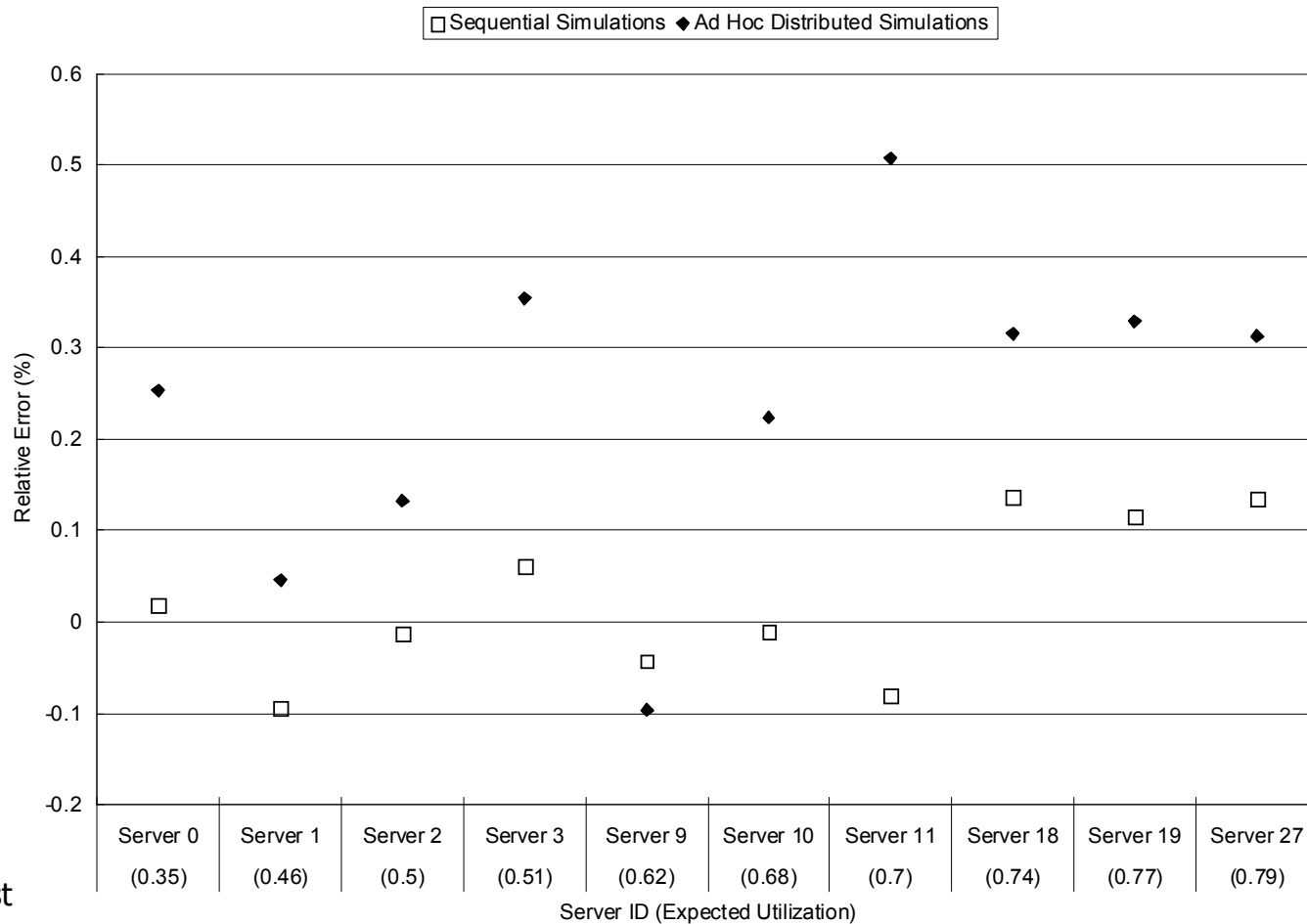
⌘ We are interested in the steady-state mean utilization and queue length of each node

☒ The data are collected after one hour in simulation time and the collection lasts for another one hour

# Results - Scenario 1



Relative Errors of Utilizations Based on 100 IID Runs under Scenario 1

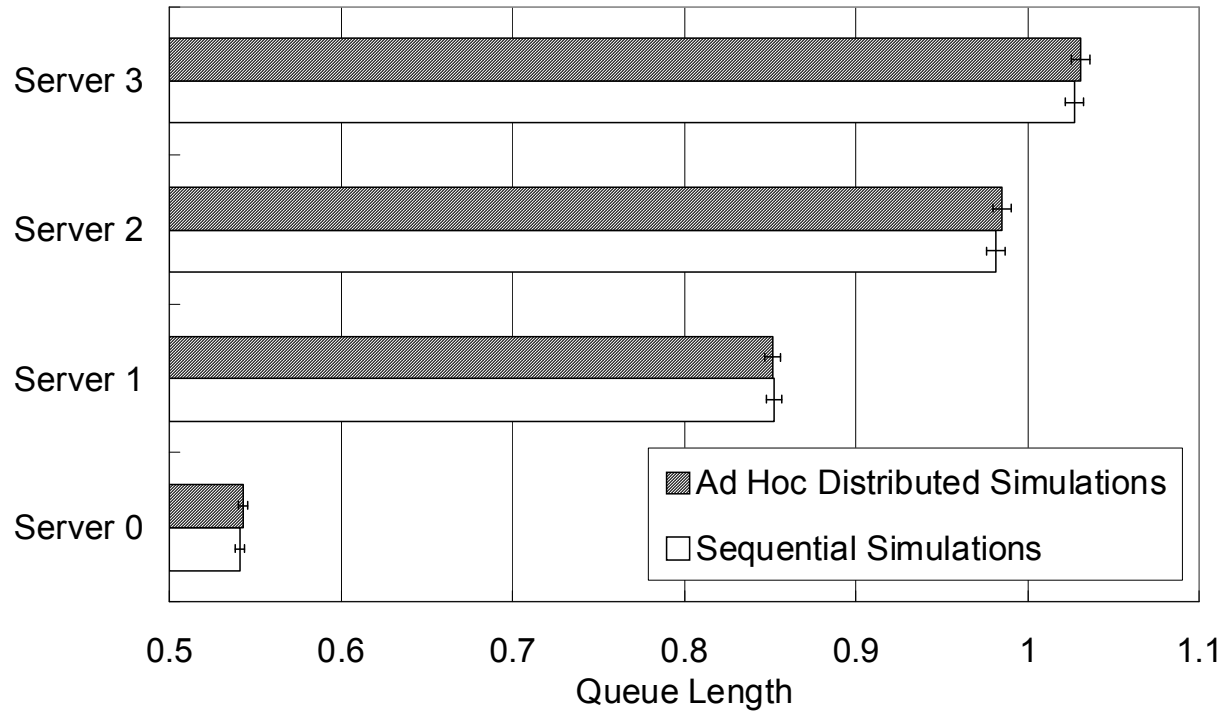


**Relative error:**  
 compared against  
 analytical solutions

# Results - Scenario 1 (*cont.*)



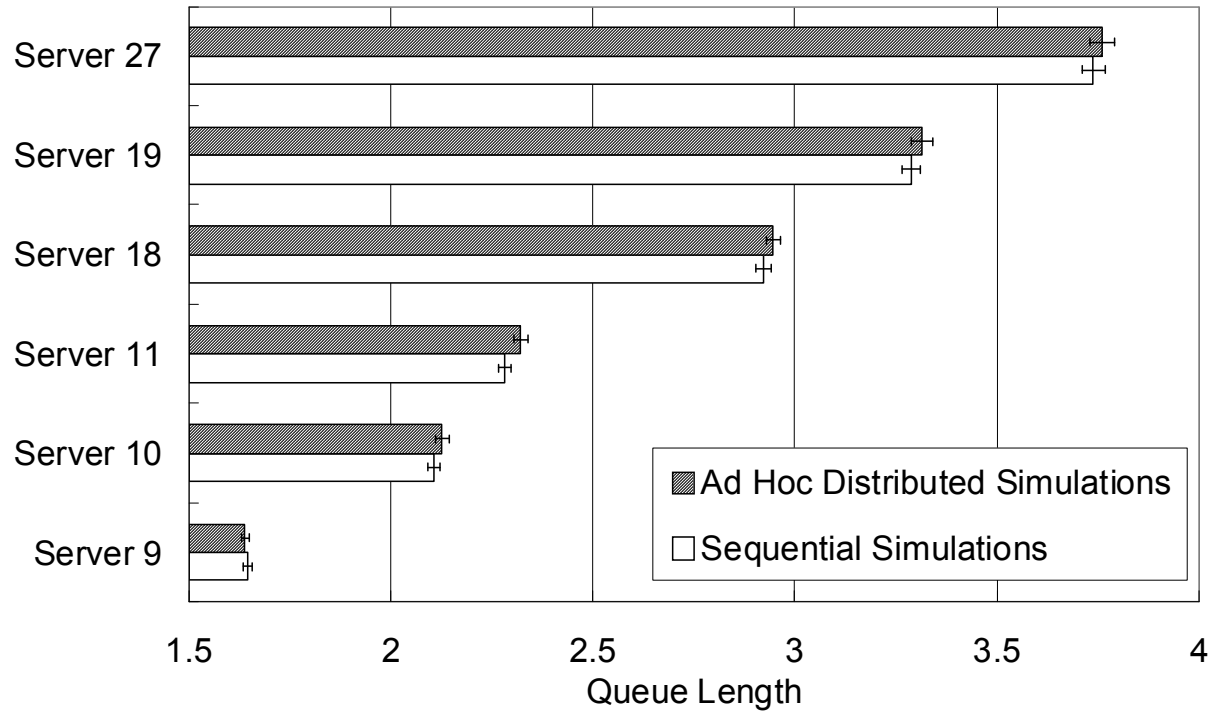
Point Estimates and 90% Confidence Intervals for Steady-state Mean Queue Length Based on 100 IID Runs under Scenario 1



# Results - Scenario 1 (*cont.*)



Point Estimates and 90% Confidence Intervals for Steady-state Mean Queue Length Based on 100 IID Runs under Scenario 1



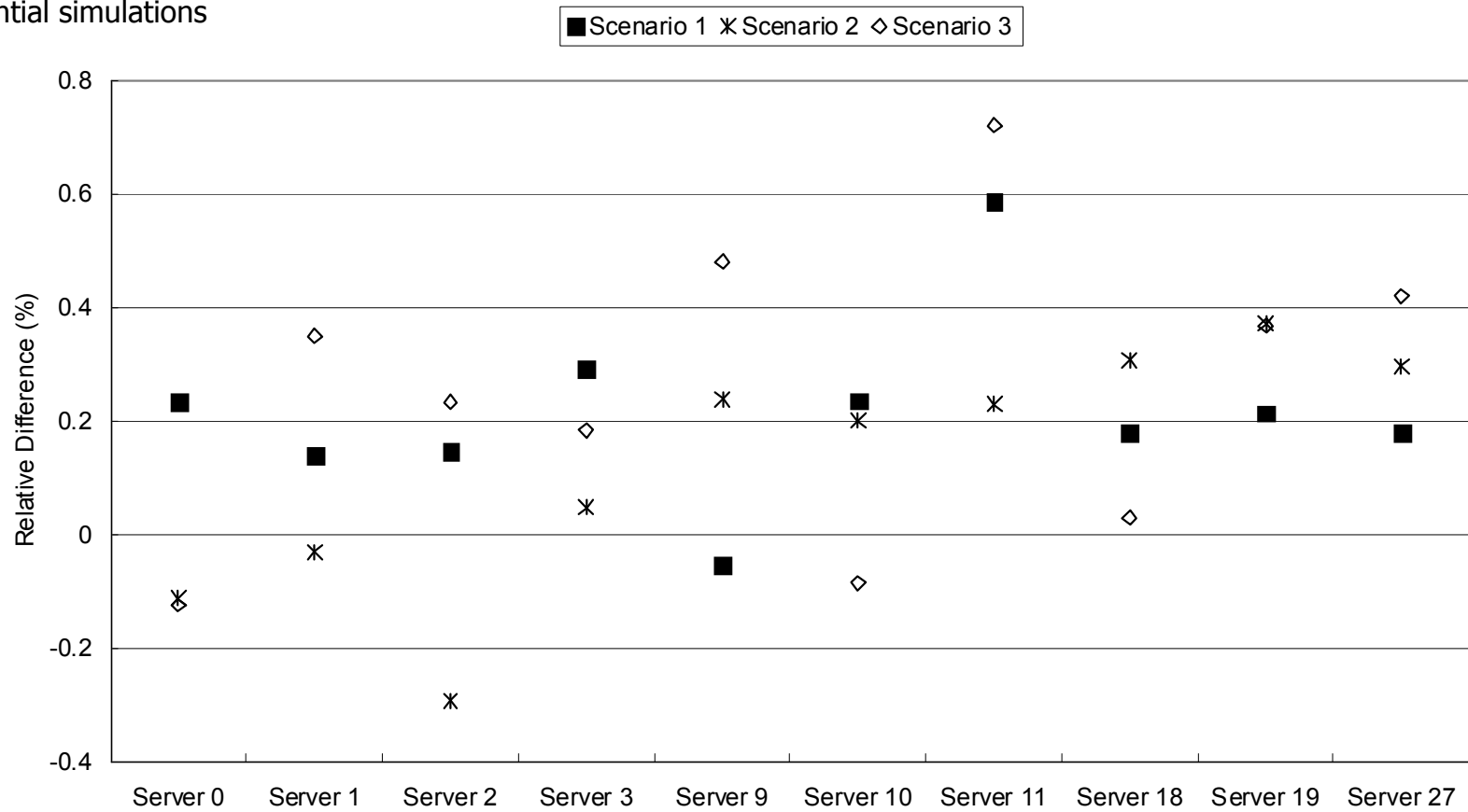
# Results - Utilization



## Relative difference:

compared against  
sequential simulations

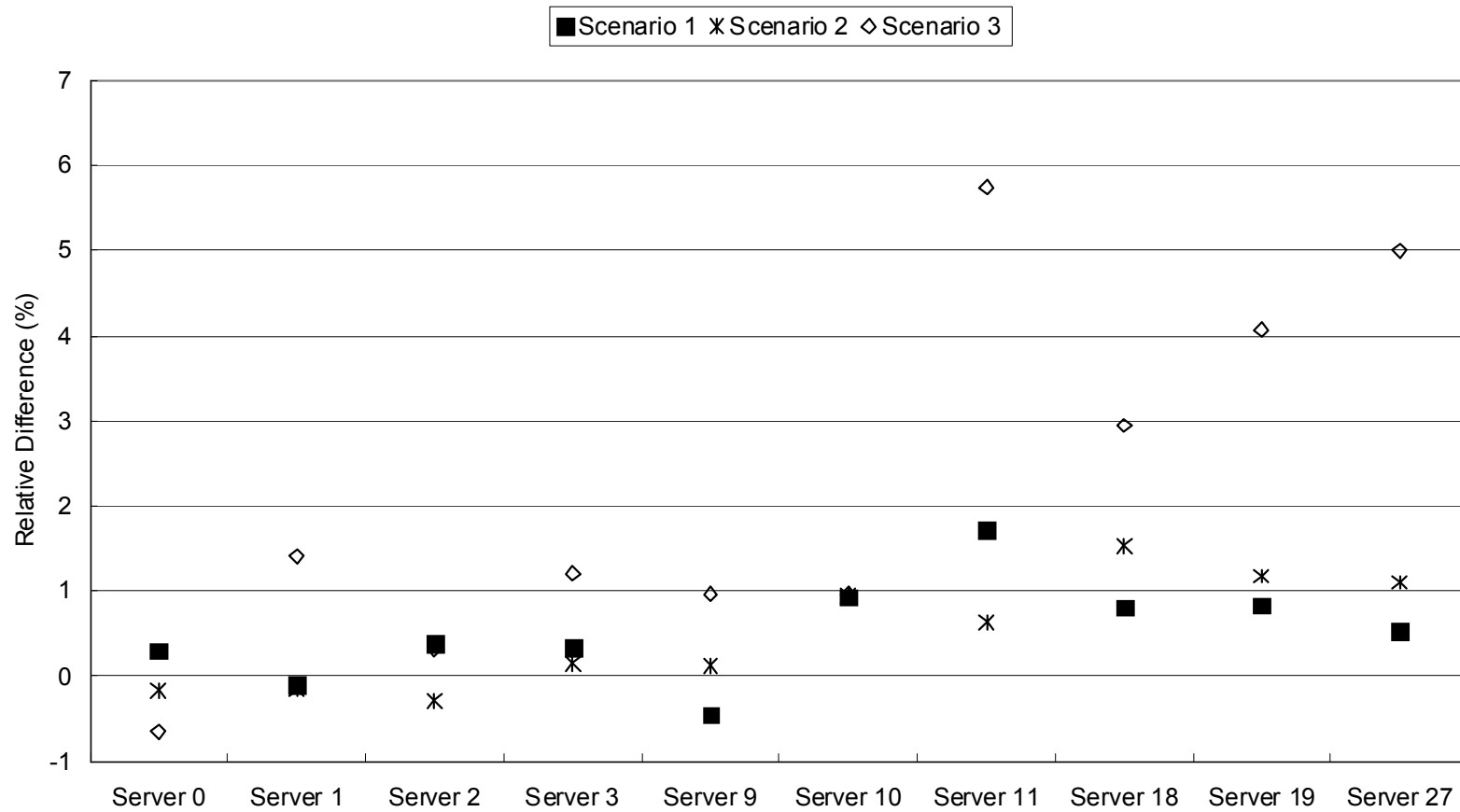
Relative Differences of Utilizations Based on 100 IID Runs



# Results - Queue Length



Relative Differences of Queue Lengths Based on 100 IID Runs



# Conclusions and Future Works



- ⌘ We generalized the ad hoc distributed simulation approach
- ⌘ We applied it to a queueing network simulation to show its capabilities and some weakness
  - ⊞ Work comparably well compared to sequential simulations especially when the variation of service times is small
  - ⊞ Reveal some issues when the variation of service time is large, an area of future work
- ⌘ Future works include
  - ⊞ Relaxing restrictions (e.g., fixed partitioning)
  - ⊞ Examining response to unexpected changes in sensor measurements
  - ⊞ Evaluating resilience to failures and errors in the underlying sensor network

# Questions?



# Appendix



# Results - Scenario 1



- ⌘ Relative errors of utilizations based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
0.25 %	0.05 %	0.13 %	0.35 %	-0.10 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.22 %	0.51 %	0.32 %	0.33 %	0.31 %

- ⌘ Relative errors of queue lengths based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
0.45 %	-0.18 %	0.25 %	0.58 %	-0.40 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.98 %	1.53 %	1.37 %	1.69 %	1.15 %

# Results - Scenario 1



- ⌘ Relative differences of utilizations based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
0.23 %	0.14 %	0.14 %	0.29 %	-0.05 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.24 %	0.59 %	0.18 %	0.21 %	0.18 %

- ⌘ Relative differences of queue lengths based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
0.30 %	-0.09 %	0.39 %	0.33 %	-0.45 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.93 %	1.72 %	0.81 %	0.83 %	0.53 %

# Results - Scenario 2



- ⌘ Relative differences of utilizations based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
-0.11 %	-0.03 %	-0.29 %	0.05 %	0.24 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.20 %	0.23 %	0.31 %	0.37 %	0.30 %

- ⌘ Relative differences of queue lengths based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
-0.17 %	-0.16 %	-0.29 %	0.16 %	0.13 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.94 %	0.63 %	1.53 %	1.17 %	1.10 %

# Results - Scenario 3



- ⌘ Relative differences of utilizations based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
-0.12 %	0.35 %	0.23 %	0.18 %	0.48 %
Server 10	Server 11	Server 18	Server 19	Server 27
-0.09 %	0.72 %	0.03 %	0.37 %	0.42 %

- ⌘ Relative differences of queue lengths based on 100 IID runs using ad hoc approach

Server 0	Server 1	Server 2	Server 3	Server 9
-0.65 %	1.41 %	0.32 %	1.21 %	0.96 %
Server 10	Server 11	Server 18	Server 19	Server 27
0.97 %	5.74 %	2.95 %	4.07 %	5.01 %