

# Performance of Computer System

## Evaluation Technique – Simulation

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# How You Should See the Problem

- What is system?
- Evaluation techniques
- Metrics to use
- Workload
- Parameters

- SIMULATION AND RELATED ISSUES

# Why simulation?

- System under study may not be available
  - ▶ Common in design and proving stages
- Simulation may be preferred alternative to measurement
  - ▶ Controlled study of wider range of workloads and environments
- Higher accuracy results than analytical modeling

Sometime simulation is not preferred

- Accurate simulation models take a long time to develop
  - ▶ Typically the evaluation strategy that takes the longest

# Evaluation techniques — Simulation

- Easy to modify and update
- Due to the cost of changing system configurations, simulation is preferred beforehand.
- With simulations, it may be possible to identify the optimal combination, but often it is not clear what the trade-off is among different parameters.
- Time consuming for a comprehensive simulation environment development, off the shelf simulations tools:
  - ▶ Smart phone application simulator/emulator: Android Studio
  - ▶ Networking simulator: NS-3, Matlab
  - ▶ Transportation simulation: Simio

# Good Tips for Evaluation

- Combining evaluation techniques is useful
  - ▶ Analytical model: find interesting range of parameters
  - ▶ Simulation: study performance within parameter range
- Until validated, all evaluation results are suspect!
  - ▶ Always validate one analysis modality with another
  - ▶ Beware of counterintuitive results!

# Common Mistakes: Too Much Detail

- Level of detail limited only by time available for development
- A detailed model may not be a better model
- Recipe for success
  - ▶ Start with less-detailed model
  - ▶ Get some results
  - ▶ Study sensitivities
  - ▶ Introduce details in key areas that affect results most

# Common Mistakes: Initial Conditions

- Initial part of a simulation is generally not representative
  - ▶ Transient behaviour rather than steady state
- Initial part of simulation should be discarded
  - ▶ Several techniques for identifying beginning of steady state



# Common Mistakes: Too Short Simulations

- Simulation run times are often very long
- Temptation is to halt simulations ASAP
- However
  - ▶ Results may be heavily dependent on initial conditions
  - ▶ may not be representative of a real system until steady state
- Correct length for simulations depends on
  - ▶ Accuracy desired (width of confidence intervals)
  - ▶ Variance of observed quantities

# Common Mistakes: Bad Random Numbers

- Bad random numbers can pollute simulation results
  - ▶ Period too short
  - ▶ Assume global randomness = local randomness
  - ▶ Rely on bit subsets: may not be as random as whole
- Use well-known generator rather than rolling your own

# Simulation Types

- Monte Carlo Simulation
- Trace-Driven Simulation
- Discrete-Event Simulations

# Monte Carlo Simulation

- Model probabilistic phenomenon that do not change over time
- It is used for evaluating nonprobabilistic expressions using probabilistic methods.

# Trace-Driven Simulation

- Trace = time ordered record of events on real system
- Advantages
  - ▶ Credibility, Easy Validation, Accurate Workload, Detailed Trade-offs, Less Randomness, Fair Comparison, Similarity to the Actual Implementation.
- Disadvantages
  - ▶ Complexity, Representativeness, Finiteness, Single Point of Validation, Detail, Trade-off

# Trace is different from Statistics

- Statistics (requests, errors, latency, etc.) are calculated based on the full volume of traces
- Examples:
  - ▶ Total requests and requests per second
  - ▶ Total errors and errors per second
  - ▶ Latency
  - ▶ Breakdown of time spent by service/type

# Discrete-Event Simulations

- Discrete-event simulations use discrete-state model of system, the simulation will have the following components
  - ▶ Event Scheduler: A list of events
  - ▶ Simulation Clock and a Time-advancing Mechanism: Absolute time and Relevant time
  - ▶ System State Variables: Global variables that describe the state of the system
  - ▶ Event Routines: Each event is simulated by its routine
  - ▶ Input Routines: Input routines typically allow a parameter to be varied in a specified manner
  - ▶ Report Generator: At the end of the simulation
  - ▶ Initialisation Routines: Initial state of the system state
  - ▶ Trace Routines: Print out intermediate variables as the simulation proceeds for debugging
  - ▶ Dynamic Memory Management: Normally automatically
  - ▶ Main Program: Brings all the routines together

# Discrete-Event Simulations

- Mostly used in Computer Sciences
  - ▶ Time driven system: auto-sync programs, alert systems, some street or home lighting systems
  - ▶ Event driven system: Almost every computer systems



# What Simulation to Use

- To model destination address reference patterns in a network traffic, given that the pattern depends upon a large number of factors.
- To model scheduling in a multiprocessor system, given that the request arrivals have a known distribution.
- To determine the value of  $\pi$

# Good Simulations

- Measuring goodness
  - ▶ Validation: are assumptions reasonable?
  - ▶ Verification: does model implement assumptions correctly?

<b>invalid, unverified</b>	<b>invalid, verified</b>
<b>valid, unverified</b>	<b>valid, verified</b>

Correctly implements bad assumptions  
Incorrectly implements good assumptions  
Correctly implements good assumptions

# Strategies for Avoiding Bugs – For Validation

- What to check?
  - ▶ Assumptions
  - ▶ Input parameter values and distributions
  - ▶ Output values and conclusions
- How to check?
  - ▶ Expert intuition: most common and practical
  - ▶ Measurements of real system
  - ▶ Theoretical results, e.g. queueing model

# Strategies for Avoiding Bugs – For Verification

- Software engineering
  - ▶ Top-down design: layered (hierarchical) system structure
  - ▶ Modularity: well-defined interfaces, unit testing
- Assertions to check invariants
  - ▶ No. packets received = No. packets sent - No. packets lost - No. in flight
- Structured walk through
- Simplified test cases with easily analysed results

# Transient Removal

- Transient state: prefix of simulation before steady state
- Steady state performance is usually that of interest
- Heuristic approaches for removing transient state
  - ▶ Long runs
  - ▶ Proper initialisation
  - ▶ Truncation
  - ▶ Moving average of independent replications
  - ▶ Initial data deletion
  - ▶ Batch means

# Long runs

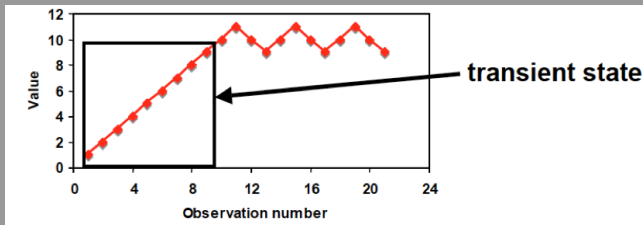
- Long run = steady state results long enough to dominate effects of initial transients
- Disadvantages
  - ▶ wastes resources (computer time and real time)
  - ▶ difficult to ensure length of run is “long enough”
- **Avoid this method if you have other choice**

# Proper initialisation

- Proper initialisation = starting simulation in state close to expected steady state
- Examples:
  - ▶ start CPU scheduling simulation with non-empty job queue
  - ▶ start WWW cache trace-driven simulation with most frequently referenced files in cache

# Truncation

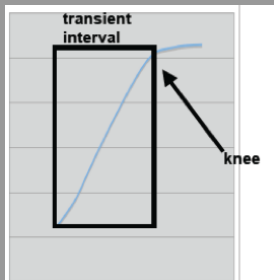
- Assumption: variability of steady state  $\neq$  transient state
- Truncation algorithm:  
input:  $n$  observations  $x_1, x_2, \dots, x_n$   
for  $k = 2, n$   
     $\min_k = \min (x_k, \dots, x_n)$   
     $\max_k = \max (x_k, \dots, x_n)$   
    if  $\min_k \neq x_k \ \&\& \ \max_k \neq x_k$ , break  
post condition: if  $k \neq n$  then  $k - 1 = \text{length of transient state}$





# Moving Average of Independent Replications

- Compute mean trajectory by averaging across replications
- Find the knee in the curve.
  - for  $k = 1$  to  $n$ 
    - plot trajectory of moving average of successive  $2k+1$  values
    - if trajectory is “sufficiently smooth”, break
- The knee gives the length of the transient phase

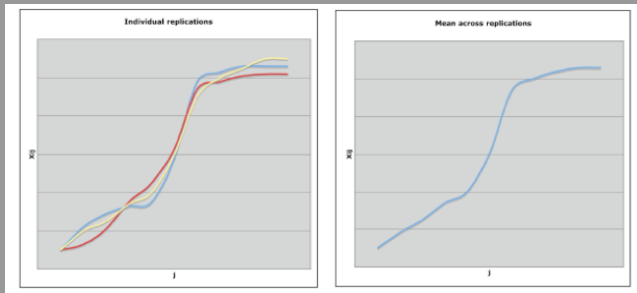


# Initial Data Deletion I

- Compute average after some of initial observations omitted
- During steady state average does not change as much as additional observations are deleted
- Problem
  - ▶ Randomness in observations causes average to change even in steady state
- Solution
  - ▶ Average across several replications

# Initial Data Deletion II

- 1 Compute mean trajectory by averaging across replications
- 2 Compute overall mean
- 3 Find knee in a curve showing the relative change in overall mean



# Initial Data Deletion III

Find knee in a curve showing the relative change in overall mean  
for  $k = 1, n - 1$

assume transient state is of length  $k$

delete first  $k$  observations from mean trajectory

compute overall mean from remaining  $n - k$  values:

$$\bar{\bar{x}} = \frac{1}{n-k} \sum_{j=k+1}^n \bar{x}_j$$

compute relative change in overall mean:

$$\text{Relative change} = \frac{\bar{\bar{x}}_k - \bar{\bar{x}}}{\bar{\bar{x}}}$$

# Batch Means

- Run a very long simulation
- Afterward, divide it into several parts of equal duration
- Each part is a batch
- Batch mean = mean of observations in each batch

# Evaluation Examples

Criterion	Analytical Modeling	Simulation	Measurement
Stage	any	any	post-prototype
Time required	small	medium	varies
Tools	analysts	programs	instrumentation
Accuracy	low	moderate	varies
Trade-off evaluation	easy	moderate	difficult
Cost	low	medium	high
Saleability	low	medium	high

