

Performance Modeling – Part III

Simulation

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Components of a Simulation Model

- Event Generation:
 - Trace-driven
 - Distribution-driven
 - Hybrid
- Event Processing
 - Calendar of Events
 - Event-handling procedures
- Transaction List (with parameters)
- Queues
- Simulation Clock
- Computation of Statistics

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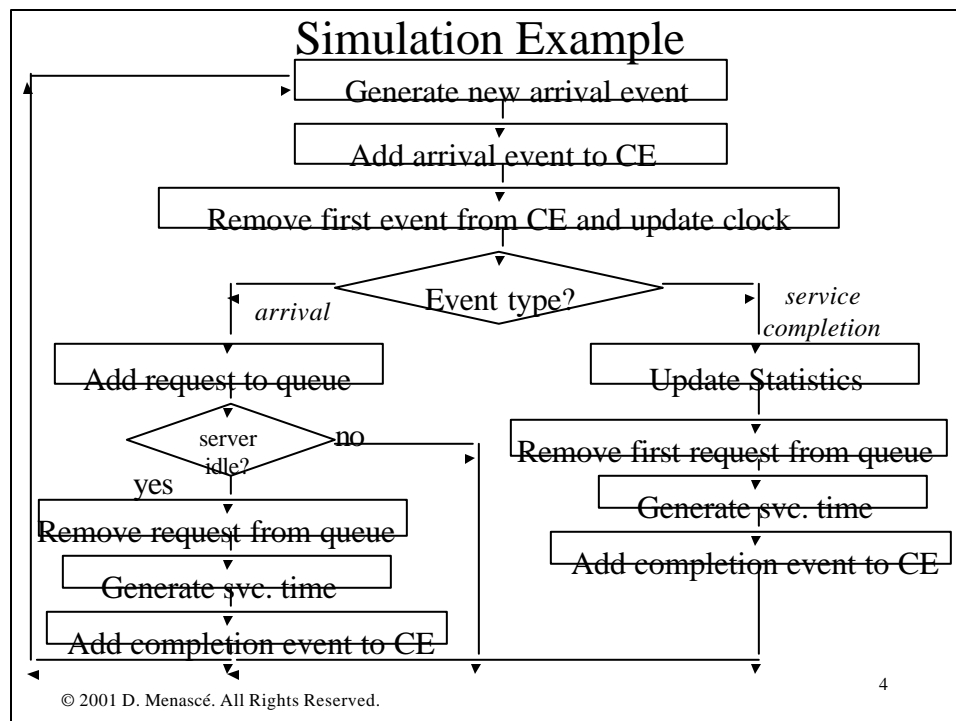
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Simulation Model Example: Single Queue

- Events:
 - Arrival of a customer
 - Service completion
- Statistics:
 - Total number of arrivals
 - Total departures
 - Total server busy time
 - Total waiting time
 - Total departures from queue
 - Total squares of waiting time

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Calendar of Events

Event Type	Event Time	Event Parameters
arrival	10.5
arrival	12.8
completion	13.1
...

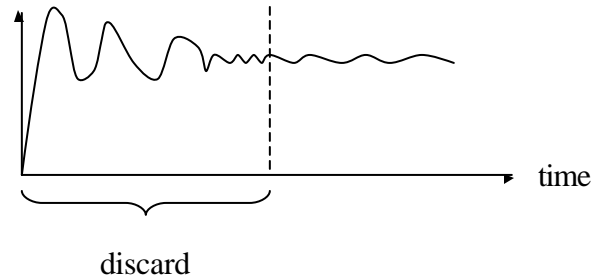
- The calendar of events is ordered in increasing chronological order.
- Parameters may include the transaction Id associated with the event.

Common Mistakes in Simulation

- Inappropriate level of detail:
 - Too detailed: more development time and higher likelihood of bugs
 - Should start with a less detailed model first and increase complexity as needed.
- Unverified Models:
 - Simulation programs are usually large and complex programs and may have bugs that invalidate the results.
- Invalid Models:
 - Incorrect assumptions may be used. Need to validate through analytic models, measurements, and or intuition.

Common Mistakes in Simulation

- Improperly Handled Initial Conditions:
 - Should discard first part of run: transient behavior.



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Common Mistakes in Simulation

- Improper simulation length.
- Poor Random Number Generator.
- Improper Selection of Seeds.

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Verifying Simulation Models

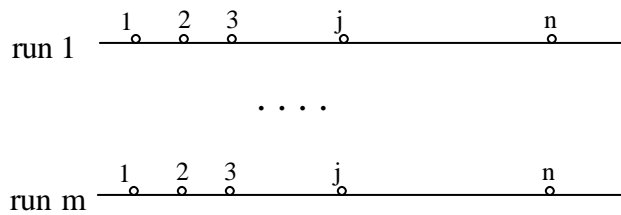
- Trace Analysis: examine traces of a few transactions as they go through the system.
- Continuity Test: small variations in the input should show small variations in the output.
- Check Extreme Values: extreme values (e.g., low loads or very high loads) should be easy to verify by crude analytic models.

Verifying Simulation Models

- Check for Basic Relationships: verify if results satisfy basic laws (e.g., Little's Law).
- Bound validation: use, if possible, existing analytic models for situations that are known to be upper or lower bounds
- Trend verification: check if the trends shown by the model match your intuition.
- Numeric range validation: check if the numerical results are within expected numerical ranges.

Transient Elimination with Independent Runs

- Run m runs of the simulation with a different seed for each run.
- Each run has n observations.
- Let $x_{i,j}$ be the j -th observation in the i -th run.



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Transient Elimination with Independent Runs

Step 1: compute average of j -th observation over all runs.

$$\bar{x}_j = \frac{1}{m} \sum_{i=1}^m x_{i,j}$$

Step 2: compute the overall average.

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

Step 3: Set the number of deleted observation, k , equal to 1.

Step 4: Compute the overall mean without the first k observations.

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

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Transient Elimination with Independent Runs

Step 5: compute the relative change Δ

$$\Delta = \frac{\bar{\bar{x}}_k - \bar{\bar{x}}}{\bar{\bar{x}}}$$

Step 6: If $|\Delta| > \text{tolerance}$ then do $k \leftarrow k + 1$ and go to step 4.

Step 7: Remove the first k observations and use $\bar{\bar{x}}_k$ as the average.

Transient Elimination with Batch Means

- Single run with N observations.
- Divide the run into m sub-samples called batches of size $n = \lfloor N/m \rfloor$.
- Let $x_{i,j}$ be the j -th observation in the i -th batch.

Transient Elimination with Batch Means

Step 1: Set $n = 10$.

Step 2: compute the average of the i -th batch.

$$\bar{x}_i = \frac{1}{n} \sum_{j=1}^n x_{i,j}$$

Step 3: compute the overall average.

$$\bar{\bar{x}} = \frac{1}{m} \sum_{i=1}^m \bar{x}_i$$

Step 4: Compute the variance of the batch means:

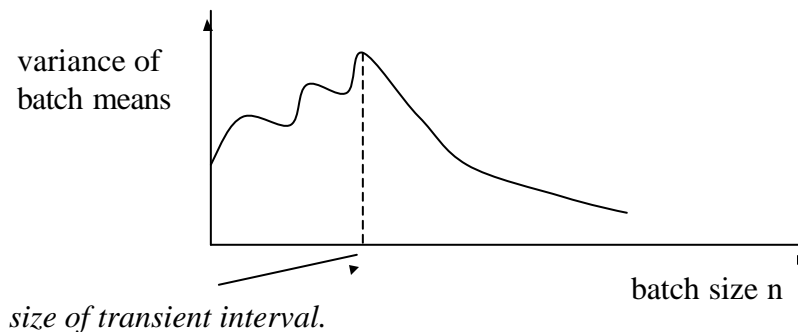
$$Var(\bar{x}) = \frac{1}{m-1} \sum_{i=1}^m (\bar{x}_i - \bar{\bar{x}})^2$$

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Transient Elimination with Batch Means

Step 5: Increase n by 10 and repeat steps 2-4 and plot the variance as a function of n . The point at which the variance starts to decrease is the length of the transient interval.



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Stopping Criteria Independent Runs

- Run m runs of the simulation with a different seed for each run.
- Each run has $n + n_o$ observations where n_o is the size of the transient phase.

Step 1: compute the mean for each replication.

$$\bar{x}_i = \frac{1}{n} \sum_{j=n_o+1}^n x_{i,j}$$

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Stopping Criteria Independent Runs

Step 2: compute the overall mean for all replications.

$$\bar{\bar{x}} = \frac{1}{m} \sum_{i=1}^m \bar{x}_i$$

Step 3: compute the variance of the replicate means.

$$\text{Var}(\bar{x}) = \frac{1}{m-1} \sum_{i=1}^m (\bar{x}_i - \bar{\bar{x}})^2$$

Step 4: compute the confidence interval for the mean as:

$$\bar{\bar{x}} \pm z_{1-\alpha/2} \text{Var}(\bar{x})$$

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Stopping Criteria Independent Runs

- Number of discarded observations: $m \times n_o$
- To reduce the number of wasted observations use a small value of m . If $m < 30$ use $t_{[1-\alpha/2; m-1]}$ instead of $z_{1-\alpha/2}$.

Stopping Criteria Batch Means

- Single run with $N + n_o$ observations where n_o is the size of the transient phase.

Step 0: Start with a small value of n (e.g., 1).

Step 1: compute the mean for each batch.

$$\bar{x}_i = \frac{1}{n} \sum_{j=1}^n x_{i,j}$$

Stopping Criteria Batch Means

Step 2: compute the overall mean for all batches.

$$\bar{\bar{x}} = \frac{1}{m} \sum_{i=1}^m \bar{x}_i$$

Step 3: compute the variance of the batch means.

$$\text{Var}(\bar{x}) = \frac{1}{m-1} \sum_{i=1}^m (\bar{x}_i - \bar{\bar{x}})^2$$

Step 4: compute the confidence interval for the mean as:

$$\bar{\bar{x}} \pm z_{1-\alpha/2} \text{Var}(\bar{x})$$

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Stopping Criteria Batch Means

Step 5: compute the auto-covariance

$$\text{Cov}(\bar{x}_i, \bar{x}_{i+1}) = \frac{1}{m-2} \sum_{i=1}^{m-1} (\bar{x}_i - \bar{\bar{x}})(\bar{x}_{i+1} - \bar{\bar{x}})$$

Step 6: Check for proper batch size: If $\text{Cov}(\bar{x}_i, \bar{x}_{i+1}) \ll \text{Var}(\bar{x})$
then stop. Otherwise, double n and go to step 1.

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Seed Selection

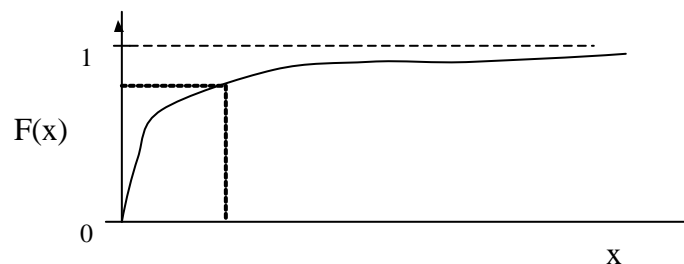
- Never use zero as a seed.
- Avoid even values.
- Reuse seed for repeatability of experiments.
- Do not use random seeds (e.g., system time) if the simulation is to be repeated.

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Generation of Random Variables

- Assume that u is a value uniformly distributed between 0 and 1.
- Method of the inverse of the CDF:



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Generation of Random Variables

- Assume that u is a value uniformly distributed between 0 and 1.
- CDF for the exponential: $1 - e^{-x/a}$
 - Inverse of the CDF: $-a \ln(u)$
- CDF for the Pareto distribution: $1 - x^{-a}$
 - Inverse of the CDF: $1/u^{1/a}$