



System Performance Evaluation



System Performance Evaluation



- Allows to obtain the highest performance at the lowest cost.
- Allows performance comparison of a number of alternative designs/solutions to find the best one.
- Gives good insights on how well a system is performing and whether any improvements need to be made.
 - Useful at any stage of the system's life cycle, i.e., design, manufacturing, use, upgrade, etc..

- Most performance problems are unique.
- Evaluation techniques used for one problem generally cannot be used for a different problem.
- Steps common to all performance evaluation projects:
 - 1. State goals and define the system under evaluation
 - Define the boundaries of the system.

2. List services and outcomes

- Each system provides a set of services
- E.g., A computer network allows its users to send packets to specific destinations
- A list of services and possible outcomes is useful in selecting the right metrics and workloads.

3. Select metrics

- Select the criteria (metrics) to compare the performance
- E.g., delay, accuracy, speed etc...



4. List of parameters that affect the performance

- System parameters (hardware and software)
- Workload parameters (depend on users' requests)



5. Select factors to study

 Some parameters will be varied during the simulation (factors) and will get different values (levels)

6. Select evaluation technique



- Analytical modeling
- Simulation
- Real test-bed

7. Select Workload

- A list of service requests to the system
- Analytical modeling: A probability of various requests
- Simulation: Trace of requests measured on a real system
- Test-bed: Scripts to be executed on the system.

8. Design Experiments

- Decide on a sequence of experiments that offer maximum information with minimal effort
 - Varying number of factors and levels to determine their relative effect.

9. Analyze and Interpret Data

- The analysis produces results (not conclusions)
- Each repetition of an experiment has a different outcome.

10. Present results

- They should be presented in a manner that is easily understood, e.g., in a graph form
 - If it is needed, redefine system boundaries, included other factors and performance metrics...(several cycles).



Selecting an evaluation technique

Criterion	Analytical modeling	Simulation	Measurement
Stage	Any	Any	Post-prototype
Time required	Small	Medium	Varies
Tools	Analysts	Computer Languages	Instrumentation
Accuracy ^a	Low	Moderate	Varies
Trade-off evaluation	Easy	Moderate	Difficult
Cost	Small	Medium	High
Scalability	Low	Medium	High





a In all cases, results may be misleading or wrong.

Selecting an evaluation technique

- Three rules of validation:
 - Do not trust the results of a simulation model until they have been validated by analytical modeling or measurements.
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 - 3. Do not trust the results of a measurement until they have been validated by simulation or analytical modeling.

- For each performance study, a set of performance criteria must be chosen.
- Time to execute a task
 - Execution time, response time, latency
- Number of tasks per day, hour, sec, ns, etc.
 - Throughput, bandwidth.

Aircraft	DC to Paris	Speed (mph)	Passengers	Throughput (pmph)
Boeing 747	6.5 hours	610	470	286,700
Concorde	3 hours	1350	132	178,200



- Flight time of Concorde vs. Boeing 747?
 - Concorde: 1350 mph / 610 mph = 2.2 times faster = 6.5 hours / 3 hours



Concorde is 2.2 times («120%») faster in terms of flight time.

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- Flight time of Concorde vs. Boeing 747:
 - Concorde: 1350 mph / 610 mph = 2.2 times faster = 6.5 hours / 3 hours
 - Concorde is 2.2 times (120%) faster in terms of flight time.



- Boeing 747 = 286,700 pmph
- Concorde = 178,200 pmph
 - Boeing 747 produces 286,700 pmph / 178,200 pmph =
 1.6 times (60%) more profit in terms of throughput.



- Global metrics: Reflect the systemwide utility
 - Resource utilization, reliability, availability.



- Individual metrics: Reflect the utility of each single user
 - Response time, throughput.
- There are cases when the decision that optimizes individual metrics is different from the one that optimizes the system metric.



- E.g.: Total vs. per node throughput
 - Keep the system throughput constant while increasing the number of packets from one source may lead to increasing its throughput, but it may also decrease someone's else throughput.
 - Using only the system throughput or the individual throughput may lead to unfair situations.



- 1. Low variability: Reduce the number of repetitions required to obtain a given level of statistical confidence.
- 2. Non-redundancy: Similar metrics should be avoided.
- 3. Completeness: All possible outcomes should be reflected in the set of performance metrics.

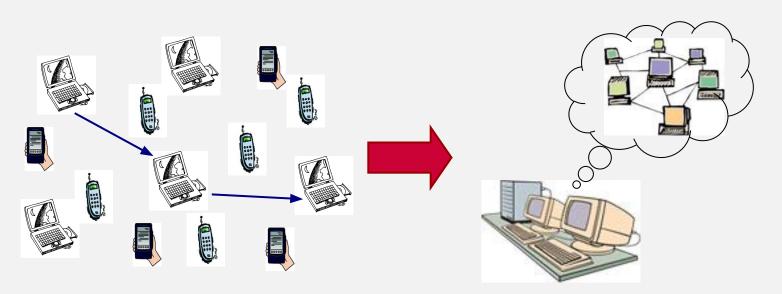




Introduction to Simulation



Introduction to Simulation





- What is a network simulator?
 - A software for modeling network applications and protocols (wired and wireless).
 - What is it used for?
 - Reproducing a system that evolves like the real system according to certain aspects, based on a model.

Simulation: When to use it

- Study and experimentation of the internal interactions of a complex system.
- System performance evaluation before the prototype.
- Verify analytical solutions.
- Common approach in research:
- Design of new protocols
- Comparison of protocols
- Traffic analysis

Simulation: Why to use it

- Only one workstation is enough to run simulations.
- Allows the study of a wide range of scenarios in a relatively short time.
- Allows realization of complex and expensive networks to be implemented in a real test-bed.
- Easy to test/check the impact of changes in a simulated solution.



Simulation: Pros & Cons

Pros

- System verification before the production of a prototype
- Easy debugging of the simulated protocol
- Possibility to analyze the system's scalability
- Identification of system vulnerabilities
- Flexibility on studying the behavior of the system.

Cons

- The design/implementation of a model and its validation require the understanding of the simulation tool.
- It is not always possible to capture the various aspects of the simulated system.

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State variables:

- The variables whose values define the state of the system
- Network simulation: number of nodes, packet queue, mac and routing protocols used etc..

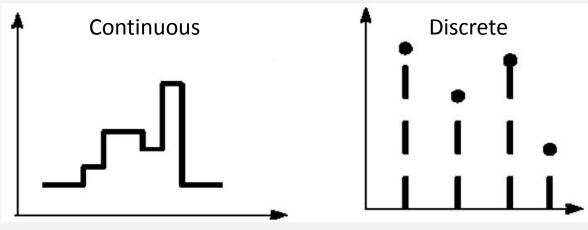
• Event:

- A change in the system state.
 - Network simulation: packet transmission, packet reception etc..

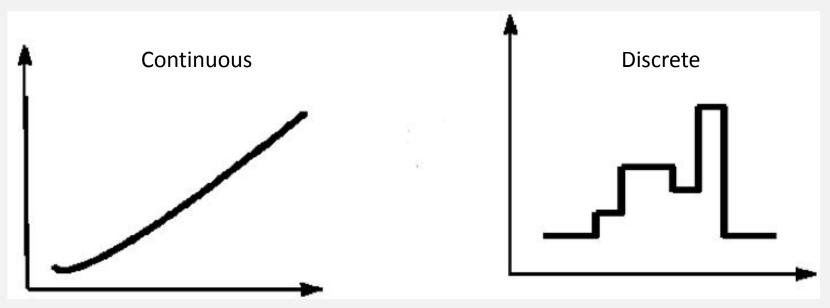


- Continuous-Time and Discrete-Time models:
 - Continuous time model: A model in which the system state is defined at all times.
 - Network simulation: number of nodes, communication among nodes is defined at any time.
 - *Discrete-Time model:* The system state is defined only at particular instants in time.
 - Classes: weekly





- Continuous-State and Discrete-State models:
 - Continuous: State variables are continuous.
 - Discrete: State variables are discrete.
 - Network simulation: number of nodes, packet queue length.







- Continuous-state models = Continuous-event models
- Discrete-state models = Discrete-event models



- Continuity of time does not imply continuity of state and vice-versa!
- Four possible combinations:
 - 1. Continuous state/continuous time
 - 2. Discrete state/discrete time

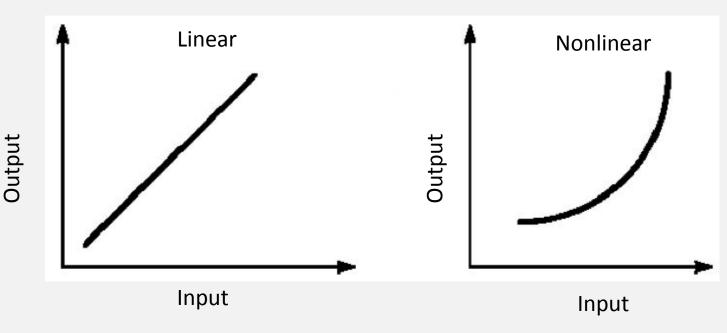


- 3. Continuous state/discrete time
- 4. Discrete state/continuous time

- Deterministic and Probabilistic models:
 - *Deterministic:* The output (results) of a model can be predicted with certainty.
 - *Probabilistic:* For the same set on input parameters, each repetition gives a different output.
- Static and Dynamic models:
 - Static: Time is not a variable.
 - Dynamic: The system state changes with time.
- Open and Closed models:
- Open: The input is external to the model and is independent of it.
- Closed: No external input.



- Linear and Nonlinear models:
 - Linear: The output parameters are a linear function of the input parameter.
 - Nonlinear: Otherwise.

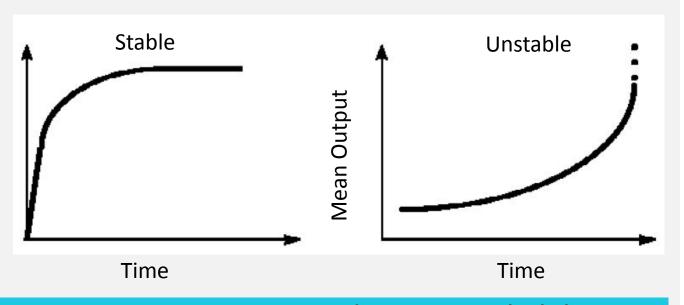




Computer system models: tempo continuo, stato discreto, probabilistico, dinamico e non lineare. Aperti o chiusi, stabili o instabili.

- **Stable and Unstable models:**
 - *Stable:* The dynamic behavior of the model settles down to a steady state.
 - Unstable: The behavior of the model is continuous changing. Mean Output







Computer system models: tempo continuo, stato discreto, probabilistico, dinamico e non lineare. Aperti o chiusi, stabili o instabili.

Simulation: Types

Model Carlo method:

- Static simulation without a time axis.
- Model probabilistic phenomena that do not change characteristics with time.

• Trace-driven:

 The simulation uses a trace as its input (a time-ordered record of events on a real system.)

• Discrete-event:

- Discrete-state model of system.
 - Network simulation: number of packets in the queue.
 - Discrete- or continuous-time values. Internet of Things A.Y. 22-23

Discrete-event Simulation

Components:

- Event scheduler: It keeps a linked list of events waiting to happen.
- Simulation clock: Each simulation has a global variable representing simulated time.
 - The scheduler is responsible for advancing this time.
 - **Unit time:** Increments time by small increment and then checks to see if there are any events that can occur.
 - **Event-driven:** Increments the time automatically to the time of the next earliest occurring time.

Discrete-event Simulation

Components:

- 3. Event routine: Each event is simulated by its routine.
- 4. Input routines: Get the model parameters.
- 5. Initialization routines: Set the initial state of the system.
- 6. Trace routines: Print out intermediate variables as the simulation proceeds; Useful on debugging.
- 7. Report generator: Output routines executed at the end of the simulation; Calculate the final result.
- 8. Main program: It brings all the routines together.



Common mistakes

1. Inappropriate level of detail

More details => Longer simulations => More bugs =>
 More computations => More parameters ≠ Higher accuracy

2. Inappropriate experimental design

Too much generic => longer simulations and less accurate

3. Unverified models

Bugs in the code

4. Invalid models

Non realistic results



Common mistakes

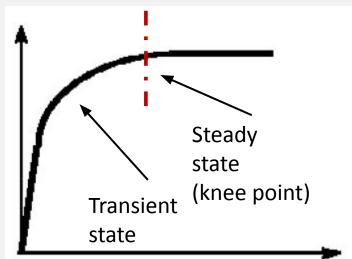
- 5. Improperly handled initial conditions
 - Generally not representative of the system behavior in a steady state.
- 6. Too short simulations
- 7. Poor random-number generators
- 8. Improper selection of seeds
 - The seed for different random-number streams should be carefully chosen to maintain independence among the simulations.

Model Verification and Validation

- PERFORMANCE EVALUATION
- Debugging: Include additional checks and output in the program that will point out the bugs (if any).
 - E.g. 1: Check if the probabilities for certain events add up to 1.
 - E.g. 2: Packets received = pkts generated pkts lost/dropped.
- 2. Structured walk-through: Explain the code to another person or a group. (It works even when the others do not understand the model!).
- 3. Run simplified cases: Easy to analyze them.
- 4. Consistency test: Check that the model produces similar results for input parameter values that have similar effects.
 - 5. Degeneracy test: Check that the model works for extreme values of system configuration or workload parameters.

Simulation Results Analysis

- In most simulations, only the steady-state performance is of interest!
- Results of the initial part of the simulation should not be included in the final computations.
- Transient removal: Identify the end of the transient state.
 - It is not possible to define exactly what constitutes the transient state and when
 - the transient state ends.
 - All methods for transient removal are heuristic.

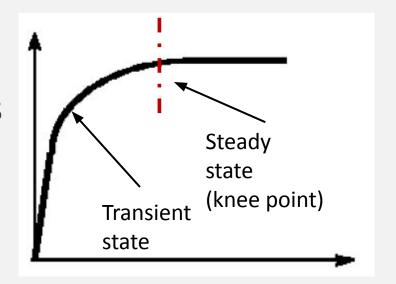


Simulation Results Analysis

Six methods for transient removal:

- 1. Long runs
- 2. Proper initialization
- 3. Truncation
- 4. Initial data deletion
- 5. Moving average of independent replications
- 6. Batch means







Terminating Simulations

- Short simulations => low degree confidence
- Long simulations => waste of resources
- There are systems that never reach a steady-state performance.
 - These systems always operate under transient conditions.
 - Such simulations are called **terminating simulations**; they do not require transient removal.
 - E.g.: A system shuts down at 5pm every day.



• To increase data confidence take the average over several independent repetitions.





Simulators for IoT Systems



What is a Simulator?



- A tool/software that realistically imitates/models the behavior of IoT systems.
- Different types of simulators; Most commonly used:
 - Trace-Driven Simulators
 - Discrete-Event Simulators



Why do we use Simulators?

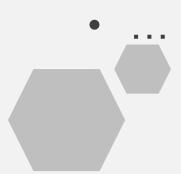
 The most common approach to develop and test new protocols/applications.



- Evaluate the performance of new solutions.
- Consider a large-scale IoT network:
 - Low cost
 - Easy(?) to implement
 - Practical

Simulators for IoT Systems

- Several simulators exist:
 - ns-3/ns-2
 - OMNeT
 - Castalia
 - GreenCastalia
 - SUNSET
 - COOJA
 - Avrora





Additional Resources



 R. Jain, "The Art of Computer Systems Performance Analysis: Techniques for Experimental Design, Measurement, Simulation, and Modeling", Wiley-Interscience, New York, NY, April 1991. (Chapters 2, 3, 24)



