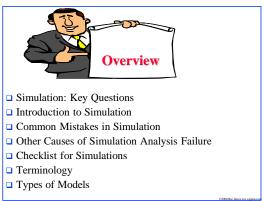
Introduction to Simulation



24-2

Simulation: Key Questions

24-1

- What are the common mistakes in simulation and why most simulations fail?
- What language should be used for developing a simulation model?
- □ What are different types of simulations?
- □ How to schedule events in a simulation?
- □ How to verify and validate a model?
- □ How to determine that the simulation has reached a steady state?
- How long to run a simulation?

24-3

Simulation: Key Questions (Cont)

- □ How to generate uniform random numbers?
- How to verify that a given random number generator is good?
- □ How to select seeds for random number generators?
- □ How to generate random variables with a given distribution?
- □ What distributions should be used and when?

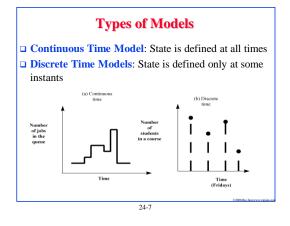
24-4

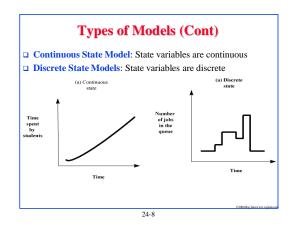
Terminology

- □ **State Variables**: Define the state of the system Can restart simulation from state variables
 - E.g., length of the job queue.
- **Event**: Change in the system state.
 - E.g., arrival, beginning of a new execution, departure

Common Mistakes in Simulation

- Inappropriate Level of Detail: More detail ⇒More time ⇒More Bugs ⇒More CPU ⇒More parameters ≠ More accurate
- 2. Improper Language
- General purpose ⇒More portable, More efficient, More time 3. Unverified Models: Bugs
- 4. Invalid Models: Model vs. reality
- 5. Improperly Handled Initial Conditions
- 6. Too Short Simulations: Need confidence intervals
- 7. Poor Random Number Generators: Safer to use a well-known generator
- 8. Improper Selection of Seeds: Zero seeds, Same seeds for all streams

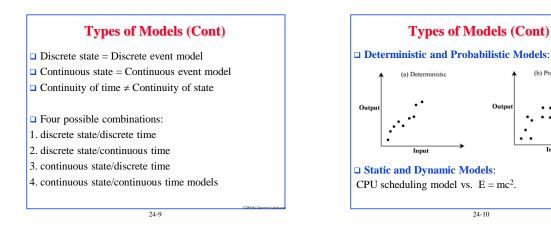


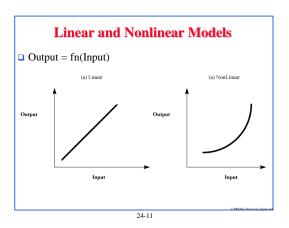


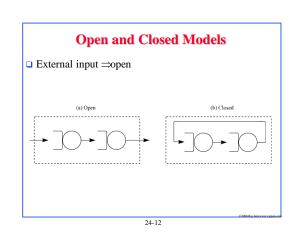
(b) Probabilistic

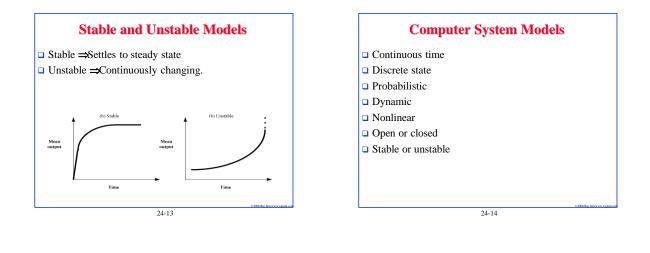
Input

Outpu









Selecting a Language for Simulation

- 1. Simulation language
- 2. General purpose
- 3. Extension of a general purpose language
- 4. Simulation package

24-15

General Purpose Language

- Analyst's familiarity
- Easy availability
- Quick startup
- □ Time for routines for event handling, random number generation
- Other Issues: Efficiency, flexibility, and portability
- Recommendation: Learn at least one simulation language.

Simulation Languages

- Save development time
- Built-in facilities for time advancing, event scheduling, entity manipulation, random variate generation, statistical data collection, and report generation
- □ More time for system specific issues
- □ Very readable modular code

24-16

Extensions of a General Purpose Language

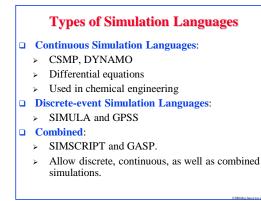
- □ Examples: GASP (for FORTRAN)
 - > Collection of routines to handle simulation tasks
 - Compromise for efficiency, flexibility, and portability.

Simulation Packages

Example: QNET4, and RESQ

- Input dialog
- Library of data structures, routines, and algorithms
- Big time savings
- □ Inflexible ⇒Simplification

24-19



24-20

Types of Simulations

- Emulation: Using hardware or firmware
 E.g., Terminal emulator, processor emulator
 Mostly hardware design issues
- 2. Monte Carlo Simulation
- 3. Trace-Driven Simulation
- 4. Discrete Event Simulation

24-21

Types of Simulation (Cont)

Monte Carlo method [Origin: after Count Montgomery de Carlo, Italian gambler and random-number generator (1792-1838).] A method of jazzing up the action in certain statistical and number-analytic environments by setting up a book and inviting bets on the outcome of a computation.

> - The Devil's DP Dictionary McGraw Hill (1981)

24-22

Monte Carlo Simulation

- □ Static simulation (No time axis)
- To model probabilistic phenomenon
- □ Need pseudorandom numbers
- Used for evaluating non-probabilistic expressions using probabilistic methods.

Monte Carlo: Example

$$I = \int_0^2 e^{-x^2} dx$$
$$x \sim \text{Uniform}(0, 2)$$

<u>.</u>

Density function $f(x) = \frac{1}{2}$ iff $0 \cdot x \cdot 2$

$$y = 2e^{-x^2}$$

Monte Carlo: Example (Cont)

$$E(y) = \int_0^2 2e^{-x^2} f(x)dx$$

$$= \int_0^2 2e^{-x^2} \frac{1}{2}dx$$

$$= \int_0^2 e^{-x^2} dx$$

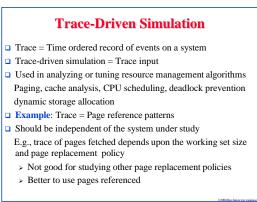
$$= I$$

$$x_i \sim \text{Uniform}(0, 2)$$

$$y_i = 2e^{-x_i^2}$$

$$I = E(y) = \frac{1}{n} \sum_{i=1}^n y_i$$

24-25



24-26

Advantages of Trace-Driven Simulations

- 1. Credibility
- 2. Easy Validation: Compare simulation with measured
- 3. Accurate Workload: Models correlation and interference
- 4. Detailed Trade-Offs:
- Detailed workload =Can study small changes in algorithms
- 5. Less Randomness:
- Trace \Rightarrow deterministic input \Rightarrow Fewer repetitions
- 6. Fair Comparison: Better than random input
- 7. Similarity to the Actual Implementation:
- Trace-driven model is similar to the system
- ⇒ Can understand complexity of implementation

24-27

Disadvantages of Trace-Driven Simulations

- 1. Complexity: More detailed
- 2. Representativeness: Workload changes with time, equipment
- 3. Finiteness: Few minutes fill up a disk
- 4. Single Point of Validation: One trace = one point
- 5. Detail
- 6. Trade-Off: Difficult to change workload

24-28

Discrete Event Simulations

- ❑ Concentration of a chemical substance ⇒Continuous event simulations
- □ Number of jobs ⇒Discrete event
- □ Discrete state \neq discrete time

Components of Discrete Event Simulations

- 1. Event Scheduler
- (a) Schedule event X at time T.
- (b) Hold event X for a time interval dt.
- (c) Cancel a previously scheduled event X.
- (d) Hold event X indefinitely
- (e) Schedule an indefinitely held event.
- 2. Simulation Clock and a Time Advancing Mechanism
- (a) Unit-time approach
- (b) Event-driven approach

Components of Discrete Events Sims (Cont)

- 3. System State Variables Global = Number of jobs
- Local = CPU time required for a job 4. Event Routines: One per event.
- E.g., job arrivals, job scheduling, and job departure
- 5. Input Routines: Get model parameters Very parameters in a range.
- 6. Report Generator
- 7. Initialization Routines: Set the initial state. Initialize seeds.
- 8. Trace Routines: On/off feature
- 9. Dynamic Memory Management: Garbage collection
- 10. Main Program

24-31



3. Monte Carlo Simulation: Static models

2.

4. Trace driven simulation: Credibility, difficult trade-offs