



Università
degli Studi di
Messina

DIPARTIMENTO DI INGEGNERIA

Dependable Computing Modeling and Simulation

Discrete event simulation

Master degree in
Engineering in Computer Science

Simulation

- Software program emulating how a system works
 - Existing systems
 - Systems in designing phase
- System behavior is described by its state

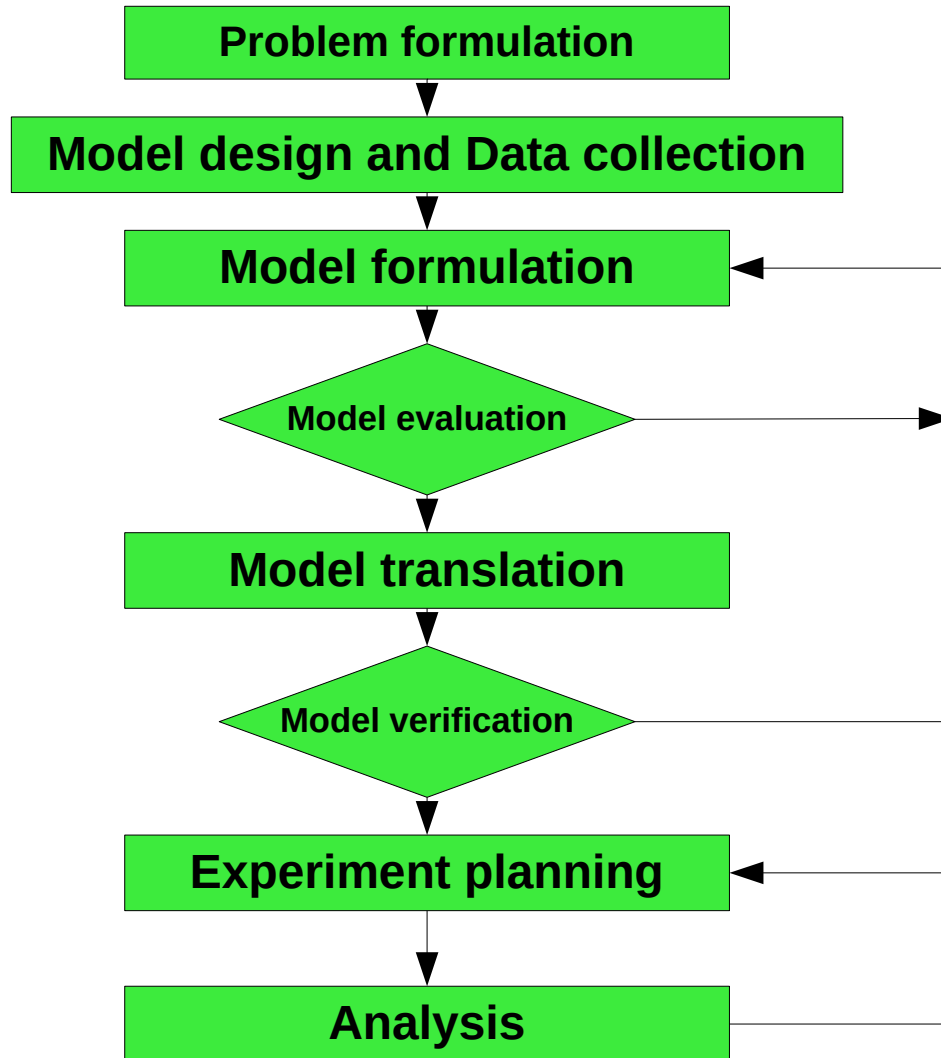
Applications

- System components interactions
- Evaluation about how the system behavior changes when subject to modifications
- Performances in different functional modes (design phase)
- Analytical models validation

Typical phases

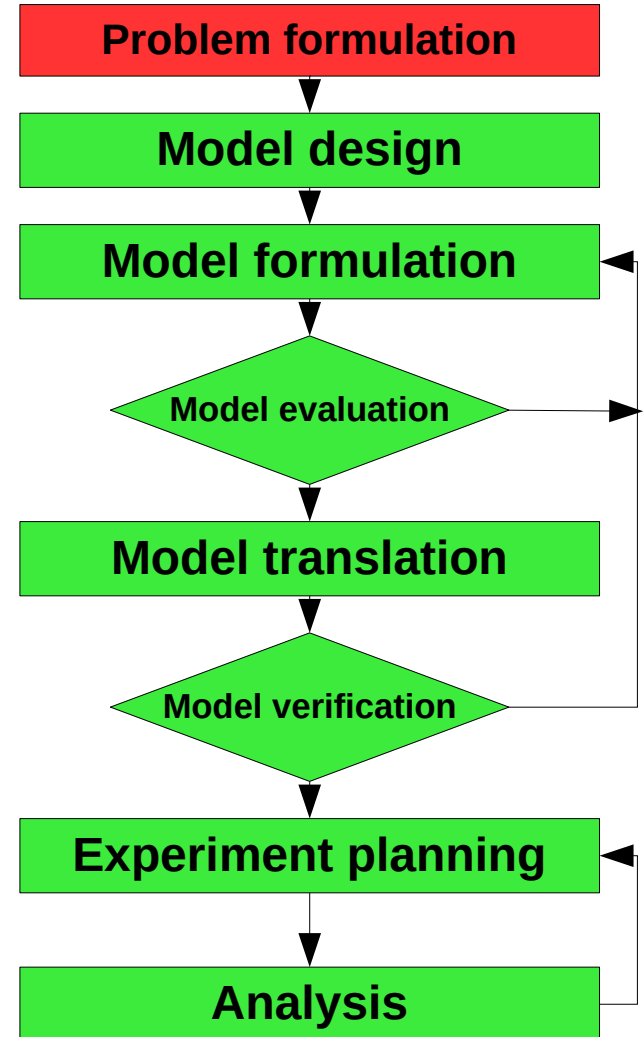
- 1) Problem formulation
- 2) Data collection
- 3) Model formulation
- 4) Model evaluation
- 5) Model translation
- 6) Model verification
- 7) Experiments planning
- 8) Analysis

Phases diagram



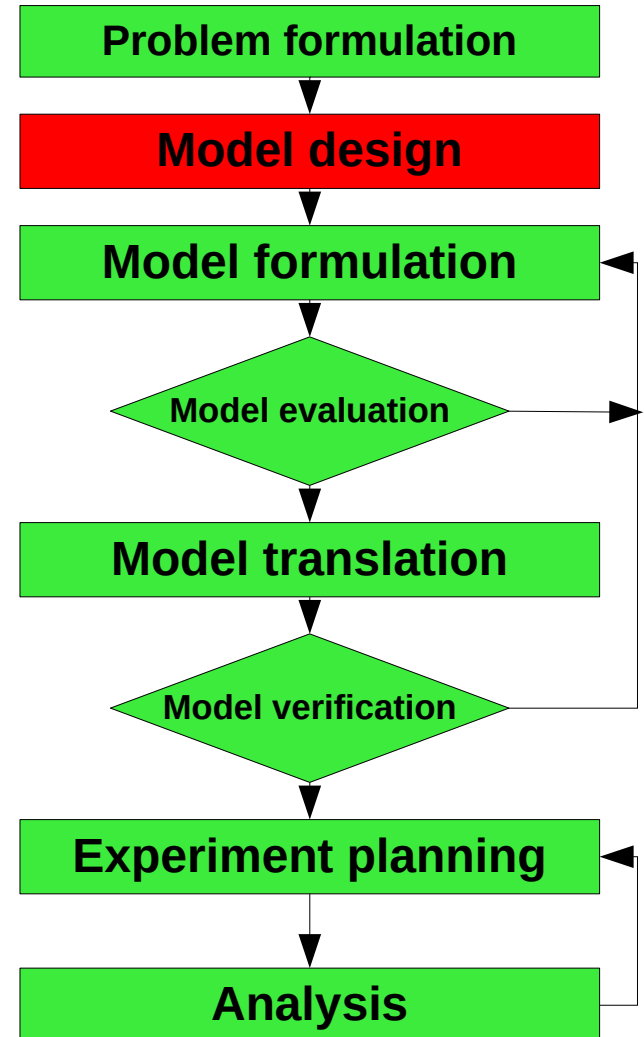
Problem formulation

- Goals
 - performances
 - input/output relations
 - optimization
 - comparison
- Measures of interest
- Input parameters



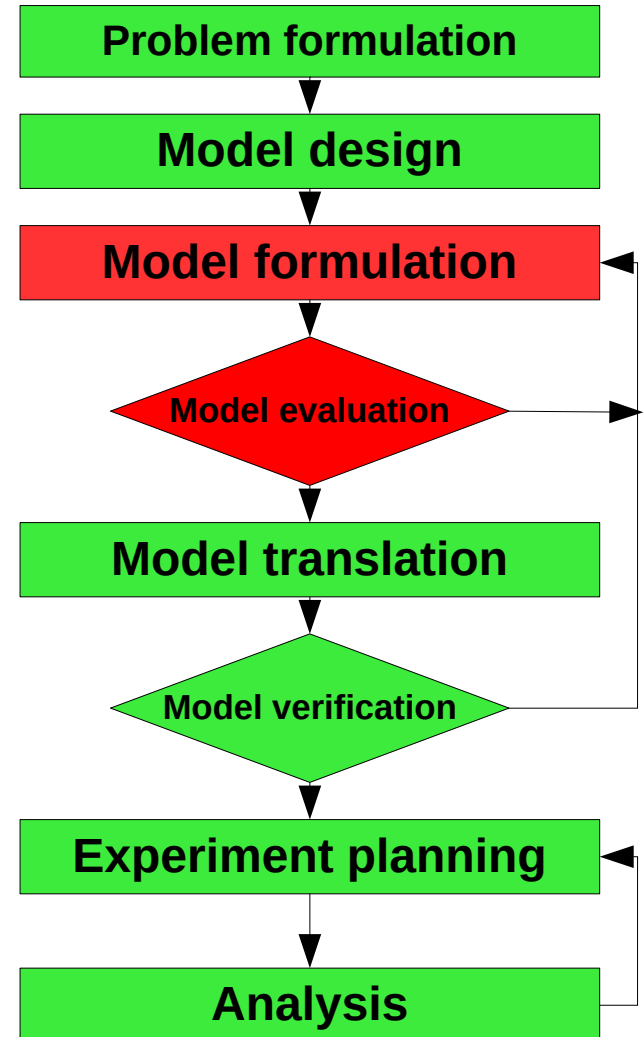
Model design and data collection

- Choice on what to represent
- Inputs
 - Service times
 - Inter-arrival times
- Experimental data
- Realistic hypothesis



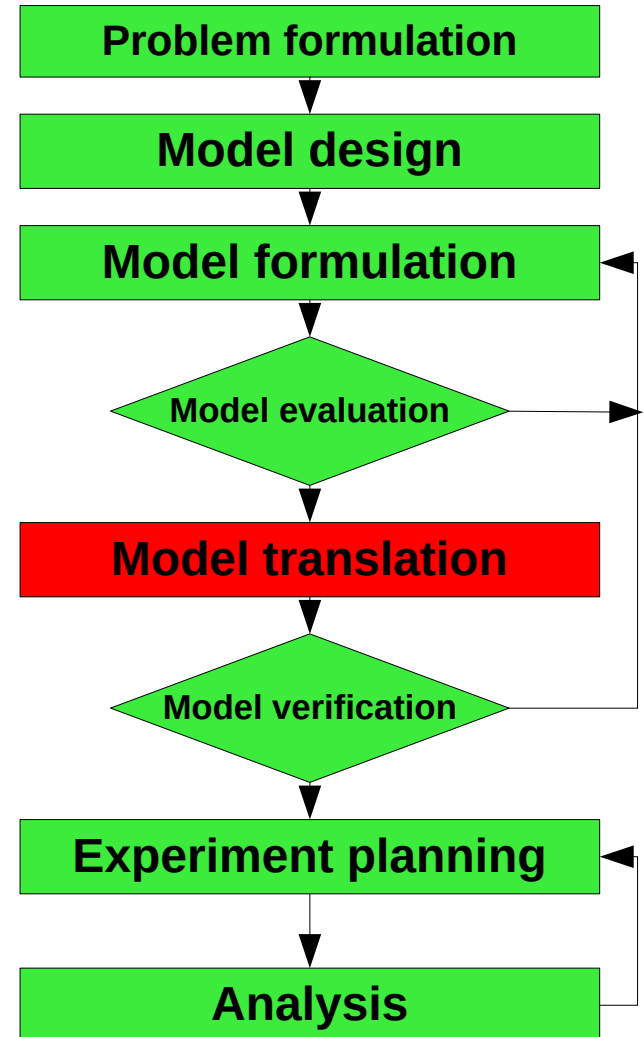
Model formulation and evaluation

- Model simplification
- Refinement
- Abstraction of main characteristics



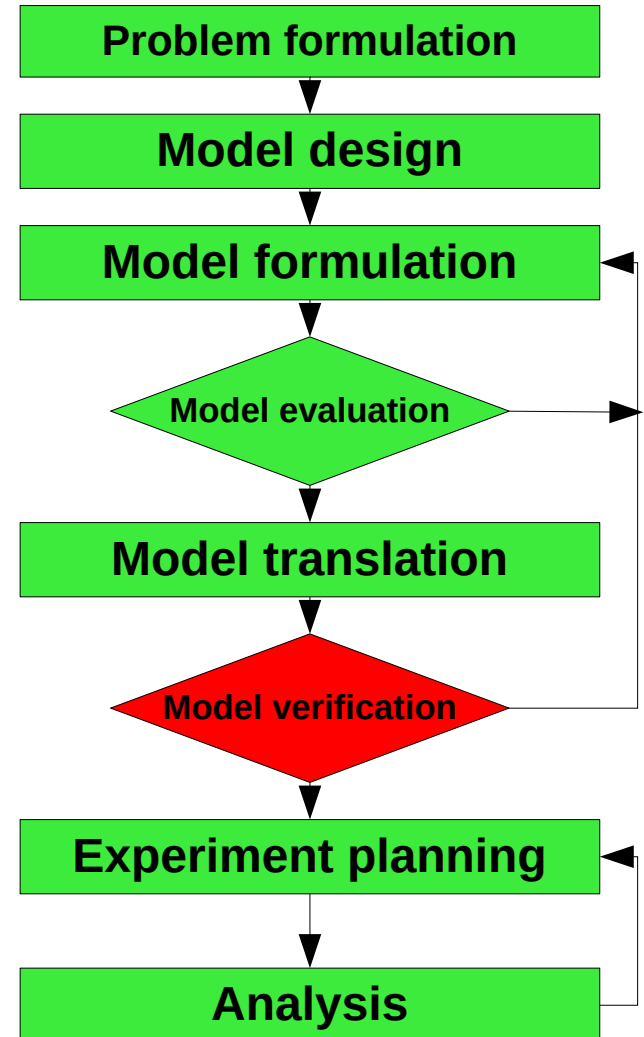
Model translation

- Programming language
 - General purpose (SIMSCRIPT, MODSIM, ...)
 - Simulation oriented



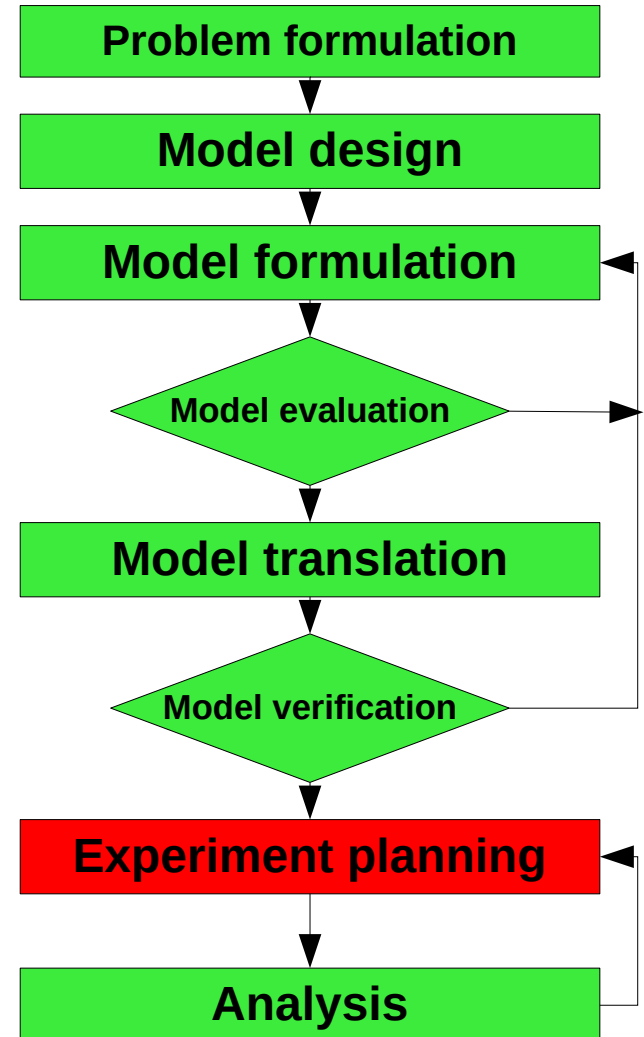
Model verification

- Correctness
 - Logic structure
 - Input and output interface modules



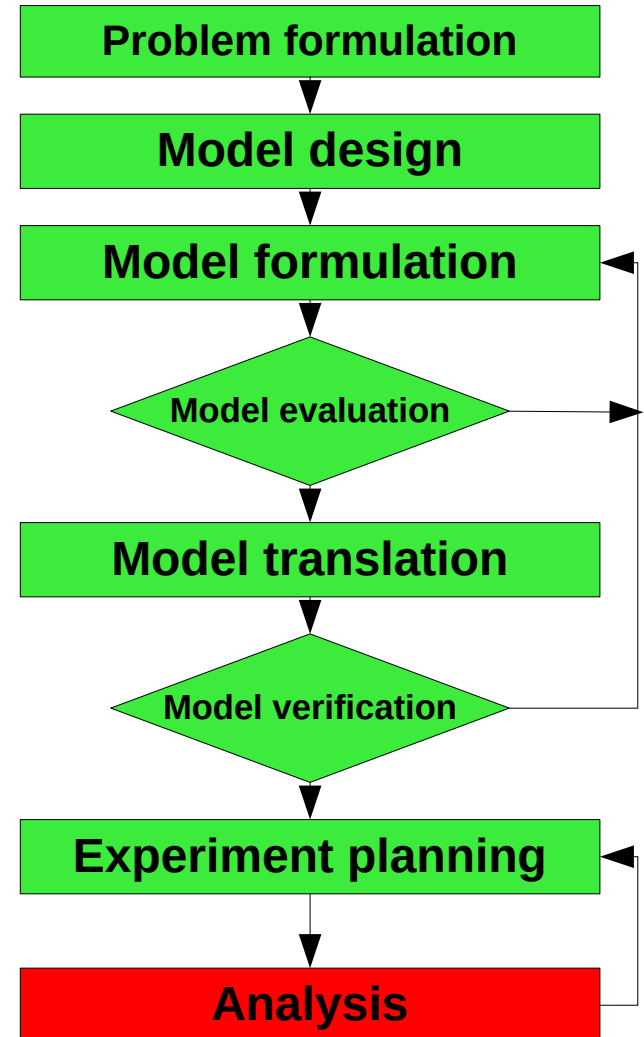
Experiment planning

- In this phase should be choosen
 - Initialization period
 - Simulation time
 - Number of simulations



Result Analysis

- Results are collected
- Goodness of results
 - More simulations can be needed



Some definitions

- *System* : set of entities interacting to achieve a goal
- *Model*: abstract representation of logical system connections; it describes the system through **state**, **activities** and their **attributes**

Some definitions (contd.)

- *System state*: the set of variables that identify the system at any time instant
- *Entity*: each system component we want to represent
- *Attributes*: characteristic of an entity
- *Events*: a fact able to change system state
 - conditioned (or dependent) event
 - primary (or independent) event

Some definitions (contd.)

- *Event calendar*: list of the happening events, ordered over the time
- *Activity*: a well identified time period
- *Delay*: an undefined time period (the length is unknown)
- *Clock*: a **variable** accounting for the *simulated* time

Events

- System state changes in a discrete way
- Type of events must be identified
- *Longitudinal* analysis
 - entities evolve in parallel
 - different objects in movement

Events

- Given an event, it influences the system state and attributes of entities
- Based on the attribute values at a given simulated time, future events are identified

Discrete event simulation

- A sequence of *system snapshots* over the time (simulated)
- A snapshot is constituted by:
 - System state
 - Active activities (a list)
 - Entity states
 - Cumulative statistics and counters
- Final state

Events

- When an event happens
 - The clock advances
 - Future events list is modified (at least one event is removed)
- System state updating
 - Data for statistics
 - Entity set
 - System state

Notation

- τ : simulation time
- x_i : system state at i -th step
- e : generic event
- $\delta(x, e)$: transition function state

Initialization

- Future event calendar is set with the happening events at time 0 (primary events)
- Implementation
 - Chained list ordered according to the scheduling time (the time when an event happens)

Future event list

- It is an important data structure
 - It is always accessed by read and insert operations
 - efficiency
- The head is removed (next event)
- Sets of event are modified

Simulation running

Time instant τ_k , states x_k

- 1) Get (e_k, τ_k) by the future event list
- 2) $\tau = \tau + \tau_k$ (time advances in a discrete way)
- 3) $x_{k+1} = \delta(x_k, e_k)$
 - Update statistics
 - impossible events in x_{k+1} are removed
 - New possible events, due to the happening of e_k are added to the future events list
- 4) Sort the list

An example: a queuing system

- Customers arrive at random inter-arrival time (Poissonian process)
- If the server is busy, customer will wait into the queue
- When the service completes, customer under service leaves the system
- The server can be into one of two states: *busy* and *free*

Basic elements

- *System state*: number of costumers in the queue, server state
- *Entities*: server, costumers, queue
 - Attributes?
- *Events*: costumer arrival, costumer departure, service starting (dependent event), simulation end
- *Activities*: inter-arrival and service times
- *Delays*: costumer waiting times

Event: *costumer arrival*

- Actions
 - next costumer arrival is planned (primary event)
 - free server
 - Server state changes to busy and the end of service event is planned
 - busy server
 - number of queued client increases

Event: *costumer departure*

- Actions
 - the server becomes **free**
 - if the queue is not empty
 - the number of queued costumers decreases
 - server becomes **busy**
 - a departure is planned

Event: *simulation end*

- Action
 - none (system state remains the same)
- Simulation ends and all the statistics are computed

Simulation run

- Simulation time?
 - Time to serve the first five costumers
 - Simulation end event
 - Unknown
- Measures
 - Expected waiting time
 - Server utilization
 - Expected queue length

Expected waiting time

- Each client has a different waiting time (θ_i) in the queue
- N costumers
- A possible estimation is

$$\hat{\theta} = \frac{1}{N} \sum_{k=1}^N \theta_k$$

Utilization

- $T_N = \tau_f - \tau_0$
- $T(i)$: overall time during which i costumers are inside the system

$$\hat{v} = \frac{1}{T_N} \sum_{i=1}^{\infty} T(i) = 1 - \frac{T(0)}{T_N}$$

Expected queue length

- $p_N(i)$: probability that the queue length is i

$$\bar{x} = \sum_{i=1}^{\infty} i p_N(i)$$

$$\hat{x} = \sum_{i=1}^{\infty} i \hat{p}_N(i) = \frac{1}{T_N} \sum_{i=0}^{\infty} i T(i)$$

Involved variables

- Computation of expected waiting time:

$\theta_i, \quad 1 \leq i \leq N \Rightarrow$ list of floats/doubles

- Computation of utilization:

$T(0) \Rightarrow$ a float/double

- Computation of expected queue length:

$T(i), \quad 1 \leq i \leq N \Rightarrow$ list of floats/doubles

— Optimization by using an accumulator

- Number of customers \Rightarrow an integer N

Events: occurrences

Costumer	Inter-arrival times	Arrival times
1		0
2	2	2
3	4	6
4	1	7
5	1	8

Costumers	Service time
1	3
2	2
3	3
4	4
5	2

Calendar

Simulated time	Event type	Client number
0	arrival	1
2	arrival	2
3	depature	1
5	depature	2
6	arrival	3
7	arrival	4
8	arrival	5
9	depature	3
13	depature	4
15	depature	5

Simulation run

- $\tau=0$: simulation starts
- Server state: busy
- Queue length: 1
- Future events
 - $\tau=2$: arrival (2)
 - $\tau=3$: departure (1)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 0$

Simulation run

- $\tau=2$: costumer 2 arrives
- Server state: busy
- Queue length: 2
- Future events
 - $\tau=3$: departure (1)
 - $\tau=6$: arrival (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 0$
 - $T(1) = 2$

Simulation run

- $\tau=3$: client 1 departs
- Server state: **busy**
- Queue length: **1**
- Future events
 - $\tau=5$: departure (2)
 - $\tau=6$: arrival (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 1$
 - $\theta_1 = 3$
 - $T(1) = 2$
 - $T(2) = 1$

Simulation run

- $\tau=5$: client 2 departs
- Server state: free
- Queue length: 0
- Future events
 - $\tau=6$: arrival (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 2$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $T(1) = 4$
 - $T(2) = 1$

Simulation run

- $\tau=6$: client 3 arrives
- Server state: **busy**
- Queue length: **1**
- Future events
 - $\tau=7$: arrival (4)
 - $\tau=9$: departure (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 2$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $T(0) = 1$
 - $T(1) = 4$
 - $T(2) = 1$

Simulation run

- $\tau=7$: client 4 arrives
- Server state: **busy**
- Queue length: **2**
- Future events
 - $\tau=8$: arrival (5)
 - $\tau=9$: departure (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 2$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $T(0) = 1$
 - $T(1) = 5$
 - $T(2) = 1$

Simulation run

- $\tau=8$: client 5 arrives
- Server state: **busy**
- Queue length: **3**
- Future events
 - $\tau=9$: departure (3)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 2$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $T(0) = 1$
 - $T(1) = 5$
 - $T(2) = 2$

Simulation run

- $\tau=9$: client 3 departs
- Server state: **busy**
- Queue length: **2**
- Future events
 - $\tau=13$: departure (4)
 - $\tau=\tau_f$: simulation end

- Statistics
 - $N = 3$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $\theta_3 = 3$
 - $T(0) = 1$
 - $T(1) = 5$
 - $T(2) = 2$
 - $T(3) = 1$

Simulation run

- $\tau=13$: client 4 departs
- Server state: **busy**
- Queue length: **1**
- Future events
 - $\tau=15$: departure (5)
 - $\tau=\tau_f$: simulation end
- Statistics
 - $N = 4$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $\theta_3 = 3$
 - $\theta_4 = 6$
 - $T(0) = 1$
 - $T(1) = 5$
 - $T(2) = 6$
 - $T(3) = 1$

Simulation run

- $\tau=15$: client 5 departs
 - Server state: free
 - Queue length: 0
 - Future events
- Statistics
 - $N = 5$
 - $\theta_1 = 3$
 - $\theta_2 = 3$
 - $\theta_3 = 3$
 - $\theta_4 = 6$
 - $\theta_5 = 7$
 - $T(0) = 1$
 - $T(1) = 7$
 - $T(2) = 6$
 - $T(3) = 1$

Measures

- Simulation ends at time 15

$$\hat{\theta} = \frac{3+3+3+6+7}{5} = 4,4$$

$$\hat{v} = 1 - \frac{T(0)}{15} = 1 - \frac{1}{15} \simeq 0.933$$

$$\hat{x} = \frac{1}{15} \sum_{i=0}^{\infty} i T(i) = \frac{(0 \times 1) + (1 \times 7) + (2 \times 6) + (3 \times 1)}{15} \simeq 1,467$$