

# Simulation: introduction

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2021-2022/Semester 2

- 1 Introduction and classification
  - Types of simulation
- 2 Discrete event simulation (DES)
  - SimPy
- 3 Single queue

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# What is simulation?

## Definition (Wikipedia)

A simulation is an **approximate imitation** of the operation of a process or system.

- A **widely-used** performance analysis method used for any stages
  - System is **not available**: to predict and compare alternatives
  - System is **available**: to compare under **wider** variety of workloads

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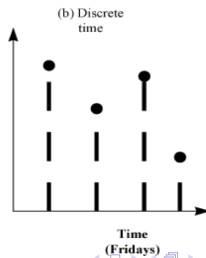
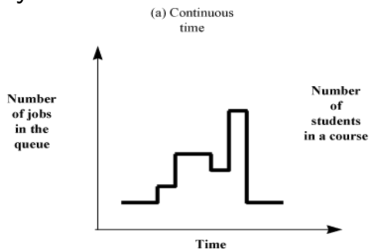
## Model developers must be proficient in

- software development
- statistical techniques

# Terminology (1)

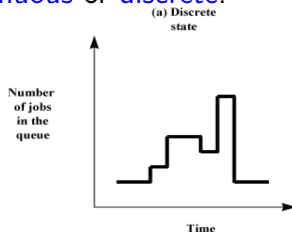
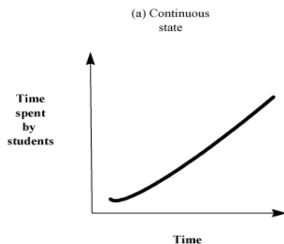
CPU scheduling is used as examples.

- **State variables:** values to define the **state** of the system  
E.g., length of job queue
- **Event:** something happens **to change the system state**  
E.g., arrival of a job; beginning of a new execution
- **Continuous-time** and **discrete-time:** whether the system state defined **at all times** or **not**.



# Terminology (2)

- **Continuous-state** and **discrete-state**: whether the state variables are **continuous** or **discrete**.

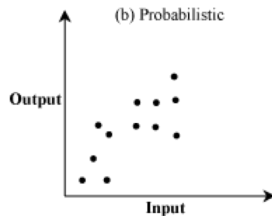
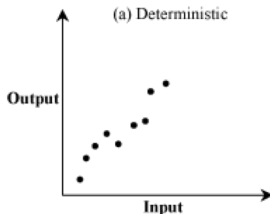


- Continuous-state model = **Continuous-event model**
- Discrete-state model = **Discrete-event model**

# Terminology (3)

## ■ **Deterministic** and **probabilistic/stochastic** models

- Deterministic model: output is **predicted with certainty**
- Probabilistic model: **different** output on repetitions for the **same** set of input parameters



## ■ **Static** and **dynamic** models

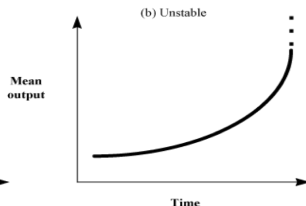
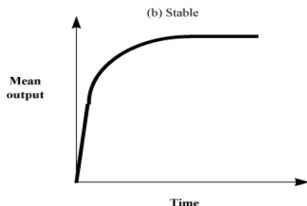
- Static model: time is **not** variable
- Dynamic model: system state **changes with time**

E.g., mass-to-energy transformation  $E = mc^2$ .

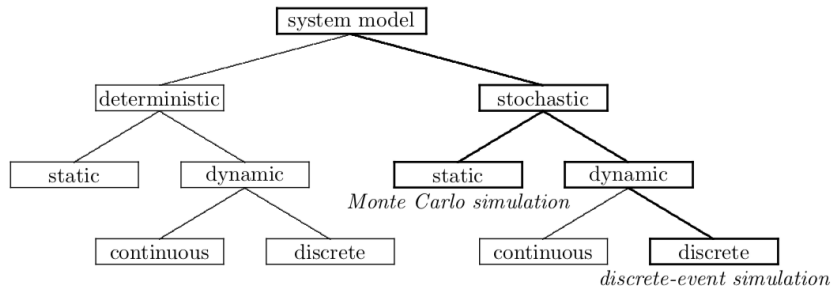


# Terminology (4)

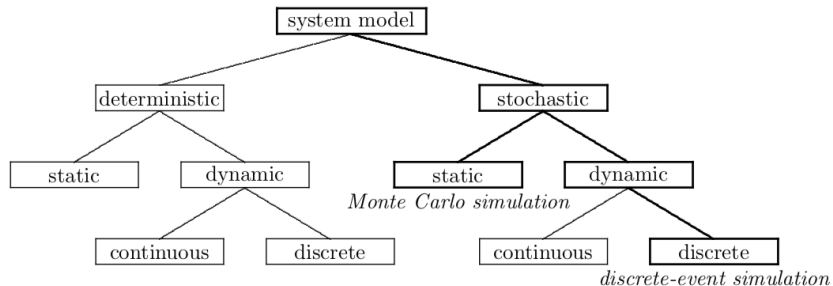
- **Open** and **closed** models: *read lecture #5 “queueing networks”*.
- **Stable** and **unstable** models
  - Stable model: dynamic behaviour settles down to a **steady state**.
  - Unstable model: behaviour is **continuously changing**.



# System model taxonomy (phân loại)



# System model taxonomy (phân loại)



## Computer system models

- Continuous time
- Discrete state
- Probabilistic
- Dynamic
- Nonlinear
- Open and closed
- Stable and unstable

# Types of simulation

There are different ways to classify simulation types. Below is only **one of them**.

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  - **Static** simulation without time axis
  - Used for evaluating **nonprobabilistic** expressions using probabilistic methods

E.g.,  $\pi$  computation

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- 3 Trace-Driven Simulation.
  - A simulation using a **trace as its input**  
E.g., using trace-driven simulation to analyse paging algorithms, cache analysis, CPU scheduling algorithms, deadlock prevention algorithms, etc.

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- 4 Discrete Event Simulation.



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## Definition

A simulation using a **discrete-state** model of the system.

# What is Discrete-Event Simulation (DES) ?

Discrete-Event Simulation is

- Discrete (in state)
- Dynamic (in time)
- Stochastic

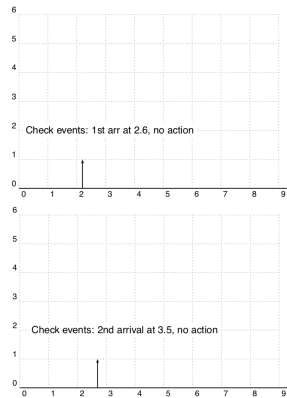
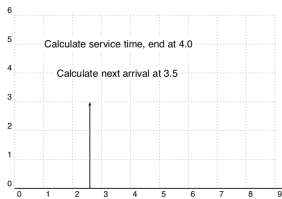
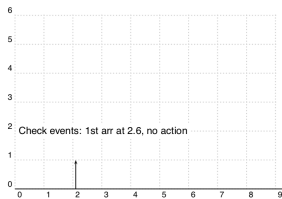
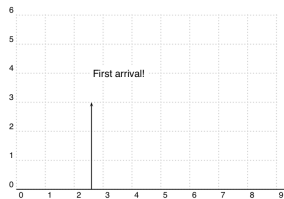
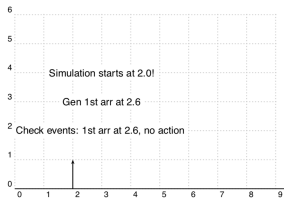
DES mostly applied to queueing systems (but not limited to)

- Factory workflow
- Freeway traffic simulation
- Network traffic simulation
- ...

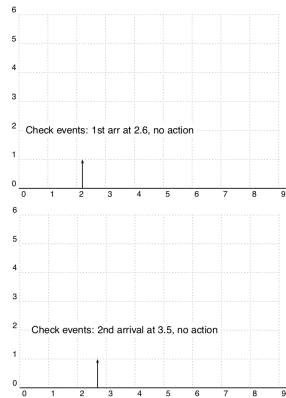
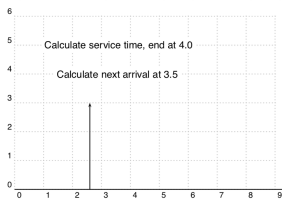
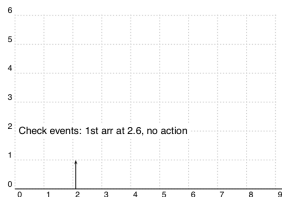
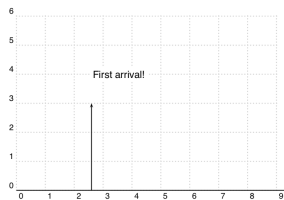
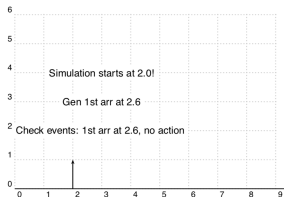
- Activity-oriented
  - fixed increment of time
  - time-consuming
- Event-oriented
  - on each event, generate next event and put into event queue and sort
  - simulation time advances to next closed event
  - faster than activity-oriented
- Process-oriented
  - abstract one object into a process
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  - **SimPy is here**

# Activity-oriented

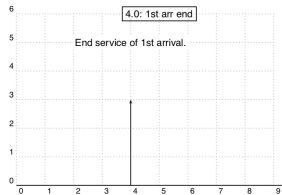
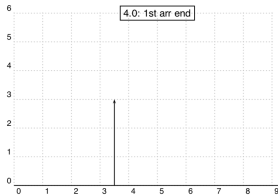
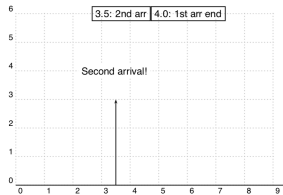
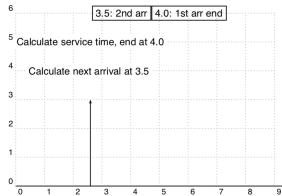
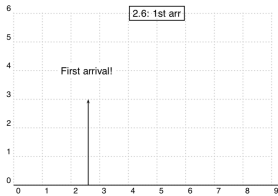
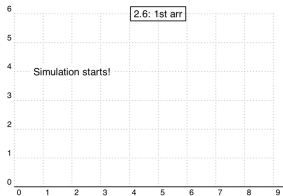


# Activity-oriented



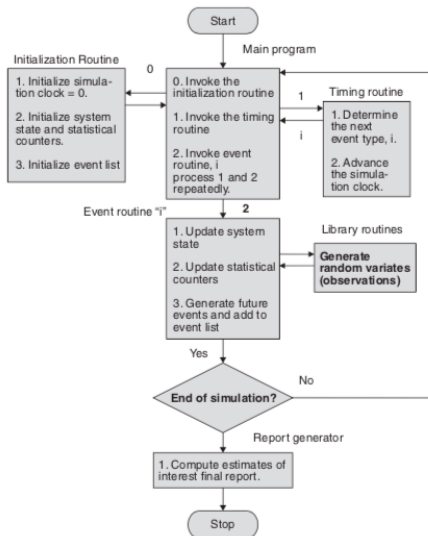
Much time for redundant checks of the unchanged state

# Event-oriented





# Event-oriented DES architecture



# Process-oriented (1)

- Based on event-oriented
- Designed into separate processes

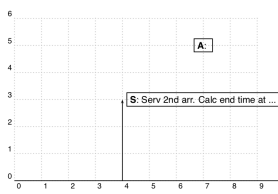
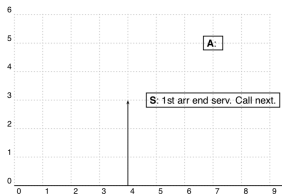
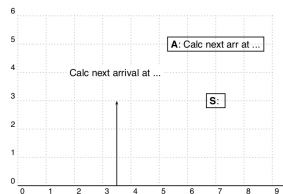
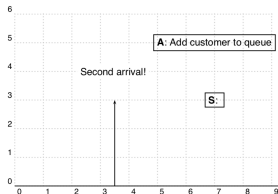
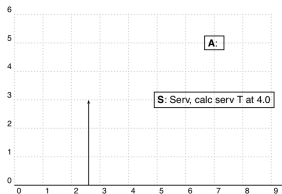
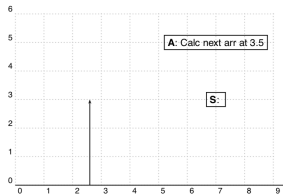
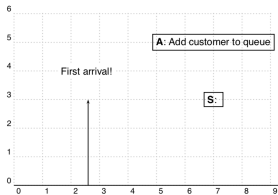
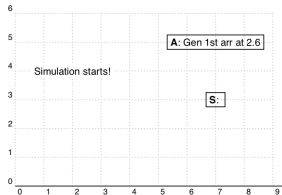
# Process-oriented (1)

- Based on event-oriented
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A single queue in process-oriented DES

- **Arrival process:** an infinite loop of the following
  - Calculate next arrival time
  - Sleep until next arrival
  - Add new job to queue
- **Service process:** an infinite loop of the following
  - Sleep until waken by new jobs
  - Serve the queued jobs on waken until there is no job in queue

# Process-oriented (2)



# What is SimPy ?

(based on slides of Stefan Scherfke)



- Environment
- Process
- Event
- Resource

```
>>> import simpy
>>>
>>> def clock(env, name, tick):
...     while True:
...         print(name, env.now)
...         yield env.timeout(tick)
...
>>> env = simpy.Environment()
>>>
>>> env.process(clock(env, 'fast', 0.5))
<Process(clock) object at 0x...>
>>> env.process(clock(env, 'slow', 1))
<Process(clock) object at 0x...>
>>>
>>> env.run(until=2)
fast 0
slow 0
fast 0.5
slow 1
fast 1.0
fast 1.5
```

Start simpy environment

Register processes with their events

Run simulation until time of 2

# Timeout and processes

(based on slides of Stefan Scherfke)

## Timeout is event



```
def speaker(env, start):  
    until_start = start - env.now  
    ● yield env.timeout(until_start)  
    ● yield env.timeout(30)
```

## Process is event, too

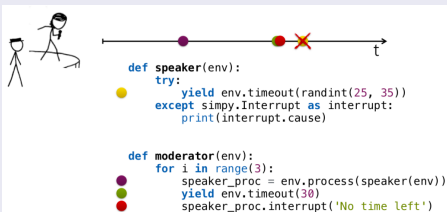


```
def speaker(env):  
    ● yield env.timeout(30)  
    return 'handout'  
  
def moderator(env):  
    for i in range(3):  
        ● val = yield env.process(speaker(env))  
        print(val)
```

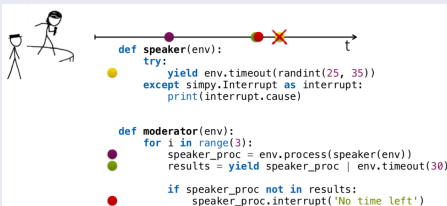
# Synchronization

(based on slides of Stefan Scherfke)

## Asynchronous interrupt



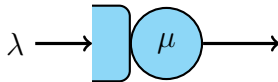
## Condition event

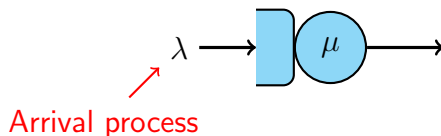


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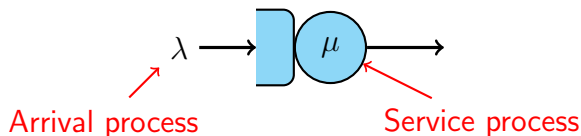


# FIFO queue





- **Job:** an object of a class
- **Queue of jobs:** list of objects
- **Arrival process:** an infinite loop of the following
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```
class Job:
    def __init__(self, name, duration):
        self.name = name
        self.duration = duration
```

More attributes could be added in order to collect statistical data, such as

- arrival time
- start time of service

# FIFO queue

A queue with single server (service process)

```
class Server:
    def __init__(self, env):
        self.Jobs = list()
        env.process( self.serve(env) )

    def serve(self, env):
        while True:
            if len( self.Jobs ) == 0 :
                self.serversleeping = env.process( self.waiting( env ) )
                yield self.serversleeping
            else:
                j = self.Jobs.pop( 0 )
                yield env.timeout( j.duration )

    def waiting(self, env):
        try:
            print( 'Server is idle at %d' % env.now )
            yield env.timeout(1000)
        except simply.Interrupt as i:
            print('A new job comes. Server waken up and works now at %d'
                  % env.now )
```

# FIFO queue

Job generator (arrival process)

```
class JobGenerator:
    jcnt = 0

    def __init__(self, env, server):
        self.server = server
        env.process( self.jobgen(env) )

    def jobgen(self, env):
        while True:
            job_interarrival = random.randint(1,5)
            yield env.timeout( job_interarrival )

            job_duration = random.randint(2,5)
            self.jcnt += 1
            self.server.Jobs.append( Job('Job %s' %self.jcnt, job_duration) )
            print( 'job %d: t = %d, l = %d' %( self.jcnt, env.now, job_duration ) )

            if not self.server.serversleeping.triggered:
                self.server.serversleeping.interrupt( 'Wake up, please.' )
```

```
env = simpy.Environment()  
MyServer = Server( env )  
MyJobGenerator = JobGenerator( env, MyServer )  
env.run( until = 20 )
```

## And results

```
Server is idle at 0  
job 1: t = 4, l = 2  
A new job comes. Server back to work at 4 by 'Wake up, please.'  
Server is idle at 6  
job 2: t = 9, l = 5  
A new job comes. Server back to work at 9 by 'Wake up, please.'  
job 3: t = 13, l = 5  
job 4: t = 14, l = 3  
job 5: t = 19, l = 4
```

## Simulation parameters

- Inter-arrival rate:  $\lambda = 5$  (jobs/time unit)
- Service rate:  $\mu = 8$  (jobs/time unit)
- Simulation time: 5000 (time unit)



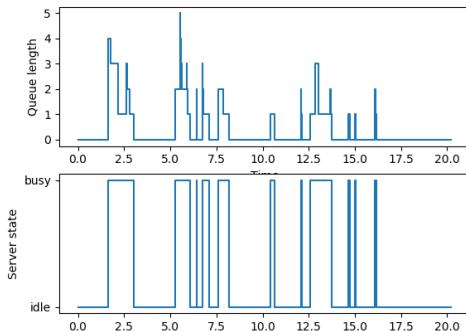
# M/M/1 queue

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- Simulation time: 5000 (time unit)

## Performance

- Analytical modeling:  $\bar{W} = 0.21$  (for validation)
- Simulation modeling:  $\bar{W} = 0.19$



- Simulation functions

- Model enhancement

- Simulation functions

- Simulation model verification/validation

- verification: develop a synthetic workload to verify the model)
    - validation: validate single queue model (M/M/1/ $\infty$ / $\infty$ /FIFO)

- Model enhancement

# M/M/1 queue

What else should be done ?

- Simulation functions
  - Simulation model verification/validation
    - verification: develop a synthetic workload to verify the model)
    - validation: validate single queue model ( $M/M/1/\infty/\infty/FIFO$ )
  - Statistical report (mean, variance)
- Model enhancement

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## ■ Model enhancement

### ■ Different types of single queue in Kendall's notation M/M/c/K/N/D

- Queueing strategies (FIFO, SJF, priority, round-robin,...)
- Different types of workload

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### ■ Different types of single queue in Kendall's notation $M/M/c/K/N/D$

- Queueing strategies (FIFO, SJF, priority, round-robin,...)
- Different types of workload
- Queueing networks