

Simulation: introduction

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- 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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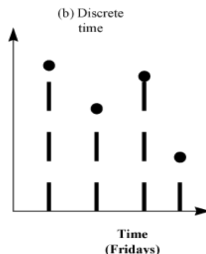
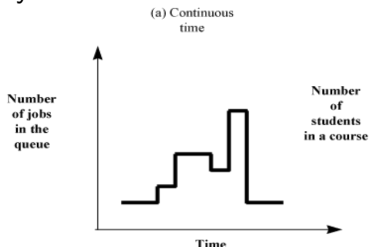
Model developers must be proficient in

- software development
- statistical techniques

Terminology (1)

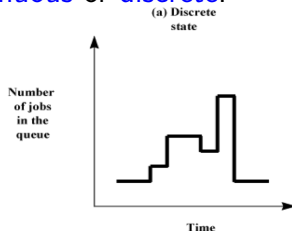
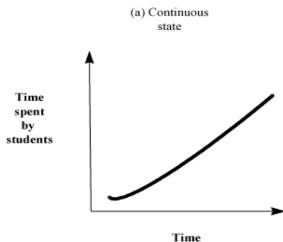
CPU scheduling is used as examples.

- **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**.

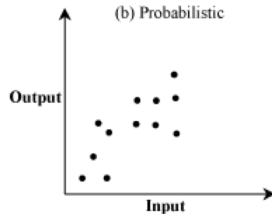
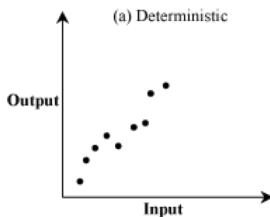


- Continuous-state model = **Continuous-event model**
- Discrete-state model = **Discrete-event model**

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



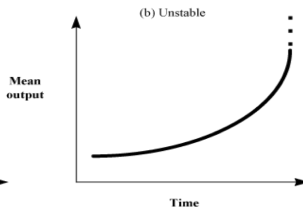
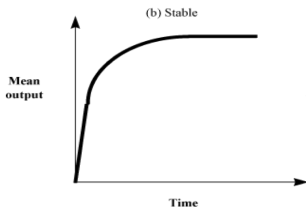
■ **Static** and **dynamic** models

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

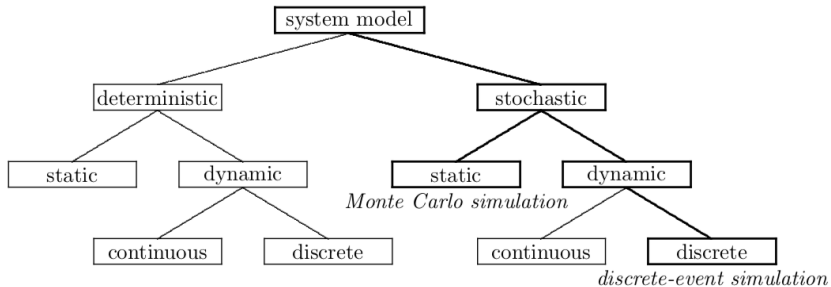
E.g., mass-to-energy transformation $E = mc^2$.

Terminology (4)

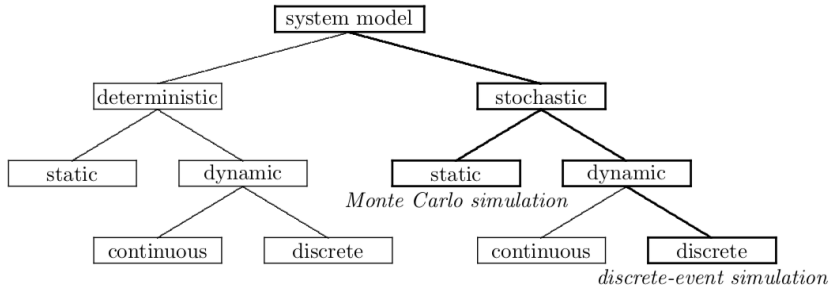
- **Open** and **closed** models: *read lecture “[queueing networks](#)”*.
- **Stable** and **unstable** models
 - Stable model: dynamic behaviour settles down to a [steady state](#).
 - Unstable model: behaviour is [continuously changing](#).



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., π 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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 - SimPy
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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
- ...

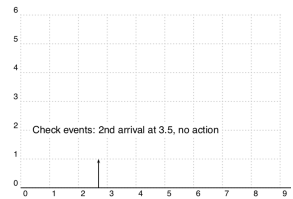
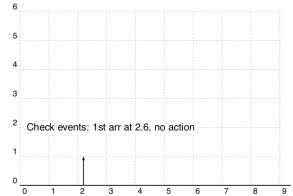
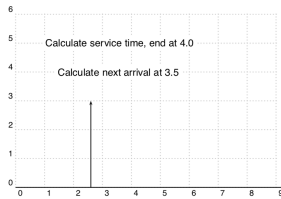
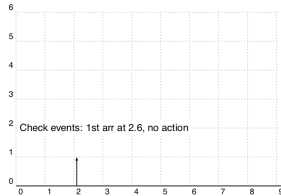
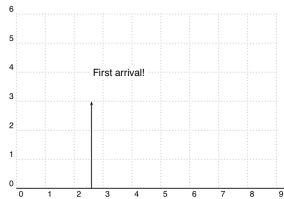
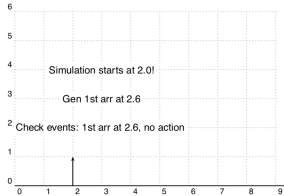
DES categories

- 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
 - easier to maintain as object-oriented approach

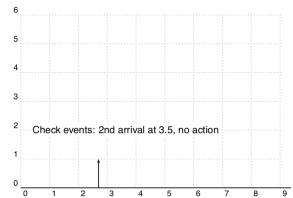
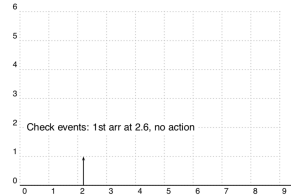
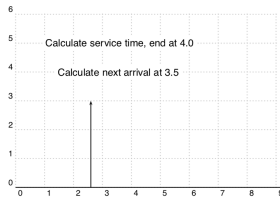
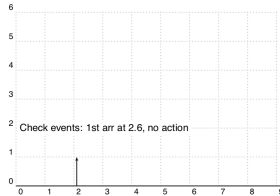
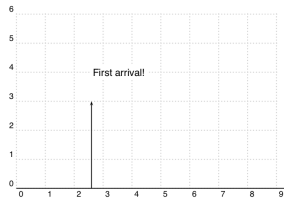
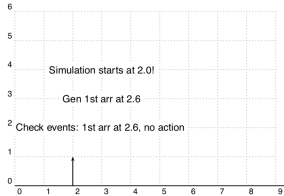
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 - **SimPy is here**

Activity-oriented

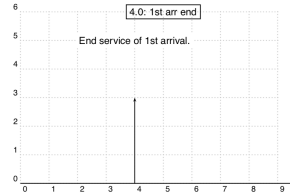
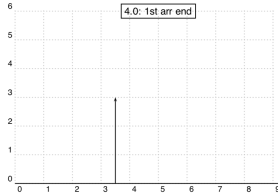
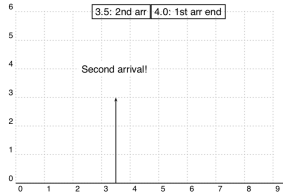
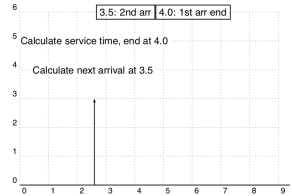
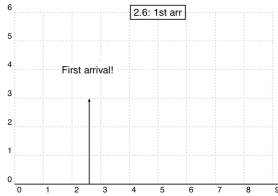
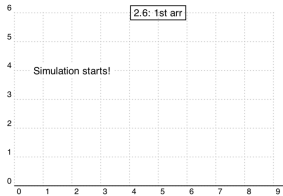


Activity-oriented

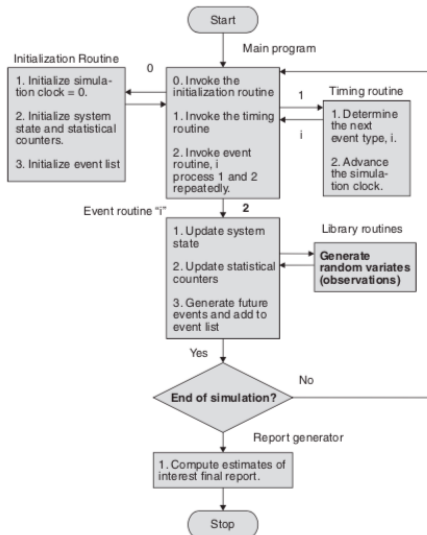


Much time for redundant checks of the unchanged state

Event-oriented



Event-oriented DES architecture



Process-oriented (1)

- Based on event-oriented
- Designed into separate processes

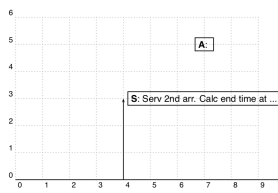
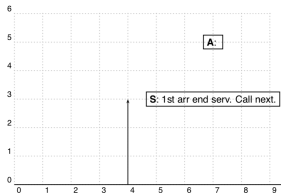
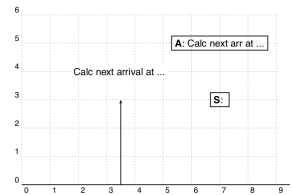
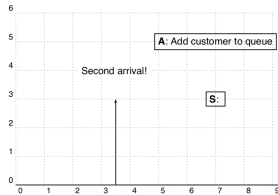
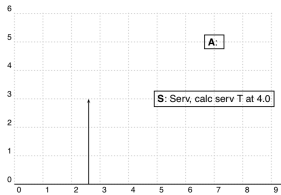
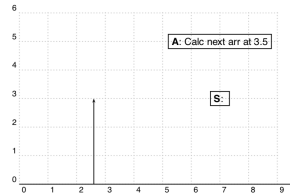
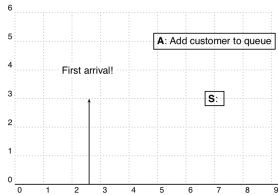
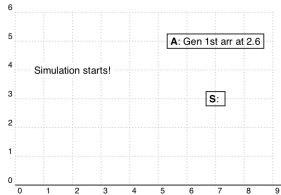
Process-oriented (1)

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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

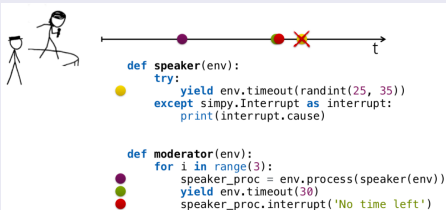


```
def speaker(env):  
    ● yield env.timeout(30)  
    return 'handout'  
  
def moderator(env):  
    for i in range(3):  
        ● val = yield env.process(speaker(env))  
        print(val)
```

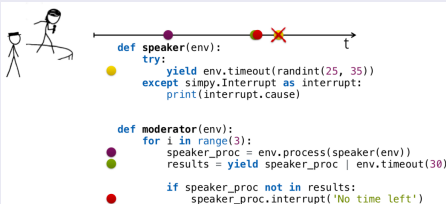
Synchronization

(based on slides of Stefan Scherfke)

Asynchronous interrupt

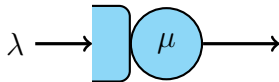


Condition event

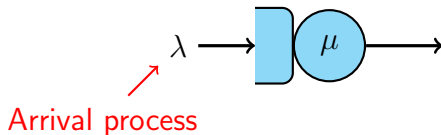


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FIFO queue

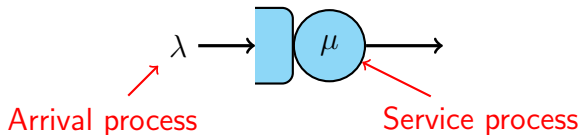


FIFO queue



- **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 simpy.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
```

M/M/1 queue

Simulation parameters

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

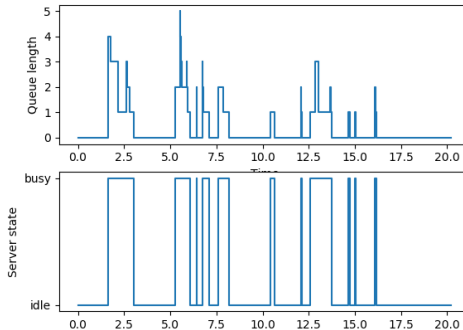
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Performance

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



M/M/1 queue

What else should be done ?

- Simulation functions

- Model enhancement

M/M/1 queue

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- Simulation functions

- Simulation model verification/validation

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

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M/M/1 queue

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 - Statistical report (mean, variance)
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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

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■ Queueing networks