

# Modeling and Simulation

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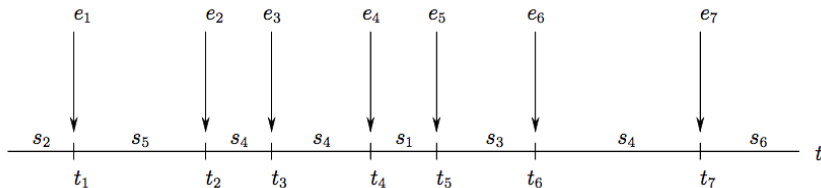


# Modeling and Simulating Discrete Event Systems

## Discrete Event Systems

### Definition

A Discrete Event System (DES) is a discrete-state, event-driven system, that is, its state evolution depends entirely on the occurrence of asynchronous discrete events over time.



Discrete State:  $s = \{s_1, s_2, s_3, s_4, s_5\}$

Events:  $e = \{e_1, e_2, e_3, e_4, e_5, e_6, e_7\}$

Time:  $t = \{t_1, t_2, t_3, t_4, t_5, t_6, t_7\}$

# Modeling and Simulating Discrete Event Systems

## Discrete Event Systems

### Discrete Event Systems are existed in:

- Queueing systems
- Operating systems and computers
- Telecommunications networks
- Distributed databases
- Automation, ...

### When Compared to Time-Driven Systems:

#### Time-Driven Systems:

- State space is continuous
- The state transition mechanism is time-driven.

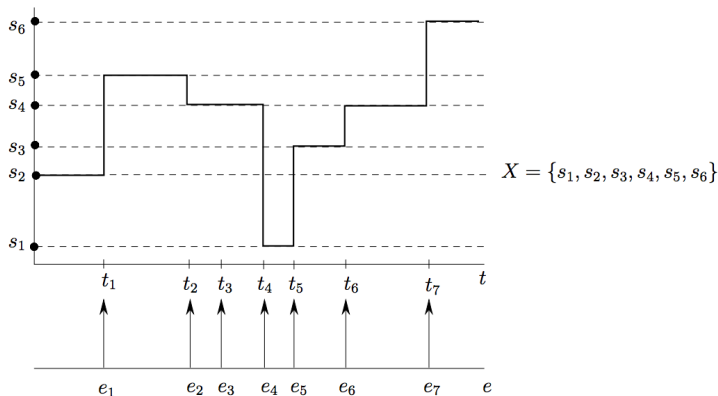
#### Discrete events-driven Systems:

- State space is discrete
- The state transition mechanism is event-driven.

# Modeling and Simulating Discrete Event Systems

## Characteristics of a Discrete-Event System

In a DES, the system state jumps from one discrete value to another whenever an event,  $e_i$  takes place.



To model such systems, a modeling/Simulation technique called **Discrete-Event Simulation** is introduced.

# Modeling and Simulating Discrete Event Systems

## What is Discrete-Event Simulation (DES)?

- A **discrete-event simulation**:
  - models a system whose state may change only at **discrete point** in time.
- **System**:
  - is composed of objects called **entities** that have certain properties called **attributes**
- **State**:
  - a collection of **attributes or state variables** that represent the **entities** of the system.
- **Event**:
  - an instantaneous occurrence in time that may alter the **state** of the system
- An **event** initiates an **activity**, which is the length of time during which **entities** engage in some operations
- **Entities, attributes, events, activities** and the interrelationships between these components are defined in the model of the system

# Modeling and Simulating Discrete Event Systems

## Discrete-Event Simulation

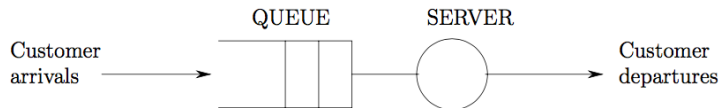
**Discrete-event simulation** is **stochastic**, **discrete**, and **dynamic**

- **Stochastic = Probabilistic**
  - Inter-arrival times and service times are random variables
  - Have cumulative distribution functions
- **Discrete = Instantaneous events are separated by intervals of time**
  - The state variables change instantaneously at separate points in time.
    - The system can change at only a countable number of points in time.
  - These points in time are the ones at which an event occurs.
- **Dynamic = Changes overtime**
  - Simulation clock.
  - mechanism to advance simulated time from one value to another

# Modeling and Simulating Discrete Event Systems

## Classical example: Customers Queue

Customer service with one operator:



Event	Timing
Arrival of a new customer	Exponentially distributed with rate $\lambda$
Customer moving from queue to service	Condition based (i.e, "when the customer is the first of the queue and the operator is free")
Customer served	Gaussian distribution with mean $\mu$ and variance $\sigma^2$ , Exponentially distributed with rate $\lambda'$



# Modeling and Simulating Discrete Event Systems

## Kendall's Notation for a Queue

Any queue system is generalized into the Kendall's Notation:  $A/B/C/K/Z$  where:

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**A:** inter-arrival time distribution

**B:** service time distribution

**C:** the number of servers

**K:** system capacity

**Z:** service discipline

---

Where **A** and **B** can be:

- M: Markovian(Exponential, Poisson)
- D: Deterministic (Constant)
- G: General distribution

**Z** is:

- FIFO
- LIFO
- ...

Default value for **K** and **Z**:

- **K**= $\infty$
- **Z**= FIFO

# Modeling and Simulating Discrete Event Systems

## Kendall's notation: examples

### One server

- $M/M/1/\infty/FIFO \Leftrightarrow M/M/1$ : Exponential inter-arrival and service times, 1 server, infinite capacity, FIFO service discipline
- $M/M/1/10$ : Exponential inter-arrival and service times, 1 server, capacity of 10 customers, FIFO service discipline
- $D/D/1$ : Deterministic (constant) inter-arrival and service times, 1 server, infinite capacity, FIFO service discipline.

### Multiple servers :

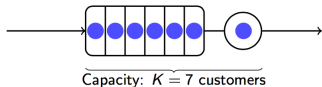
- $M/M/8/100$ : exponential inter-arrival and service times, 8 servers, capacity of 100 customers, FIFO service discipline
- $G/G/10$ : General inter-arrival and service times, 10 servers, FIFO service discipline.

# Modeling and Simulating Discrete Event Systems

## Kendall's notation: examples

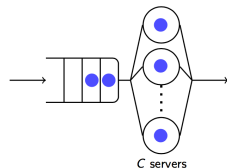
Define the Kendall's notation of the following queuing systems:

Poisson arrivals, Exponential service times



**M/M/1/7 queue**

Poisson arrivals, Exponential service times

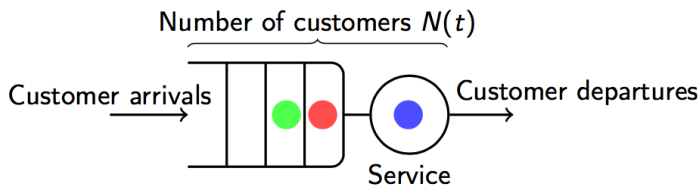


**M/M/C queue**

# Modeling and Simulating Discrete Event Systems

## Introduction to the M/M/1 Queue

**Example of M/M/1 Queue:** Consider the following queuing system with:

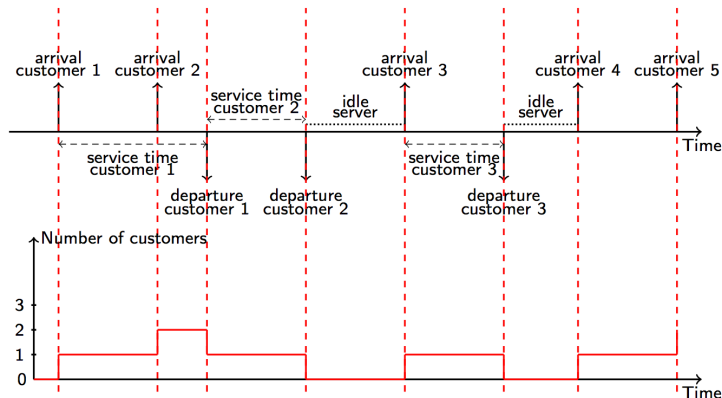


- Customer arrival process (e.g.: Poisson arrivals ( $\lambda$ )),
- Customer departure process (e.g.: Exponential service times ( $\mu$ )),
- 1 server,
- $\infty$  buffer capacity,
- FIFO (First In First Out) service discipline

# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Evolution of the number of customers, $N(t)$ , in the Queue

- Arrivals:  $\uparrow$
- Departures:  $\downarrow$
- idle: the system is empty

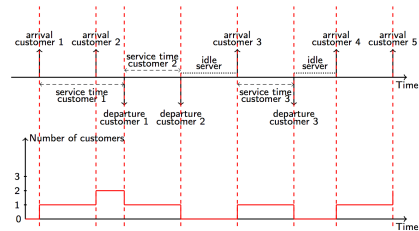


# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Evolution of the number of customers, $N(t)$ , in the Queue

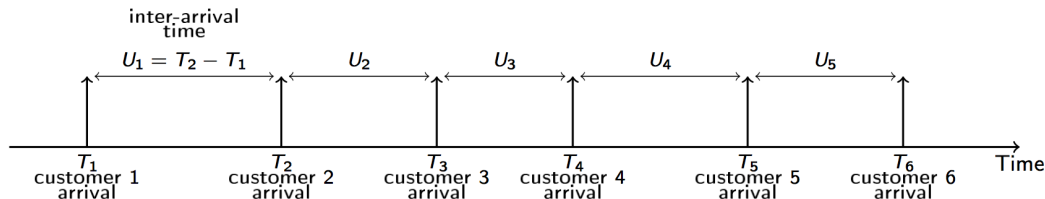
### Mathematical modeling:

- The evolution over time of  $N(t)$  in the system (waiting or being served) is modeled by a random process.
- The stationary state of this random process is studied in order to estimate average performance
  - Average waiting time in Queues,
  - Average number of customers in systems
  - Average number of customers in Queues
  - Mean utilization rate, etc....



# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Customer Arrivals

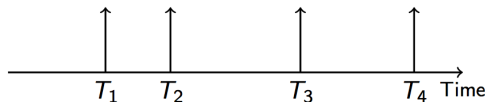
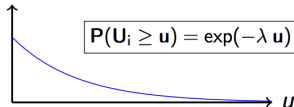


- $T_i$ : Arrival time of customer number  $i$ .
- $U_i = T_{i+1} - T_i$ , inter-arrival time between customers  $i$  and  $i + 1$
- In simple models, we assume that the successive inter-arrival times  $U_1, U_2$ , etc...
  - are **independent**
  - follow the same **probability distribution**
- The customer arrival process is then fully characterized by the law of **inter-arrival**

# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Customer Arrivals

- The **Poisson process** is an important particular case of **arrival processes**.
  - The successive inter-arrival times  $U_1, U_2, U_3$ , etc... are independent, and distributed according to the **exponential law** with parameter  $\lambda$
  - Therefore, the **arrival process** is said to be a **Poisson process** of rate  $\lambda$
- **Arrival rate  $\lambda$** 
  - The arrival rate  $\lambda$  is the average number of arrivals per time unit (e.g., customers/second).
  - The average inter-arrival time equals  $1/\lambda$  (time units, e.g. seconds).



Exponential interarrival times



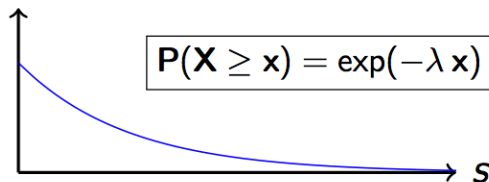
Poisson arrival process



# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Customer Arrivals - Exponential distribution

- Let us consider  $X \sim \text{Exp}(\lambda)$ .



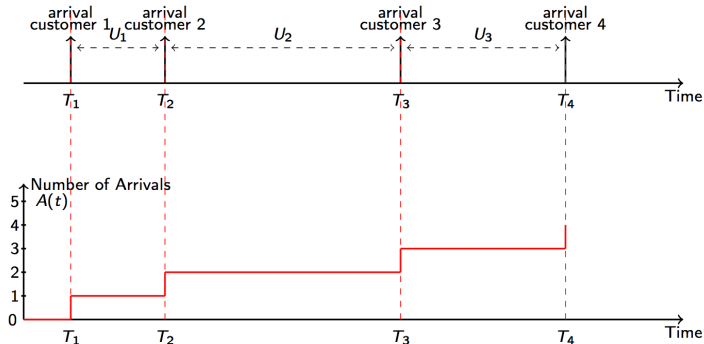
- Mean, variance and squared coefficient of variation:**

Mean:	$E(X) = 1/\lambda$
Variance:	$\text{var}(X) = E(X^2) - (E(X))^2 = E((X - E(X))^2) = 1/\lambda^2$
Squared Coefficient of Variation:	$Cv^2 = \text{var}(X) = 1, \text{ for all } \lambda$

# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Customer Arrivals

- Characterizing Arrivals with the Counting Process

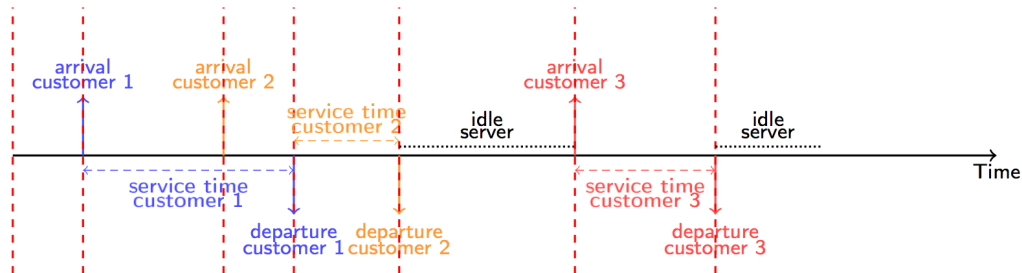


$$\begin{aligned} P(A(t + \Delta t) = n + 1 | A(t) = n) &= \lambda \Delta t \\ P(A(t + \Delta t) = n | A(t) = n) &= 1 - \lambda \Delta t \end{aligned} \quad (1)$$

# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Service Duration

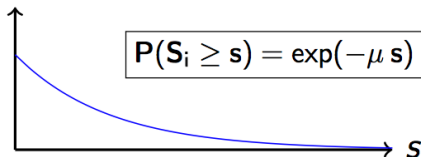
The service time  $S_i$  of customer  $i$  is the time during which a server is busy serving customer  $i$ .



# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Customer Departures - Exponential service duration

- In general, it is assumed that:
  - service times are independent and identically distributed;
  - they are characterized by their probability distribution.
- The **exponential distribution** is an important particular case of service duration.



- **Service rate  $\mu$ :**
  - The **service rate**  $\mu$  is the average number of clients served per time unit if the server is always busy (e.g. customers/second).
  - The **mean service duration** is  $1/\mu$  (time units, e.g. seconds).

# Modeling and Simulating Discrete Event Systems

## M/M/1 Queue: Offered load

- Notations:
  - The **arrival rate**  $\lambda$  is the average number of arrivals per time unit.
  - The **service rate**  $\mu$  is the average number of customers a server is able to handle if it is always busy. Equivalently,  $1/\mu$  is the average service duration.
- Average performance depends on the **offered load**  $\rho$ , defined as

$$\rho = \frac{\lambda}{\mu}$$

- The **offered load** is an important parameter. It gives an indication about the steady state behavior of the system (System Stability)
  - Under what conditions is a steady state established ?
  - The stability condition is:

$$\lambda < \mu$$

- The queue is stable if  $\rho < 1$ , otherwise it is **overloaded**.

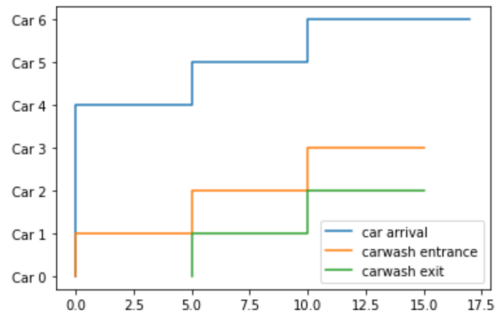
# Modeling and Simulating Discrete Event Systems

## Carwash Simulation Example with Simpy

```
1 # Install a pip package in the current Jupyter kernel
2 import sys
3 ![sys.executable} -m pip install simpy
4 import simpy
```

### Carwash

Car 0 arrives at the carwash at 0.00.  
Car 1 arrives at the carwash at 0.00.  
Car 2 arrives at the carwash at 0.00.  
Car 3 arrives at the carwash at 0.00.  
Car 0 enters the carwash at 0.00.  
Car 4 arrives at the carwash at 5.00.  
Carwash removed 97% of Car 0's dirt.  
Car 0 leaves the carwash at 5.00.  
Car 1 enters the carwash at 5.00.  
Car 5 arrives at the carwash at 10.00.  
Carwash removed 65% of Car 1's dirt.  
Car 1 leaves the carwash at 10.00.  
Car 2 enters the carwash at 10.00.  
Carwash removed 64% of Car 2's dirt.  
Car 2 leaves the carwash at 15.00.



# End of Lecture 7:

## Modeling and Simulating Discrete Event Systems