

# **CS-417**

# **COMPUTER SYSTEMS MODELING**

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**(LECTURE # 9)**

**FAKHRA AFTAB**  
**LECTURER**

**DEPARTMENT OF COMPUTER & INFORMATION SYSTEMS ENGINEERING**  
**NED UNIVERSITY OF ENGINEERING & TECHNOLOGY**



# Recap of Lecture # 8

Poisson Distribution

Hypergeometric, Uniform, Constant, Indicator & Multinomial Distribution

Exponential Distribution

Rare Events & their Inter-Arrival Times

Memoryless Property



## Chapter # 3 (Cont'd)

# REVIEW OF PROBABILITY THEORY



# Conditional Law of Probability

- It is sometimes useful to interpret  $P[A]$  as our knowledge of the occurrence of event  $A$  before an experiment takes place.
- Sometimes, we refer to  $P[A]$  as the *a priori* probability of  $A$ .

Given the two events  $E$  and  $F$  with  $P\{F\} > 0$ , the conditional probability of  $E$  given  $F$ ,  $P\{E|F\}$ , is defined as

$$P\{E|F\} = \frac{P\{EF\}}{P\{F\}}, \quad P\{F\} > 0$$

If  $E$  is a subset of (i.e., contained in)  $F$ , then  $P\{EF\} = P\{E\}$ .

The two events,  $E$  and  $F$ , are *independent* if, and only if,

$$P\{E|F\} = P\{E\}$$

In this case, the conditional probability law reduces to

$$P\{EF\} = P\{E\}P\{F\}$$



# Conditional Law of Probability

## ➤ Example 1:

You are playing a game in which another person is rolling a die. You cannot see the die, but you are given information about the outcomes. Your job is to predict the outcome of each roll. Determine the probability that the outcome is a 6, given that you are told that the roll has turned up an even number.

Let  $E = \{6\}$ , and define  $F = \{2, 4, \text{ or } 6\}$ . Thus,

$$P\{E|F\} = \frac{P\{EF\}}{P\{F\}} = \frac{P\{E\}}{P\{F\}} = \left(\frac{1/6}{1/2}\right) = \frac{1}{3}$$

Note that  $P\{EF\} = P\{E\}$  because  $E$  is a subset of  $F$ .



# Memoryless Property

- Let  $T$  be a random variable denoting service time of a server.
- Suppose that a job currently with the server has already consume service time  $t$ .
- We are interested in the probability that the job will stay with the server for additional time  $s$ ; i.e. we wish to calculate the conditional probability  $P[T > t + s \mid T > t]$

$$P[T > t + s \mid T > t] = \frac{P[T > t + s]}{P[T > t]} = \frac{e^{-\lambda(t+s)}}{e^{-\lambda t}} = e^{-\lambda s}$$



# Recall Formulae

- If  $\lambda$  is the rate at which Poisson events occur, then the distribution of time between successive arrivals,  $t$ ,  $f(t) = \lambda e^{-\lambda t}, t \geq 0$ . This is the Probability Density Function (PDF).
- $1/\lambda$  is the average time interval between successive events.
- The Cumulative Distribution Function (CDF) of Exponential Distribution is given by:

$$\begin{aligned} P\{t \leq T\} &= \int_0^T \lambda e^{-\lambda t} dt \\ &= 1 - e^{-\lambda T} \end{aligned}$$

- Additionally,  $P(t > T) = e^{-\lambda T}$
- $P(a < t \leq b) = e^{-\lambda a} - e^{-\lambda b}$
- Memoryless Property:  $P(X > x + k | X > x) = P(X > k)$



# Percentile

- A **percentile** is a measure used in statistics indicating the value below which a given percentage of observations in a group of observations falls.
- For example, the 20th percentile is the value (or score) below which 20% of the observations may be found. Equivalently, 80% of the observations are found above the 20th percentile.
- The 25th percentile is also called the first quartile.
- The 50th percentile is generally the median.
- The 75th percentile is also called the third quartile.





# rth Percentile Value

The rth percentile value  $\Pi(r)$ , for any random variable  $X$ , is defined by:

$$P[X \leq \Pi(r)] = r/100$$

Thus the 90<sup>th</sup> percentile value of an exponential random variable is defined by:

$$P[X \leq \Pi(90)] = 0.9 \quad \text{or} \quad 1 - e^{-\lambda \Pi(90)} = 0.9$$

Hence,

$$e^{-\lambda \Pi(90)} = 0.1$$

Taking natural logarithm on both sides,

$$-\lambda \Pi(90) = \ln(0.1)$$

$$\Pi(90) = \frac{\ln(0.1)^{-1}}{\lambda}$$

$$\Pi(90) = E[X] \ln(10)$$

$$\Pi(90) = 2.3 E[X]$$



# Example Problem 1

Personnel of the “Farout Engineering Company” use an online terminal to make routine engineering calculations. If the time each engineer spends in the session at a terminal has an exponential distribution with an average value of 36 minutes, find:

- a) The probability that engineer will spend 30 minutes or less at the terminal.
- b) The probability that an engineer will use it for more than an hour.
- c) If the engineer has already been at the terminal for 30 minutes, what is the probability that he will spend more than another hour at the terminal?
- d) 90% of the session ends in “R” minutes. What is R ?



# Solution

$E[X] = 36 \text{ min}$       Therefore,  $\lambda = 1/36$

1)  $P(t \leq 30) = 1 - e^{-30/36} = \underline{0.5654}$

2)  $P(t > 60) = e^{-60/36} = \underline{0.1888}$

3) It does not matter. If the engineer is using the terminal already, according to memoryless property, it has no impact that he will use it for another 1 hour. Hence  $P(t > 60) = \underline{0.1888}$

4)  $R = \Pi(90) = 2.3 E[X] = 2.3 * 36 = \underline{82.8 \text{ minutes}}$



# Example Problem 2

Requests arrive at a database server randomly, every 12 msec on the average, following negative-exponential distribution. Determine the probability that the inter-arrival time of requests:

- does not exceed 4 msec
- does exceed 4 msec
- falls between 3 msec and 6 msec (inclusive)

## Answers:

- 0.2834
- 0.7165
- 0.17227



## Example Problem 3

Consider a web server with Poisson arrival stream at an average rate of 60/hour. Calculate the probability that the inter-arrival time is:

- longer than 4 min,
- between 2 and 6 min,
- shorter than 8 min.

### Answers:

- 0.0183
- 0.1328
- 0.9996



# Example Problem 4

## Example

Cars arrive at a gas station randomly every 2 minutes, on the average. Determine the probability that the interarrival time of cars does not exceed 1 minute.

The desired probability is of the form  $P\{x \leq A\}$ , where  $A = 1$  minute in the present example. The determination of this probability is the same as computing the CDF of  $x$ —namely,

$$\begin{aligned} P\{x \leq A\} &= \int_0^A \lambda e^{-\lambda x} dx \\ &= -e^{-\lambda x} \Big|_0^A \\ &= 1 - e^{-\lambda A} \end{aligned}$$

The arrival rate for the example is computed as

$$\lambda = \frac{1}{2} \text{ arrival per minute}$$

Thus,

$$P\{x \leq 1\} = 1 - e^{-(\frac{1}{2})(1)} = .3934$$



# Tasks

Q1) Derive the generalized formula for  $r$ th Percentile value for the random variable  $X$  of Exponential Distribution. Also determine the value of  $\Pi_x(95)$ .

Q2) A modem transmits a data packet over phone line every 50 milliseconds, on the average, following negative-exponential distribution. Calculate the probability (up to 4 decimal places) that the inter-arrival time of transmissions

- does not exceed 30 msec.
- exceeds 50 msec.
- is between 30 msec and 50 msec.



# Bayes' Theorem

- Bayes' theorem can be used to calculate the probability that a certain event will occur or that a certain proposition is true, given that we already know a related piece of information.
- $P(B)$  is called the **prior probability** of  $B$ .
- $P(B | A)$ , being called the conditional probability, is also known as the **posterior probability** of  $B$ .
- Let us briefly examine how Bayes' theorem is derived: We can deduce a further equation from the following **product rule**:-

$$P(A \wedge B) = P(A | B)P(B)$$

- Note that due to the commutativity of  $\wedge$ , we can also write

$$P(A \wedge B) = P(B | A)P(A)$$

- Hence, we can deduce:-

$$P(B | A)P(A) = P(A | B)P(B)$$





# Bayes' Theorem (Cont'd)

- This can then be rearranged to give Bayes' theorem:-

$$P(B|A) = \frac{P(A|B) \cdot P(B)}{P(A)}$$

## ➤ Example: Medical Diagnosis

- Let us examine a simple example to illustrate the use of Bayes' theorem for the purposes of medical diagnosis.
- When one has a cold, one usually has a high temperature (let us say, 80% of the time).
  - $A$ : "I have a high temperature" and
  - $B$ : "I have a cold"
- Therefore, we can write this statement of posterior probability as

$$P(A | B) = 0.8$$



## Bayes' Theorem (Cont'd)

- Note that in this case, we are using  $A$  and  $B$  to represent pieces of data that could each either be a hypothesis or a piece of evidence.
- Now, let us suppose that we also know that at any one time around 1 in every 10,000 people has a cold, and that 1 in every 1000 people has a high temperature.
- We can write these prior probabilities as
  - $P(A) = 0.001$
  - $P(B) = 0.0001$



## Bayes' Theorem (Cont'd)

- Now suppose that you have a high temperature.
- What is the likelihood that you have a cold?
- This can be calculated very simply by using Bayes' theorem:

$$\begin{aligned}P(B|A) &= \frac{P(A|B) \cdot P(B)}{P(A)} \\&= \frac{0.8 \cdot 0.0001}{0.001} \\&= 0.008\end{aligned}$$

- Hence, we have shown that just because you have a high temperature does not necessarily make it very likely that you have a cold—in fact; the chances of cold are just 8 in 1000.
- **Task: Recall the Axioms of Probability**



