

## **Lecture 09 - 10**

### **Chapter 02: Random Variables**

#### **Math 4441: Probability and Statistics**

Reference: Goodman & Yates – Introduction to Probability and Stochastic Process, 3<sup>rd</sup> Edition

## Expected Values

Example: Suppose you play a game, where in each play you lose \$1 with probability 0.60 and you win \$1, \$2 and \$3 with probabilities 0.30, 0.08 and 0.02, respectively. Find your net gain in each play of the game.

- If you play only a few times, your gain might depend on your luck
- If you play a large number of times – your luck no longer determines the total gain – rather it will be multiple of average gain

Say, you play the game  $n$  times, approximately in  $0.6 \times n$  games you will lose \$1 per game and will win \$1, \$2 and \$3 in  $0.3 \times n$ ,  $0.08 \times n$  and  $0.02 \times n$  games, respectively.



Your total gain is

$$(0.6) \times n \cdot (-1) + (0.3) \times n \cdot (1) + (0.08) \times n \cdot (2) + (0.02) \times n \cdot (3) = (-0.08) \times n.$$

Average loss of \$0.08 per game

The more you play, the less luck interferes and your loss comes to \$0.08 per game

- Let  $X$  denote the gain in one play
- Then -0.08 is the expected value of  $X$
- $E[X], \mu_X$ , or  $\mu$  is the expected value of  $X$

$x$	-1	1	2	3	
$P_X(x)$	0.60	0.30	0.08	0.02	1.00
$x \cdot P_X(x)$	-0.60	0.30	0.16	0.06	-0.08

$$P_X(x) = \begin{cases} 0.60, & x = -1 \\ 0.30, & x = 1 \\ 0.08, & x = 2 \\ 0.02, & x = 3 \\ 0, & o/w \end{cases}$$

$$-1 \times 0.6n + 1 \times 0.3n + 2 \times 0.08n + 3 \times 0.02n = -0.08n$$

$$\text{Or, } -1 \times 0.6 + 1 \times 0.3 + 2 \times 0.08 + 3 \times 0.02 = -0.08$$

$$\text{Or, } -1 \times P_X(-1) + 1 \times P_X(1) + 2 \times P_X(2) + 3 \times P_X(3) = -0.08$$

$$E[X] = \sum_{x \in S_X} x \cdot P_X(x)$$

Definition (Expected Value): The *expected value* of a discrete random variable  $X$  with the set of possible values  $S_X$  and probability mass function  $P_X(x)$  is defined by

$$E[X] = \sum_{x \in S_X} x P_X(x)$$

$E[X]$  exists if this sum converges

The expected value of a random variable  $X$  is also called the **mean**, or the **mathematical expectation**, or simply the **expectation** of  $X$ . It is denoted by  $E[X]$ ,  $\mu_X$ , or simply  $\mu$ .

**Example:** Three packets are sent from one computer to another. Each packet is successfully sent with a probability of 0.8 irrespective of the others. Each successfully sent packet results in a reward of 1 unit whereas 1 unit penalty (negative reward) is imposed if none of the packets is sent successfully. Find the expected net reward.





Let  $X$  is the net reward.  $X$  is a discrete random variable with values

$$S_X = \{-1, 1, 2, 3\}$$

$$P[X = -1] = 0.008$$

$$P[X = 1] = 0.096$$

$$P[X = 2] = 0.384$$

$$P[X = 3] = 0.512$$

$$E[X] = -1 \times 0.008 + 1 \times 0.096 + 2 \times 0.384 + 3 \times 0.512$$

Example: A college mathematics department sends 8 to 12 professors to the annual meeting of the American Mathematical Society, which lasts five days. The hotel at which the conference is held offers a bargain rate of  $a$  dollars per day per person if reservations are made 45 or more days in advance, but charges a cancellation fee of  $2a$  dollars per person. The department is not certain how many professors will go. However, from past experience it is known that the probability of the attendance of  $i$  professors is  $1/5$  for  $i = 8, 9, 10, 11,$  and  $12$ . If the regular rate of the hotel is  $2a$  dollars per day per person, should the department make any reservations? If so, how many?

$$E(X_8) = (40a)\frac{1}{5} + (50a)\frac{1}{5} + (60a)\frac{1}{5} + (70a)\frac{1}{5} + (80a)\frac{1}{5} = 60a$$



$$E(X_9) = (42a)\frac{1}{5} + (45a)\frac{1}{5} + (55a)\frac{1}{5} + (65a)\frac{1}{5} + (75a)\frac{1}{5} = 56.4a,$$

$$E(X_{10}) = (44a)\frac{1}{5} + (47a)\frac{1}{5} + (50a)\frac{1}{5} + (60a)\frac{1}{5} + (70a)\frac{1}{5} = 54.2a,$$

$$E(X_{11}) = (46a)\frac{1}{5} + (49a)\frac{1}{5} + (52a)\frac{1}{5} + (55a)\frac{1}{5} + (65a)\frac{1}{5} = 53.4a,$$

$$E(X_{12}) = (48a)\frac{1}{5} + (51a)\frac{1}{5} + (54a)\frac{1}{5} + (57a)\frac{1}{5} + (60a)\frac{1}{5} = 54a.$$

$X_{11}$  has the smallest expected value. Therefore, making 11 reservation is the most reasonable policy

## Derived Random Variable:

If  $x$  is a sample value of a random variable  $X$ , and  $y$  is a sample value of a random variable  $Y$ . If we have a function  $Y = g(X)$  and we get the values of  $Y$  from the values of  $X$ . Because we obtain  $Y$  from another random variable, we refer to  $Y$  as a **derived random variable**.

**Definition 2.4** (Derived Random Variable). If  $X$  and  $Y$  are two random variables, and each sample value  $y$  of  $Y$  is a mathematical function  $g(x)$  of  $X$ , then the relationship of these two random variables is given by

$$Y = g(X),$$

and the random variable  $Y$  is called a derived random variable.

What will be the PMF of  $Y$ ?

For a discrete random variable  $Y$ , if there exists another random variable  $X$ , where  $Y = g(X)$ , then the PMF of  $X$  is

$$P_Y(y) = \sum_{x:g(x)=y} P_X(x)$$

Expected Value of a derived random variable





$$E[X + Y] = ?$$

$$E[aX + b] = ?$$

$n$ -th Moment of a Random Variable

Variance: Variance measures the average magnitude of the fluctuations of a random variable from its expectations.

$$E[X - E[X]] = E[X - \mu] = E[X] - \mu = E[X] - E[X] = 0.$$

Hence,  $E[X - E[X]]$  is not an appropriate measure for the variance.

if we consider  $E(|X - E[X]|)$  instead, the problem of negative and positive deviations canceling each other disappears.

mathematically,  $E(|X - E[X]|)$  is difficult to handle

The quantity  $E[(X - E[X])^2]$  is used to measure the average amount of fluctuations, and is called the variance of  $X$

The square root of  $E[(X - E[X])^2]$  is called the **standard deviation** of  $X$ .

**Definition 2.7.** Let  $X$  be a discrete random variable with a set of possible values  $S_X$ , probability mass function  $P_X(x)$ , and  $E(X) = \mu$ . Then  $\sigma_X$  and  $Var[X]$ , called the **standard deviation** and the **variance** of  $X$ , respectively, and are defined by

$$\sigma_X = \sqrt{E[(X - \mu)^2]},$$

and

$$Var[X] = E[(X - \mu)^2].$$

Note that by this definition, we have

$$Var[X] = E[(X - \mu)^2] = \sum_{x \in S_X} (x - \mu)^2 P_X(x).$$

$$\text{Var}[X] = E[X^2] - (E[X])^2.$$

*Proof.*

$$\begin{aligned}\text{Var}[X] &= E[(X - \mu)^2] \\ &= E[X^2 - 2\mu X + \mu^2] \\ &= E[X^2] - 2\mu E[X] + \mu^2 \\ &= E[X^2] - 2\mu^2 + \mu^2 = E[X^2] - \mu^2 \\ &= E[X^2] - (E[X])^2.\end{aligned}$$

$Var[X] = 0$  if and only if  $X$  is a constant with probability 1.

$$Var[aX + b] = a^2 Var[X]$$

$$P_X(x) = \begin{cases} 0.60, & x = -1 \\ 0.30, & x = 1 \\ 0.08, & x = 2 \\ 0.02, & x = 3 \\ 0, & o/w \end{cases}$$

$$E[X] = \mu = 0.08$$

$$V[X] = [-1 - (-0.08) \times 0.60 + (1 - 0.08) \times 0.30 + (2 - 0.08) \times 0.08 + (3 - .08) \times 0.02$$



## n-th Central Moment

## Expected Value and Variance for Continuous Random Variables

$$E[X] = \int_{-\infty}^{+\infty} x \cdot f_X(x) dx$$

$$V[X] = \int_{-\infty}^{+\infty} (x - \mu)^2 \cdot f_X(x) dx$$

Example: The time elapsed, in minutes, between the placement of an order of pizza and its delivery is random with pdf

$$f(x) = \begin{cases} 1/15 & \text{if } 25 < x < 40 \\ 0 & \text{otherwise.} \end{cases}$$

$$E(X) = \int_{25}^{40} x \frac{1}{15} dx = 32.5,$$

$$E(X^2) = \int_{25}^{40} x^2 \frac{1}{15} dx = 1075.$$

$$V[X] = 1075 - (32.5)^2 = 18.75$$

$$\sigma_X = \sqrt{V[X]} = 4.33$$

Sketch the graph of the function

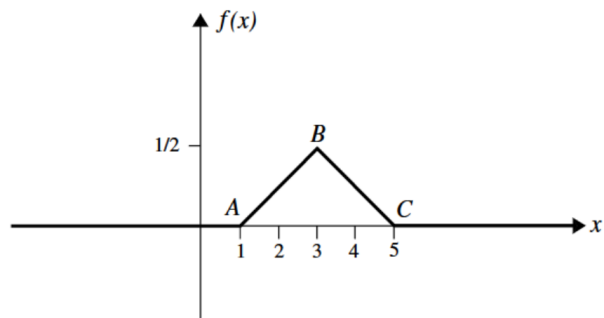
$$f(x) = \begin{cases} \frac{1}{2} - \frac{1}{4}|x - 3| & 1 \leq x \leq 5 \\ 0 & \text{otherwise,} \end{cases}$$

and show that it is the probability density function of a random variable  $X$ .

Find  $F$ , the distribution function of  $X$ , and show that it is continuous.

Find  $E[X]$  and  $V[X]$ .

$$f(x) = \begin{cases} 0 & x < 1 \\ \frac{1}{2} + \frac{1}{4}(x - 3) & 1 \leq x < 3 \\ \frac{1}{2} - \frac{1}{4}(x - 3) & 3 \leq x < 5 \\ 0 & x \geq 5. \end{cases}$$



For  $t < 1$ ,

$$F(t) = \int_{-\infty}^t 0 \, dx = 0;$$

for  $1 \leq t < 3$ ,

$$F(t) = \int_{-\infty}^t f(x) \, dx = \int_1^t \left[ \frac{1}{2} + \frac{1}{4}(x-3) \right] dx = \frac{1}{8}t^2 - \frac{1}{4}t + \frac{1}{8};$$

for  $3 \leq t < 5$ ,

$$\begin{aligned} F(t) &= \int_{-\infty}^t f(x) \, dx = \int_1^3 \left[ \frac{1}{2} + \frac{1}{4}(x-3) \right] dx + \int_3^t \left[ \frac{1}{2} - \frac{1}{4}(x-3) \right] dx \\ &= \frac{1}{2} + \left( -\frac{1}{8}t^2 + \frac{5}{4}t - \frac{21}{8} \right) = -\frac{1}{8}t^2 + \frac{5}{4}t - \frac{17}{8}; \end{aligned}$$

and for  $t \geq 5$ ,

$$\begin{aligned} F(t) &= \int_{-\infty}^t f(x) \, dx = \int_1^3 \left[ \frac{1}{2} + \frac{1}{4}(x-3) \right] dx + \int_3^5 \left[ \frac{1}{2} - \frac{1}{4}(x-3) \right] dx \\ &= \frac{1}{2} + \frac{1}{2} = 1. \end{aligned}$$