# Lecture No. 10 Functions of a Random Variable

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- In many practical situations, we observe sample values of a random variable and use these sample values to compute other quantities.
- One example that occurs frequently is an experiment in which the procedure is to measure the power level of the received signal in a cellular telephone.
- An observation is x, the power level in units of milliwatts. Frequently engineers convert the measurements to decibels by calculating  $y = 10 \log_{10} x \, dBm$  (decibels with respect to one milliwatt).

• If *x* is a sample value of a random variable *X*, implies that *y* is a sample value of a random variable *Y*.

• Because we obtain *Y* from another random variable, we refer to *Y* as a derived random variable.

#### Definition Derived Random Variable

Each sample value y of a **derived random variable** Y is a mathematical function g(x) of a sample value x of another random variable X. We adopt the notation Y = g(X) to describe the relationship of the two random variables.

- If Y = g(X) is a function of a random variable X, then Y is also a random variable, since it provides a numerical value for each possible outcome.
- This is because every outcome in the sample space defines a numerical value x for X and hence also the numerical value y = g(x) for Y.
- If X is discrete with PMF  $P_X$ , then Y is also discrete, and its PMF  $P_Y$  can be calculated using the PMF of X.
- In particular, to obtain  $P_Y(y)$  for any y, we add the probabilities of all values of x such that g(x) = y:

$$p_Y(y) = \sum_{\{x \mid g(x)=y\}} p_X(x).$$

#### Example

• Let Y = |X| and let us apply the preceding formula for the PMF  $P_Y$  to the case where

$$p_X(x) = \begin{cases} 1/9 & \text{if } x \text{ is an integer in the range } [-4,4], \\ 0 & \text{otherwise.} \end{cases}$$

• The possible values of Y are y = 0, 1, 2, 3, 4. To compute  $P_Y(y)$  for some given value y from this range, we must add  $P_X(x)$  over all values x such that |x| = y. In particular, there is only one value of X that corresponds to y = 0, namely x = 0.

#### Example

$$p_Y(0) = p_X(0) = \frac{1}{9}.$$

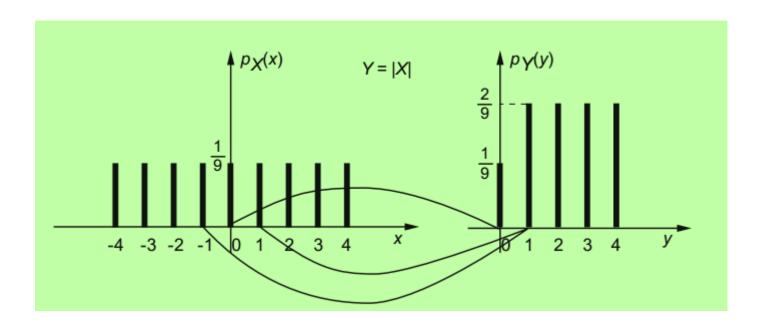
Also, there are two values of *X* that correspond to each *y*= 1, 2, 3, 4 so for example,

$$p_Y(1) = p_X(-1) + p_X(1) = \frac{2}{9}.$$

Thus, the PMF of Y is

$$p_Y(y) = \begin{cases} 2/9 & \text{if } y = 1, 2, 3, 4, \\ 1/9 & \text{if } y = 0, \\ 0 & \text{otherwise.} \end{cases}$$

#### Example



PMF of X and 
$$Y = |X|$$

## **Expected Value for Functions of Random Variables**

#### **Expected Value Rule for Functions of Random Variables**

• Let X be a random variable with PMF  $P_X(x)$ , and let g(X) be a real valued function of X. Then, the expected value of the random variable g(X) is given by

$$\mathbf{E}[g(X)] = \sum_{x} g(x) p_X(x).$$

Monitor three phone calls and observe whether each one is a voice call or a data call.

The random variable N is the number of voice calls. Assume N has **PMF** 

$$P_N(n) = \begin{cases} 0.1 & n = 0, \\ 0.3 & n = 1, 2, 3, \\ 0 & otherwise. \end{cases}$$

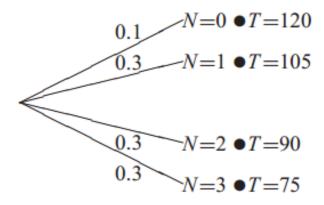
Voice calls cost 25 cents each and data calls cost 40 cents each. T cents is the cost of the three telephone calls monitored in the experiment.

(1) Express T as a function of N. (2) Find PT (t) and E[T].

(1) As a function of N, the cost T is

$$T = 25N + 40(3 - N) = 120 - 15N$$

(2) To find the PMF of T, we can draw the following tree:



From the tree, we can write down the PMF of T:

$$P_T(t) = \begin{cases} 0.3 & t = 75, 90, 105 \\ 0.1 & t = 120 \\ 0 & \text{otherwise} \end{cases}$$

From the PMF  $P_T(t)$ , the expected value of T is

$$E[T] = 75P_T(75) + 90P_T(90) + 105P_T(105) + 120P_T(120)$$
$$= (75 + 90 + 105)(0.3) + 120(0.1) = 62$$

#### Theorem 1

For any random variable X

$$E[X-\mu_X]=0$$

**Proof** Defining  $g(X) = X - \mu_X$ 

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

The first term on the right side is  $\mu_X$  by definition. In the second term,  $\sum_{x \in S_X} P_X(x) = 1$ , so both terms on the right side are  $\mu_X$  and the difference is zero.

**Theorem 2** For any random variable X

$$E[aX+b]=aE[X]+b$$

Theorem 3

The variance of random variable X

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

$$\begin{split} E[(X - \mu_X)^2] &= \sum_x (x - \mu_X)^2 \, p_X(x) \\ &= \sum_x (x^2 + \mu_X^2 - 2x\mu_X) \, p_X(x) \\ &= \sum_x x^2 p_X(x) + \mu_X^2 \sum_x p_X(x) - 2\mu_X \sum_x x p_X(x) \\ &= E[X^2] + \mu_X^2(1) - 2\mu_X E[X] \\ &= E[X^2] + \mu_X^2 - 2\mu_X^2 \\ &= E[X^2] - (E[X])^2 \end{split}$$

#### **Moments** For any random variable X

- (a) The nth moment is  $E[X^n]$ .
- (b) The nth central moment is  $E[(X \mu_X)^n]$

**Theorem 4** For any random variable X

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

## **Proof**: $Var[aX + b] = a^2 Var[X]$

Let 
$$Y = aX + b$$
 
$$Var[Y] = E[Y^2] - (E[Y])^2$$

$$E[Y^2] = E[a^2X^2 + 2abX + b^2] = a^2E[X^2] + 2ab\mu_X + b^2$$

$$E[Y] = aE[X] + b$$

$$(E[Y])^2 = (aE[X] + b)^2$$

$$\mu_Y^2 = a^2\mu_X^2 + 2ab\mu_X + b^2$$

$$Var[Y] = a^2E[X^2] + 2ab\mu_X + b^2 - (a^2\mu_X^2 + 2ab\mu_X + b^2)$$

$$Var[Y] = a^2E[X^2] - a^2\mu_Y^2$$

 $= a^{2}(E[X^{2}] - \mu_{Y}^{2}) = a^{2}Var[X]$ 

The amplitude V (volts) of a sinusoidal signal is a random variable with PMF

$$P_{V}(v) = \begin{cases} 1/7 & v = -3, -2, \dots, 3, \\ 0 & \text{otherwise.} \end{cases}$$

A new voltmeter records the amplitude U in millivolts.

- (1) Express U as a function of V.
- (2) What is the variance of U?

(1) Express U as a function of V.

$$U = 1000V$$

#### (2) What is the variance of U?

We first find the variance of V. The expected value of the amplitude is

$$\mu_V = 1/7[-3 + (-2) + (-1) + 0 + 1 + 2 + 3] = 0$$
 volts.

The second moment is

$$E[V^2] = 1/7[(-3)^2 + (-2)^2 + (-1)^2 + 0^2 + 1^2 + 2^2 + 3^2] = 4 \text{ volts}^2$$

Therefore the variance is

$$Var[V] = E[V^2] - \mu_V^2 = 4 \text{ volts}^2$$

$$Var[U] = 1000^2 Var[V] = 4,000,000 millivolts^2$$

## Homework Problem

X is a Gaussian random variable with E[X] = 0 and  $P[|X| \le 10] = 0.1$ .

What is the standard deviation  $\sigma_X$ ?

## Homework Problem

X is a Gaussian random variable with zero mean but unknown variance. We do know, however, that

$$P[|X| \le 10] = 0.1\tag{1}$$

We can find the variance Var[X] by expanding the above probability in terms of the  $\Phi(\cdot)$  function.

$$P[-10 \le X \le 10] = F_X(10) - F_X(-10) = 2\Phi\left(\frac{10}{\sigma_X}\right) - 1 \tag{2}$$

This implies  $\Phi(10/\sigma_X) = 0.55$ . Using Table for the Gaussian CDF, we find that  $10/\sigma_X = 0.15$  or  $\sigma_X = 66.6$ .