

# 8<sup>th</sup> Material Subject: Joint Probability Distribution (Discrete)

Undergraduate of Telecommunication Engineering

**MUH1F3 - PROBABILITY AND STATISTICS**

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# السلام عليكم ورحمة الله وبركاته

## WELCOME

### TABLE OF CONTENTS:

1. **Joint Probability Mass Functions**
2. **Marginal Probability Mass Functions**
3. **Conditional Probability Distribution**
4. **Covariance and Correlation**

### LEARNING OBJECTIVES:

After careful study of this chapter, student should be able to do the following:

1. **Use joint probability mass functions to calculate probabilities**
2. **Calculate marginal and conditional probability distributions from joint probability distributions**
3. **Interpret and calculate covariance and correlations between random variables**

# JOINT PROBABILITY MASS FUNCTION

For simplicity, we begin by considering random experiments in which only two random variables, called **Bi-variate**. The **Joint Probability Mass Function** of the discrete random variables **X** and **Y**, denoted as  $f_{XY}(\mathbf{xy})$ , satisfies:

$$f_{XY}(\mathbf{xy}) \geq 0 \quad (1)$$

$$\sum_{\mathbf{x}} \sum_{\mathbf{y}} f_{XY}(\mathbf{xy}) = 1 \quad (2)$$

$$f_{XY}(\mathbf{xy}) = P(\mathbf{X} = \mathbf{x} \text{ and } \mathbf{Y} = \mathbf{y}) = P(\mathbf{X} = \mathbf{x}) \cap P(\mathbf{Y} = \mathbf{y}) \quad (3)$$

The **Marginal Probability Mass Function** of the discrete random variables  $\mathbf{X}$  and  $\mathbf{Y}$ , denoted as  $\mathbf{f_X(x)}$  or  $\mathbf{f_Y(y)}$ , satisfies:

$$\mathbf{f_X(x)} = \mathbf{P(X = x)} = \sum_y \mathbf{f_{XY}(x, y)} \quad (4)$$

$$\mathbf{f_Y(y)} = \mathbf{P(Y = y)} = \sum_x \mathbf{f_{XY}(x, y)} \quad (5)$$

Remember that, for a random variable  $\mathbf{X}$ , we define the CDF as  $\mathbf{F_X(x) = P(X \leq x)}$ . Now, if we have two random variables  $\mathbf{X}$  and  $\mathbf{Y}$  and we would like to study them jointly, we can define the **Joint Cumulative Function** as follows:

$$\mathbf{F_{XY}(x, y) = P(X \leq x \text{ and } Y \leq y) = P(X \leq x) \cap P(Y \leq y)} \quad (6)$$

# INDEPENDENT BIVARIATE

The random variable **X** and **Y** become **independent**, if only:

$$f_{XY}(x, y) = P(X = x) \cdot P(Y = y) = f_X(x) \cdot f_Y(y) \quad (7)$$

or:

$$F_{XY}(x, y) = P(X \leq x) \cdot P(Y \leq y) = F_X(x) \cdot F_Y(y) \quad (8)$$



## EXAMPLE

**Example:** Will randomly pick two balls from a box that contains of three blue, two red and three green ball. If:

**X** = Random variables are declared elected as a blue ball

**Y** = Random variables are declared elected as a red ball

- a. Determine range of random variable **X**
- b. Determine range of random variable **Y**
- c. Determine range of joint random variable **X** and **Y**
- d. Determine the joint PMF of **X** and **Y**
- e. Determine the marginal PMF of **X**
- f. Determine the marginal PMF of **Y**
- g. Are the random variables **X** and **Y** independent?
- h. If your answers are not independent, specify **Cov(XY)]** and  $\rho_{XY}$



## EXAMPLE

**Answer:**

- a. Since **X** is a random variables declared elected as a blue ball, the range of **X** will:

$$R_X = \{0, 1, 2\}$$

- b. Since **Y** is a random variables declared elected as a red ball, the range of **Y** will:

$$R_Y = \{0, 1, 2\}$$

- c. And range of joint random variable **X** and **Y**

$$R_{XY} = \{(0, 0), (0, 1), (0, 2), (1, 0), (1, 1), (2, 0)\}$$

## EXAMPLE

d. The joint PMF of  $X$  and  $Y$

Suppose  $X = 0$  and  $Y = 0$ , meaning that no blue or red balls are drawn. The two balls are taken from green balls. So that:

$$f_{XY}(0, 0) = \frac{{}^3C_0 \cdot {}^2C_0 \cdot {}^3C_2}{{}^8C_2} = \frac{3}{28}$$

While  $X = 0$  and  $Y = 1$ , meaning that no blue drawn. The two balls are taken from 1 red and 1 green ball. So that:

$$f_{XY}(0, 1) = \frac{{}^3C_0 \cdot {}^2C_1 \cdot {}^3C_1}{{}^8C_2} = \frac{6}{28}$$

And  $X = 0$  and  $Y = 2$ , meaning that the two balls are taken from red. So that:

$$f_{XY}(0, 2) = \frac{{}^3C_0 \cdot {}^2C_2 \cdot {}^3C_0}{{}^8C_2} = \frac{1}{28}$$

In the same way, it can be calculated for  $f_{XY}(1, 0)$ ,  $f_{XY}(1, 1)$  and  $f_{XY}(2, 0)$ .

## EXAMPLE

The joint PMF for **X** and **Y** shown below:

		Y		
		0	1	2
X	0	3/28	6/28	1/28
	1	9/28	6/28	0
	2	3/28		0

e. The marginal PMF of **X**

$$f_X(0) = \frac{3}{28} + \frac{6}{28} + \frac{1}{28} = \frac{10}{28}$$

$$f_X(1) = \frac{9}{28} + \frac{6}{28} = \frac{15}{28} \quad \text{and} \quad f_X(2) = \frac{3}{28} = \frac{3}{28}$$

## EXAMPLE

f. Determine the marginal PMF of **Y**

$$f_Y(0) = \frac{3}{28} + \frac{9}{28} + \frac{3}{28} = \frac{15}{28} \quad , \quad f_Y(1) = \frac{6}{28} + \frac{6}{28} = \frac{12}{28} \quad \text{and} \quad f_Y(2) = \frac{1}{28}$$

g. Random variables **X** and **Y** independent if  $f_{XY}(xy) = f_X(x) \cdot f_Y(y)$

$$f_{XY}(0, 0) = f_X(0) \cdot f_Y(0)$$

$$\frac{3}{28} \neq \frac{10}{28} \cdot \frac{15}{28}$$

So, Random variables **X** and **Y** are not independent

## EXAMPLE

i. The **Cov(XY)** and  $\rho_{XY}$  are

$$E(X) = \sum x \cdot f_X(x) = \left(0 \cdot \frac{10}{28}\right) + \left(1 \cdot \frac{15}{28}\right) + \left(2 \cdot \frac{3}{28}\right) = \frac{21}{28}$$

$$E(Y) = \sum y \cdot f_Y(y) = \left(0 \cdot \frac{15}{28}\right) + \left(1 \cdot \frac{12}{28}\right) + \left(2 \cdot \frac{1}{28}\right) = \frac{14}{28}$$

$$\begin{aligned} E(XY) &= \sum x \cdot y \cdot f_{XY}(xy) = \left(0 \cdot 0 \cdot \frac{3}{28}\right) + \left(0 \cdot 1 \cdot \frac{6}{28}\right) + \left(0 \cdot 2 \cdot \frac{1}{28}\right) + \left(1 \cdot 0 \cdot \frac{9}{28}\right) \\ &+ \left(1 \cdot 1 \cdot \frac{6}{28}\right) + (1 \cdot 2 \cdot 0) + \left(2 \cdot 0 \cdot \frac{3}{28}\right) + (2 \cdot 1 \cdot 0) + (2 \cdot 2 \cdot 0) = \frac{6}{28} \end{aligned}$$

$$\text{Cov}(XY) = E(XY) - (E(X) \cdot E(Y)) = \frac{6}{28} - \left(\frac{21}{28} \cdot \frac{14}{28}\right) = -\frac{9}{56}$$

## EXAMPLE

$$E(X^2) = \sum x^2 \cdot f_X(x) = \left(0^2 \cdot \frac{10}{28}\right) + \left(1^2 \cdot \frac{15}{28}\right) + \left(2^2 \cdot \frac{3}{28}\right) = \frac{27}{28}$$

$$\text{Var}(X) = \sigma_x^2 = E(X^2) - (E(X))^2 = \frac{27}{28} - \left(\frac{21}{28}\right)^2 = \frac{45}{112} \quad \text{then } \sigma_x = \sqrt{\frac{45}{112}}$$

$$E(Y^2) = \sum y^2 \cdot f_Y(y) = \left(0^2 \cdot \frac{15}{28}\right) + \left(1^2 \cdot \frac{12}{28}\right) + \left(2^2 \cdot \frac{1}{28}\right) = \frac{16}{28}$$

$$\text{Var}(Y) = \sigma_y^2 = E(Y^2) - (E(Y))^2 = \frac{14}{28} - \left(\frac{21}{28}\right)^2 = \frac{9}{112} \quad \text{then } \sigma_y = \sqrt{\frac{9}{112}}$$

$$\rho_{XY} = \frac{-\frac{9}{56}}{\sqrt{\frac{45}{112}} \cdot \sqrt{\frac{9}{112}}} = -0.894$$

*Thank You*