# Statistical and Mathematical Methods for Data Analysis

Dr. Syed Faisal Bukhari

**Associate Professor** 

Department of Data Science

Faculty of Computing and Information Technology

University of the Punjab

#### **Textbooks**

- ☐ Probability & Statistics for Engineers & Scientists,
  Ninth Edition, Ronald E. Walpole, Raymond H.
  Myer
- ☐ Elementary Statistics: Picturing the World, 6<sup>th</sup> Edition, Ron Larson and Betsy Farber
- ☐ Elementary Statistics, 13<sup>th</sup> Edition, Mario F. Triola

#### Reference books

- ☐ Probability and Statistical Inference, Ninth Edition, Robert V. Hogg, Elliot A. Tanis, Dale L. Zimmerman
- ☐ Probability Demystified, Allan G. Bluman
- □ Practical Statistics for Data Scientists: 50 Essential Concepts, Peter Bruce and Andrew Bruce
- ☐ Schaum's Outline of Probability, Second Edition, Seymour Lipschutz, Marc Lipson
- ☐ Python for Probability, Statistics, and Machine Learning, José Unpingco

#### References

Readings for these lecture notes:

- □ Schaum's Outline of Probability, Second Edition (Schaum's Outlines)
  - by by Seymour Lipschutz, Marc Lipson
- ☐ Probability & Statistics for Engineers & Scientists,
  Ninth Edition, Ronald E. Walpole, Raymond H.
  Myer
- ☐ Introduction to Probability SECOND EDITION Dimitri P. Bertsekas and John N. Tsitsiklis
- □ https://en.wikipedia.org/wiki/Joint\_probability\_distribution

These notes contain material from the above resources.

☐ Given random variables X,Y,..., that are defined on a probability space, the **joint probability distribution** for X,Y,..., is a **probability distribution** that gives the probability that each of X,Y,..., falls in any particular range or discrete set of values specified for that variable.

☐ In the case of only **two random variables**, this is called a **bivariate distribution**, but the concept generalizes to **any number of random variables**, giving a **multivariate distribution**.

□ The joint probability distribution can be expressed either in terms of a joint cumulative distribution function or in terms of a joint probability density function (in the case of continuous variables) or joint probability mass function (in the case of discrete variables).

☐ These in turn can be used to find two other types of distributions: the marginal distribution giving the probabilities for any one of the variables with no reference to any specific ranges of values for the other variables, and the conditional probability distribution giving the probabilities for any subset of the variables conditional on particular values of the remaining variables

- □ Our study of random variables and their probability distributions in previous lectures is restricted to one-dimensional sample spaces, in that we recorded outcomes of an experiment as values assumed by a single random variable.
- ☐ There will be situations, however, where we may find it desirable to record the **simultaneous outcomes** of **several random variables**.

- □ For example, we might measure the amount of precipitate P and volume V of gas released from a controlled chemical experiment, giving rise to a two-dimensional sample space consisting of the outcomes (p, v), or
- we might be interested in the hardness *H* and tensile strength *T* of cold-drawn copper, resulting in the outcomes (*h*, *t*).

□ In a study to determine the likelihood of success in college based on high school data, we might use a three dimensional sample space and record for each individual his or her aptitude test score, high school class rank, and grade-point average at the end of freshman year in college.

- □ If X and Y are two discrete random variables, the probability distribution for their simultaneous occurrence can be represented by a function with values f(x, y) for any pair of values (x, y) within the range of the random variables X and Y.
- ☐ It is customary to refer to this function as the **joint probability distribution** of **X** and **Y**. Hence, in the discrete case,

$$f(x, y) = P(X = x, Y = y);$$

that is, the values f(x, y) give the probability that outcomes x and y occur at the same time.

The function f(x, y) is a joint probability distribution or probability mass function of the discrete random variables X and Y if

- 1.  $f(x, y) \ge 0$  for all (x, y),
- $2. \sum_{x} \sum_{y} f(x, y) = 1,$
- 3. P(X = x, Y = y) = f(x, y).

For any region A in the xy plane,

$$P[(X, Y) \in A] = \sum_{A} f(x, y).$$

Example: Two ballpoint pens are selected at random from a box that contains 3 blue pens, 2 red pens, and 3 green pens. If X is the number of blue pens selected and Y is the number of red pens selected, find

(a) the joint probability function f(x, y),

(b)  $P[(X, Y) \in A]$ , where A is the region  $\{(x, y) \mid x + y \le 1\}$ .

#### **Solution**

a) The possible pairs of values (x, y) are (0, 0), (0, 1), (1, 0), (1, 1), (0, 2), and (2, 0).

The joint probability distribution of

$$f(x, y) = \frac{\binom{3}{2}\binom{0}{2}\binom{0}{3}\binom{0}{3}\binom{0}{2-x-y}}{\binom{8}{2}}, \text{ for } x = 0, 1, 2; y = 0, 1, 2; \text{ and } 0 \le x + y \le 2.$$

$$f(0, 0) = \frac{\binom{3}{3}\binom{0}{2}\binom{2}{3}\binom{3}{3}\binom{2}{2-0-0}}{\binom{8}{2}} = \frac{3}{28}$$

$$f(0, 1) = \frac{\binom{3}{3}\binom{0}{2}\binom{2}{2}\binom{3}{3}\binom{3}{3}\binom{2}{2-0-1}}{\binom{8}{2}\binom{2}{3}} = \frac{6}{28}$$

$$f(1, 0) = \frac{\binom{3}{3}\binom{1}{2}\binom{2}{3}\binom{3}{3}\binom{2}{2-1-0}}{\binom{3}{2}} = \frac{9}{28}$$

$$f(1, 1) = \frac{\binom{3}{3}\binom{1}{2}\binom{2}{3}\binom{3}{3}\binom{2}{2-1-1}}{\binom{8}{2}} = \frac{6}{28}$$

$$f(0, 2) = \frac{\binom{3}{3}\binom{0}{2}\binom{2}{3}\binom{3}{3}\binom{2}{2-0-2}}{\binom{8}{2}} = \frac{1}{28}$$

$$f(2, 0) = \frac{\binom{3}{2}\binom{2}{2}\binom{2}{2}\binom{3}{3}\binom{3}{2}\binom{2}{2-2-0}}{\binom{3}{2}} = \frac{3}{28}$$

f(x, y)		X			Danistatala
		0 1		2	Row totals
	0	3	9	3_	<u>15</u>
y		28	28	28	28
	1	<u>6</u>	<u>6</u>	0	<u>12</u>
		28	28		28
	2	1_	0	0	<u>1</u>
		28			28
Column totals		<u>10</u>	<u>15</u>	3_	<b>28</b> = <b>1</b>
		28	28	28	28

(b)  $P[(X, Y) \in A]$ , where A is the region  $\{(x, y) | x + y \le 1\}$ . The probability that (X, Y) fall in the region A is

$$P[(X, Y) \in A] = P(X + Y \le 1)$$
$$= f(0, 0) + f(0, 1) + f(1, 0)$$

$$=\frac{3}{28}+\frac{6}{28}+\frac{9}{28}$$

$$=\frac{18}{28}$$

$$=\frac{9}{14}$$

#### **Joint Density Function**

The function f(x, y) is a **joint density function** of the **continuous random variables** X and Y if

1. 
$$f(x, y) \ge 0$$
, for all  $(x, y)$ ,

$$2.\int_{-\infty}^{\infty}\int_{-\infty}^{\infty}f(x,y)\,dxdy=1,$$

3.  $P[(X, Y) \in A] = \int \int_A f(x, y) dxdy$ , for any region A in the xy plane.

**Example:** A privately owned business operates both a drive-in facility and a walk-in facility. On a randomly selected day, let *X* and *Y*, respectively, be the proportions of the time that the drive-in and the walk-in facilities are in use, and suppose that the joint density function of these random variables is

$$f(x, y) = \begin{cases} \frac{2}{5} (2x + 3y), & 0 \le x \le 1, 0 \le y \le 1. \\ 0, & \text{elsewhere} \end{cases}$$

(a) Verify  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$ .

(b) 
$$P[(X, Y) \in A]$$
, where  $A = \{(x, y) | 0 < x < \frac{1}{2}, \frac{1}{4} < y < \frac{1}{2}\}$ 

#### Solution

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \, dx dy = \int_{y=0}^{1} \{ \int_{x=0}^{1} \frac{2}{5} (2x + 3y) \, dx \} dy$$

$$= \frac{2}{5} \int_{y=0}^{1} |\frac{(2)x^{2}}{2} + 3xy|_{x=0}^{1} dy$$

$$= \frac{2}{5} \int_{0}^{1} \{ 1^{2} + 3(1)y \} dy - 0$$

$$= \frac{2}{5} |y + 3\frac{y^{2}}{2}|_{0}^{1}$$

$$= \frac{2}{5} (1 + \frac{3}{2}) - 0$$

$$= 1$$

$$P[(X, Y) \in A] = P(0 < x < \frac{1}{2}, \frac{1}{4} < y < \frac{1}{2})$$

$$= \int_{y=\frac{1}{4}}^{\frac{1}{2}} \{ \int_{x=0}^{\frac{1}{2}} \frac{2}{5} (2x + 3y) dx \} dy$$

$$= \frac{2}{5} \{ \int_{y=\frac{1}{4}}^{\frac{1}{2}} | \frac{2x^{2}}{2} + 3xy |_{x=0}^{\frac{1}{2}} dy \}$$

$$= \frac{2}{5} \int_{y=\frac{1}{4}}^{\frac{1}{2}} | x^{2} + 3xy |_{x=0}^{\frac{1}{2}} dy$$

$$= \frac{2}{5} \int_{\frac{1}{4}}^{\frac{1}{2}} \{ \frac{1}{4} + 3(\frac{1}{2})y \} dy$$

$$= \frac{2}{5} \int_{\frac{1}{4}}^{\frac{1}{2}} \{ \frac{1}{4} + \frac{3}{2}y \} dy$$

$$= \frac{2}{5} | \frac{1}{4}y + \frac{3}{4}y^{2} |_{\frac{1}{4}}^{\frac{1}{2}}$$

$$= \frac{2}{5} \left| \frac{1}{4} y + \frac{3}{4} y^2 \right|_{\frac{1}{4}}^{\frac{1}{2}}$$

$$= \frac{2}{5} \left\{ \frac{1}{8} + \left( \frac{3}{4} \right) \left( \frac{1}{2} \right)^2 - \frac{1}{16} - \left( \frac{3}{4} \right) \left( \frac{1}{4} \right)^2 \right\}$$

$$= \frac{2}{5} \left( \frac{1}{8} + \frac{3}{16} - \frac{1}{16} - \frac{3}{64} \right)$$

$$= \frac{2}{5} \left( \frac{2+12-4+3}{64} \right)$$

$$= \frac{(2)(13)}{(5)(64)}$$

$$= \frac{13}{160}$$

Given the joint probability distribution f(x, y) of the discrete random variables X and Y, the probability distribution g(x) of X alone is obtained by summing f(x, y) over the values of Y.

 $\square$  Similarly, the probability distribution h(y) of Y alone is obtained by summing f(x, y) over the values of X. We define g(x) and h(y) to be the marginal distributions of X and Y, respectively.

□ When *X* and *Y* are continuous random variables, summations are replaced by integrals.

$$g(x) = \sum_{y} f(x, y)$$
 and  $h(y) = \sum_{x} f(x, y)$  for discrete case, and

$$g(x) = \int_{y=-\infty}^{\infty} f(x, y) dy \text{ and } h(y) = \int_{x=-\infty}^{\infty} f(x, y) dx$$

for the continuous case.

Note: The term *marginal* is used here because, in the discrete case, the values of g(x) and h(y) are just the marginal totals of the respective columns and rows when the values of f(x, y) are displayed in a rectangular table.

□ Example: Show that the column and row totals of the table in the coming slide give the marginal distribution of X alone and of Y alone.

f(x, y)		X			Danistatala
		0 1		2	Row totals
	0	3	9	3_	<u>15</u>
y		28	28	28	28
	1	<u>6</u>	<u>6</u>	0	<u>12</u>
		28	28		28
	2	1_	0	0	<u>1</u>
		28			28
Column totals		<u>10</u>	<u>15</u>	3_	<b>28</b> = <b>1</b>
		28	28	28	28

**Solution :** For the random variable X, we see that

$$g(x) = \sum_{y} f(x, y)$$

$$g(0) = f(0, 0) + f(0, 1) + f(0, 2)$$

$$= \frac{3}{28} + \frac{6}{28} + \frac{1}{28} = \frac{10}{28} = \frac{5}{14}$$

$$g(1) = f(1, 0) + f(1, 1) + f(1, 2)$$

$$= \frac{9}{28} + \frac{6}{28} + 0 = \frac{15}{28}$$

$$g(2) = f(2, 0) + f(2, 1) + f(2, 2)$$

$$= \frac{3}{28} + 0 + 0 = \frac{3}{28}$$

## **Marginal Distribution of x**

x = 0	0	1	2	Total
g(x)	10	15	3	<b>28</b> = 1
	<b>28</b>	<b>28</b>	<b>28</b>	28

For the random variable y, we see that

$$h(y) = \sum_{x} f(x, y)$$

$$h(0) = f(0, 0) + f(1, 0) + f(2, 0)$$

$$= \frac{3}{28} + \frac{9}{28} + \frac{3}{28} = \frac{15}{28}$$

$$h(1) = f(0, 1) + f(1, 1) + f(2, 1)$$

$$= \frac{6}{28} + \frac{6}{28} + 0 = \frac{12}{28}$$

$$h(2) = f(0, 2) + f(1, 2) + f(2, 2)$$

$$= \frac{1}{28} + 0 + 0 = \frac{1}{28}$$

## **Marginal Distribution of y**

y = 0	0	1	2	Total
h(y)	15	12	1	$\frac{28}{28} = 1$
	<del>28</del>	<b>28</b>	<del>28</del>	28

**Example :** Find g(x) and h(y) for the joint density function

$$f(x, y) = \begin{cases} \frac{2}{5} (2x + 3y), & 0 \le x \le 1, 0 \le y \le 1. \\ 0, & \text{elsewhere} \end{cases}$$

### **Marginal Density of x**

#### Solution

$$g(x) = \int_{y=-\infty}^{\infty} f(x, y) dy$$

$$= \int_{y=0}^{1} \frac{2}{5} (2x + 3y) dy$$

$$= \left| \frac{2}{5} (2xy + \frac{3y^2}{2}) \right|_{y=0}^{1}$$

$$= \frac{2}{5} \left\{ 2x(1) + \frac{3(1)^2}{2} \right\} - 0$$

$$= \frac{2}{5} \left( \frac{4x + 3}{2} \right)$$

$$= \frac{4x + 3}{5}$$

## **Marginal Density of y**

$$h(y) = \int_{x=-\infty}^{\infty} f(x, y) dx$$

$$= \int_{x=0}^{1} \frac{2}{5} (2x + 3y) dx$$

$$= \frac{2}{5} \left| \frac{2x^{2}}{2} + 3xy \right|_{x=0}^{1}$$

$$= \frac{2}{5} \{1 + 3(1)y\} - 0$$

$$= \frac{2}{5} (1 + 3y)$$