CHAPTER 5 JOINT PROBABILITY DISTRIBUTION

L1 - Discrete variables





Learning Objectives:

At the end of the lecture, you will be able to:

- define the joint probability distributions for discrete variables.
- calculate the marginal probability distributions.
- find the mean and the variance of X and Y of the joint probability distributions.

Definition of JPD for discrete variables



If X and Y are discrete random variables, then p(x,y) is called the joint probability distribution (JPD) or joint probability mass function if it satisfies the following properties:

1.
$$p(x, y) \ge 0$$

$$2.\sum_{x}\sum_{y}p(x,y)=1$$

$$3.p(x, y) = p(X = x, Y = y)$$



Marginal Probability Distributions (MPD)

If X and Y are discrete random variables with joint probability mass function, p(x,y), then the marginal probability mass functions for X is:

$$p(x) = p(X = x) = \sum_{y} p(x, y)$$

the marginal probability mass functions for *Y* is:

$$p(y) = p(Y = y) = \sum_{x} p(x, y)$$



Mean for X and Y

If X and Y are discrete random variables with joint probability mass function, p(x,y), marginal probabilities p(x), then the mean or expected values for X is:

$$E(x) = \mu_x = \sum_x x p(x)$$

If X and Y are discrete random variables with joint probability mass function, p(x,y), marginal probabilities p(y), then the mean or expected values for Y is:

$$E(y) = \mu_y = \sum_{v} y p(Y = y)$$





If X and Y are discrete random variables with joint probability mass function, p(x,y), marginal probabilities p(x), then the variance for X is:

$$V(x) = \sigma_x^2 = \sum_{x} (x - \mu_x)^2 p(x) = \sum_{x} x^2 p(x) - (\sum_{x} x p(x))^2$$

If X and Y are discrete random variables with joint probability mass function, p(x,y), marginal probabilities p(y), then the variance for Y is:

$$V(y) = \sigma_y^2 = \sum_y (y - \mu_y)^2 p(y) = \sum_y y^2 p(y) - (\sum_y y p(y))^2$$





A bus ferrying employees travels from Tronoh to Ipoh routeand back over the same route each day. There are three bus stops on this route. Let *X* be the number of bus stops that bus must stop on the way from Tronoh to Ipoh. Let *Y* be the number of bus stops the bus must stop on the way back from Ipoh to Tronoh. The JPD of *X* and *Y* is given below:

		X			
		0	1	2	3
	0	0.01	0.02	0.07	0.01
Y	1	0.03	0.06	0.10	0.06
·	2	0.05	0.12	0.15	0.08
	3	0.02	0.09	0.08	0.05

- a. Find the marginal distribution for *X* and *Y*.
- b. What is the expected number of bus stops the bus must stop over the entire road, that is going to Ipoh and back to Tronoh?
- c. Find also the variance of X and the variance of Y.

Solution



The JPD of *X* and *Y* is given below:

		X			
		0	1	2	3
	0	0.01	0.02	0.07	0.01
Υ	1	0.03	0.06	0.10	0.06
•	2	0.05	0.12	0.15	0.08
	3	0.02	0.09	0.08	0.05

a. The marginal distribution for *X* is:

$$P(x) = \sum_{y} p(x, y), \quad x = 0, 1, 2, 3$$

$$P(0) = \sum_{y=0}^{3} p(0, y) = P(0,0) + p(0,1) + p(0,2) + p(0,3)$$
$$= 0.01 + 0.03 + 0.05 + 0.02 = 0.11$$



$$P(1) = \sum_{y=0}^{3} p(1, y) = P(1,0) + p(1,1) + p(1,2) + p(1,3)$$

$$= 0.02 + 0.06 + 0.12 + 0.09 = 0.29$$

$$P(2) = \sum_{y=0}^{3} p(2, y) = P(2,0) + p(2,1) + p(2,2) + p(2,3)$$

$$= 0.07 + 0.10 + 0.15 + 0.08 = 0.4$$

$$P(3) = \sum_{y=0}^{3} p(3, y) = P(3,0) + p(3,1) + p(3,2) + p(3,3)$$

$$= 0.01 + 0.06 + 0.08 + 0.05 = 0.2$$



The JPD of *X* and *Y* is given below:

		X			
		0	1	2	3
	0	0.01	0.02	0.07	0.01
Y	1	0.03	0.06	0.10	0.06
•	2 3	0.05	0.12	0.15	0.08
		0.02	0.09	0.08	0.05

a. The marginal distribution for X is:

X	0	1	2	3
P(x)	0.11	0.29	0.40	0.20





The JPD of *X* and *Y* is given below:

		X			
		0	1	2	3
	0	0.01	0.02	0.07	0.01
Υ	1	0.03	0.06	0.10	0.06
	2	0.05	0.12	0.15	0.08
	3	0.02	0.09	0.08	0.05

The marginal distribution for *Y* is:

$$P(y) = \sum_{x} p(x, y), \quad y = 0, 1, 2, 3$$

$$P(0) = \sum_{x=0}^{3} p(x, 0) = P(0, 0) + p(1, 0) + p(2, 0) + p(3, 0)$$

$$= 0.01 + 0.02 + 0.07 + 0.01 = 0.11$$



$$P(1) = \sum_{x=0}^{3} p(x,1) = P(0,1) + p(1,1) + p(2,1) + p(3,1)$$

$$= 0.03 + 0.06 + 0.10 + 0.06 = 0.25$$

$$P(2) = \sum_{x=0}^{3} p(x,2) = P(0,2) + p(1,2) + p(2,2) + p(3,2)$$

$$= 0.05 + 0.12 + 0.15 + 0.08 = 0.4$$

$$P(3) = \sum_{x=0}^{3} p(x,3) = P(0,3) + p(1,3) + p(2,3) + p(3,3)$$

$$= 0.02 + 0.09 + 0.08 + 0.05 = 0.24$$



The JPD of *X* and *Y* is given below:

		X			
		0	1	2	3
	0	0.01	0.02	0.07	0.01
Y	1	0.03	0.06	0.10	0.06
•	2 3	0.05	0.12	0.15	0.08
		0.02	0.09	0.08	0.05

a. The marginal distribution for Y is:

У	0	1	2	3
P(y)	0.11	0.25	0.40	0.24





b. The expected of *X* is:

X	0	1	2	3
P(x)	0.11	0.29	0.40	0.20

$$E(x) = \mu(x) = \sum xp(x) = 0(0.11) + 1(0.29) + 2(0.40) + 3(0.20)$$
$$= 0 + 0.29 + 0.80 + 0.60) = 1.69$$

The expected of *Y* is:

у	0	1	2	3
P(y)	0.11	0.25	0.40	0.24

$$E(y) = \mu(y) = \sum yp(y) = 0(0.11) + 1(0.25) + 2(0.40) + 3(0.24)$$
$$= 0 + 0.25 + 0.80 + 0.72 = 1.77$$





c. The variance of X is:

X	0	1	2	3
P(x)	0.11	0.29	0.40	0.20
xp(x)	0	0.29	0.80	0.60
X2p(x)	0	0.29	1.60	1.80

$$E(x) = \mu(x) = \sum xp(x) = 0(0.11) + 1(0.29) + 2(0.40) + 3(0.20)$$
$$= 0 + 0.29 + 0.80 + 0.60) = 1.69$$

$$E(x^2) = \sum x^2 p(x) = 0^2 (0.11) + 1^2 (0.29) + 2^2 (0.40) + 3^2 (0.20)$$
$$= 0 + 0.29 + 1.60 + 1.80 = 3.69$$

$$V(x) = E(x^2) - (E(x))^2 = 3.69 - (1.69)^2 = 3.69 - 2.856$$
$$= 0.834$$





c. The variance of Y is:

У	0	1	2	3
P(y)	0.11	0.25	0.40	0.24
yp(y)	0	0.25	0.80	0.72
<i>y</i> 2 <i>p</i> (<i>y</i>)	0	0.25	1.60	2.16

$$E(y) = \mu(y) = \sum yp(y) = 0(0.11) + 1(0.25) + 2(0.40) + 3(0.24)$$
$$= 0 + 0.25 + 0.80 + 0.72 = 1.77$$

$$E(y^2) = \sum y^2 p(y) = 0^2 (0.11) + 1^2 (0.25) + 2^2 (0.40) + 3^2 (0.24)$$
$$= 0 + 0.25 + 1.60 + 2.16 = 4.01$$

$$V(y) = E(y^2) - (E(y))^2 = 4.01 - (1.77)^2 = 4.01 - 3.133$$
$$= 0.877$$





An assignment which consists of two sections was given to students who took the FEM1063 Statistics and Applications course. Let *X* be the total marks for section 1 and *Y* be the total marks for section 2. Assume that the JPD for *X* and *Y* is given below:

		Y		
	0	5	10	
X	0	0.02	0.06	0.12
^	5	0.04	0.15	0.30
	10	0.01	0.15	0.15

- a. Find the marginal distribution for X and Y.
- b. What is the expected number of X and the expected number of Y?
- c. Find also the variance of X and the variance of Y.















Chapter 5 JOINT PROBABILITY DISTRIBUTIONS

■ L2 - Continuous variables





Learning Objectives:

At the end of the lecture, you will be able to:

- define the joint probability distributions for continuous variables.
- calculate the marginal probability distributions.
- find the mean and the variance of X and Y of the joint probability distributions.

Definition of JPD for continuous variables



If X and Y are continuous random variables, then f(x,y) is called the joint probability distribution (JPD) or joint probability densenty function if it satisfies the following properties:

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

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

$$3.P(a < x < b, c < y < d) = \int_{c}^{d} \int_{a}^{b} f(x, y) dx dy$$

where a, b, c, d are constants





If X and Y are continuous random variables with joint probability density function, f(x,y), then the marginal probability density functions for X is:

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

the marginal probability density functions for Y is:

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



Mean for X and Y

If X and Y are continuous random variables with joint probability density function, f(x,y), marginal probabilities f(x), then the mean or expected values for X is:

$$E(x) = \mu_x = \int_{0}^{\infty} x f(x) dx$$

If X and Y are continuous random variables with joint probability density function, f(x,y), marginal probabilities f(y), then the mean or expected values for Y is:

$$E(y) = \mu_y = \int_{-\infty}^{\infty} y f(y) dy$$

Variance for X and Y



If X and Y are continuous random variables with joint probability density function, f(x,y), marginal probabilities f(x), then the variance for X is:

$$V(x) = \sigma_x^2 = \int_{-\infty}^{\infty} (x - \mu_x)^2 f(x) dx = \int_{-\infty}^{\infty} x^2 f(x) dx - (\int_{-\infty}^{\infty} x f(x))^2$$

If X and Y are continuous random variables with joint probability density function, f(x,y), marginal probabilities f(y), then the variance for Y is:

$$V(y) = \sigma_y^2 = \int_{-\infty}^{\infty} (y - \mu_y)^2 f(y) dx = \int_{-\infty}^{\infty} y^2 f(y) dy - (\int_{-\infty}^{\infty} y f(y))^2$$

Example



The JPD for *X* and *Y* is given by

$$f(x,y) = \begin{cases} \frac{1}{16}xy, & 0 \le x \le 2, \ 0 \le y \le 4 \\ 0, & \text{otherwise} \end{cases}$$

- a. Find the marginal distribution for *X* and *Y*.
- b. What is the mean of *X* and the mean of *Y*?
- c. Find also the variance of *X* and the variance of *Y*.

Solution



a. The marginal distribution for *X* is:

$$f(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_{0}^{4} \frac{1}{16} xy dy = \frac{xy^{2}}{32} \Big|_{0}^{4} = \frac{x}{2}, \ 0 \le x \le 2$$

The marginal distribution for *Y* is:

$$f(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_{0}^{2} \frac{1}{16} xy dx = \frac{x^{2} y}{32} \Big|_{0}^{2} = \frac{y}{8}, \ 0 \le y \le 4$$





b. The mean of X is:

$$\mu(x) = E(X) = \int_{-\infty}^{\infty} xf(x)dx = \int_{0}^{2} x \frac{x}{2} dx = \frac{x^{3}}{6} \Big|_{0}^{2} = \frac{8}{6} = 1.33$$

The mean of *Y* is:

$$\mu(y) = E(Y) = \int_{0}^{4} yf(y)dy = \int_{0}^{4} y \frac{y}{8} dy = \frac{y^{3}}{24} \Big|_{0}^{4} = \frac{64}{24} = 2.67$$

100



c. The variance of *X* is:

$$V(X) = E(X^2) - (E(X))^2$$

$$E(X^{2}) = \int_{-\infty}^{\infty} x^{2} f(x) dx = \int_{0}^{2} x^{2} \frac{x}{2} dx = \frac{x^{4}}{8} \Big|_{0}^{2} = \frac{16}{8} = 2$$

$$\therefore V(X) = 2 - (1.33)^2 =$$





The variance of *Y* is:

$$V(Y) = E(Y^2) - (E(Y))^2$$

$$E(Y^{2}) = \int_{-\infty}^{\infty} y^{2} f(y) dy = \int_{0}^{4} y^{2} \frac{y}{8} dy = \frac{y^{4}}{32} \Big|_{0}^{4} = \frac{256}{32} = 8$$

$$\therefore V(Y) = 8 - (2.67)^2 =$$

Exercise



The JPD for *X* and *Y* is given by

$$f(x, y) = \begin{cases} kxy, & 0 \le x \le 1, \ 0 \le y \le 1 \\ 0, & \text{otherwise} \end{cases}$$

- a. Find the value of the constant k.
- b. Find the marginal distribution for *X* and *Y*.
- c. What is the mean of *X* and the mean of *Y*?
- d. Find also the variance of *X* and the variance of *Y*.











