Probability and Statistics (IT302) 26th August 2020 Wednesday 11:15 AM-11:45 AM Class 10

Conditional Probability Distribution

It is extremely important that we make use of the special type of distribution of the form f(x, y)/g(x) in order to be able to effectively compute conditional probabilities. This type of distribution is called a **Conditional Probability Distribution**; the formal definition follows.

Let X and Y be two random variables, discrete or continuous. The **conditional distribution** of the random variable Y given that X = x is

$$f(y|x) = \frac{f(x,y)}{g(x)}$$
, provided $g(x) > 0$.

Similarly, the conditional distribution of X given that Y = y is

$$f(x|y) = \frac{f(x,y)}{h(y)}$$
, provided $h(y) > 0$.

Conditional Probability Distribution Contd.

If we wish to find the probability that the discrete random variable X falls between a and b when it is known that the discrete variable Y = y, we evaluate

$$P(a < X < b \mid Y = y) = \sum_{a < x < b} f(x|y),$$

where the summation extends over all values of X between a and b. When X and Y are continuous, we evaluate

$$P(a < X < b \mid Y = y) = \int_{a}^{b} f(x|y) dx.$$

Example 3.14

Example 3.14: 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 $\{(x,y)|x+y \le 1\}$.

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

(a) Now, f(0,1), for example, represents the probability that a red and a green pens are selected. The total number of equally likely ways of selecting any 2 pens from the 8 is $\binom{8}{2} = 28$. The number of ways of selecting 1 red from 2 red pens and 1 green from 3 green pens is $\binom{2}{1}\binom{3}{1} = 6$. Hence, f(0,1) = 6/28 = 3/14. Similar calculations yield the probabilities for the other cases, which

Example 3.14 Contd.

5, it will become clear that the joint probability distribution of Table 3.1 can be represented by the formula

$$f(x,y) = \frac{\binom{3}{x}\binom{2}{y}\binom{3}{2-x-y}}{\binom{8}{2}},$$

for x = 0, 1, 2; y = 0, 1, 2; and $0 \le x + y \le 2$.

(b) The probability that (X,Y) fall in the region A is

$$P[(X,Y) \in A] = P(X+Y \le 1) = f(0,0) + f(0,1) + f(1,0)$$
$$= \frac{3}{28} + \frac{3}{14} + \frac{9}{28} = \frac{9}{14}.$$

Table 3.1: Joint Probability Distribution for Example 3.14

			x		Row
	f(x,y)	0	1	2	Totals
	0	$\frac{3}{28}$	$\frac{9}{28}$	$\frac{3}{28}$	$\frac{15}{28}$ $\frac{3}{7}$
y	1	$\begin{array}{r} \frac{3}{28} \\ \frac{3}{14} \end{array}$	$\frac{\frac{9}{28}}{\frac{3}{14}}$	O	$\frac{3}{7}$
	2	$\frac{1}{28}$	0	O	$\frac{1}{28}$
Column Totals		$\frac{5}{14}$	$\frac{15}{28}$	$\frac{3}{28}$	1

Conditional Probability Distribution Example 3.18

Referring to Example 3.14, find the Conditional Distribution of X, given that Y = 1, and use it to determine $P(X = 0 \mid Y = 1)$.

Solution: We need to find f(x|y), where y = 1. First, we find that

$$h(1) = \sum_{x=0}^{2} f(x,1) = \frac{3}{14} + \frac{3}{14} + 0 = \frac{3}{7}.$$

Now

$$f(x|1) = \frac{f(x,1)}{h(1)} = \left(\frac{7}{3}\right)f(x,1), \quad x = 0, 1, 2.$$

Conditional Probability Distribution Example 3.18 Contd.

Therefore,

$$\begin{split} f(0|1) &= \left(\frac{7}{3}\right) f(0,1) = \left(\frac{7}{3}\right) \left(\frac{3}{14}\right) = \frac{1}{2}, \ f(1|1) = \left(\frac{7}{3}\right) f(1,1) = \left(\frac{7}{3}\right) \left(\frac{3}{14}\right) = \frac{1}{2}, \\ f(2|1) &= \left(\frac{7}{3}\right) f(2,1) = \left(\frac{7}{3}\right) (0) = 0, \end{split}$$

and the conditional distribution of X, given that Y = 1, is

Finally,

$$P(X = 0 \mid Y = 1) = f(0|1) = \frac{1}{2}.$$

Therefore, if it is known that 1 of the 2 pen refills selected is red, we have a probability equal to 1/2 that the other refill is not blue.

Conditional Probability Distribution Example 3.19

Example 3.19: The joint density for the random variables (X, Y), where X is the unit temperature change and Y is the proportion of spectrum shift that a certain atomic particle produces, is

 $f(x,y) = \begin{cases} 10xy^2, & 0 < x < y < 1, \\ 0, & \text{elsewhere.} \end{cases}$

- (a) Find the marginal densities g(x), h(y), and the conditional density f(y|x).
- (b) Find the probability that the spectrum shifts more than half of the total observations, given that the temperature is increased by 0.25 unit.

Solution: (a) By definition,

$$g(x) = \int_{-\infty}^{\infty} f(x,y) \ dy = \int_{x}^{1} 10xy^{2} \ dy$$

$$= \frac{10}{3}xy^{3} \Big|_{y=x}^{y=1} = \frac{10}{3}x(1-x^{3}), \ 0 < x < 1,$$

$$h(y) = \int_{-\infty}^{\infty} f(x,y) \ dx = \int_{0}^{y} 10xy^{2} \ dx = 5x^{2}y^{2} \Big|_{x=0}^{x=y} = 5y^{4}, \ 0 < y < 1.$$

Conditional Probability Distribution Example 3.19 Contd.

Now

$$f(y|x) = \frac{f(x,y)}{g(x)} = \frac{10xy^2}{\frac{10}{3}x(1-x^3)} = \frac{3y^2}{1-x^3}, \ 0 < x < y < 1.$$

(b) Therefore,

$$P\left(Y > \frac{1}{2} \mid X = 0.25\right) = \int_{1/2}^{1} f(y \mid x = 0.25) \ dy = \int_{1/2}^{1} \frac{3y^2}{1 - 0.25^3} \ dy = \frac{8}{9}.$$

Conditional Probability Distribution Example 3.20

Example 3.20: Given the joint density function

$$f(x,y) = \begin{cases} \frac{x(1+3y^2)}{4}, & 0 < x < 2, \ 0 < y < 1, \\ 0, & \text{elsewhere,} \end{cases}$$

Conditional Probability Distribution Example 3.20 Contd.

find g(x), h(y), f(x|y), and evaluate $P(\frac{1}{4} < X < \frac{1}{2} \mid Y = \frac{1}{3})$. **Solution:** By definition of the marginal density. for 0 < x < 2,

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_{0}^{1} \frac{x(1 + 3y^{2})}{4} dy$$
$$= \left(\frac{xy}{4} + \frac{xy^{3}}{4}\right)\Big|_{y=0}^{y=1} = \frac{x}{2},$$

and for 0 < y < 1,

$$h(y) = \int_{-\infty}^{\infty} f(x,y) dx = \int_{0}^{2} \frac{x(1+3y^{2})}{4} dx$$
$$= \left(\frac{x^{2}}{8} + \frac{3x^{2}y^{2}}{8}\right) \Big|_{x=0}^{x=2} = \frac{1+3y^{2}}{2}.$$

Therefore, using the conditional density definition, for 0 < x < 2,

$$f(x|y) = \frac{f(x,y)}{h(y)} = \frac{x(1+3y^2)/4}{(1+3y^2)/2} = \frac{x}{2},$$

and

$$P\left(\frac{1}{4} < X < \frac{1}{2} \mid Y = \frac{1}{3}\right) = \int_{1/4}^{1/2} \frac{x}{2} dx = \frac{3}{64}.$$

Statistical Independence Introduction

If f(x|y) does not depend on y, as is the case for Example 3.20, then f(x|y) = g(x) and f(x,y) = g(x)h(y). The proof follows by substituting

$$f(x,y) = f(x|y)h(y)$$

into the marginal distribution of X. That is,

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

If f(x|y) does not depend on y, we may write

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

Now

$$\int_{-\infty}^{\infty} h(y) \ dy = 1,$$

since h(y) is the probability density function of Y. Therefore,

$$g(x) = f(x|y)$$
 and then $f(x,y) = g(x)h(y)$.

Statistical Independence Definition

Let X and Y be two random variables, discrete or continuous, with joint probability distribution f(x, y) and marginal distributions g(x) and h(y), respectively. The random variables X and Y are said to be **statistically independent** if and only if

$$f(x,y) = g(x)h(y)$$

for all (x, y) within their range.

If you can find any point (x, y) for which f(x, y) is defined such that $f(x, y) \neq g(x)h(y)$, the discrete variables X and Y are not statistically independent.

Statistical Independence Example 3.21

Show that the random variables of Example 3.14 are not statistically independent.

Proof: Let us consider the point (0,1). From Table 3.1 we find the three probabilities f(0,1), g(0), and h(1) to be

$$f(0,1) = \frac{3}{14},$$

$$g(0) = \sum_{y=0}^{2} f(0,y) = \frac{3}{28} + \frac{3}{14} + \frac{1}{28} = \frac{5}{14},$$

$$h(1) = \sum_{x=0}^{2} f(x,1) = \frac{3}{14} + \frac{3}{14} + 0 = \frac{3}{7}.$$

Clearly,

$$f(0,1) \neq g(0)h(1),$$

and therefore X and Y are not statistically independent.

Mutually Statistically Independent Random Variable Definition

Let $X_1, X_2, ..., X_n$ be n random variables, discrete or continuous, with joint probability distribution $f(x_1, x_2, ..., x_n)$ and marginal distribution $f_1(x_1), f_2(x_2), ..., f_n(x_n)$, respectively. The random variables $X_1, X_2, ..., X_n$ are said to be mutually **statistically independent** if and only if

$$f(x_1, x_2, \dots, x_n) = f_1(x_1) f_2(x_2) \cdots f_n(x_n)$$

for all (x_1, x_2, \ldots, x_n) within their range.

Mutually Statistically Independent Random Variable Example 3.22

Example 3.22: Suppose that the shelf life, in years, of a certain perishable food product packaged in cardboard containers is a random variable whose probability density function is given by

$$f(x) = \begin{cases} e^{-x}, & x > 0, \\ 0, & \text{elsewhere.} \end{cases}$$

Let X_1 , X_2 , and X_3 represent the shelf lives for three of these containers selected independently and find $P(X_1 < 2, 1 < X_2 < 3, X_3 > 2)$.

Solution: Since the containers were selected independently, we can assume that the random variables X_1, X_2 , and X_3 are statistically independent, having the joint probability density

$$f(x_1, x_2, x_3) = f(x_1)f(x_2)f(x_3) = e^{-x_1}e^{-x_2}e^{-x_3} = e^{-x_1-x_2-x_3},$$

for $x_1 > 0$, $x_2 > 0$, $x_3 > 0$, and $f(x_1, x_2, x_3) = 0$ elsewhere. Hence

$$P(X_1 < 2, 1 < X_2 < 3, X_3 > 2) = \int_2^\infty \int_1^3 \int_0^2 e^{-x_1 - x_2 - x_3} dx_1 dx_2 dx_3$$
$$= (1 - e^{-2})(e^{-1} - e^{-3})e^{-2} = 0.0372.$$