# CSL003P1M: Probability and Statistics Lecture 34 (Joint Distribution Of A Function Of Continuous Random Variables)

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Let X be uniformly distributed over (0,1). Find the density of the random variable Y, defined by  $Y=X^n$ .

Solution: For  $0 \le y \le 1$ ,

$$F_Y(y) = P\{Y \le y\}$$

$$= P\{X^n \le y\}$$

$$= P\{X \le y^{1/n}\}$$

$$= F_X(y^{1/n})$$

$$= y^{1/n}$$

So,

$$f_Y(y) = \begin{cases} \frac{1}{n} y^{1/n-1} & 0 \le y \le 1\\ 0 & \text{otherwise} \end{cases}$$

If X is a continuous random variable with probability density  $f_X$ , then find the distribution of  $Y = X^2$ .

Solution: For  $y \ge 0$ ,

$$F_Y(y) = P\{Y \le y\}$$

$$= P\{X^2 \le y\}$$

$$= P\{-\sqrt{y} \le X \le \sqrt{y}\}$$

$$= F_X(\sqrt{y}) - F_X(-\sqrt{y})$$

Differentiation yields

$$f_Y(y) = \frac{1}{2\sqrt{y}}[f_X(\sqrt{y}) + f_X(-\sqrt{y})] \quad y \ge 0$$



If X has a probability density  $f_X$ , then has a density function of Y = |X|.

Solution: For y > 0,

$$F_Y(y) = P\{Y \le y\}$$
  
=  $P\{|X| \le y\}$   
=  $P\{-y \le X \le y\}$   
=  $F_X(y) - F_X(-y)$ 

Differentiation yields

$$f_Y(y) = f_X(y) + f_X(-y)$$
  $y \ge 0$ 



## Joint Probability Distribution of Functions of Random Variables

- Let  $X_1$  and  $X_2$  be jointly continuous random variable with joint probability density function  $f_{X_1,X_2}$ .
- It is sometimes necessary to obtain the joint distribution of the random variables  $Y_1$  and  $Y_2$ , which arise as functions of  $X_1$  and  $X_2$ .
- Specifically, suppose that  $Y_1 = g_1(X_1, X_2)$  and  $Y_2 = g_2(X_1, X_2)$  for some functions  $g_1$  and  $g_2$ .

## Joint Probability Distribution of Functions of Random Variables

#### Assumptions:

- The equations  $y_1 = g_1(x_1, x_2)$  and  $y_2 = g_2(x_1, x_2)$  can be uniquely solved for  $x_1$  and  $x_2$  in terms of  $y_1$  and  $y_2$ , with solutions given by, say,  $x_1 = h_1(y_1, y_2)$ ,  $x_2 = h_2(y_1, y_2)$ .
- ② The functions  $g_1$  and  $g_2$  have continuous partial derivatives at all points  $(x_1, x_2)$  and are such that the  $2 \times 2$  determinant

$$J(x_1, x_2) = \begin{vmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} \end{vmatrix} = \frac{\partial g_1}{\partial x_1} \frac{\partial g_2}{\partial x_2} - \frac{\partial g_1}{\partial x_2} \frac{\partial g_2}{\partial x_1} \neq 0$$

at all points  $(x_1, x_2)$ .



## Joint Probability Distribution of Functions of Random Variables

Under the previous two assumptions, it can be shown that the random variables  $Y_1$  and  $Y_2$  are jointly continuous with joint density function given by

$$f_{Y_1,Y_2}(y_1,y_2) = f_{X_1,X_2}(x_1,x_2)|J(x_1,x_2)|^{-1}$$

where  $x_1 = h_1(y_1, y_2)$ ,  $x_2 = h_2(y_1, y_2)$ .

Let  $X_1$  and  $X_2$  be jointly continuous random variables with probability density function  $f_{X_1,X_2}$ . Let  $Y_1=X_1+X_2$ ,  $Y_2=X_1-X_2$ . Find the joint density function of  $Y_1$  and  $Y_2$  in terms of  $f_{X_1,X_2}$ .

Let 
$$g_1(x_1, x_2) = x_1 + x_2$$
 and  $g_2(x_1, x_2) = x_1 - x_2$ . Then

$$J(x_1,x_2) = \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2$$

Now, since the equations  $y_1 = x_1 + x_2$  and  $y_2 = x_1 - x_2$  have  $x_1 = (y_1 + y_2)/2$ ,  $x_2 = (y_1 - y_2)/2$  as their solution, it follows that the desired density is

$$f_{Y_1,Y_2}(y_1,y_2) = \frac{1}{2} f_{X_1,X_2}\left(\frac{y_1+y_2}{2},\frac{y_1-y_2}{2}\right)$$



If X and Y are independent gamma random variables with parameters  $(\alpha, \lambda)$  and  $(\beta, \lambda)$ , respectively, compute the joint density of U = X + Y and V = X/(X + Y).

Solution: The joint density of X and Y is given by

$$f_{X,Y}(x,y) = \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} \frac{\lambda e^{-\lambda y} (\lambda y)^{\beta-1}}{\Gamma(\beta)}$$
$$= \frac{\lambda^{\alpha+\beta}}{\Gamma(\alpha)\Gamma(\beta)} e^{-\lambda(x+y)} x^{\alpha-1} y^{\beta-1}$$

Now, if  $g_1(x, y) = x + y$ ,  $g_2(x, y) = x/(x + y)$ , then

$$\frac{\partial g_1}{\partial x} = \frac{\partial g_1}{\partial y} = 1 \quad \frac{\partial g_2}{\partial x} = \frac{y}{(x+y)^2} \quad \frac{\partial g_2}{\partial y} = -\frac{x}{(x+y)^2}$$

So,

$$J(x,y) = \begin{vmatrix} \frac{1}{y} & \frac{1}{-x} \\ \frac{1}{(x+y)^2} & \frac{1}{(x+y)^2} \end{vmatrix} = -\frac{1}{x+y}$$

Finally, as u = x + y, v = x/(x + y), so x = uv, y = u(1 - v).



Therefore,

$$f_{U,V}(u,v) = f_{X,Y}[uv, u(1-v)]u$$

$$= \left(\frac{\lambda e^{-\lambda u}(\lambda u)^{\alpha+\beta-1}}{\Gamma(\alpha+\beta)}\right) \left(\frac{v^{\alpha-1}(1-v)^{\beta-1}\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}\right)$$

Hence X+Y and X/(X+Y) are independent, with X+Y having a gamma distribution with parameters  $(\alpha+\beta,\lambda)$  and X/(X+Y) having a beta distribution with parameters  $(\alpha,\beta)$ .

### Thank You