Jointly Distributed Random Variables

Joint Distribution of Functions of Random Variables [Ross S6.7]

Let X and Y have joint pdf $f_{XY}(x, y)$.

In some examples we computed the distribution of $Z=g(X,Y), {\rm e.g.}$

- in Example 23.2 we computed the cdf of $D = \sqrt{X^2 + Y^2}$
- in Example 23.3 we computed the pdf of Z = X/Y.

Now, consider

$$Y_1 = g_1(X_1, X_2)$$

 $Y_2 = g_2(X_1, X_2)$

and we want the joint pdf of Y_1 and Y_2 .

We make the following assumptions on g_1 and g_2 :

• The system of equations

$$y_1 = g_1(x_1, x_2)$$

 $y_2 = g_2(x_1, x_2)$

can be uniquely solved for x_1 and x_2 in terms of y_1 and y_2 :

$$x_1 = h_1(y_1, y_2)$$

 $x_2 = h_2(y_1, y_2).$

• g_1 and g_2 have continuous partial derivates such that the determinant

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

Under these conditions, the pdf of Y_1 and Y_2 can be shown to be:

$$f_{Y_1Y_2}(y_1, y_2) = f_{X_1X_2}(x_1, x_2) |J(x_1, x_2)|^{-1}$$
(29.1)

where

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

Example 29.1: Let

$$Y_1 = X_1 + X_2 Y_2 = X_1 - X_2$$

Find the joint pdf $f_{Y_1Y_2}(y_1, y_2)$ in terms of $f_{X_1X_2}(x_1, x_2)$. *Solution:* Solving

$$y_1 = x_1 + x_2$$
$$y_2 = x_1 - x_2$$

we get

$$x_1 = \frac{1}{2}y_1 + \frac{1}{2}y_2$$
$$x_2 = \frac{1}{2}y_1 - \frac{1}{2}y_2$$

Also

$$J(x_1, x_2) = \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2 \neq 0$$

Hence, from (29.6)

$$f_{Y_1Y_2}(y_1, y_2) = \frac{1}{2} f_{X_1X_2} \left(\frac{1}{2} y_1 + \frac{1}{2} y_2, \frac{1}{2} y_1 - \frac{1}{2} y_2 \right)$$
 (29.2)

Example 29.2: Let R and Θ be two random variables with joint pdf $f_{R\Theta}(r,\theta)$. Consider the change of variables

$$X = R\cos\Theta$$
$$Y = R\sin\Theta.$$

Find $f_{R\Theta}(r,\theta)$ in terms of $f_{XY}(x,y)$. [Hard]

Note: This is Problem T9.1; see also textbook Example 6.7b for a different tedious approach.

Solution: We have the system of equations

$$x = g_1(r, \theta) = r \cos \theta$$
$$y = g_2(r, \theta) = r \sin \theta$$

which is solved by

$$r = h_1(x, y) = \sqrt{x^2 + y^2}$$

$$\theta = h_2(x, y) = \begin{cases} \tan^{-1}(y/x) & x > 0, y > 0 \\ \tan^{-1}(y/x) + \pi & x < 0 \\ \tan^{-1}(y/x) + 2\pi & x > 0, y < 0 \end{cases}$$

where $h_2(x, y)$ is the angle of the vector (x, y).

Computing the Jacobian determinant

$$J(r,\theta) = \begin{vmatrix} \frac{\partial g_1}{\partial r} & \frac{\partial g_1}{\partial \theta} \\ \frac{\partial g_2}{\partial r} & \frac{\partial g_2}{\partial \theta} \end{vmatrix}$$
$$= \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix}$$
$$= r \cos^2 \theta + r \sin^2 \theta$$
$$= r$$

So, the pdf $f_{XY}(x,y)$ and $f_{R\Theta}(r,\theta)$ are related by

$$f_{XY}(x,y) = f_{R\Theta}(r,\theta) |J(r,\theta)|^{-1}$$
 (29.3)

$$= f_{R\Theta}(r,\theta)/r \tag{29.4}$$

or equivalently:

$$f_{R\Theta}(r,\theta) = f_{XY}(x,y)r \tag{29.5}$$

$$= f_{XY}(r\cos\theta, r\sin\theta)r \tag{29.6}$$

So, to compute the probability that $(R, \Theta) \in A$:

$$P[(R,\Theta) \in A] = \iint_{(r,\theta)\in A} f_{R\Theta}(r,\theta) dr d\theta$$
$$= \iint_{(r,\theta)\in A} f_{XY}(r\cos\theta, r\sin\theta) r dr d\theta$$