5-7. TRANSFORMATION OF TWO-DIMENSIONAL RANDOM VARIABLE

In this section we shall consider the problem of change of variables in the two-dimensional case. Let r.v.'s U and V be transformed to the r.v's X and Y by the transformation u = u(x, y), v = v(x, y), where u and v are continuously differentiable

functions, for which Jacobian of transformation:
$$J = \frac{\partial(x,y)}{\partial(u,v)} = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial y}{\partial u} \\ \frac{\partial x}{\partial v} & \frac{\partial y}{\partial v} \end{vmatrix}, \dots (5.23)$$

is either > 0 or < 0 throughout the (x, y) plane so that the inverse transformation is uniquely given by x = x(u, v), y = y(u, v).

Theorem 5.5. The joint p.d.f.
$$g_{UV}(u, v)$$
 of the transformed variables U and V is: $g_{UV}(u, v) = f_{XY}(x, y) |J|$, ... (5.24)

where |J| is the modulus value of the Jacobian of transformation and f(x, y) is expressed in terms of u and v.

Proof.
$$P(x < X \le x + dx, y < Y \le y + dy) = P(u < U \le u + du, v < V \le + dv)$$
$$\Rightarrow f_{XY}(x, y) dx dy = g_{UV}(u, v) du dv$$

$$\Rightarrow g_{UV}(u,v) du dv = f_{XY}(x,y) \left| \frac{\partial(x,y)}{\partial(u,v)} \right| du dv$$

$$\therefore g_{UV}(u,v) = f_{XY}(x,y) \left| \frac{\partial(x,y)}{\partial(u,v)} \right| = f_{XY}(x,y) |J|$$
Theorem 5.6 (a) If $x \in S$

Theorem 5.6. (a) If X and Y are independent continuous r.v.'s, then the p.d.f. of $U = X + Y \text{ is given by :} \qquad h(u) = \int_{-\infty}^{\infty} f_X(v) f_Y(u - v) dv \qquad \dots (5.25)$

Proof. Let $f_{XY}(x, y)$ be the joint p.d.f. of independent continuous r.v.'s X and Yand let us make the transformation: u = x + y, $v = x \implies x = v$, y = u - v.

$$J = \frac{\partial (x, y)}{\partial (u, v)} = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial y}{\partial u} \\ \frac{\partial x}{\partial v} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ 1 & -1 \end{vmatrix} = -1$$

Thus the joint p.d.f. of r.v.'s U and V is given by :

$$g_{UV}(u, v) = f_{XY}(x, y) |J| = f_X(x) f_Y(y) |J|$$
 (Since X and Y are independent)

The marginal density of U is given by :

$$h(u) = \int_{-\infty}^{\infty} g_{UV}(u, v) dv = \int_{-\infty}^{\infty} f_X(v) f_Y(u - v) dv$$

Remark. The function h (.) is given a special name and is said to be the *convolution of* f_X (.) and $f_Y(.)$ and we write $h(.) = f_X(.) * f_Y(.)$.

Example 5.48. Let (X, Y) be a two-dimensional non-negative continuous r.v. having the joint density:

$$f(x,y) = \begin{cases} 4xy e^{-(x^2 + y^2)}; x \ge 0, y \ge 0\\ 0, elsewhere \end{cases}$$

Prove that the density function of $U = \sqrt{X^2 + Y^2}$ is:

$$h\left(u\right) \ = \left\{ \begin{array}{ll} 2u^3 \, e^{-u^2} \ , \, 0 \leq u < \infty \\ 0 \ , \, elsewhere \end{array} \right.$$

Solution. Let us make the transformation : $u = \sqrt{x^2 + y^2}$ and v = x

$$v \ge 0$$
, $u \ge 0$ and $u \ge v$ or $u \ge 0$ and $0 \le v \le u$.

The Jacobian of transformation J is given by:

$$\frac{1}{J} = \frac{\partial(u, v)}{\partial(x, y)} = \begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial v}{\partial x} \\ \frac{\partial u}{\partial y} & \frac{\partial v}{\partial y} \end{vmatrix} = \begin{vmatrix} \frac{x}{\sqrt{x^2 + y^2}} & 1 \\ \frac{y}{\sqrt{x^2 + y^2}} & 0 \end{vmatrix} = -\frac{\frac{y}{\sqrt{x^2 + y^2}}}{\sqrt{x^2 + y^2}}$$

The joint p.d.f. of U and V is given by:

The joint p.a.f. of
$$u$$
 and $v = g$

$$g(u, v) = f(x, y) \mid J \mid = 4xy e^{-(x^2 + y^2)} \left| -\frac{\sqrt{x^2 + y^2}}{y} \right| = 4x \sqrt{x^2 + y^2} e^{-(x^2 + y^2)}$$

$$= \begin{cases} 4vu. e^{-u^2}; & u \ge 0, 0 \le v \le u \\ 0, & \text{otherwise} \end{cases}$$

Hence the marginal density function of $U = \sqrt{X^2 + Y^2}$ is:

$$h(u) = \int_0^u g(u, v) dv = 4u e^{-u^2} \int_0^u v dv = \begin{cases} 2u^3 e^{-u^2}, u \ge 0 \\ 0, \text{ elsewhere} \end{cases}$$

Example 5-49. Let the p.d.f. of the random variable (x, y) be :

$$f(x,y) = \begin{cases} \alpha^{-2} e^{-(x+y)/\alpha}; x, y > 0, \alpha > 0 \\ 0, elsewhere \end{cases}$$

Find the distribution of $\frac{1}{2}(X - Y)$.

solution. Consider the transformation:

$$u = \frac{1}{2}(x - y)$$
 and $v = y \implies x = 2u + v$ and $y = v$

The Jacobian of the transformation is :
$$J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 2 & 1 \\ 0 & 1 \end{vmatrix} = 2$$

Thus, the joint p.d.f. of the random variables (U, V) is given by :

$$g(u,v) = \begin{cases} \frac{2}{\alpha^2} e^{-(2/\alpha)(u+v)} & ; -\infty < u < \infty, v > -2u, \text{ if } u < 0; \\ v > 0 \text{ if } u \ge 0 \text{ and } \alpha > 0 \end{cases}$$

$$0 & \text{, elsewhere}$$

The marginal p.d.f. of U is given by:

$$g_{U}(u) = \begin{cases} \int_{-2u}^{\infty} \frac{2}{\alpha^{2}} \exp\left\{-(2/\alpha)(u+v)\right\} dv = \frac{1}{\alpha} e^{2u/\alpha}, u < 0 \\ \int_{0}^{\infty} \frac{2}{\alpha^{2}} \exp\left\{-(2/\alpha)(u+v)\right\} dv = \frac{1}{\alpha} e^{-2u/\alpha}, u \ge 0 \end{cases}$$

Hence
$$g_U(u) = \frac{1}{\alpha} exp \left\{ -\frac{2}{\alpha} |u| \right\} - \infty < u < \infty.$$