



8 Joint Probability Distribution of Function of Random Variables

X_1, X_2 : jointly continuous random variables with joint density f_{X_1, X_2}

Suppose that $Y_1 = g_1(X_1, X_2)$, $Y_2 = g_2(X_1, X_2)$ for some functions g_1, g_2

Assume that,

1. $y_1 = g_1(x_1, x_2)$, $y_2 = g_2(x_1, x_2)$ can be uniquely solved for x_1 and x_2 in terms of y_1 & y_2 , say, $x_1 = h_1(y_1, y_2)$, $x_2 = h_2(y_1, y_2)$.
2. g_1 & g_2 have continuous partial derivatives at all points (x_1, x_2) and are such that at all (x_1, x_2) ,

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

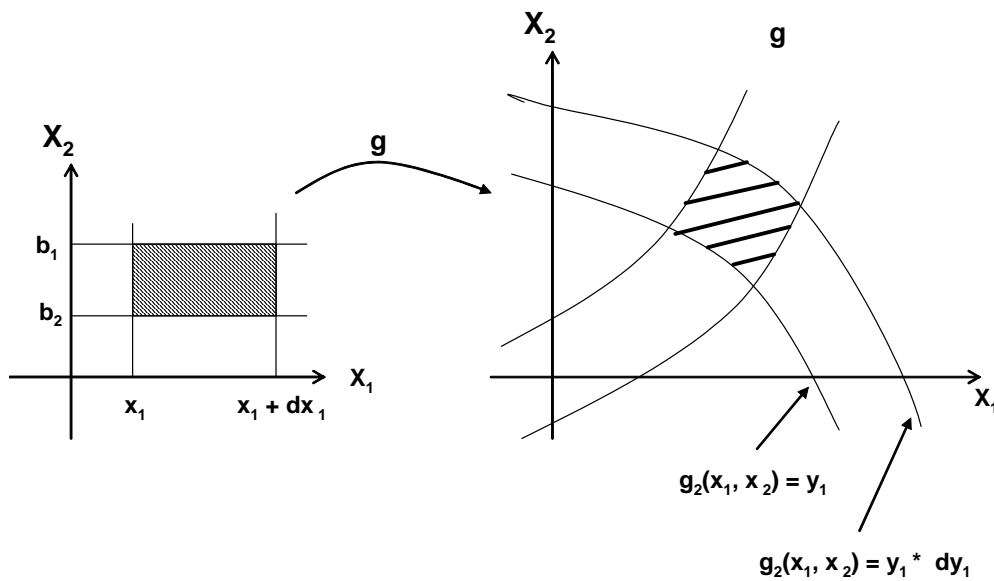
Then, Y_1 & Y_2 are jointly continuous with joint density function,

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

Proof.

$$P\{Y_1 \leq y_1, Y_2 \leq y_2\} = \int \int_{(x_1, x_2): g_1(x_1, x_2) \leq y_1, g_2(x_1, x_2) \leq y_2} f_{X_1, X_2}(x_1, x_2) dx_1 dx_2$$

differentiate w.r.t. y_1 & y_2



$$f_{X_1, X_2}(x_1, x_2) dx_1 dx_2 \cdot |J| \approx f_{Y_1, Y_2}(y_1, y_2) dy_1 dy_2$$

Example .

(X, Y) : random point in the plane
 X, Y : independent unit normal random variables
 Joint distribution of R, Θ ?

Solution.

$$r = g_1(x, y) = \sqrt{x^2 + y^2}$$

$$\theta = g_2(x, y) = \tan^{-1} \frac{x}{y}$$

$$J = \begin{vmatrix} \frac{x}{\sqrt{x^2+y^2}} & \frac{y}{\sqrt{x^2+y^2}} \\ \frac{-y}{x^2+y^2} & \frac{x}{x^2+y^2} \end{vmatrix} = \frac{1}{\sqrt{x^2 + y^2}} = \frac{1}{r}$$

$$f(x, y) = \frac{1}{2\pi} e^{-\frac{1}{2}(x^2+y^2)}$$

$$f(r, \theta) = \frac{1}{2\pi} r e^{-\frac{1}{2}r^2} \quad 0 < \theta < 2\pi, \quad 0 < r < \infty$$

Θ : uniform over $(0, 2\pi)$.

R : Rayleigh distribution $f(r) = re^{-\frac{r^2}{2}}$
example) X, Y: horizontal/vertical miss distance
absolute value of the error \sim Rayleigh.

Example .

(X, Y) : random point in the plane
X, Y : independent unit normal random va
Joint distribution of R^2, Θ ?

Solution.

$$g_1(x, y) = x^2 + y^2$$

$$J = \left| \begin{array}{cc} 2x & 2y \\ \frac{-y}{x^2+y^2} & \frac{x}{x^2+y^2} \end{array} \right| = 2$$

$$f(d, \theta) = \frac{1}{4\pi} e^{-\frac{d}{2}}$$

$R^2 \sim$ exponential distribution with parameter $\frac{1}{2}$.
 $R^2 = X^2 + Y^2 \Rightarrow R^2 \sim \chi^2$ distribution with 2 d.o.f. \therefore The exp distribution with parameter $\frac{1}{2} = \chi^2$ distribution with 2 d.o.f.

• We can simulate (generate) normal random variables by making a suitable transformation on uniform random variables.

U_1, U_2 : independent random variables each uniform over $(0, 1)$

X_1, X_2 : two independent unit normal random variables.

$R^2 = X_1^2 + X_2^2, \Theta \Rightarrow$ exponential distribution with $\lambda = \frac{1}{2}$

$$P\{-2 \log U_1 < x\} = P\{\log U_1 > -\frac{x}{2}\} = P\{U_1 > e^{-\frac{x}{2}}\} = 1 - e^{-\frac{x}{2}}$$

$\therefore -2 \log U_1$ as an exp distribution with $\lambda = \frac{1}{2}$.

$2\pi U_2$: uniform over $(0, 2\pi)$, use it to generate Θ

$\therefore X_1 = \sqrt{-2 \log U_1} \cos(2\pi U_2), X_2 = \sqrt{-2 \log U_1} \sin(2\pi U_2)$, independent standard normal.

• When the joint density function of the n random variables X_1, X_2, \dots, X_n is given and we want to compute the joint density function of Y_1, \dots, Y_n , where

$$Y_1 = g_1(X_1, \dots, X_n)$$

$$Y_n = g_n(X_1, \dots, X_n)$$

Assuming g_i 's have continuous partial derivatives and that, at all points (x_1, \dots, x_n)

$$J(x_1, x_2) = \begin{vmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_n}{\partial x_1} & \dots & \frac{\partial g_n}{\partial x_n} \end{vmatrix} \neq 0.$$

Suppose that $y_1 = g_1(x_1, \dots, x_n)$, $y_n = g_n(x_1, \dots, x_n)$ have a unique solution, say $x_1 = h_1(y_1, \dots, y_n)$, $x_n = h_n(y_1, \dots, y_n)$, then

$$f_{Y_1, \dots, Y_n}(y_1, \dots, y_n) = f_{X_1, \dots, X_n}(x_1, \dots, x_n) |J(x_1, \dots, x_n)|^{-1}$$

where $x_i = h_i(y_1, \dots, y_n)$, $i = 1, \dots, n$.

Example .

X_1, X_2, X_3 : independent unit normal random variables.

$$Y_1 = X_1 + X_2 + X_3$$

$$Y_2 = X_1 - X_2$$

$$Y_3 = X_1 - X_3$$

↓

$$X_1 = \frac{1}{3}(Y_1 + Y_2 + Y_3)$$

$$X_2 = \frac{1}{3}(Y_1 - 2Y_2 + Y_3)$$

$$X_3 = \frac{1}{3}(Y_1 + Y_2 - 2Y_3)$$

$$J = \begin{vmatrix} 1 & 1 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{vmatrix} = 3$$

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = \frac{1}{(2\pi)^{\frac{3}{2}}} e^{-\sum_{i=1}^3 \frac{x_i^2}{2}}$$

$$f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) = \frac{1}{3} \frac{1}{(2\pi)^{\frac{3}{2}}} \exp \left[-\frac{1}{2} \left[\left(\frac{y_1 + y_2 + y_3}{3} \right)^2 + \left(\frac{y_1 - 2y_2 + y_3}{3} \right)^2 + \left(\frac{y_1 + y_2 - 2y_3}{3} \right)^2 \right] \right]$$