

## 1 Problem Manual 6.2.1

**Question :**

Let:

$$X1 \sim \mathcal{N}(0, 1) \text{ and } X2 \sim \mathcal{N}(0, 1)$$

Evaluate the joint PDF of  $X1, X2$ , given by

$$p_{X1, X2}(x1, x2) = p_{X1}(x1)p_{X2}(x2)$$

**Solution :**

Normal distribution is defined as :

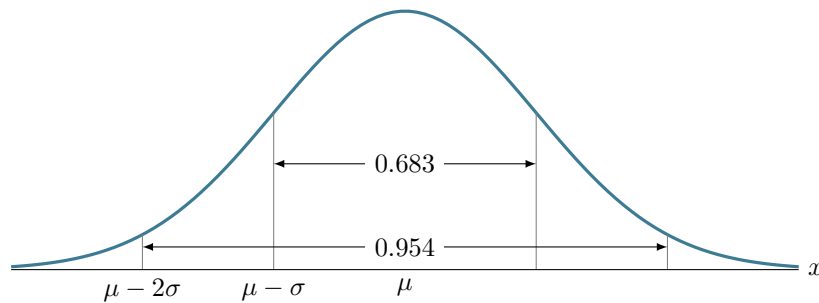
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \dots\dots\dots(1)$$

Where :

$\mu = \text{Mean}$

$\sigma = \text{Standard deviation}$

Given by the curve :



Given in the question that  $\mu = 0$  and  $\sigma = 1$  equation(1) transforms to

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

Similarly the individual PDF of  $X1$  and  $X2$  would be defined as :

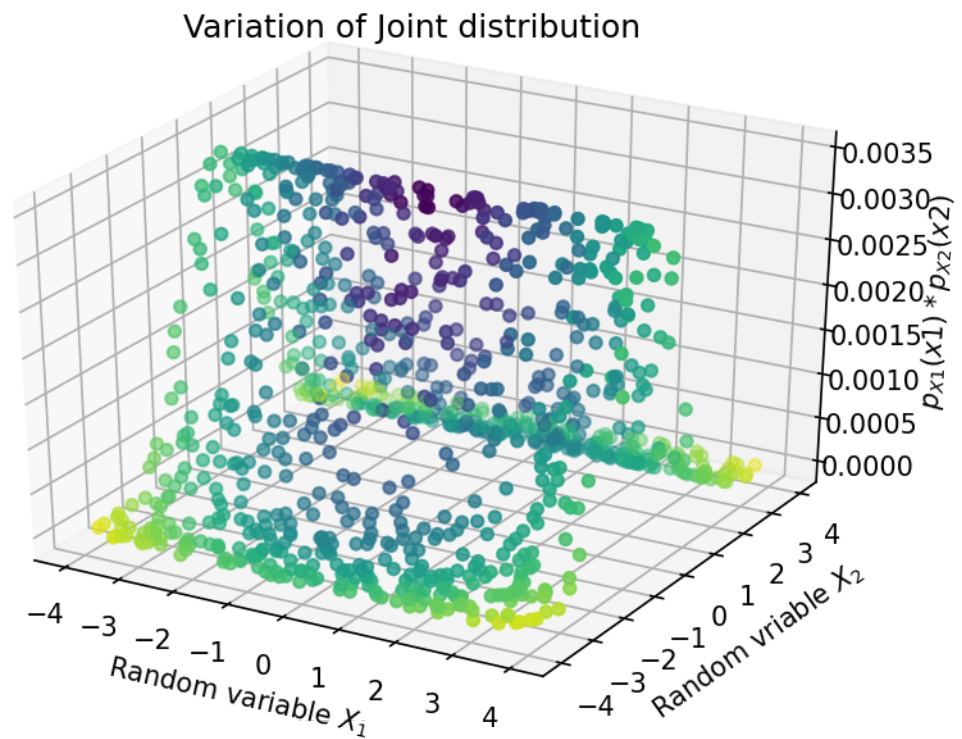
$$p_{X1}(x1) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x1^2}{2}} \dots\dots\dots(2)$$

$$p_{X_2}(x_2) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x_2^2}{2}} \dots\dots\dots(3)$$

The joint probability distribution given by  $p_{X_1, X_2}(x_1, x_2)$  is obtained by multiplying (1) and (2) is :

$$p_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{(x_1+x_2)^2}{2}} \dots\dots\dots(4)$$

Graphically equivalent to :



[https://github.com/harshal9876/AI5002/blob/main/Joint\\_distribution\\_assignment1\(1\).png](https://github.com/harshal9876/AI5002/blob/main/Joint_distribution_assignment1(1).png)

*Joint Bivariate Gaussian distribution is given as :*

$$p_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp -\frac{z}{2(1-\rho^2)} \dots\dots\dots (5)$$

Where

$$z = (x_1 - \mu_1)^2 \frac{1}{\sigma_1^2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} - \frac{2\rho(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2}$$

and

$$\rho = \text{cor}(x_1, x_2) = \frac{V_{12}}{\sigma_1\sigma_2}$$

where :

$x_1, x_2 = \text{random variables}$

$\mu_1 = \text{Mean of random variable } X_1$

$\mu_2 = \text{Mean of random variable } X_2$

$\sigma_1 = \text{Standard deviation of random variable } X_1$

$\sigma_2 = \text{Standard deviation of random variable } X_2$

$V_{12} = \text{covariance}(x_1, x_2)$

For a Bivariable Normal distribution having independent random variables

$$\mu_1 = \mu_2 = 0$$

$$\sigma_1 = \sigma_2 = 1$$

$$V_{12} = 0$$

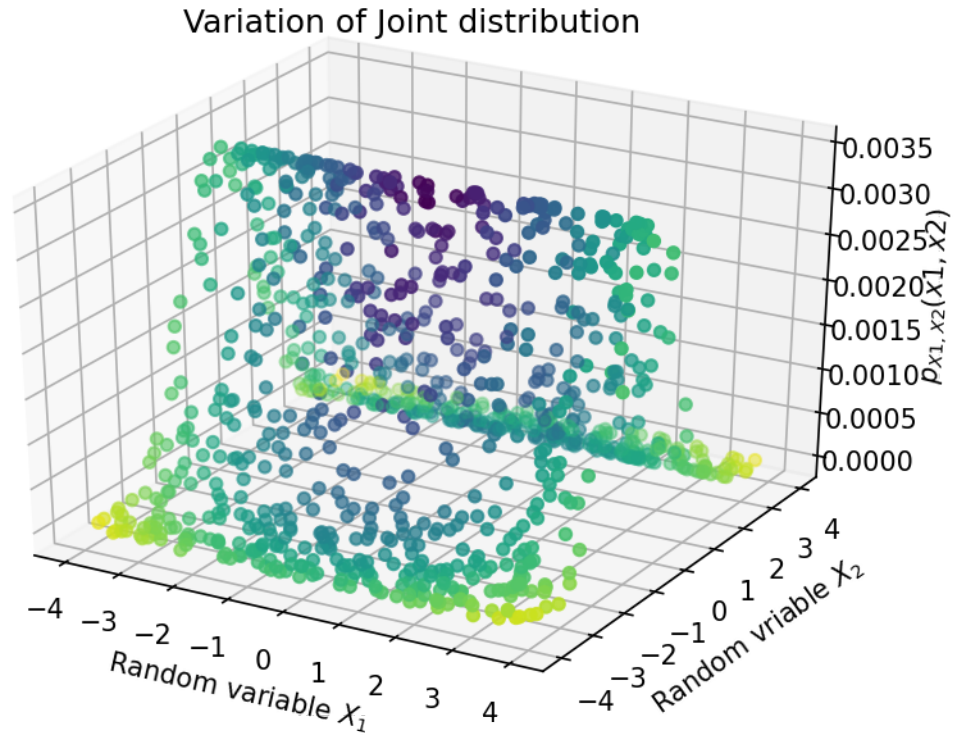
The equation transforms to :

$$z = x_1^2 + x_2^2$$

Thus (4) transforms to

$$p_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{(x_1^2 + x_2^2)}{2}}$$

Graphically equivalent to



Source: [https://github.com/harshal9876/AI5002/blob/main/Joint\\_distribution\\_Assignment1.png](https://github.com/harshal9876/AI5002/blob/main/Joint_distribution_Assignment1.png)

Which is the same as of equation (4), thus proving that for two normal random distribution

$$p_{X_1, X_2}(x_1, x_2) = p_{X_1}(x_1)p_{X_2}(x_2)$$

## 2 Problem Manual 6.2.2

**Question :**

Let

$$X1 = \sqrt{V} \cos \theta$$

$$X2 = \sqrt{V} \sin \theta$$

Evaluate the Jacobian

$$J = \begin{vmatrix} \frac{\partial X1}{\partial V} & \frac{\partial X2}{\partial V} \\ \frac{\partial X1}{\partial \theta} & \frac{\partial X2}{\partial \theta} \end{vmatrix}$$

**Solution:**

Using partial derivative

$$\frac{\partial X1}{\partial V} = \frac{1}{2\sqrt{V}} \cos \theta$$

$$\frac{\partial X2}{\partial V} = \frac{1}{2\sqrt{V}} \sin \theta$$

$$\frac{\partial X1}{\partial \theta} = -\sqrt{V} \sin \theta$$

$$\frac{\partial X2}{\partial \theta} = \sqrt{V} \cos \theta$$

Substituting the resultanat Jacobian would be

$$J = \begin{vmatrix} \frac{1}{2\sqrt{V}} \cos \theta & \frac{1}{2\sqrt{V}} \sin \theta \\ -\sqrt{V} \sin \theta & \sqrt{V} \cos \theta \end{vmatrix}$$