#### 1

# Random Numbers

# \*Govinda Rohith Y

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Abstract—This manual provides solutions to the Assignment on Random Numbers

# 1 Uniform Random Numbers

Let U be a uniform random variable between 0 and 1.

1.1 Generate  $10^6$  samples of U using a C program and save into a file called uni.dat .

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/1.1.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 1.1.c \$ ./a.out

1.2 Load the uni.dat file into python and plot the empirical CDF of *U* using the samples in uni.dat. The CDF is defined as

$$F_U(x) = \Pr\left(U \le x\right) \tag{1.1}$$

**Solution:** The following code plots Fig. 1.2

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/1.2.py

Download the above files and execute the following commands to produce Fig.1.2

\$ python3 1.2.py

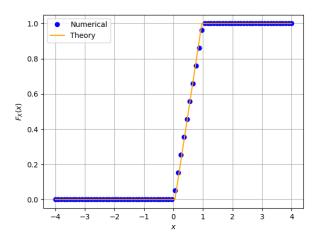


Fig. 1.2: The CDF of U

1.3 Find a theoretical expression for  $F_U(x)$ . Solution: Given U is a uniform Random Variable

$$p_U(x) = 1 \text{ for}$$
 (1.2)

$$F_U(x) = \int_{-\infty}^{\infty} p_U(x) dx \qquad (1.3)$$

$$\implies F_U(x) = \begin{cases} 0 & x \le 0 \\ x & 0 < x < 1 \\ 1 & x \ge 1 \end{cases}$$
 (1.4)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.5)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.6)

Write a C program to find the mean and variance of U.

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/1.4.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.7}$$

**Solution:** 

$$\operatorname{var}[U] = E[U - E[U]]^{2}$$

$$(1.8)$$

$$\Rightarrow \operatorname{var}[U] = E[U^{2}] - E[U]^{2}$$

$$(1.9)$$

$$E[U] = \int_{-\infty}^{\infty} x dF_{U}(x)$$

$$(1.10)$$

$$E[U] = \int_0^1 x$$
 (1.11)

$$\Longrightarrow E[U] = \frac{1}{2}$$
 (1.12)

$$E\left[U^{2}\right] = \int_{-\infty}^{\infty} x^{2} dF_{U}(x) \tag{1.13}$$

$$E[U^{2}] = \int_{0}^{1} x^{2} dF_{U}(x)$$
 (1.14)

$$\implies E\left[U^2\right] = \frac{1}{3} \tag{1.15}$$

$$\implies \left| \text{var} [U] = \frac{1}{12} = 0.0833 \right|$$
 (1.16)

### 2 Central Limit Theorem

2.1 Generate 10<sup>6</sup> samples of the random variable

$$X = \sum_{i=1}^{12} U_i - 6 \tag{2.1}$$

using a C program, where  $U_i$ , i = 1, 2, ..., 12 are a set of independent uniform random variables between 0 and 1 and save in a file called gau.dat

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/2.1.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 2.1.c \$ ./a.out

2.2 Load gau.dat in python and plot the empirical CDF of *X* using the samples in gau.dat. What properties does a CDF have?

**Solution:** The CDF of X is plotted in Fig. 2.2 using the code below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/2.2.py

Download the above files and execute the following commands to produce Fig.2.2

\$ python3 2.2.py

Some of the properties of CDF

- a)  $\lim_{x\to\infty} F_X(x) = 1$
- b)  $F_X(x)$  is non decreasing function.
- c) Symmetric about one point.
- 2.3 Load gau.dat in python and plot the empirical PDF of *X* using the samples in gau.dat. The PDF of *X* is defined as

$$p_X(x) = \frac{d}{dx} F_X(x) \tag{2.2}$$

What properties does the PDF have?

**Solution:** The PDF of X is plotted in Fig. 2.3 using the code below

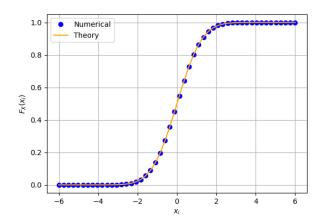


Fig. 2.2: The CDF of X

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/2.3.py

Download the above files and execute the following commands to produce Fig.2.3

\$ python3 2.3.py

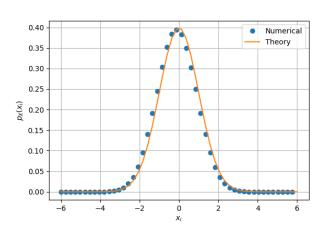


Fig. 2.3: The PDF of X

Some of the properties of the PDF:

- a) Symmetric about  $x = \mu$  in this case
- b) Decreasing function for  $x > \mu$  and increasing for  $x < \mu$  and attains maximum at  $x = \mu$
- c) Area under the curve is unity.
- 2.4 Find the mean and variance of *X* by writing a C program.

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/2.4.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 2.4.c \$ ./a.out

# 2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, (2.3)$$

repeat the above exercise theoretically.

# **Solution:**

1) CDF is given by

$$F_X(x) = \int_{-\infty}^{\infty} p_X(x) dx \qquad (2.4)$$

$$F_X(x) = 1 \tag{2.5}$$

2) Mean is given by

$$E(x) = \int_{-\infty}^{\infty} x p_X(x) dx$$
 (2.6)

$$\Longrightarrow \boxed{E(x) = 0} \tag{2.7}$$

3) Variance is given by

$$var[U] = E(U^2) - (E(U))^2$$
 (2.8)

$$Ex^2 = \int_{-\infty}^{\infty} x^2 p_X(x) dx \tag{2.9}$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} x^2 exp\left(-\frac{x^2}{2}\right) dx \qquad (2.10)$$

$$= \frac{1}{\sqrt{2\pi}} \left( x \int x exp\left(-\frac{x^2}{2}\right) dx \right) \quad (2.11)$$

$$-\frac{1}{\sqrt{2\pi}} \int \int \left( x exp\left(-\frac{x^2}{2}\right) \right) dx. dx \tag{2.12}$$

 $= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} exp\left(-\frac{x^2}{2}\right) dx \qquad (2.13)$ 

$$=\frac{\sqrt{2\pi}}{\sqrt{2\pi}}\tag{2.14}$$

$$= 1 \implies \boxed{\operatorname{var}[U] = 1} \tag{2.15}$$

# 3 From Uniform to Other

# 3.1 Generate samples of

$$V = -2\ln(1 - U) \tag{3.1}$$

and plot its CDF.

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/3.1.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

The CDF of *V* is plotted in Fig. 3.1 using the code below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/3.1pyth.py

Download the above files and execute the following commands to produce Fig.3.1

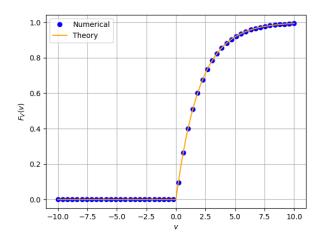


Fig. 3.1: The CDF of V

3.2 Find a theoretical expression for  $F_V(x)$ .

**Solution:** If Y = g(X), we know that  $F_Y(y) = F_X(g^{-1}(y))$ , here

$$V = -2\ln(1 - U) \tag{3.2}$$

$$1 - U = e^{\frac{-V}{2}} \tag{3.3}$$

$$U = 1 - e^{\frac{-V}{2}} \tag{3.4}$$

$$F_V(x) = F_U(1 - e^{\frac{-x}{2}})$$
 (3.5)

$$\implies F_V(x) = \begin{cases} 0 & x < 0 \\ 1 - e^{\frac{-x}{2}} & x \ge 0 \end{cases}$$
 (3.6)

### 4 Triangular Distribution

### 4.1 Generate

$$T = U_1 + U_2 \tag{4.1}$$

**Solution:** Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/4.1.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 4.1.c \$ ./a.out

4.2 Find the CDF of T.

**Solution:** The CDF of *T* is plotted in Fig. 4.2 using the code below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/4.5cdf.py

Download the above files and execute the following commands to produce Fig.4.2

\$ python3 4.5cdf.py

### 4.3 Find the PDF of T.

**Solution:** The PDF of T is plotted in Fig. 4.2 using the code below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/4.5pdf.py

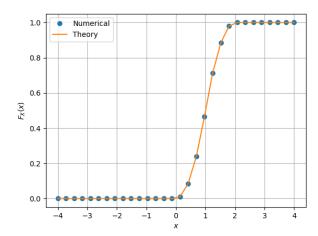


Fig. 4.2: The CDF of T

Download the above files and execute the following commands to produce Fig.4.2

\$ python3 4.5pdf.py

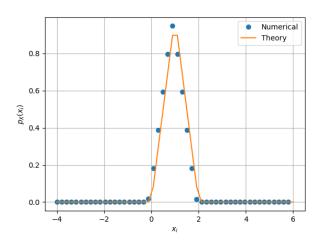


Fig. 4.3: The PDF of T

4.4 Find the Theoreotical Expression for the PDF and CDF of *T* 

# **Solution:**

$$T = U_1 + U_2 (4.2)$$

$$\implies p_T(t) = \int_{-\infty}^t p_{U1}(x)p_{U2}(y)dx \qquad (4.3)$$

$$As, p_{U1}(x) = p_{U1}(y) = p_{U}(u)$$
 (4.4)

$$\implies p_T(t) = \int_{-\infty}^t p_U(u) p_U(t-u) du \quad (4.5)$$

a) Theoretical PDF

i)  $t \le 1$ 

$$p_T(t) = \int_0^t p_U(t - u) du$$
 (4.6)

$$\implies p_T(t) = \int_0^t du = t \tag{4.7}$$

ii) t > 1

$$p_T(t) = \int_0^1 p_U(t - u) du$$
 (4.8)

$$\implies p_T(t) = \int_{t-1}^1 du = 2 - t$$
 (4.9)

$$\implies P_T(t) = \begin{cases} t & 0 \le t \le 1\\ 2 - t & 1 < t \le 2\\ 0 & t < 0 \text{ or } t > 2 \end{cases}$$

b) Theoretical CDF

$$F_T(t) = \int_{-\infty}^t p_T(u) du \tag{4.10}$$

$$\implies F_T(t) = \begin{cases} 0 & t < 0 \\ \frac{t^2}{2} & 0 \le t \le 1 \\ 2t - 1 - \frac{t^2}{2} & 1 < t \le 2 \\ 1 & t > 2 \end{cases}$$

4.5 Verify your results through a plot **Solution:** The Results are verified in the plots Fig 4.2 and Fig 4.3

# 5 Guasssian to Other

5.1 Generate equiprobable  $X \in \{1, -1\}$ . Solution: Download the following files and execute the C program.

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/5.1.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 5.1.c \$ ./a.out 5.2 Generate

$$Y = AX + N, (5.1)$$

Where A=5 dB and  $N \sim \mathcal{N}(0, 1)$ 

**Solution:**:

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/5.2.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands

\$ gcc 5.2.c

\$ ./a.out

5.3 Plot Y using a scatter plot.

**Solution:** The CDF of *V* is plotted in Fig. 5.5 using the code below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/5.3.py

Download the above files and execute the following commands to produce Fig.5.5

\$ python3 5.3.py

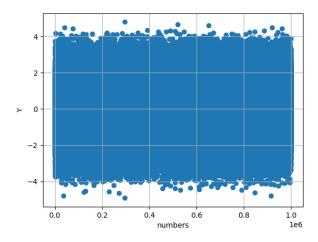


Fig. 5.5: The Scatter Plot

5.4 Guess how to estimate *X* from *Y*. **Solution:** :

- i) If Y < 0 then probably X = -1
- ii) If Y > 0 then probably X = 1

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.2)

and

$$P_{e|1} = \Pr(\hat{X} = 1|X = -1)$$
 (5.3)

**Solution:**:

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/5.5.c

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/source.h

Download the above files and execute the following commands to get the result

\$ gcc 5.5.c

\$ ./a.out

$$P_{e|0} = 0.499033 \tag{5.4}$$

$$P_{e|1} = 0.500138 \tag{5.5}$$

5.6 Find  $P_e$  assuming that X has equiprobable symbols.

**Solution:** 

$$P_e = P(X = 1)P_{e|0} + P(X = -1)P_{e|1}$$
 (5.6)

Since X is equiprobable

$$P(X = 1) = P(X = -1) = 0.5$$
 (5.7)

$$\implies P_e = \frac{P_{e|0} + P_{e|1}}{2} \tag{5.8}$$

$$\Longrightarrow \boxed{P_e = 0.499585} \tag{5.9}$$

5.7 Verify by plotting the theoretical  $P_e$  with respect to A from 0 to 10 dB.

**Solution:**:

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.10)

$$P_{e|0} = \Pr(AX + N < 0|X = 1)$$
 (5.11)

$$P_{e|0} = \Pr(N < -A)$$
 (5.12)

$$P_{e|0} = \int_{-\infty}^{-A} \frac{e^{\frac{-x^2}{2}}}{\sqrt{2\pi}}$$
 (5.13)

$$P_{e|0} = \int_{A}^{\infty} \frac{e^{\frac{-x^2}{2}}}{\sqrt{2\pi}}$$
 (5.14)

$$P_{e|0} = Q_N(A) (5.15)$$

Similarly, 
$$P_{e|1} = Q_N(A)$$
 (5.16)

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/5.6.py

Download the above files and execute the following commands to produce Fig.5.5

\$ python3 5.6.py

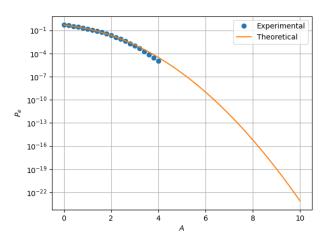


Fig. 5.5:  $P_e(A)$  with semilog-y axis

5.8 Now, consider a threshold  $\delta$  while estimating X from Y. Find the value of  $\delta$  that maximizes the theoretical  $P_e$ .

**Solution:** To estimate X from Y, we now consider the following:

$$X = \begin{cases} 1, & Y > \delta \\ -1, & Y < \delta \end{cases}$$
 (5.17)

Therefore,

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.18)

= 
$$Pr(AX + N < \delta | X = 1)$$
 (5.19)

$$\implies P_{e|0} = \Pr(N < \delta - A) \tag{5.20}$$

$$= \int_{-\infty}^{\delta - A} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$
 (5.21)

$$= \int_{4-\delta}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$
 (5.22)

$$\implies P_{e|0} = Q_N(A - \delta) \tag{5.23}$$

Similarly,

$$P_{e|1} = Q_N(A + \delta)$$
 (5.24)  

$$P_e = P_{e|0} \Pr(X = 1) + P_{e|1} \Pr(X = -1)$$
 (5.25)  

$$= \frac{Q_N(A - \delta) + Q_N(A + \delta)}{2}$$
 (5.26)

Differentiating the above equation wrt  $\delta$ :

$$0 = \frac{d}{d\delta} \left( \frac{Q_N(A - \delta) + Q_N(A + \delta)}{2} \right)$$

$$= \frac{1}{2} \left( \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta - A)^2}{2}} - \frac{1}{\sqrt{2\pi}} e^{-\frac{(A + \delta)^2}{2}} \right)$$
(5.28)

Therefore,

$$(\delta - A)^2 = (\delta + A)^2 \tag{5.29}$$

$$\implies \boxed{\delta = 0} \tag{5.30}$$

5.9 Repeat the above exercise when

$$p_X(0) = p (5.31)$$

**Solution:** Using Eq. (5.25), we have:

$$P_{e} = P_{e|0}p + P_{e|1}(1-p)$$

$$= pQ_{N}(A-\delta) + (1-p)Q_{N}(A+\delta)$$
(5.32)
(5.33)

Differentiating as before, we get:

$$0 = p \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta - A)^2}{2}} - (1 - p) \frac{1}{\sqrt{2\pi}} e^{-\frac{(A + \delta)^2}{2}}$$
(5.34)

$$e^{\frac{(\delta+A)^2 - ((\delta-A))^2}{2}} = \frac{1-p}{p}$$

$$\implies \delta = 0$$
(5.35)

5.10 Repeat the above exercise using the MAP criterion.

**Solution:** Assume that Pr(X = -1) = p, and Pr(X = 1) = (1 - p).

From Total Probability Theorem, we have:

$$p_Y(y) = p_{Y|X=-1}(y|-1) \Pr(X = -1) + p_{Y|X=1}(y|1) \Pr(X = 1)$$
 (5.37)

$$p_Y(y) = p \times p_{(-A+N)}(y)$$
 (5.38)

$$+(1-p) \times p_{(A+N)}(y)$$
 (5.39)

where  $p_Y(y)$  is the pdf of Y. Now,  $p_{(-A+N)}$  is just the pdf of a shifted normal distribution, and therefore:

$$p_Y(y) = p \frac{e^{-\frac{(y+A)^2}{2}}}{\sqrt{2\pi}} + (1-p) \frac{e^{-\frac{(y-A)^2}{2}}}{\sqrt{2\pi}}$$
 (5.40)

To use the MAP criterion, we must find  $p_{X|Y}(x|y)$ . To do this, we use the Theorem of Conditional Probability:

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x) \times p_X(x)}{p_Y(y)}$$
 (5.41)

When X = 1, we have:

$$p_{X|Y}(1|y) = \frac{p_{Y|X}(y|1) \times p_X(1)}{p_Y(y)}$$
 (5.42)

$$= \frac{(1-p)\frac{e^{\frac{-(y-A)^2}{2}}}{\sqrt{2\pi}}}{p\frac{e^{-\frac{(y+A)^2}{2}}}{\sqrt{2\pi}} + (1-p)\frac{e^{-\frac{(y-A)^2}{2}}}{\sqrt{2\pi}}}$$
(5.43)

$$= \frac{(1-p)e^{2yA}}{p+(1-p)e^{2yA}}$$
 (5.44)

Similarly, when X = -1, we get:

$$p_{X|Y}(-1|y) = \frac{p}{p + (1-p)e^{2yA}}$$
 (5.45)

Therefore, when  $p_{X|Y}(1|y) > p_{X|Y}(-1|y)$ , we have:

$$\frac{(1-p)e^{2yA}}{p+(1-p)e^{2yA}} > \frac{p}{p+(1-p)e^{2yA}}$$
 (5.46)

$$e^{2yA} > \frac{p}{(1-p)} \tag{5.47}$$

$$y > \frac{1}{2A} \ln \frac{p}{(1-p)}$$
 (5.48)

Therefore, when Eq. (5.48), we can assert that X = 1, and X = -1 otherwise. Now, consider when  $p = \frac{1}{2}$ . We have:

$$y > \frac{1}{2A} \ln \frac{p}{(1-p)} \tag{5.49}$$

$$= \frac{1}{2A} \ln 1 \tag{5.50}$$

$$=0 (5.51)$$

Therefore, when y > 0, we choose X = 1, and we choose X = -1 otherwise.

### 6 Guasssian to Other

6.1 Let  $X_1 \sim \mathcal{N}(0, 1)$  and  $X_2 \sim \mathcal{N}(0, 1)$ . Plot the CDF and PDF of

$$V = X_1^2 + X_2^2 \tag{6.1}$$

### **Solution:**

The CDF and PDF of *V* is plotted in Fig. 6.5 and 6.5 using the codes below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/6.1cdf.py

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/6.1pdf.py

Download the above files and execute the following commands to produce Fig.6.5 and 6.5

\$ python3 6.1cdf.py

\$ python3 6.1pdf.py

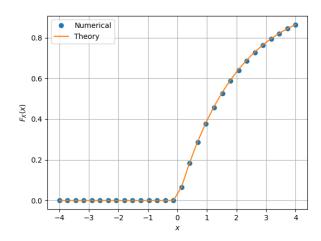


Fig. 6.5: The CDF of V

6.2 If

$$F_V(x) = \begin{cases} 1 - e^{-\alpha x} & x \ge 0\\ 0 & x < 0, \end{cases}$$
 (6.2)

find  $\alpha$ 

**Solution:** Assuming  $X_1$  and  $X_2$  are i.i.d

$$X_1 = r\cos\Theta \tag{6.3}$$

$$X_2 = r \sin \Theta \tag{6.4}$$

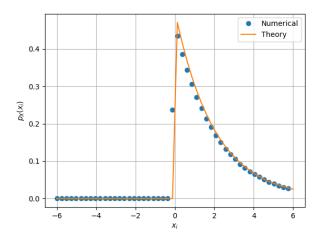


Fig. 6.5: The PDF of V

The Jacobian matrix transforming  $r, \Theta$  to  $X_1, X_2$  is defined as

$$\mathbf{J} = \begin{pmatrix} \frac{\partial X_1}{\partial r} & \frac{\partial X_1}{\partial \Theta} \\ \frac{\partial X_2}{\partial r} & \frac{\partial X_2}{\partial \Theta} \end{pmatrix} \tag{6.5}$$

$$\implies \mathbf{J} = \begin{pmatrix} \cos\Theta & -r\sin\Theta \\ \sin\Theta & r\cos\Theta \end{pmatrix} \tag{6.6}$$

$$\implies |\mathbf{J}| = r \tag{6.7}$$

(6.8)

$$f_{X_1,X_2}(x_1,x_2) = f_{X_1}(x_1)f_{X_2}(x_2)$$
 (6.9)

$$\implies f_{X_1,X_2}(x_1,x_2) = \frac{e^{\frac{-x_1^2 - x_2^2}{2}}}{2\pi}$$
 (6.10)

$$\implies f_{X_1,X_2}(x_1,x_2) = \frac{e^{\frac{-r^2}{2}}}{2\pi}$$
 (6.11)

$$f_{R,\Theta}(r,\theta) = f_{X_1,X_2}(x_1,x_2) |\mathbf{J}|$$
 (6.12)

$$\implies f_{R,\Theta}(r,\theta) = \frac{re^{\frac{-r^2}{2}}}{2\pi} \tag{6.13}$$

$$f_{R}(r) = \int_{-r}^{2\pi} f_{R,\Theta}(r,\theta) d\theta$$

$$f_R(r) = \int_0^{2\pi} f_{R,\Theta}(r,\theta) d\theta$$
(6.14)

$$f_R(r) = \int_0^{2\pi} \frac{re^{\frac{-r^2}{2}}}{2\pi} d\theta$$
 (6.15)

$$\implies \boxed{f_R(r) = re^{\frac{-r^2}{2}}}$$

$$F_V(r) = F_{X_1^2 + X_2^2}(r)$$
(6.16)
(6.17)

$$F_V(r) = F_{X_1^2 + X_2^2}(r)$$
 (6.17)

$$\implies F_V(r) = F_{R^2}(r) \tag{6.18}$$

$$\implies F_V(r) = \Pr(R^2 \le x)$$
 (6.19)

$$\implies F_V(r) = \Pr(R \le \sqrt{x})$$
 (6.20)

But 
$$F_V(x) = \Pr(R \le x)$$
 (6.21)

$$\implies F_V(x) = \int_0^x f_R(r)dr \tag{6.22}$$

$$F_V(x) = 1 - e^{\frac{-x^2}{2}} (6.23)$$

$$\implies F_V(r) = \begin{cases} 0 & r < 0 \\ 1 - e^{\frac{-r}{2}} & r \ge 0 \end{cases}$$
 (6.24)

$$\implies \boxed{\alpha = \frac{1}{2}} \tag{6.25}$$

6.3 Plot the CDF and PDF of

$$A = \sqrt{V} \tag{6.26}$$

**Solution:** : For  $r \ge 1$ 

$$F_A(r) = \Pr\left(A \le r\right) \tag{6.27}$$

$$= \Pr\left(\sqrt{V} \le r\right) \tag{6.28}$$

$$= \Pr\left(V \le r^2\right) \tag{6.29}$$

$$= F_V(r^2) = 1 - e^{-\frac{r^2}{2}}$$
 (6.30)

and so on differentiating

$$f_A(r) = re^{-\frac{r^2}{2}}$$
 (6.31)

Thus, the CDF and PDF of A is given by

$$F_V(r) = \begin{cases} 1 - e^{-\frac{r^2}{2}} & r \ge 0\\ 0 & r < 0 \end{cases}$$
 (6.32)

$$f_V(r) = \begin{cases} re^{-\frac{r}{2}} & r \ge 0\\ 0 & r < 0 \end{cases}$$
 (6.33)

The CDF and PDF of A is plotted in Fig. 6.5 and 6.5 using the codes below

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/6.3cdf.py

wget https://github.com/GovindaRohith/ Assignments/blob/main/Randomnum/ codes/6.3pdf.py

Download the above files and execute the following commands to produce Fig.6.5 and

\$ python3 6.3cdf.py

\$ python3 6.3pdf.py

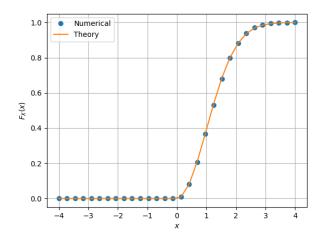


Fig. 6.5: The CDF of V

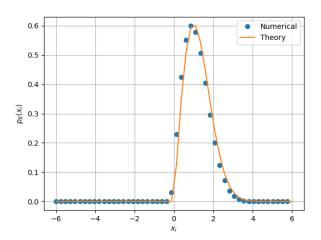


Fig. 6.5: The PDF of V