1

Random Numbers

AI1110: Probability and Random Variables

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30 Jun 2022

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 C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/1.1.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

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: Download the following Python code that plots Fig. 1.2

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/1.2.py

Run the code by executing

1.3 Find a theoretical expression for $F_U(x)$

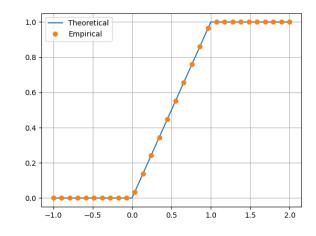


Fig. 1.2. The CDF of U

Solution: The PDF of *U* is given by

$$p_U(x) = \begin{cases} 1 & x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$
 (1.2)

The CDF of U is given by

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

If x < 0.

$$\int_{-\infty}^{x} p_U(x) \, dx = \int_{-\infty}^{x} 0 \, dx = 0 \qquad (1.4)$$

If c,

$$\int_{-\infty}^{x} p_U(x) \, dx = \int_{-\infty}^{0} 0 \, dx + \int_{0}^{x} 1 \, dx \quad (1.5)$$

$$= 0 + x \tag{1.6}$$

$$= x \tag{1.7}$$

If x > 1,

$$\int_{-\infty}^{x} p_{U}(x) dx$$

$$= \int_{-\infty}^{0} 0 dx + \int_{0}^{1} 1 dx + \int_{1}^{x} 0 dx \quad (1.8)$$

$$\int_{-\infty}^{x} p_U(x) \, dx = 0 + 1 + 0 \qquad (1.9)$$

$$= 1 \qquad (1.10)$$

Therefore, we obtain the CDF of U as

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \le x \le 1 \\ 1 & x > 1 \end{cases}$$
 (1.11)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.12)

and its variance as

$$Var[U] = E[U - E[U]]^2$$
 (1.13)

Write a C program to find the mean and variance of \boldsymbol{U}

Solution: Download the C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/1.4.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

The output of the code is

$$\mu_{\rm emp} = 0.500007 \tag{1.14}$$

$$\mu_{\text{the}} = 0.500000 \tag{1.15}$$

$$\sigma_{\rm emp}^2 = 0.083301 \tag{1.16}$$

$$\sigma_{\text{the}}^2 = 0.083333 \tag{1.17}$$

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} \mathrm{d}F_{U}(x) \tag{1.18}$$

Solution: The mean of U is given by

$$E[U] = \int_{-\infty}^{\infty} x \, \mathrm{d}F_U(x) \tag{1.19}$$

On differentiating the CDF of U, we get

$$dF_U(x) = \begin{cases} 0 & x < 0 \\ dx & 0 \le x \le 1 \\ 0 & x > 1 \end{cases}$$
 (1.20)

$$\therefore E[U] = \int_0^1 x \, dx = \frac{1}{2} = 0.5 \quad (1.21)$$

Similarly,

$$\therefore E[U^2] = \int_0^1 x^2 \, dx = \frac{1}{3}$$
 (1.22)

Now, the variance of U is given by

$$Var[U] \tag{1.23}$$

$$= E [U - E [U]]^{2}$$
 (1.24)

$$= E \left[U^2 - 2UE[U] + (E[U])^2 \right]$$
 (1.25)

By linearity of expectation, we have

$$E[U^{2}] + E[-2UE[U]] + E[(E[U])^{2}]$$
 (1.26)

$$= E[U^{2}] - 2E[U]E[U] + (E[U])^{2}$$
 (1.27)

$$= E[U^{2}] - (E[U])^{2}$$
 (1.28)

$$=\frac{1}{3} - \left(\frac{1}{2}\right)^2 \tag{1.29}$$

$$=\frac{1}{12}\approx 0.083333\tag{1.30}$$

2. Central Limit Theorem

2.1 Generate 10⁶ samples of the random variable

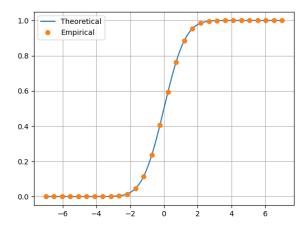
$$X = \sum_{i=1}^{12} U_i - 6 \tag{2.1}$$

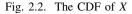
using a C program, where U_i , i = 1, 2, ..., 12 are a set of independent uniform random variables between 0 and 1 and save in a file called gau.dat

Solution: Download the C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/2.1.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h





Compile and run the C program by executing the following

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: Download the following Python code that plots Fig. 2.2

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/2.2.py

Run the code by executing

Every CDF is monotone increasing and right-continuous. Furthermore,

$$\lim_{x \to -\infty} F_X(x) = 0 \qquad \lim_{x \to \infty} F_X(x) = 1 \qquad (2.2)$$

Thus, every CDF is bounded between 0 and 1 and hence, convergent.

In this case, the CDF is also left-continuous. Therefore, *X* is a continuous random variable.

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

$$p_X(x) = \frac{\mathrm{d}}{\mathrm{d}x} F_X(x) \tag{2.3}$$

What properties does the PDF have?

Solution: Download the following Python code that plots Fig. 2.2

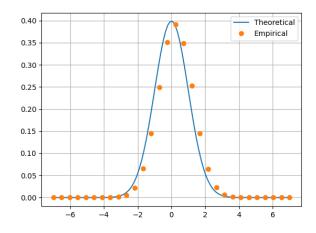


Fig. 2.3. The PDF of X

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/2.3.py

Run the code by executing

python 2.3.py

Every PDF is bounded between 0 and 1 and

$$\int_{-\infty}^{\infty} p_X(x) \, \mathrm{d}x = 1 \tag{2.4}$$

In this case, the PDF is symmetric about x = 02.4 Find the mean and variance of X by writing a C program

Solution: Download the C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/2.4.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

The output of the code is

$$\mu_{\rm emp} = 0.000294 \tag{2.5}$$

$$\mu_{\text{the}} = 0.000000$$
 (2.6)

$$\sigma_{\rm emp}^2 = 0.999560 \tag{2.7}$$

$$\sigma_{\text{the}}^2 = 1.000000 \tag{2.8}$$

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.9)$$

repeat the above exercise theoretically

Solution: The mean of *X* is given by

$$E[X] = \int_{-\infty}^{\infty} x p_X(x) dx$$
 (2.10)

$$= \int_{-\infty}^{\infty} \frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \qquad (2.11)$$

Now, let

$$g(x) = \frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \tag{2.12}$$

$$\implies g(-x) = \frac{-x}{\sqrt{2\pi}} \exp\left(-\frac{(-x)^2}{2}\right) \qquad (2.13)$$

$$= -\frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \qquad (2.14)$$

$$= -g(x) \tag{2.15}$$

Thus, g(x) is an odd function

$$\therefore E[X] = \int_{-\infty}^{\infty} g(x) dx = 0 \qquad (2.16)$$

Now,

$$E\left[X^{2}\right] = \int^{\infty} x^{2} p_{X}(x) dx \qquad (2.17)$$

$$= \int_{-\infty}^{\infty} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \qquad (2.18)$$

$$=2\int_0^\infty \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \qquad (2.19)$$

since $\frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$ is an even function Using integration by parts,

$$E\left[X^{2}\right] = \sqrt{\frac{2}{\pi}} \int_{0}^{\infty} x \cdot x \exp\left(-\frac{x^{2}}{2}\right) dx \quad (2.20)$$

$$= \sqrt{\frac{2}{\pi}} \left(x \int x \exp\left(-\frac{x^2}{2}\right) dx \right) \Big|_0^{\infty}$$
$$- \sqrt{\frac{2}{\pi}} \int_0^{\infty} 1 \cdot \int x \exp\left(-\frac{x^2}{2}\right) dx \quad (2.21)$$

Substitute $t = \frac{x^2}{2} \implies dt = xdx$

$$\int x \exp\left(-\frac{x^2}{2}\right) dx = \int \exp(-t) dt \qquad (2.22)$$

$$= -\exp(-t) \tag{2.23}$$

$$= -\exp\left(-\frac{x^2}{2}\right) \qquad (2.24)$$

Now,

$$-x \exp\left(-\frac{x^2}{2}\right)\Big|_0^\infty = 0 - 0 = 0$$
 (2.25)

$$\lim_{x \to \infty} x \exp\left(-\frac{x^2}{2}\right) = \lim_{x \to \infty} \frac{x}{\exp\left(\frac{x^2}{2}\right)} = 0 \quad (2.26)$$

as exponential function grows much faster than a polynomial function Also,

$$\int_0^\infty -\exp\left(-\frac{x^2}{2}\right) \mathrm{d}x \tag{2.27}$$

$$\stackrel{x=t\sqrt{2}}{\longleftrightarrow} \int_0^\infty -\exp(-t^2) dt \sqrt{2}$$
 (2.28)

$$= -\sqrt{2} \int_0^\infty \exp(-t^2) dt \qquad (2.29)$$

$$=-\sqrt{2}\frac{\sqrt{\pi}}{2}\tag{2.30}$$

$$=-\sqrt{\frac{\pi}{2}}$$
 (2.31)

Therefore,

$$E[X^2] = 0 - \sqrt{\frac{2}{\pi}} \left(-\sqrt{\frac{\pi}{2}} \right)$$
 (2.32)

$$= 1 \tag{2.33}$$

:. Var
$$[X] = E[X^2] - (E[X])^2$$
 (2.34)

$$=1-0$$
 (2.35)

$$= 1 \tag{2.36}$$

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 C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/3.1.c

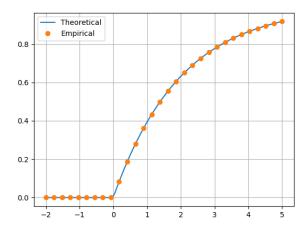


Fig. 3.1. The CDF of V

Compile and run the C program by executing the following

Download the following Python code that plots Fig. 3.1

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/3.1.py

Run the code by executing

python 3.1.py

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

Solution: We have

$$F_{V}(x) = \Pr(V \le x)$$

$$= \Pr(-2 \ln (1 - U) \le x)$$

$$= \Pr\left(\ln (1 - U) \ge -\frac{x}{2}\right)$$

$$= \Pr\left(1 - U \ge \exp\left(-\frac{x}{2}\right)\right)$$

$$= \Pr\left(U \le 1 - \exp\left(-\frac{x}{2}\right)\right)$$
(3.2)
(3.3)
(3.4)
(3.5)

(3.7)

Now,

$$0 \le 1 - \exp\left(-\frac{x}{2}\right) < 1 \qquad \text{if } x \ge 0 \qquad (3.8)$$
$$1 - \exp\left(-\frac{x}{2}\right) < 0 \qquad \text{if } x < 0 \qquad (3.9)$$

 $=F_U\left(1-\exp\left(-\frac{x}{2}\right)\right)$

Therefore,

$$F_V(x) = \begin{cases} 1 - \exp\left(-\frac{x}{2}\right) & x \ge 0\\ 0 & x < 0 \end{cases}$$
 (3.10)

4. Triangular Distribution

4.1 Generate

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

Solution: Download the C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/4.1.c

Compile and run the C program by executing the following

4.2 Find the CDF of T

Solution: The CDF of T is given by

$$F_T(t) = \Pr(T \le t) = \Pr(U_1 + U_2 \le t)$$
 (4.2)

Since $U_1, U_2 \in [0, 1] \implies U_1 + U_2 \in [0, 2]$ Therefore, if $t \ge 2$, then $U_1 + U_2 \le t$ is always true and if t < 0, then $U_1 + U_2 \le t$ is always false.

Now, fix the value of U_1 to be some x

$$x + U_2 \le t \implies U_2 \le t - x$$
 (4.3)

If $0 \le t \le 1$, then x can take all values in [0, t]

$$F_T(t) = \int_0^t \Pr(U_2 \le t - x) \, p_{U_1}(x) \mathrm{d}x \quad (4.4)$$

$$= \int_0^t F_{U_2}(t-x)p_{U_1}(x)\mathrm{d}x \tag{4.5}$$

$$0 \le x \le t \implies 0 \le t - x \le t \le 1$$
 (4.6)

$$\implies F_{U_2}(t-x) = t - x \qquad (4.7)$$

$$F_T(t) = \int_0^t (t - x) \cdot 1 \cdot \mathrm{d}x \tag{4.8}$$

$$= tx - \frac{x^2}{2} \bigg|_0^t \tag{4.9}$$

$$=\frac{t^2}{2}$$
 (4.10)

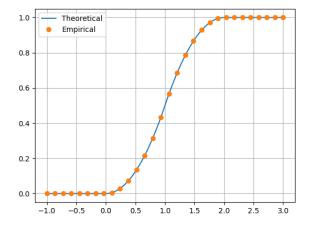


Fig. 4.2. The CDF of T

If 1 < t < 2, x can only take values in [0, 1] as $U_1 \le 1$

$$F_T(t) = \int_0^1 F_{U_2}(t - x) \cdot 1 \cdot dx$$
 (4.11)

$$0 \le x \le t - 1 \implies 1 \le t - x \le t$$

$$t - 1 \le x \le 1 \implies 0 < t - 1 \le t - x \le 1$$

$$(4.12)$$

$$(4.13)$$

$$F_T(t) = \int_0^{t-1} 1 dx + \int_{t-1}^1 (t - x) dx$$
 (4.14)

$$= t - 1 + t(1 - (t - 1)) - \frac{1}{2} + \frac{(t - 1)^2}{2}$$
 (4.15)

$$= t - 1 + 2t - t^2 - \frac{1}{2} + \frac{t^2}{2} + \frac{1}{2} - t$$
 (4.16)

$$= -\frac{t^2}{2} + 2t - 1$$
 (4.17)

Therefore,

$$F_T(t) = \begin{cases} 0 & t < 0 \\ \frac{t^2}{2} & 0 \le t \le 1 \\ 2t - \frac{t^2}{2} - 1 & 1 < t < 2 \\ 1 & t \ge 2 \end{cases}$$
 (4.18)

4.3 Find the PDF of T

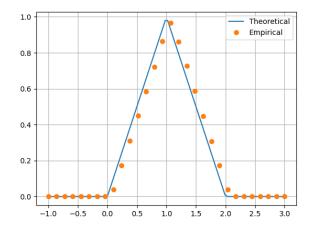


Fig. 4.3. The PDF of T

Solution: The PDF of *T* is given by

$$p_T(t) = \frac{\mathrm{d}}{\mathrm{d}t} F_T(t) \tag{4.19}$$

$$\therefore p_T(t) = \begin{cases} 0 & t < 0 \\ t & 0 \le t \le 1 \\ 2 - t & 1 < t < 2 \\ 0 & t \ge 2 \end{cases}$$
 (4.20)

4.4 Find the theoretical expressions for the PDF and CDF of *T*

Solution: The theoretical expressions for the CDF and PDF have been found in problems 4.2 and 4.3 respectively

4.5 Verify your results through a plot

Solution: Download the following Python codes that plot Fig. 4.2 and Fig. 4.3

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/4.2.py

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/4.3.py

Run the codes by executing

python 4.2.py python 4.3.py

5. Maximum Likelihood

5.1 Generate equiprobable $X \in \{-1, 1\}$ Solution: Download the C source code by executing the following commands wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/5.1.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

5.2 Generate

$$Y = AX + N \tag{5.1}$$

where A = 5 dB, $X \in \{-1, 1\}$ is Bernoulli and $N \sim \mathcal{N}(0, 1)$

Solution: Download the C source code by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/5.2.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

5.3 Plot *Y*

Solution: Download the following Python code that plots Fig. 5.3

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/5.3.py

Run the code by executing

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

$$X = \begin{cases} 1 & Y > 0 \\ -1 & Y < 0 \end{cases}$$
 (5.2)

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.3)

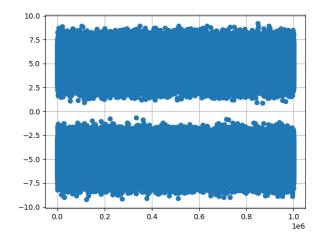


Fig. 5.3. Plot of *Y*

and

$$P_{e|1} = \Pr(\hat{X} = 1|X = -1)$$
 (5.4)

Solution:

$$\Pr(\hat{X} = -1|X = 1) = \Pr(Y < 0|X = 1)$$
 (5.5)

$$= \Pr(A + N < 0)$$
 (5.6)

$$= \Pr(N < -5)$$
 (5.7)

$$= 1 - \Pr(N > -5) \quad (5.8)$$

$$= 1 - Q(-5) \tag{5.9}$$

$$= Q(5) \tag{5.10}$$

where Q(x) = Pr(N > x) is the Q-function

$$Q(x) = 1 - Q(-x) \qquad \forall x \in \mathbb{R} \tag{5.11}$$

$$\Pr(\hat{X} = 1|X = -1) = \Pr(Y > 0|X = -1)$$
(5.12)

$$= Pr(-A + N > 0)$$
 (5.13)

$$= \Pr(N > 5)$$
 (5.14)

$$= Q(5) \tag{5.15}$$

5.6 Find P_e assuming that X has equiprobable symbols

Solution:

$$P_e = \Pr(X = -1) P_{e|0} + \Pr(X = 1) P_{e|1}$$
 (5.16)

Since X has equiprobable symbols, $Pr(X = -1) = Pr(X = 1) = \frac{1}{2}$

$$P_e = \frac{1}{2}Q(5) + \frac{1}{2}Q(5) \tag{5.17}$$

$$=Q(5) \tag{5.18}$$

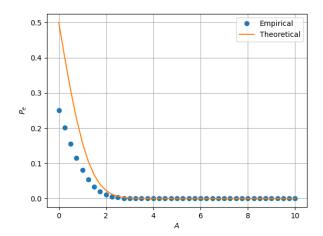


Fig. 5.7. Plot of P_e

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

Solution: Download the following Python code that plots Fig. 5.7

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/5.7.py

Run the code by executing

5.8 Now, consider a threshold δ while estimating X from Y. Find the value of δ that minimizes the theoretical P_e

Solution:

$$X = \begin{cases} 1 & Y > \delta \\ -1 & Y < \delta \end{cases} \tag{5.19}$$

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.20)

$$= \Pr\left(Y < \delta | X = 1\right) \tag{5.21}$$

$$= \Pr\left(A + N < \delta\right) \tag{5.22}$$

$$= \Pr\left(N < \delta - 5\right) \tag{5.23}$$

$$= Q(5 - \delta) \tag{5.24}$$

$$P_{e|1} = \Pr(\hat{X} = 1|X = -1)$$
 (5.25)

=
$$\Pr(Y > \delta | X = -1)$$
 (5.26)

$$= \Pr\left(-A + N > \delta\right) \tag{5.27}$$

$$= \Pr\left(N > \delta + 5\right) \tag{5.28}$$

$$= Q(5+\delta) \tag{5.29}$$

Now, P_e is given by

$$P_e = \Pr(X = -1) P_{e|0} + \Pr(X = 1) P_{e|1}$$
 (5.31)

$$= \frac{1}{2}Q(5-\delta) + \frac{1}{2}Q(5+\delta)$$
 (5.32)

$$= \frac{Q(5-\delta) + Q(5+\delta)}{2}$$
 (5.33)

$$= g(\delta) \tag{5.34}$$

On differentiating g with respect to δ , we get

$$g'(\delta) = \frac{Q'(5+\delta) - Q'(5-\delta)}{2}$$
 (5.35)

Recall the definition of Q(x)

$$Q(x) = \int_{x}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u^2}{2}\right) du \quad (5.36)$$

$$\implies Q'(x) = -\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \tag{5.37}$$

Thus.

$$g'(\delta) = \frac{\exp\left(-\frac{(5-\delta)^2}{2}\right) - \exp\left(-\frac{(5+\delta)^2}{2}\right)}{2\sqrt{2\pi}}$$
(5.38)

$$g'(\delta) = 0 \implies (5 - \delta)^2 = (5 + \delta)^2$$
 (5.39)

$$\implies |5 - \delta| = |5 + \delta| \tag{5.40}$$

$$\implies \delta = 0$$
 (5.41)

$$g''(\delta) = \frac{(5-\delta)}{2\sqrt{2\pi}} \exp\left(-\frac{(5-\delta)^2}{2}\right) + \frac{(5+\delta)}{2\sqrt{2\pi}} \exp\left(-\frac{(5+\delta)^2}{2}\right)$$
 (5.42)

$$g''(0) = \frac{5}{\sqrt{2\pi}} \exp\left(-\frac{25}{2}\right) > 0 \tag{5.43}$$

Therefore, $\hat{\delta} = 0$ is a minima and it is what minimizes P_e

5.9 Repeat the above exercise when

$$p_X(-1) = p (5.44)$$

Solution:

$$P_e = p_X(1)P_{e|0} + p_X(-1)P_{e|1}$$
 (5.45)

$$= (1 - p)Q(5 - \delta) + pQ(5 + \delta)$$
 (5.46)

$$= g(\delta) \tag{5.47}$$

On differentiating g with respect to δ , we get

$$g'(\delta) = \frac{(1-p)\exp(-\frac{(5-\delta)^2}{2}) - p\exp(-\frac{(5+\delta)^2}{2})}{\sqrt{2\pi}}$$
(5.48)

 $g'(\delta) = 0$ when

$$(1-p)\exp\left(-\frac{(5-\delta)^2}{2}\right) = p\exp\left(-\frac{(5+\delta)^2}{2}\right)$$
(5.49)

$$\implies \exp\left(\frac{(5+\delta)^2 - (5-\delta)^2}{2}\right) = \frac{p}{1-p}$$
 (5.50)

$$\implies \exp(10\delta) = \frac{p}{1-p} \tag{5.51}$$

$$\therefore \hat{\delta} = \frac{1}{10} \ln \frac{p}{1 - p} \tag{5.52}$$

5.10 Repeat the above exercise using the MAP criterion

Solution: The PDF of X|Y is given by

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x)p_X(x)}{p_Y(y)}$$
 (5.53)

Assuming *X* has equiprobable symbols, $p_X(x) = \frac{1}{2}$, x = -1, 1

$$p_{Y}(y) = p_{X}(-1)p_{Y|X}(y|-1) + p_{X}(1)p_{Y|X}(y|1)$$

$$= \frac{1}{2}\Pr(-A+N=y) + \frac{1}{2}\Pr(A+N=y)$$
(5.54)
$$= \frac{p_{N}(y+5) + p_{N}(y-5)}{2}$$
(5.56)

$$= \frac{\exp\left(-\frac{(y+5)^2}{2}\right) + \exp\left(-\frac{(y-5)^2}{2}\right)}{2\sqrt{2\pi}}$$
 (5.57)

Now,

$$p_{X|Y}(1|y) = \frac{\Pr(A + N = y) p_X(1)}{p_Y(y)}$$
(5.59)

$$= \frac{p_N(y - 5)p_X(1)}{p_Y(y)}$$
(5.60)

$$= \frac{\exp\left(-\frac{(y - 5)^2}{2}\right)}{2\left(\exp\left(-\frac{(y + 5)^2}{2}\right) + \exp\left(-\frac{(y - 5)^2}{2}\right)\right)}$$
(5.61)

$$= \frac{1}{2\left(1 + \exp\left(\frac{(y - 5)^2 - (y + 5)^2}{2}\right)\right)}$$
(5.62)

$$= \frac{1}{2\left(1 + \exp\left(-\frac{(y - 5)^2 - (y + 5)^2}{2}\right)\right)}$$
(5.63)

Similarly,

$$p_{X|Y}(-1|y) = \frac{1}{2(1 + \exp(10y))}$$
 (5.64)

Now,

$$\iff \frac{p_{X|Y}(1|y) > p_{X|Y}(-1|y)}{2(1 + \exp(-10y))} > \frac{1}{2(1 + \exp(10y))}$$
(5.65)
(5.66)

$$\iff \exp(-10y) < \exp(10y) \tag{5.67}$$

$$\iff y > 0 \tag{5.68}$$

And $p_{X|Y}(1|y) < p_{X|Y}(-1|y) \iff y < 0$ Therefore, X = 1 is more probable than X = -1when Y > 0 and vice versa Consider now a general Bernoulli random variable X with $p_X(-1) = p$, $p_X(1) = 1 - p$

$$p_{Y}(y) = p_{X}(-1)p_{Y|X}(y|-1) + p_{X}(1)p_{Y|X}(y|1)$$

$$= pp_{N}(y+5) + (1-p)p_{N}(y-5) \quad (5.70)$$

$$= \frac{p\exp\left(-\frac{(y+5)^{2}}{2}\right) + (1-p)\exp\left(-\frac{(y-5)^{2}}{2}\right)}{\sqrt{2\pi}}$$
(5.71)

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

$$= \frac{(1-p)\exp\left(-\frac{(y-5)^2}{2}\right)}{p\exp\left(-\frac{(y+5)^2}{2}\right) + (1-p)\exp\left(-\frac{(y-5)^2}{2}\right)}$$

$$= \frac{1-p}{1-p+p\exp(-10y)}$$
(5.74)

Similarly,

$$p_{X|Y}(-1|y) = \frac{p}{p + (1-p)\exp(10y)}$$
 (5.75)

Now,

$$p_{X|Y}(1|y) > p_{X|Y}(-1|y)$$

$$\iff \frac{1-p}{1-p+p\exp(-10y)} > \frac{p}{p+(1-p)\exp(10)}$$

$$\iff 1+\frac{p}{1-p}\exp(-10y) < 1+\frac{1-p}{p}\exp(10y)$$

$$\iff \exp(20y) > \left(\frac{p}{1-p}\right)^2$$
 (5.79)

$$\iff y > \frac{1}{10} \ln \frac{p}{1-p} = \hat{\delta}$$
 (5.80)

and

$$p_{X|Y}(1|y) < p_{X|Y}(-1|y)$$
 (5.81)

$$\iff \frac{1-p}{1-p+p\exp(-10y)} < \frac{p}{p+(1-p)\exp(10)}$$
 (5.82)

$$\iff 1+\frac{p}{1-p}\exp(-10y) > 1+\frac{1-p}{p}\exp(10y)$$
 (5.83)

$$\iff \exp(20y) < \left(\frac{p}{1-p}\right)^2$$
 (5.84)

$$\iff y < \frac{1}{10} \ln \frac{p}{1-p} = \hat{\delta}$$
 (5.85)

Therefore, X = 1 is more probable than X = -1 when $Y > \hat{\delta}$ and vice versa



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: Download the C source code to generate the data by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/6.1.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

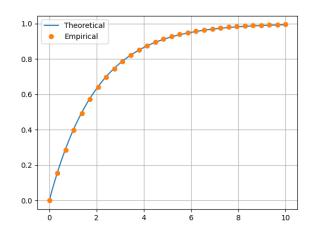


Fig. 6.1. CDF of *V*

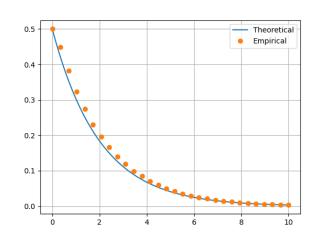


Fig. 6.1. PDF of *V*

Compile and run the C program by executing the following

cc -lm 6.1.c ./a.out

Download the following Python code that plots Fig. 6.1

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/6.1.py

Run the code by executing

python 6.1.py

6.2 If

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

find α

Solution: Let $R \ge 0, \Theta \in [0, 2\pi]$

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

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

such that $V = X_1^2 + X_2^2 = R^2$ The Jacobian matrix transforming R, Θ to X_1, X_2 is defined as

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

$$= \begin{pmatrix} \cos \Theta & -R \sin \Theta \\ \sin \Theta & R \cos \Theta \end{pmatrix} \tag{6.6}$$

$$\implies |\mathbf{J}| = R\cos^2\Theta + R\sin^2\Theta = R$$
 (6.7)

Then,

$$p_{R,\Theta}(r,\theta) = p_{X_1,X_2}(x_1,x_2) |\mathbf{J}|$$
 (6.8)

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

$$= r p_{X_1}(x_1) p_{X_2}(x_2) \tag{6.10}$$

since X_1 and X_2 are independent

$$p_{R,\Theta}(r,\theta) = r \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x_1^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x_2^2}{2}\right)$$
(6.11)

$$= \frac{r}{2\pi} \exp\left(-\frac{(x_1 + x_2)^2}{2}\right) \tag{6.12}$$

$$= \frac{r}{2\pi} \exp\left(-\frac{r^2}{2}\right) \tag{6.13}$$

The marginal distribution of R is given by

$$p_R(r) = \int_0^{2\pi} p_{R,\Theta}(r,\theta) \,\mathrm{d}\theta \tag{6.14}$$

$$= \frac{r}{2\pi} \exp\left(-\frac{r^2}{2}\right) \int_0^{2\pi} d\theta \qquad (6.15)$$

$$= r \exp\left(-\frac{r^2}{2}\right) \quad \text{for } r \ge 0 \tag{6.16}$$

The CDF of R is thus given by

$$F_R(r) = \Pr\left(R \le r\right) \tag{6.17}$$

$$= \int_0^r p_R(r) \mathrm{d}r \tag{6.18}$$

$$= \int_0^r r \exp\left(-\frac{r^2}{2}\right) dr \tag{6.19}$$

$$= -\int_0^r \exp\left(-\frac{r^2}{2}\right)(-r\mathrm{d}r) \tag{6.20}$$

$$= -\int_0^r \exp\left(-\frac{r^2}{2}\right) d\left(-\frac{r^2}{2}\right) \tag{6.21}$$

$$= -\exp\left(-\frac{r^2}{2}\right)\Big|_0^r \tag{6.22}$$

$$= 1 - \exp\left(-\frac{r^2}{2}\right) \quad \text{for } r \ge 0 \qquad (6.23)$$

Now, $V = R^2$, hence the CDF of V is given by

$$F_V(x) = \Pr\left(V \le x\right) \tag{6.24}$$

$$= \Pr\left(R^2 \le x\right) \tag{6.25}$$

$$=\Pr\left(|R| \le \sqrt{x}\right) \tag{6.26}$$

$$= \Pr\left(R \le \sqrt{x}\right) \qquad (\because R \ge 0)$$

(6.27)

$$= p_R(\sqrt{x}) \tag{6.28}$$

$$= 1 - \exp\left(-\frac{x}{2}\right) \quad \text{for } x \ge 0 \tag{6.29}$$

And the PDF of V is given by

$$p_V(x) = \frac{\mathrm{d}}{\mathrm{d}x} F_V(x) \tag{6.30}$$

$$= \frac{\mathrm{d}}{\mathrm{d}x} F_V(x) \left(1 - \exp\left(-\frac{x}{2}\right) \right) \tag{6.31}$$

$$= \frac{1}{2} \exp\left(-\frac{x}{2}\right) \tag{6.32}$$

Therefore,

$$F_V(x) = \begin{cases} 1 - \exp\left(-\frac{x}{2}\right) & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$
 (6.33)

$$p_V(x) = \begin{cases} \frac{1}{2} \exp\left(-\frac{x}{2}\right) & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$
 (6.34)

$$\therefore \alpha = \frac{1}{2} \tag{6.35}$$

6.3 Plot the CDF and PDF of

$$A = \sqrt{V} \tag{6.36}$$

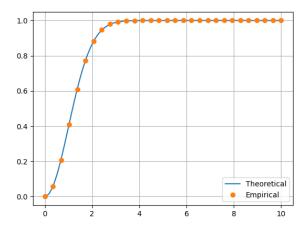


Fig. 6.3. CDF of A

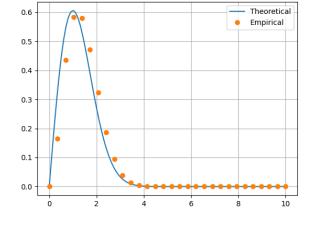


Fig. 6.3. PDF of *A*

Solution: Download the C source code to generate the data by executing the following commands

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/6.3.c

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/header.h

Compile and run the C program by executing the following

cc -lm 6.3.c ./a.out

Download the following Python code that plots Fig. 6.3

wget https://github.com/Ankit-Saha-2003/ AI1110/raw/main/Random-Numbers/ codes/6.3.py

Run the code by executing

python 6.3.py

The CDF of A for $x \ge 0$ is given by

$$F_A(x) = \Pr(A \le x) \tag{6.37}$$

$$= \Pr\left(\sqrt{V} \le x\right) \tag{6.38}$$

$$= \Pr\left(V \le x^2\right) \tag{6.39}$$

$$= F_V(x^2) (6.40)$$

$$=1-\exp\left(-\frac{x^2}{2}\right) \tag{6.41}$$

The PDF of A is given by

$$p_A(x) = \frac{\mathrm{d}}{\mathrm{d}x} F_A(x) \tag{6.42}$$

$$= \frac{\mathrm{d}}{\mathrm{d}x} \left(1 - \exp\left(-\frac{x^2}{2} \right) \right) \tag{6.43}$$

$$= x \exp\left(-\frac{x^2}{2}\right) \tag{6.44}$$

Therefore,

$$F_A(x) = \begin{cases} 1 - \exp\left(-\frac{x^2}{2}\right) & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$

$$p_A(x) = \begin{cases} x \exp\left(-\frac{x^2}{2}\right) & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$

$$(6.45)$$

$$p_A(x) = \begin{cases} x \exp\left(-\frac{x^2}{2}\right) & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$
 (6.46)