1

Random Numbers

Abstract—This manual provides a simple introduction to the generation of 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:

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/exrand.c

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/coeffs.h

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

https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/cdf_plot.py

1.3 Find a theoretical expression for $F_U(x)$. Solution:

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

$$F_U(x) = \int_{-\infty}^{x} f_U(x) \ dx \tag{1.3}$$

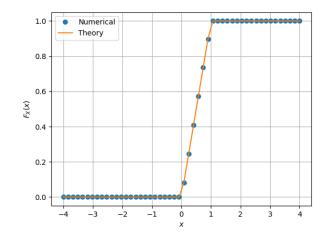


Fig. 1.2: The CDF of U

Hence, If $x \le 0$,

$$F_U(x) = \int_{-\infty}^x f_U(x) \ dx \tag{1.4}$$

$$F_U(x) = \int_{-\infty}^x 0 \, dx \tag{1.5}$$

$$=0 (1.6)$$

If 0 < x < 1,

$$F_U(x) = \int_0^x f_U(x) \ dx$$
 (1.7)

$$F_U(x) = \int_0^x 1 \, dx \tag{1.8}$$

$$= x \tag{1.9}$$

If $x \ge 1$,

$$F_U(x) = \int_1^x f_U(x) \ dx$$
 (1.10)

$$F_U(x) = \int_1^x 0 \, dx \tag{1.11}$$

$$=0 (1.12)$$

Hence,

$$F_U(x) = \begin{cases} x, & x \in (0,1) \\ 0, & \text{otherwise} \end{cases}$$
 (1.13)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.14)

and its variance as

codes/coeffs.h

$$var[U] = E[U - E[U]]^2$$
 (1.15)

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

Solution:

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/mean_variance.c wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/

Mean obtained =
$$0.500007$$
 (1.16)

Variance obtained =
$$0.083301$$
 (1.17)

1.5 Verify your result theoretically given that

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

Solution:

$$E[U^{k}] = 0 + \int_{0}^{1} x^{k} dF_{U}(x) + 0 \qquad (1.19)$$

$$=\frac{1}{k+1}$$
 (1.20)

$$\implies E[U] = \frac{1}{2} \tag{1.21}$$

$$= 0.50$$
 (1.22)

$$E\left[U^2\right] = \frac{1}{3} \tag{1.23}$$

Hence,

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

$$=\frac{1}{3} - \left(\frac{1}{2}\right)^2 \tag{1.25}$$

$$=\frac{1}{12}$$
 (1.26)

$$= 0.0833$$
 (1.27)

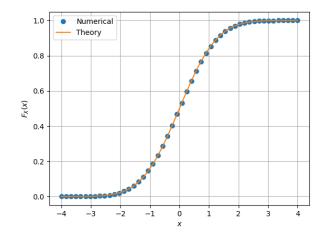


Fig. 2.2: The CDF of X

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:**

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/exrand.c

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/coeffs.h

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 code

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/cdf_plot.py

The CDF of a random variable U has the following properties:

a) $F_U(x)$ is a non decreasing function of x where $-\infty < x < \infty$

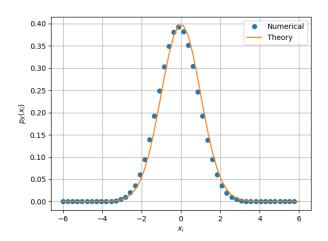


Fig. 2.3: The PDF of X

- b) $F_U(x)$ ranges from 0 to 1
- c) $F_U(x) = 0$ as $x \to -\infty$
- d) $F_U(x) = 1$ as $x \to \infty$
- 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

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/pdf plot.py

The PDF of a random variable X has the following properties:

- a) The probability density function is nonnegative for all the possible values.
- b) $\int_{-\infty}^{\infty} f(x) dx = 1$ c) f(x) = 0 as $x \to -\infty$
- d) f(x) = 0 as $x \to \infty$
- 2.4 Find the mean and variance of X by writing a C program.

Solution:

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/mean variance.c

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/coeffs.h

Mean obtained =
$$0.000326$$
 (2.3)

Variance obtained =
$$1.000907$$
 (2.4)

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, (2.5)$$

repeat the above exercise theoretically.

Solution: By definition,

$$p_X(x) dx = dF_U(x) (2.6)$$

Hence,

$$E[U] = \int_{-\infty}^{\infty} x p_X(x) \ dx \tag{2.7}$$

$$= \int_{-\infty}^{\infty} x \frac{1}{\sqrt{2\pi}} exp\left(-\frac{x^2}{2}\right) dx \qquad (2.8)$$

$$=0 (2.9)$$

Also,

variance =
$$E[U^2] - (E[U])^2$$
 (2.10)

$$E\left[U^{2}\right] = \int_{-\infty}^{\infty} x^{2} p_{X}(x) dx \qquad (2.11)$$

$$= \int_{-\infty}^{\infty} x^2 \frac{1}{\sqrt{2\pi}} exp\left(-\frac{x^2}{2}\right) dx$$
(2.12)

$$= 1 \tag{2.13}$$

$$\Rightarrow$$
 variance = 1 (2.14)

Also,

$$F_U(x) = \int_{-\infty}^x p_X(x) \ dx \tag{2.15}$$

$$= \int_{0.0}^{x} e^{-\frac{x^2}{2}} dx \tag{2.16}$$

3 From Uniform to Other

3.1 Generate samples of

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

and plot its CDF.

Solution:

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/exrand.c

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/coeffs.h

The CDF of *X* is plotted in Fig. 3.2 using the code below

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/cdf plot.py

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

Solution: Let, V = g(U)

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

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

Now,

$$F_V(x) = P(g(U) \le x) \tag{3.4}$$

$$= P(X < g^{-1}(V))$$
 (3.5)

$$= F_U(g^{-1}(V))$$
 (3.6)

$$= F_U \left(1 - e^{-\frac{V}{2}} \right) \tag{3.7}$$

$$F_{U}\left(1 - e^{-\frac{V}{2}}\right) = \begin{cases} 1 - e^{-\frac{V}{2}}, & V \in (0, \infty) \\ 0, & \text{otherwise} \end{cases}$$
(3.8)

4 Triangular Distribution

4.1 Generate

$$T = U_1 + U_2 (4.1)$$

Solution:

wget https://github.com/tejalkul/AI1110-Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/exrand.c

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/coeffs.h

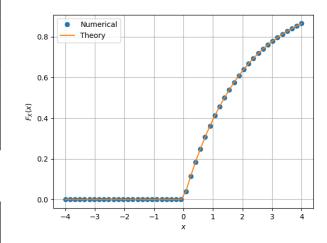


Fig. 3.2: CDF of V

- 4.2 Find the CDF of T.
- 4.3 Find the PDF of *T*.
- 4.4 Find the theoretical expressions for the PDF and CDF of *T*.

Solution:

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

$$\implies f_T(T) = \int_{-\infty}^{\infty} f_{U_1}(T - U_2) f_{U_2}(U_2) \ dU_2$$
(4.3)

since U_1 and U_2 are independent Also,

$$f_{U_1}(U_1) = f_{U_2}(U_2) = \begin{cases} 1, & x \in (0,1) \\ 0, & \text{otherwise} \end{cases}$$
 (4.4)

Hence (4.3) becomes,

$$f_T(T) = \int_0^1 f_{U_1}(T - U_2) \ dU_2$$
 (4.5)

If $0 \le T \le 1$,

$$f_T(T) = \int_0^T dU_2 \tag{4.6}$$

$$=T\tag{4.7}$$

If $1 < T \le 2$,

$$f_T(T) = \int_{T-1}^1 dU_2 \tag{4.8}$$

$$=2-T\tag{4.9}$$

$$f_T(T) = \begin{cases} T, & 0 \le T \le 1\\ 2 - T, & 1 < T \le 2\\ 0, & \text{otherwise} \end{cases}$$
 (4.10)

Now,

$$F_T(T) = \int_{-\infty}^T f_T(T) dT \qquad (4.11)$$

Hence, if T < 0,

$$F_T(T) = \int_{-\infty}^T f_T(T) dT \qquad (4.12)$$

$$=\int_{-\infty}^{T}0\,dT\tag{4.13}$$

$$=0 (4.14)$$

If $0 \le T \le 1$,

$$F_T(T) = \int_0^T f_T(T) dT$$
 (4.15)

$$= \int_0^T T \, dT \tag{4.16}$$

$$=\frac{T^2}{2}\tag{4.17}$$

If $1 < T \le 2$,

$$F_T(T) = \int_0^1 f_T(T) dT + \int_1^T f_T(T) dT$$
(4.18)

$$= \int_0^1 T \, dT + \int_1^T (2 - T) \, dT \quad (4.19)$$

$$=\frac{1}{2}+\frac{4T-T^2-3}{2}\tag{4.20}$$

$$=1-\frac{(2-T)^2}{2}\tag{4.21}$$

If T > 2,

$$F_T(T) = \int_2^T f_T(T) dT$$
 (4.22)

$$= \int_{2}^{T} 0 \, dT \tag{4.23}$$

$$=0 (4.24)$$

$$F_T(T) = \begin{cases} \frac{T^2}{2}, & 0 \le T \le 1\\ 1 - \frac{(2-T)^2}{2}, & 1 < T \le 2\\ 0, & \text{otherwise} \end{cases}$$
 (4.25)

4.5 Verify your results through a plot.

Solution: The CDF of T is plotted in Fig. 4.5

below

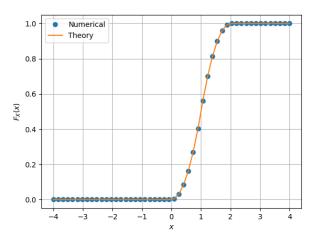


Fig. 4.5: CDF of T

The PDF of T is plotted in Fig. 4.5 below

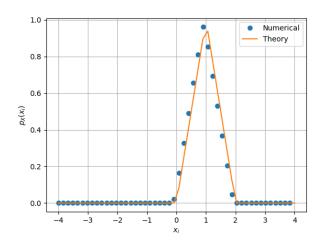


Fig. 4.5: PDF of T

Using codes

wget https://github.com/tejalkul/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/cdf_plot.py wget https://github.com/tejalkul/AI1110—

Assignments/blob/main/AI1110— Assignments/blob/main/AI1110%20 Random%20Variables%20Assignment/ codes/pdf_plot.py

5 Maximul Likelihood

5.1 Generate

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

where A = 5 dB, $X_1\{1, -1\}$, is Bernoulli and where $N \sim 01$.

- 5.2 Plot *Y*.
- 5.3 Guess how to estimate *X* from *Y*.
- 5.4 Find

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

and

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

- 5.5 Find P_e .
- 5.6 Verify by plotting the theoretical P_e .

6 Gaussian to Other

6.1 Let $X_1 \sim 01$ and $X_2 \sim 01$. Plot the CDF and PDF of

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

6.2 If

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

find α .

6.3 Plot the CDF and PDf of

$$A = \sqrt{V} \tag{6.3}$$

7 CONDITIONAL PROBABILITY

7.1

7.2 Plot

$$P_e = \Pr(\hat{X} = -1|X = 1) \tag{7.1}$$

for

$$Y = AX + N, (7.2)$$

where *A* is Raleigh with $E[A^2] = \gamma, N \sim 01, X \in (-1, 1)$ for $0 \le \gamma \le 10$ dB.

- 7.3 Assuming that N is a constant, find an expression for P_e . Call this $P_e(N)$
- 7.4 For a function g,

$$E[g(X)] = \int_{-\infty}^{\infty} g(x)p_X(x) dx \qquad (7.3)$$

Find $P_e = E[P_e(N)]$.

7.5 Plot P_e in problems 7.2 and 7.4 on the same graph w.r.t γ . Comment.

8 Two Dimensions

Let

$$\mathbf{y} = A\mathbf{x} + \mathbf{n},\tag{8.1}$$

$$x \in (\mathbf{s}_0, \mathbf{s}_1), \mathbf{s}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{s}_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$
 (8.2)

$$\mathbf{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix}, n_1, n_2 \sim 01. \tag{8.3}$$

8.1 Plot

$$\mathbf{y}|\mathbf{s}_0 \text{ and } \mathbf{y}|\mathbf{s}_1$$
 (8.4)

on the same graph using a scatter plot.

- 8.2 For the above problem, find a decision rule for detecting the symbols s_0 and s_1 .
- 8.3 Plot

$$P_e = \Pr(\hat{\mathbf{x}} = \mathbf{s}_1 | \mathbf{x} = \mathbf{s}_0) \tag{8.5}$$

with respect to the SNR from 0 to 10 dB.

8.4 Obtain an expression for P_e . Verify this by comparing the theory and simulation plots on the same graph.