1

Random Numbers

CS21BTECH11044

CONTENTS

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: Download the following files and execute the C program.

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/exrand.c

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/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. ??

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/cdf plot uni.py

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

Solution: As we know

Probability density function of U:

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

$$F_u(x) = \int_{-\infty}^x \Pr(U = a) da$$
$$= \int_{-\infty}^0 (0) da + \int_0^x (1) da$$
$$= x$$

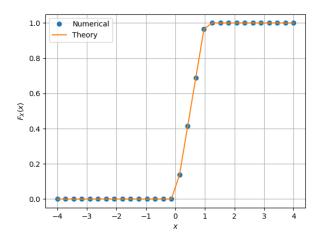


Fig. 1.2: The CDF of U

$$F_u(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \in (0, 1) \\ 1 & \text{for } x > 1 \end{cases}$$
 (1.3)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.4)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.5)

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

Solution:

The Mean and Variance of X is plotted in Fig. ??

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/q1_4.c wget https://github.com/Pranav-Varma-03/

wget https://github.com/Pranav – Varma – 03/ Ai 1110 – Assig/blob/main/Random %20 Numbers/codes/coeffs.h ubuntu@ubuntuXPS:~/Desktop/P&RV/A27_06/Q1\$ gcc q1_4.c -lm ubuntu@ubuntuXPS:~/Desktop/P&RV/A27_06/Q1\$./a.out nean = 0.500007 variance = 0.083301

Fig. 1.4: The Mean and Variance of X

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.6}$$

Solution:

1) k = 1 : E[U] is the mean

2) $k = 2 : E[U^2]$ where, $var[U] = E[U^2] - (E[U])^2$

i) Verifying Mean of sample of 10⁶ Uniform Random Variable

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

$$As, F_U(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \in (0, 1) \\ 1 & \text{for } x > 1 \end{cases}$$
 (1.7)

$$E[U] = 0 + \int_0^1 x d(x) + 0$$

$$E[U] = \frac{x^2}{2} \Big|_0^1$$

$$E[U] = 0.5$$

$$E[U] \approx 0.5 \tag{1.8}$$

ii) Verifying Variance of sample of 10⁶ Uniform Random Variable

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

$$As, F_{U}(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \in (0, 1) \\ 1 & \text{for } x > 1 \end{cases}$$

$$E\left[U^{2}\right] = 0 + \int_{0}^{1} x^{2} d(x) + 0$$

$$E\left[U^{2}\right] = \frac{x^{3}}{3} \Big|_{0}^{1}$$

$$E\left[U^{2}\right] = \frac{1}{3} = 0.33$$

Thus,

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

 $Var[U] = (\frac{1}{3}) - (\frac{1}{2})^2$
 $Var[U] = 0.0833$
 $Var[U] \approx 0.0833$ (1.9)

2 Central Limit Theorem

2.1 Generate 10^6 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/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/q1_4.c wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/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 required python file can be downloaded using

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/q2 4.c

The CDF of *X* is plotted in Fig. **??** Properties of CDF:

- cdf is non decreasing fucntion
- $\lim_{x\to -\infty} f(x) = 0$
- $\lim_{x\to +\infty} f(x) = 1$

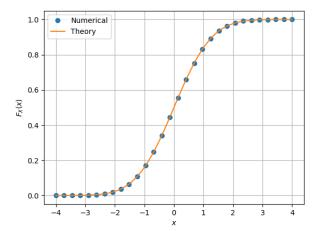


Fig. 2.2: The CDF of X

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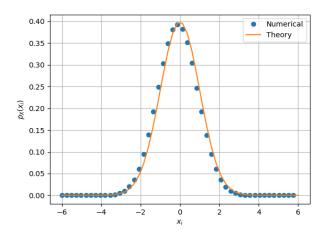


Fig. 2.3: The PDF of X

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. ?? using the code below

wget https://github.com/Pranav-Varma-03/ Ai1110-Assig/blob/main/Random%20 Numbers/codes/pdf plot gau.py

Properties of PDF:

- It is symmetric about y-axis
- Area under curve is 1 i.e., $\int_{-\infty}^{\infty} P_X(x) = 1$

2.4 Find the mean and variance of *X* by writing a C program.

Solution: Execute the C program given below:

https://github.com/Pranav-Varma-03/Ai1110 -Assig/blob/main/Random%20Numbers/ codes/q2_4.c

??

ubuntu@ubuntuXPS:~/Desktop/P&RV/A27_06/Q1\$ gcc q2_4.c -lm ubuntu@ubuntuXPS:~/Desktop/P&RV/A27_06/Q1\$./a.out mean = 0.000326 variance = 1.000907

Fig. 2.4: The Mean and Variance of X

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:**

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

- a) k = 1 : E[U] is the mean
- b) $k = 2 : E[U^2]$

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

i) Verifying Mean by using (??):

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

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

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

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

$$E[U] = 0$$

$$E[U] \approx 0 \tag{2.4}$$

ii) Verifying Variance by using (??):

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

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

$$E\left[U^{2}\right] = 2 \int_{0}^{\infty} x^{2} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^{2}}{2}\right) dx$$

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

$$E\left[U^{2}\right] = \sqrt{\frac{2}{\pi}} \left[0 + \sqrt{\frac{\pi}{2}}\right]$$

$$E\left[U^{2}\right] = 1$$

$$\therefore Var[U] = E\left[U^2\right] - (E\left[U\right])^2$$

$$= 1 - 0$$

$$= 1$$

$$Var[U] \approx 1 \qquad (2.5)$$

3 From Uniform to Other

3.1 Generate samples of

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

and plot its CDF.

Solution: Execute the following python code

The output CDF is plotted in Figure (??).

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

Solution:

Need to find $F_{\nu}(x)i.e.$, $\Pr(V \le x)$:

$$= \Pr(V \le x)$$

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

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

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

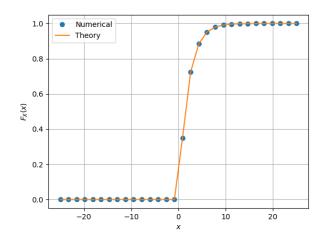


Fig. 3.1: The CDF of X

As we know,
$$Pr(U \le k) = k$$
:

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

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

$$As, F_V(x) = \begin{cases} 1 - exp\left(\frac{-x}{2}\right) & \text{for } x > 0\\ 0 & \text{for } x < 0 \end{cases}$$
(3.2)