#### 1

# Random Numbers

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#### 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/Pradeep8802/

Random\_numbers/blob/main/1.1/exrand.c wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/1.1/coeffs.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:** The following code plots Fig. 1.2

wget https://github.com/Pradeep8802/ Random\_numbers/tree/main/1.2/cdf\_U. py

Run the code by executing the below command

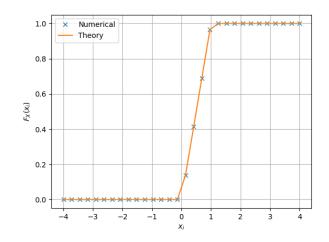


Fig. 1.2. The CDF of U

1.3 Find a theoretical expression for  $F_U(x)$ **Solution:** Here U is a uniform random variable in [0,1]. The PDF of U is given as follows

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

The CDF of U is

$$F_U(x) = \Pr\left(U \le x\right) \tag{1.3}$$

Since,

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

$$\therefore F_U(x) = \int_{-\infty}^x p_U(x) \, \mathrm{d}x$$

If x < 0,

$$\int_{-\infty}^{x} p_{U}(x) \, dx = \int_{-\infty}^{x} 0 \, dx \qquad (1.5)$$

$$=0 (1.6)$$

(1.9)

$$F_U(x) = 0 ag{1.7}$$

If  $0 \le x \le 1$ ,

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

$$= x \tag{1.10}$$

$$F_U(x) = x \tag{1.11}$$

If x > 1,

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

$$= \int_{-\infty}^{0} 0 dx + \int_{0}^{1} 1 dx + \int_{1}^{x} 0 dx \quad (1.12)$$

$$\int_{-\infty}^{x} p_U(x) \, \mathrm{d}x = 0 + 1 + 0 \tag{1.13}$$

$$= 1 \tag{1.14}$$

$$F_U(x) = 1 (1.15)$$

Hence the CDF of U is

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \le x \le 1 \\ 1 & x > 1 \end{cases}$$
 (1.16)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.17)

and its variance as

$$Var[U] = E[(U - E[U])^{2}]$$
 (1.18)

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

**Solution:** Download the following codes and execute the C program

wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/1.4/

mean variance.c

wget https://github.com/Pradeep8802/

Random numbers/blob/main/1.1/coeffs.h

Run the code by executing the below command

cc mean\_variance.c -lm ./a.out

The output of the code is

$$\mu = 0.500007 \tag{1.19}$$

$$\sigma^2 = 0.083301 \tag{1.20}$$

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} \mathrm{d}F_{U}(x) \tag{1.21}$$

**Solution:** The mean of U is given by

$$E[U] = \int_{-\infty}^{\infty} x \, \mathrm{d}x \tag{1.22}$$

The CDF of U is

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \le x \le 1 \\ 1 & x > 1 \end{cases}$$
 (1.23)

On differentiating the above CDF, we have

$$dF_U(x) = \begin{cases} d(0) & x < 0 \\ d(x) & 0 \le x \le 1 \\ d(1) & x > 1 \end{cases}$$
 (1.24)

$$dF_U(x) = \begin{cases} 0 & x < 0 \\ dx & 0 \le x \le 1 \\ 0 & x > 1 \end{cases}$$
 (1.25)

$$E[U] = \int_{-\infty}^{\infty} x \, \mathrm{d}F_U(x) \tag{1.26}$$

$$E[U^k] = \int_{-\infty}^{\infty} x^k d(x)$$
 (1.27)

$$E[U] = \int_0^1 x d(x) = \frac{1}{2}$$
 (1.28)

$$E[U^2] = \int_0^1 x^2 dx = \frac{1}{3}$$
 (1.29)

$$Var(U) = E[U^2] - (E[U])^2 = \frac{1}{12}$$
 (1.30)

Hence,

$$E[U] = \frac{1}{3} \approx 0.5 \tag{1.31}$$

$$\mu = 0.5$$
 (1.32)

$$Var(U) = \frac{1}{12} \approx 0.083333 \tag{1.33}$$

$$\sigma = 0.083333 \tag{1.34}$$

$$\Delta\mu \approx 0.000007 \tag{1.35}$$

$$\Delta \sigma \approx 0.000032 \tag{1.36}$$

## 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 codes and execute the C program

wget https://github.com/Pradeep8802/ Random\_numbers/blob/main/2.1/2.1.c wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/2.1/coeffs .h

Run the code by executing the below command

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 code and execute the Python program

Run the code by executing the below command

Properties of CDF:

- (a) It is always continuous.
- (b) It always lies in [0,1]
- (c) Q function is defined as:

$$Q_X(x) = \Pr(x < X) \tag{2.2}$$

$$\therefore F_X(x) = 1 - Q_X(x) \qquad (2.3)$$

- (d) CDF ( $F_X(x)$ ) is non-decreasing function as it is the sum of non-negative probabilities.
- 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

$$p_X(x) = \frac{d}{dx} F_X(x) \tag{2.4}$$

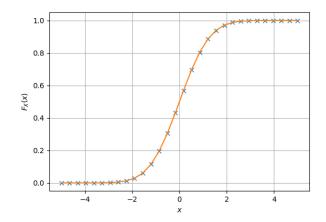


Fig. 2.5. The CDF of X

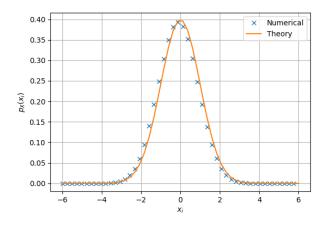


Fig. 2.5. The PDF of X

What properties does the PDF have? **Solution:** Download the following code and execute the Python program

wget https://github.com/Pradeep8802/ Random\_numbers/blob/main/2.3/ pdf\_gaussian.py

Run the code by executing the below command

python3 pdf gaussian.py

Properties of PDF:

- (a) PDF( $p_X(x)$ ) is bounded between [0,1]
- (b) PDF is a bell-shaped curve
- (c) PDF graph has symmetry around the mean of the distribution
- (d) If  $\mu$  is the mean of the distribution, here  $p_X(x)$  is symmetric with respect to 0,

since here  $\mu = 0$  ( $\mathcal{N}(0, 1)$ )

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

**Solution:** Download the following code and execute the C program

wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/2.4/coeffs .h

wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/2.4/

mean\_variance\_gaussian.c

Run the code by executing the below command

cc mean\_variance\_gaussian.c -lm ./a.out

$$\mu = 0.000326 \tag{2.5}$$

$$\sigma = 1.000906$$
 (2.6)

$$Mean = 0.000326 (2.7)$$

Variance = 
$$1.000906$$
 (2.8)

## 2.5 Given that

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

repeat the above exercise theoretically

**Solution:** As  $p_X(x)$  is given, Hence, the mean of X is

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

$$E[X] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} x \exp\left(-\frac{x^2}{2}\right) dx \quad (2.11)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{0} x \exp\left(-\frac{x^2}{2}\right) dx$$
$$+ \frac{1}{\sqrt{2\pi}} \int_{0}^{\infty} x \exp\left(-\frac{x^2}{2}\right) dx \tag{2.12}$$

Let 
$$g(x) = xe^{-\frac{-x^2}{2}}$$
  
Consider  $\int_{-\infty}^{0} g(x) dx$ 

Substituting z=-x, we get

$$\int_{-\infty}^{0} g(x) dx = \int_{0}^{0} g(-x) d(-x)$$
 (2.13)

$$= -\int_{\infty}^{0} g(-x)d(x)$$
 (2.14)

$$= \int_0^\infty g(-x)\mathrm{d}(x) \tag{2.15}$$

$$g(-x) = -xe^{-\frac{-(-x)^2}{2}}$$
 (2.16)

$$g(-x) = -(xe^{-\frac{-(x)^2}{2}})$$
 (2.17)

$$g(-x) = g(x) \tag{2.18}$$

$$\int_{-\infty}^{0} g(x) dx = -\int_{0}^{\infty} g(x) d(x)$$
 (2.19)  
=  $\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} x \exp\left(-\frac{x^{2}}{2}\right) dx$  (2.20)  
=  $\int_{0}^{\infty} g(x) d(x) - \int_{0}^{\infty} g(x) d(x) = 0$  (2.21)

 $\therefore$  Mean = 0

Let

$$h(x) = \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \tag{2.22}$$

$$h(-x) = \frac{(-x)^2}{\sqrt{2\pi}} \exp\left(-\frac{(-x)^2}{2}\right)$$
 (2.23)

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

(2.25)

$$\therefore h(-x) = h(x)$$

$$\therefore \int_{-\infty}^{0} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx = \int_{0}^{\infty} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx$$
(2.26)

$$E\left[X^{2}\right] = \int_{-\infty}^{\infty} \frac{x^{2}}{\sqrt{2\pi}} \exp\left(-\frac{x^{2}}{2}\right) dx \qquad (2.27)$$

$$=2\int_0^\infty \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \quad (2.28)$$

Using integration by parts,

$$E\left[X^{2}\right] = \sqrt{\frac{2}{\pi}} \int_{0}^{\infty} (x) \cdot (x \exp\left(-\frac{x^{2}}{2}\right)) dx$$
(2.29)

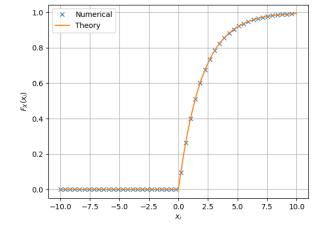
$$= \sqrt{\frac{2}{\pi}} \left( x \int x \exp\left(-\frac{x^2}{2}\right) dx \right) \Big|_0^{\infty}$$
$$- \sqrt{\frac{2}{\pi}} \int_0^{\infty} 1 \cdot \left( \int x \exp\left(-\frac{x^2}{2}\right) dx \right) (2.30)$$

Substituting 
$$z = \frac{x^2}{2} \implies dz = xdx$$
 (2.31)

$$\int x \exp\left(-\frac{x^2}{2}\right) dx = \int \exp(-z) dz \quad (2.32)$$

$$= -\exp(-z) \tag{2.33}$$

$$= -\exp\left(-\frac{x^2}{2}\right) \tag{2.34}$$



(2.34) Fig. 3.5. The CDF of V

Substituting  $x = \sqrt{2}t$ 

$$\int_0^\infty \exp(-t^2) \sqrt{2} dt = \sqrt{2} \int_0^\infty \exp(-t^2) dt$$
(2.35)

$$=\sqrt{\frac{\pi}{2}}\tag{2.36}$$

$$E[X^{2}] = 0 - \sqrt{\frac{2}{\pi}} \left( -\sqrt{\frac{\pi}{2}} \right)$$
 (2.37)

$$= 1 \tag{2.38}$$

:. 
$$Var[X] = E[X^2] - (E[X])^2$$
(2.39)

$$= 1 - 0$$
 (2.40)

$$= 1 \tag{2.41}$$

$$Var[X] = 1 \tag{2.42}$$

## 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 code and execute the C program

wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/3.1/3.1.c wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/3.1/coeffs.h

Run the code by executing the below command

Download the following code and execute the Python program

wget https://github.com/Pradeep8802/ Random\_numbers/blob/main/3.1/cdf\_3.1. py

Run the code by executing the below command

3.6 Find a theoretical expression for  $F_V(x)$  Solution: As  $F_V(x)$  is the CDF of V,

$$F_V(x) = P(V \le x) \tag{3.2}$$

$$= P(-2ln(1-U) \le x) \tag{3.3}$$

$$= -2ln(1 - U) \le x \tag{3.4}$$

$$= ln(1 - U) \ge -x/2 \tag{3.5}$$

Since ln(x) and -x/2 are monotonic, we can write

$$1 - U \ge e^{-x/2} \implies 1 - e^{-x/2} \le U$$

Hence,

$$F_V(x) = P(-2\log(1-U) \le x) = P(1 - e^{\frac{-x}{2}} \ge U)$$
(3.6)

Since,

$$p_U(x) = \begin{cases} 1 & x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$
 (3.7)

$$P(U < x) = \int_0^x dx = x$$
 (3.8)

$$P(U \le 1 - e^{\frac{-x}{2}}) = 1 - e^{\frac{-x}{2}}, \forall x \ge 0$$

#### 4. Triangular Distribution

#### 4.1 Generate

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

**Solution:** Download the following code and execute the C program

wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/4.1/4.1.c wget https://github.com/Pradeep8802/

Random\_numbers/blob/main/4.1/coeffs .h

Run the code by executing the below command

#### 4.2 Find the CDF of T

**Solution:** Download the following code and execute the Python program

wget https://github.com/Pradeep8802/ Random\_numbers/blob/main/4.2/cdf\_4 .2\_num.py

Run the code by executing the below command

$$T = U1 + U2$$
 (4.2)

$$p_T(x) = p_{U_1 + U_2}(x) \tag{4.3}$$

Consider the convolution of the two distributions,

$$p_T(x) = p_{U_1 + U_2}(x) (4.4)$$

$$= p_{U_1}(x) * p_{U_2}(x) d\tau$$
 (4.5)

$$p_T(x) = \int_{-\infty}^{\infty} p_{U_1}(\tau) p_{U_2}(x - \tau) d\tau \qquad (4.6)$$

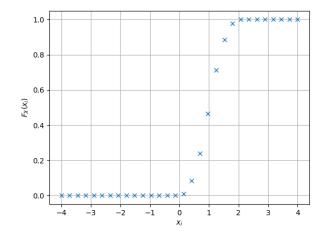


Fig. 4.6. The Numerical CDF of T

For 
$$\tau \in (-\infty, 0) \cup (1, \infty)$$
  

$$p_{U_1}(\tau) = 0 \qquad (4.7)$$

Hence,

$$p_T(x) = \int_{-\infty}^{\infty} p_{U_1}(\tau) p_{U_2}(x - \tau) d\tau \qquad (4.8)$$

$$= \int_0^1 p_{U_2}(x-\tau) d\tau$$
 (4.9)

$$p_T(x) = \begin{cases} 0 & x \le 0\\ \int_0^x 1d\tau & 0 < x < 1\\ \int_{x-1}^1 1d\tau & 1 \le x < 2\\ 0 & x > 2 \end{cases}$$
 (4.10)

Hence

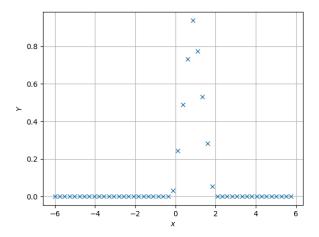
$$p_T(x) = \begin{cases} 0 & x \le 0 \\ x & 0 < x < 1 \\ 2 - x & 1 \le x < 2 \\ 0 & x > 2 \end{cases}$$
 (4.11)

Expression for CDF can be obtained by Integrating  $p_T(x)$  w.r.t. x

$$F_T(x) = \begin{cases} 0 & x \le 0\\ \frac{x^2}{2} & 0 < x < 1\\ -\frac{x^2}{2} + 2x - 1 & 1 \le x < 2\\ 1 & x > 2 \end{cases}$$
(4.12)

### 4.3 Find the PDF of T

**Solution:** Download the following code and execute the Python program





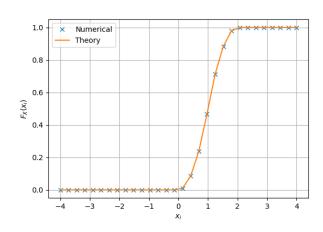


Fig. 4.6. The CDF of T

Run the code by executing the below command

The PDF of T is given by

$$p_T(t) = \frac{d(F_T(t))}{dt} \tag{4.13}$$

$$\therefore p_T(t) = \begin{cases} 0 & t < 0 \\ t & 0 \le t \le 1 \\ 2 - t & 1 < t < 2 \\ 0 & t \ge 2 \end{cases}$$
 (4.14)

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

### **Solution:**

$$P_T(t) = \begin{cases} 0 & t < 0 \\ t & 0 \le t \le 1 \\ 2 - t & 0 < t \le 2 \\ 0 & t > 2 \end{cases}$$
 (4.15)

$$F_T(t) = \begin{cases} 0 & t < 0 \\ \frac{t^2}{2} & 0 \le t \le 1 \\ 2t - \frac{t^2}{2} - 1 & 1 < t \le 2 \\ 1 & t > 2 \end{cases}$$
 (4.16)

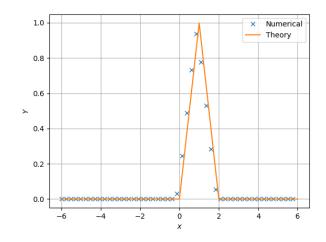


Fig. 4.6. The PDF of T

4.5 Verify your results through a plot **Solution:** Download the following code and execute the Python programs

Run the codes by executing the below commands for cdf and pdf respectively