1

Random Numbers

SADINENI ABHINAY - CS21BTECH11055

6

Contents

1	Uniform Random Numbers	1
2	Central Limit Theorem	2
3	From Uniform to Other	3
4	Triangular Distribution	4
5	Maximum Likelihood	5

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

Gaussian to Other

6

1.0 Numerical Theory 0.8 0.6 0.7 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.9 0.

Fig. 1.2: The CDF of U

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

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/1.1.c

Now compile and exceute

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/karna-rash/A11110
-Assignment/blob/master/codes/1.2.py
\$ python3 1.2.py

1.3 Find a theoretical expression for $F_U(x)$. **Solution:** Since it is uniform distribution then pdf is given by:

$$p_U(x) = \begin{cases} 1 & 0 \le x \le 1 \\ 0 & otherwise \end{cases}$$
 (1.2)

for 0 < x < 1:

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

$$= \int_0^x 1.dx \tag{1.4}$$

$$=x\tag{1.5}$$

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \le x \le 1 \\ 1 & 1 < x \end{cases}$$
 (1.6)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.7)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.8)

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

Solution: Download and excecute the code

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/1.3-1.4.c

Now compile and excecute

Experimental values:

$$Mean = 0.499772 (1.9)$$

Variance =
$$0.083368$$
 (1.10)

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.11}$$

Solution: Mean and variance are given by :

$$E[U] = \int_0^1 x.dx = 0.5 \tag{1.12}$$

$$Var[U] = E[U^2] - E[U]^2 = \int_0^1 x^2 dx - 0 \quad (1.13)$$

$$=\frac{1}{3}-\frac{1}{4}\tag{1.14}$$

$$=\frac{1}{12}$$
 (1.15)

By comparing these values obtained with results (1.9) and (1.10) we can conclude that simulation varies with theoritical analysis.

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

Soltion: Download

\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/2.1.c
\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/2.1.py

Now compile the c-code to generate the data

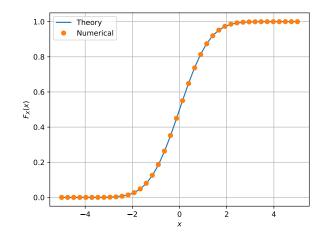


Fig. 2.2: The CDF of X

\$ gcc -o 2.1 2.1.c \$./2.1

Now excecute the python code which take the data from output of c code:

\$ python3 2.1.py

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 Properties:

- a) it is symmetric about (0, 0.5)
- b) its tends to one as we approach x = 6

\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/2.2.py
\$ python3 2.2.py

 $Q(x) = \int_{x}^{\infty} exp\left(\frac{-x^{2}}{2}\right) dx \qquad (2.2)$

$$erf(x) = \int_0^x exp\left(\frac{-x^2}{2}\right) dx \qquad (2.3)$$

$$F_X(x) = \int_{-\infty}^x exp\left(\frac{-x^2}{2}\right) dx \qquad (2.4)$$

$$F_X(x) = 1 - Q(x)$$
 (2.5)

$$Q(x) = \frac{1}{2} \left(1 - erf\left(\frac{x}{\sqrt{2}}\right) \right) \tag{2.6}$$

2.3 Load gau.dat in python and plot the empirical PDF of *X* using the samples in gau.dat. The

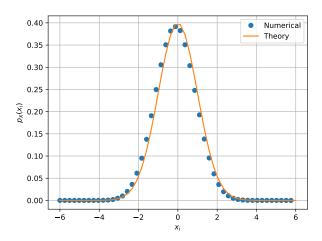


Fig. 2.3: The PDF of X

PDF of X is defined as

$$p_X(x) = \frac{d}{dx} F_X(x) \tag{2.7}$$

What properties does the PDF have?

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

Properties:

- a) it is symmetric about x = 0
- b) its tends to zero as we approach x = 6

\$ wget https://github.com/karna-rash/A11110
 -Assignment/blob/master/codes/2.3.py
\$ python3 2.3.py

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

Solution:

wget https://github.com/karna-rash/A11110-Assignment/blob/master/codes/gau_mean.

compile and execute:

Experimental values

$$Mean = -0.000354 \tag{2.8}$$

Variance =
$$1.001179$$
 (2.9)

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty,$$
(2.10)

repeat the above exercise theoretically.

Solution:

a) Mean:

$$E(X) = \int_{-\infty}^{\infty} x p_X(x) dx$$
 (2.11)

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

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

$$= 0$$
 (2.14)

b) variance:

$$E\left(X^{2}\right) = \int_{-\infty}^{\infty} x^{2} p_{X}(x) dx \qquad (2.15)$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} x^{2} exp\left(-\frac{x^{2}}{2}\right) dx \quad (2.16)$$

$$= \frac{1}{\sqrt{2\pi}} \left(x \int x exp\left(-\frac{x^{2}}{2}\right) dx\right) \qquad (2.17)$$

$$-\frac{1}{\sqrt{2\pi}} \int \int \left(x exp\left(-\frac{x^{2}}{2}\right)\right) dx . dx \qquad (2.18)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} exp\left(-\frac{x^{2}}{2}\right) dx \qquad (2.19)$$

$$= \frac{\sqrt{2\pi}}{\sqrt{2\pi}} \qquad (2.20)$$

$$= 1 \qquad (2.21)$$

By comparing these values obatined with results (2.8) and (2.9) we can conclude that simulation varies with theoritical analysis.

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 codes 3.1

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/3.1.c

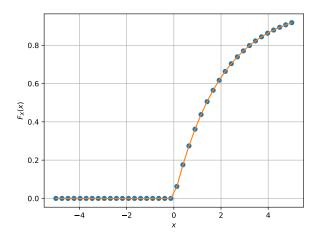


Fig. 3.1: The PDF of V

\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/3.1.py

Compile and execute:

3.2 Find a theoretical expression for $F_V(x)$. Soltion: Let V = G(U) then

$$F_V(x) = F_U(G^{-1}(x))$$
 (3.2)

$$= F_U(1 - e^{-\frac{x}{2}}) \tag{3.3}$$

(3.4)

$$F_V(x) = \begin{cases} G^{-1}(x) & 0 \le 1 - e^{-\frac{x}{2}} \le 1\\ 0 & 1 - e^{-\frac{x}{2}} < 0 \end{cases}$$
 (3.5)

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

4 Triangular Distribution

4.1 Generate

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

Solution: Download and excecute the code

\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/4.1.c
\$ wget https://github.com/karna-rash/A11110
-Assignment/blob/master/codes/functions.
h
\$ gcc 4.1.c -lm

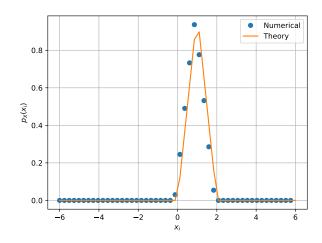


Fig. 4.2: The PDF verify of T

\$./a.out

4.2 Find the CDF of T.

Solution: Download and excecute the code .it gives the following plot Fig.4.3

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/4.2.py \$ python 4.2.py

$$F_T(t) = \begin{cases} 0 & t \le 0\\ \frac{t^2}{2} & 0 \le t < 1\\ \frac{(2-t)^2}{2} & 1 \le t \le 2\\ 1 & 2 \le t \end{cases}$$
(4.2)

4.3 Find the PDF of T.

Solution: Download and excecute the code. It gives the following plot Fig.4.2

\$ wget https://github.com/karna-rash/A11110
 -Assignment/blob/master/codes/4.3.py
\$ python 4.3.py

$$p_T(t) = \begin{cases} t & 0 \le t < 1\\ 2 - t & 1 \le t \le 2\\ 0 & otherwise \end{cases}$$
 (4.3)

Solution: Download and excecute the code. It gives the following plot Fig.4.3

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

Solution: Let us use convulsion of

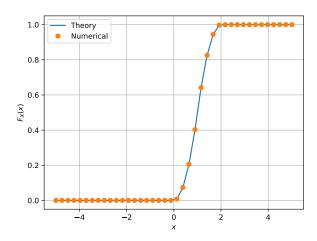


Fig. 4.3: The CDF Verify of T

 $p_{U_1}(x)$ and $p_{U_2}(x)$

$$p_T(t) = \int_{-\infty}^{\infty} p_{U_1}(T - \alpha) p_{U_2}(\alpha) d\alpha \qquad (4.4)$$

$$p_{T}(t) = \begin{cases} \int_{0}^{t} 1.d\alpha & 0 \le t < 1\\ \int_{t-1}^{1} 1.d\alpha & 1 \le t \le 2\\ 0 & otherwise \end{cases}$$
 (4.5)

$$= \begin{cases} t & 0 \le t < 1 \\ 2 - t & 1 \le t \le 2 \\ 0 & otherwise \end{cases}$$
 (4.6)

Now we can get CDF from PDF

$$F_T(t) = \Pr(T \le t) \tag{4.7}$$

$$= \int_{-\infty}^{t} p_T(x) dx \tag{4.8}$$

$$F_{T}(t) = \begin{cases} 0 & t \leq 0 \\ \int_{0}^{t} x.dx & 0 \leq t < 1 \\ \int_{1}^{t} 2 - x.dx + \frac{1}{2} & 1 \leq t \leq 2 \\ 1 & 2 \leq t \end{cases}$$
(4.9) 5.6 Find P_{e} assuming that X has equiprobable symbols. 5.7 Verify by plotting the theoretical P_{e} with respect to A from $0 \leq t \leq 10$.

$$= \begin{cases} 0 & t \le 0\\ \frac{t^2}{2} & 0 \le t < 1\\ \frac{(2-t)^2}{2} & 1 \le t \le 2\\ 1 & 2 \le t \end{cases}$$
 (4.10)

4.5 Verify your results through a plot.

theorical plots in 4.3 and 4.2

5 Maximum Likelihood

5.1 Generate equiprobable $X \in \{1, -1\}$. Solution: Download and excecute the code

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/5.1.c

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/functions.

\$ gcc 5.1.c -lm

\$./a.out

5.2 Generate

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

where A = 5 dB, and $N \sim \mathcal{N}(0, 1)$. Solution: Download and excecute the code

\$ wget https://github.com/karna-rash/A11110

-Assignment/blob/master/codes/5.2.c

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/functions.

\$ gcc 5.2.c -lm

\$./a.out

5.3 Plot Y using a scatter plot. Download and execute the code

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/5.3.py \$ python 5.3.py

- 5.4 Guess how to estimate X from Y.
- 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)

- spect to A from 0 to 10 dB.
- 5.8 Now, consider a threshold δ while estimating X from Y. Find the value of δ that maximizes the theoretical P_e .
- 5.9 Repeat the above exercise when

$$p_X(0) = p \tag{5.4}$$

Solution: After Compare the empirical and 5.10 Repeat the above exercise using the MAP criterion.

6 Gaussian 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: Download and excecute the code

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/5.1.c

\$ wget https://github.com/karna-rash/A11110 -Assignment/blob/master/codes/functions. h

\$ gcc 6.1.c -lm

\$./a.out

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: Since X1 and X2 are independent normal distribution:

For $x \ge 0$ Let $X1 = R\cos(\theta)$ and $X2 = R\sin(\theta)$ and And the jacobian matrix J:

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

$$= \begin{pmatrix} \cos \theta & -R \sin \theta \\ \sin \theta & R \cos \theta \end{pmatrix}$$
(6.3)

$$= \begin{pmatrix} \cos \theta & -R \sin \theta \\ \sin \theta & R \cos \theta \end{pmatrix} \tag{6.4}$$

Now let us find the

$$p_{R,\theta}(r,\theta) = p_{X1,X2}(x1,x2)|J|$$
 (6.5)

$$= p_{X1}(x1) p_{X2}(x2) \times r \qquad (6.6)$$

$$=\frac{r}{2\pi}e^{-\frac{x_1^2}{2}}e^{-\frac{x_2^2}{2}}\tag{6.7}$$

$$=\frac{r}{2\pi}e^{-\frac{r^2}{2}}\tag{6.8}$$

Remove θ and get in R:

$$P_R(r) = \int_0^{2\pi} p_{R,\theta}(r,\theta) . d\theta \qquad (6.9)$$

$$= \int_{0}^{2\pi} \frac{r}{2\pi} e^{-\frac{r^2}{2}} d.\theta \tag{6.10}$$

$$=2\pi.\frac{r}{2\pi}e^{-\frac{r^2}{2}} \tag{6.11}$$

$$= re^{-\frac{r^2}{2}} \tag{6.12}$$

Now transform into Y:

$$p_Y(y) = p_R(r) \cdot \frac{dr}{dy}$$
 (6.13)

$$= \sqrt{y}e^{-\frac{y}{2}} \cdot \frac{1}{2\sqrt{y}} \tag{6.14}$$

$$=\frac{e^{-\frac{y}{2}}}{2} \tag{6.15}$$

from pdf we can get cdf:

For $0 \le x \le 1$

$$F_V(x) = \int_0^x \frac{e^{-\frac{u}{2}}}{2} du \tag{6.16}$$

$$=1-e^{-\frac{x}{2}} \tag{6.17}$$

(6.18)

$$F_V(x) = \begin{cases} 1 - e^{-\frac{x}{2}} & 0 \le x \le 1\\ 0 & x < 0 \end{cases}$$
 (6.19)

So Thus confirmed that $\alpha = \frac{1}{2}$

6.3 Plot the CDF and PDf of

$$A = \sqrt{V} \tag{6.20}$$