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 following files and execute the C program.

wget https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex1/1.1/exrand.c we get https://github.com/mohilmukundareddy /Assignment1/blob/main/ex1/1.1/coeffs.h

Use the below command in the terminal to run the code

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 graph 1.2 is obtained by running the below code

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex1/1.2/main.py

Run the following command in the terminal to run the code.

python3 main.py

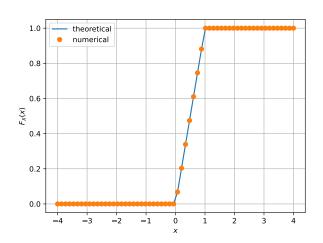


Fig. 1.2: The CDF of U

1.3 Find a theoretical expression for $F_U(x)$. **Solution:** Given U is uniform random variable so

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

$$F_X(x) = \int_{-\infty}^x f_X(x) dx$$
$$= 0 + \int_0^x 1 dx$$
$$= \begin{cases} 1 & x > 1 \\ x & 0 \le x \le 1 \\ 0 & x < 0 \end{cases}$$

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.2)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.3)

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

Solution:

wget https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex1/1.4/exrand.c weget https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex1/1.4/coeffs.h

Use below command to run file,

Fig. 1.4: Caption

1.5 Verify your result theoretically given that

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

$$E[U] = \int_{-\infty}^{\infty} x^k f_X(x) dx$$
$$= \int_0^1 x \times 1 dx$$
$$= \left[\frac{x^2}{2}\right]_0^1 = \frac{1}{2}$$

if k=2

$$E[U^{2}] = \int_{0}^{1} x^{2} \times 1 dx$$
$$= \left[\frac{x^{3}}{3}\right]_{0}^{1} = \frac{1}{3}$$

variance =
$$E[u - E[u]]^2$$

= $E[U^2] - E^2[U]$
= $\frac{1}{3} - \frac{1}{4} = 0.0833$

- 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/mohilmukundareddy/ Assignment1/blob/main/ex2/2.1/exrand.c wget hhttps://github.com/mohilmukundareddy/ Assignment1/blob/main/ex2/2.1/coeffs.h

Running the above codes generates uni.dat and gau.dat file. Use the 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: The CDF of *X* is plotted in plot,Properties of the CDF:

- $F_X(x) = P(X \le x)$
- $Q_X(x) = P(X > x)$
- $F_X(x) = 1 Q_X(x)$ This can be used to calculate F (x).

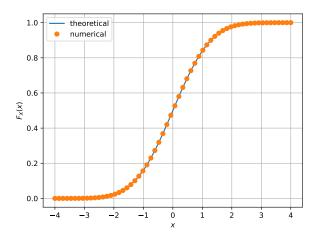


Fig. 2.2: The CDF of X

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?

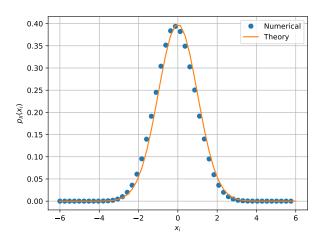


Fig. 2.3: The PDF of X

Solution: The PDF of X is plotted using the code below

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex2/2.3/main.py

Use the below command to run the code:

python3 main.py

Properties of PDF:

- PDF is symmetric about $x \approx 0$
- graph is similar to bell shaped
- mean of graph is situated at the symmetrical point
- 2.4 Find the mean and variance of *X* by writing a C program.

Solution: Running the below code gives Mean = 0.000326 Variance= 1.000906

wget https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex2/2.4/exrand.c

Command used:

gcc exrand.c -lm ./a.out

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, -\infty < x < \infty$$

by property of probability

$$\int_{-\infty}^{\infty} p_X(x) dx = 1$$

$$F_X(x) = \int_{-\infty}^x p_X(x)dx$$
$$= \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$E[X] = \int_{-\infty}^{\infty} x p(x) dx$$
$$= \int_{-\infty}^{\infty} \frac{x}{\sqrt{2\pi}} e^{\frac{-x^2}{2}} dx$$

 $\frac{x}{\sqrt{2\pi}}e^{\frac{-x^2}{2}}$ is an odd function so integral is zero i.e E[X] = 0.

$$E[X^{2}] = \int_{-\infty}^{\infty} x^{2} p(x) dx$$

$$= x \int_{-\infty}^{\infty} x p(x) dx - \int_{-\infty}^{\infty} \left(\int x p(x) dx \right) dx$$

$$= \left[-x \frac{1}{\sqrt{2\pi}} e^{\frac{-x^{2}}{2}} \right]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} -\frac{1}{\sqrt{2\pi}} e^{\frac{-x^{2}}{2}} dx$$

$$= 0 - (-1)$$

we know that

$$\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} = \sqrt{2\pi}$$

by series expansion $\frac{x}{e^{\frac{x^2}{2}}} = \frac{x}{1 + \frac{x^2}{2} + \frac{x^4}{8} + \dots}$ putting $x = \infty$, we get $\frac{1}{\infty} = 0$ Similarly when $x = -\infty$ we get 0 $var(x) = E[x^2] - E[x] = 1 - 0 = 1$

3 From Uniform to Other

3.1 Generate samples of

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

and plot its CDF.

Solution:

Running the below code generates samples of V from file uni.dat(U).

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex3/main.py

Use the below command in the terminal to run the code:

python3 main.py

Now these samples are used to plot by running the below code

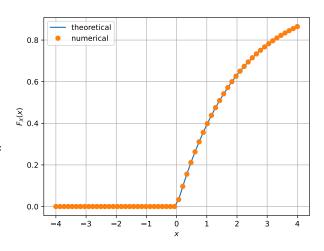


Fig. 3.5: CDF for (3)

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex3/cdf.py

Use the below command to run the code:

python3 cdf.py

3.2 Theoretical expression for $F_V(x)$

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

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

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

$$= F_{U}\{1 - e^{(-\frac{x}{2})}\}$$

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

4 Triangular Distribution

4.1 Generate

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

we get the code

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex4/1/main. py

run the command

python3 main.py

4.2 Find the CDF of T.

$$F_T(t) = P\{T \le t\}$$

= $P\{U1 + U2 \le t\}$

let us take two cases if $0 \le t \le 1$ and 5 Verify your results through a plot Take the code for $1 < t \le 2$

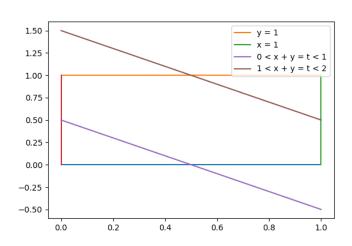


Fig. 4.5: def plot

The above graph is produced by

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex4/2/find. py

Run the code in terminal

python3 find.py

from the figures it is evident that $P(U1 + U2 < t, 0 \le t < 1) = \frac{t^2}{2}$ $P(U1 + U2 < t, 1 \le t \le 2) = 1 - \frac{(2-t)^2}{2}$

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

4.3 Find the PDF of T Solution:

$$P_T(t) = \frac{d(F_T(t))}{dt}$$

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

4.4 Find the theoretical expressions for the PDF and CDF of *T*. **Solution:** As we have already showed

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex4/5/ main1.py

Run in terminal

python3 main1.py

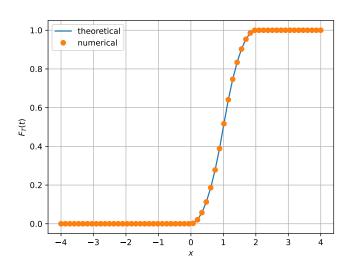


Fig. 4.5: t-cdf

Take the code for pdf

https://github.com/mohilmukundareddy/ Assignment1/blob/main/ex4/5/ main2.py

Run in terminal

python3 main2.py

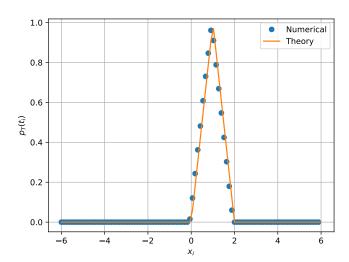


Fig. 4.5: t-pdf

5 Maximul Likelihood

5.1 Generate

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

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

- 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

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

Solution: The sum of squares of n independent standard random normal variables is χ^2 distribution with n degrees of freedom.

$$P_{\chi^2}(x|n) = \frac{x^{\frac{n}{2}-1}e^{\frac{-x}{2}}}{2^{\frac{n}{2}}\Gamma(\frac{n}{2})}, \forall x \ge 0$$

Here k=2,

$$P_{\chi^2}(x|2) = P_V(v) = \frac{e^{\frac{-v}{2}}}{2}$$

For the cumulative distribution

$$F_V(v) = \int_0^v \frac{e^{\frac{-v}{2}}}{2} dv$$
$$= 1 - e^{\frac{-v}{2}}$$

To generate data for V, run the following code,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/1/ main.py

Run the below command in terminal,

python3 main.py

The PDF plot of the $\chi^2(2)$ can be obtained by running the code below,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/1/ main1.py

Use the following command in the terminal to run the code

python3 main1.py

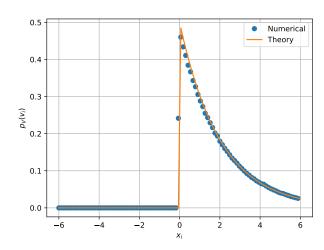


Fig. 6.5: PDF plot

The CDF plot of the $\chi^2(2)$ can be obtained by running the code below,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/1/ main2.py

Use the following command in the terminal to run the code

python3 main2.py

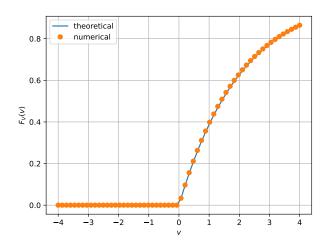


Fig. 6.5: CDF plot

6.2 If

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

find α .

Solution: From 6.1 we know that $\alpha = 0.5$

6.3 Plot the CDF and PDf of

$$A = \sqrt{V}$$

Solution:

To generate data for A, run the following code,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/3/ main.py

Run the below command in terminal,

python3 main.py

The PDF plot of A can be obtained by running the code below,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/3/ main1.py

Use the following command in the terminal to run the code

python3 main1.py

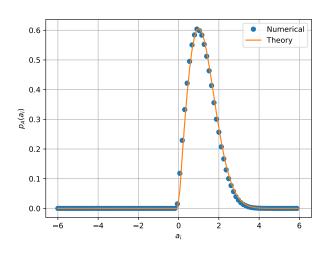


Fig. 6.5: PDF

The CDF plot of the A can be obtained by running the code below,

https://github.com/ mohilmukundareddy/ Assignment1/blob/main/ex6/3/ main2.py

Use the following command in the terminal to run the code

python3 main2.py

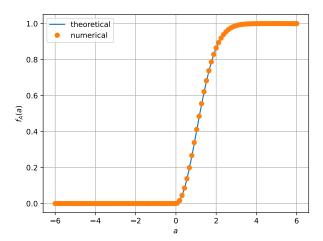


Fig. 6.5: CDF

7 CONDITIONAL PROBABILITY

7.1 7.2 Plot

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

for

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

where A is Raleigh with $E\left[A^2\right] = \gamma, N \sim 01, X \in (-1, 1)$ 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,

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

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

7.5 Plot P_e in problems 7.7.5 and 7.7.5 on the same graph w.r.t γ . Comment.

8 Two Dimensions

Let

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

where

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

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

8.1 Plot

$$\mathbf{y}|\mathbf{s}_0$$
 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 \mathbf{s}_0 and \mathbf{s}_1 .
- 8.3 Plot

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

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.