1

Random Numbers

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Uniform 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/Pranavb060504/
Random_numbers/blob/main/1.1/exrand.c
wget https://github.com/Pranavb060504/
Random_numbers/blob/main/1.1/coeffs.h

Use the below command in the terminal to run the code

cc exrand.c -lm ./a.out

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/Pranavb060504/ Random_numbers/blob/main/1.2/uni_cdf. py

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

python3 uni_cdf.py

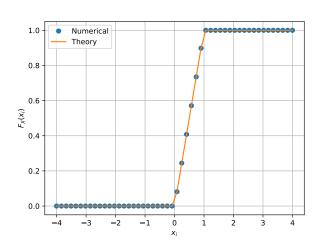


Fig. 1.2: The CDF of U

1.3 Find a theoretical expression for $F_U(x)$. **Solution:** Since U is an uniform random variable distribution, $P_U(x_i) = P_U(x_j) = k, \forall i, j$ CDF of $P_U(x) = F_U(x)$

$$= \int P_U(x)dx \tag{1.2}$$

$$= \int k dx \tag{1.3}$$

$$\operatorname{wkt} \int_0^1 k dx = 1 \tag{1.4}$$

$$\therefore k = 1 \tag{1.5}$$

$$\therefore F_U(x) = x \tag{1.6}$$

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:

wget https://github.com/Pranavb060504/ Random numbers/blob/main/1.4/ mean var.c

Use below command to run file,

running the code gives us Mean =0.500137, Variance =0.083251

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.9}$$

$$dF_U(x) = dx (1.10)$$

$$\therefore E[U^k] = \int_{-\infty}^{\infty} x^k dx \tag{1.11}$$

$$E[U] = \int_0^1 x dx = \frac{1}{2}$$
 (1.12)

$$E[U^2] = \int_0^1 x^2 dx = \frac{1}{3}$$
 (1.13)

$$\therefore P_X(x) = 0, \forall x \in (1, \infty) \cap (-\infty, 0)$$
 (1.14)

$$Var(X) = E[U^2] - (E[U])^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$
 (1.15)

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, ..., 12are a set of independent uniform random variables between 0 and 1 and save in a file called gau.dat

Solution:

wget https://github.com/Pranavb060504/ Random numbers/blob/main/1.1/exrand.c wget https://github.com/Pranavb060504/ Random numbers/blob/main/1.1/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 2.2,Properties of the CDF:

•
$$\Phi(x) = P(Z \le x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x} \exp\left\{-\frac{u^2}{2}\right\} du$$

• $\lim_{x \to \infty} \Phi(x) = 1$, $\lim_{x \to -\infty} \Phi(x) = 0$

•
$$\lim_{x \to \infty} \Phi(x) = 1$$
, $\lim_{x \to -\infty} \overline{\Phi}(x) = 0$

•
$$\Phi(0) = \frac{1}{2}$$

•
$$\Phi(-x) = 1 - \Phi(x)$$

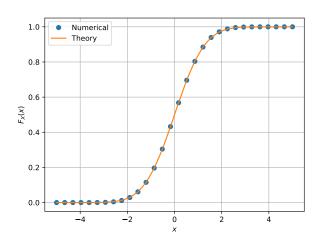


Fig. 2.2: The CDF of X

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?

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

https://github.com/Pranavb060504/ Random numbers/blob/main/2.3/pdf.py

Use the below command to run the code:

python3 pdf.py

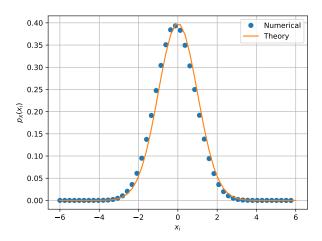


Fig. 2.3: The PDF of X

Properties of PDF:

- PDF is symmetric about x = 0
- graph is bell shaped
- mean of graph is situated at the apex point of the bell
- 2.4 Find the mean and variance of *X* by writing a C program.

Solution: Running the below code gives Mean = -0.000417 Variance= 0.999902

wget https://github.com/Pranavb060504/ Random_numbers/blob/main/2.4/ mean_var(gau).c

Command used:

cc mean_var(gau).c -lm
./a.out

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.3)$$

repeat the above exercise theoretically.

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

$$E[x] = \int_{-\infty}^{\infty} x p_X(x) dx \tag{2.4}$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} x e^{-\frac{-x^2}{2}}$$
 (2.5)

$$\therefore xe^{-\frac{-x^2}{2}}$$
 is a odd function, (2.6)

E[x] = 0

$$E[x^{2}] = \int_{-\infty}^{\infty} x^{2} p_{X}(x) dx$$
 (2.7)

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} x(xe^{-\frac{-x^2}{2}}) dx$$
 (2.8)

Using integration by parts:

$$= x \int xe^{-\frac{-x^2}{2}} dx - \int \frac{d(x)}{dx} \int xe^{-\frac{-x^2}{2}} dx \qquad (2.9)$$

$$I = \int xe^{-\frac{-x^2}{2}} \tag{2.10}$$

$$Let \frac{x^2}{2} = t \tag{2.11}$$

$$\implies xdx = dt$$
 (2.12)

$$\Longrightarrow = \int e^{-t}dt = -e^{-t} + c \tag{2.13}$$

$$\therefore \int xe^{-\frac{-x^2}{2}} = -e^{-\frac{-x^2}{2}} + c \tag{2.14}$$

Using (2.14) in (2.9)

$$= -xe^{-\frac{-x^2}{2}} + \int e^{-\frac{-x^2}{2}} dx \tag{2.15}$$

Also,
$$\int_{-\infty}^{\infty} e^{-\frac{-x^2}{2}} dx = \sqrt{2\pi}$$
 (2.16)

$$\therefore$$
 substituting limits we get, $E[x^2] = 1$ (2.17)

$$Var(X) = E[x^2] - (E[x])^2 = 1 - 0$$
 (2.18)

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/Pranavb060504/ Random numbers/blob/main/3.1/V.py

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

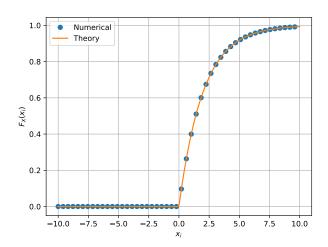


Fig. 3.1: CDF for (3)

python3 V.py

Now these samples are used to plot (3.1) by running the below code,

https://github.com/Pranavb060504/ Random_numbers/blob/main/3.1/V_cdf. py

Use the below command to run the code:

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

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

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

$$= P(1 - e^{\frac{-x}{2}} \ge U) \tag{3.4}$$

$$P(U < x) = \int_0^x dx = x$$
 (3.5)

$$\therefore P(1 - e^{\frac{-x}{2}} \ge U) = 1 - e^{\frac{-x}{2}}, \forall x \ge 0 \quad (3.6)$$

4 Triangular Distribution

4.1 Generate

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

Solution: Run the below code to generate T.dat

Run the command below in the terminal

4.2 Find the CDF of T.

$$F_T(t) = P(T < t) \tag{4.2}$$

$$= P(U_1 + U_2 < t) \tag{4.3}$$

we know that $0 \le U_1 \le 1$ and $0 \le U_2 \le 1$

$$0 \le U_1 + U_2 \le 2$$
, so

$$\forall t > 2, P(U_1 + U_2 < t) = 1$$

$$\forall t < 0, P(U_1 + U_2 < t) = 0$$

for $0 \le t \le 2$ let us split it into 2 cases, for $0 \le t \le 1$ and $1 < t \le 2$

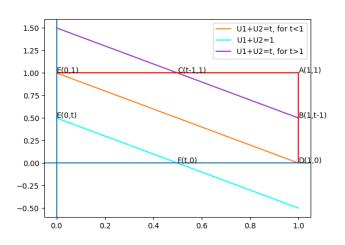


Fig. 4.2: Plot

The above figure is produced by the following code

https://github.com/Pranavb060504/ Random_numbers/blob/main/4.2/T_plot. py

Run the following command in the terminal to run the code

From Fig (4.2)

$$P(U_1 + U_2 < t, 0 \le t \le 1) = \frac{\Delta(EOF)}{\Delta(AEOD)}$$
 (4.4)

$$=\frac{t^2}{2}\tag{4.5}$$

$$P(U_1 + U_2 < t, 1 \le t \le 2) = \frac{\Delta(ABC)}{\Delta(AEOD)}$$
 (4.6)

$$=1-\frac{(2-t)^2}{2}\tag{4.7}$$

$$\therefore F_T(t) = P(U_1 + U_2 < t) = \begin{cases} 0 & t < 0 \\ \frac{t^2}{2} & 0 \le t \\ 1 - \frac{(2-t)^2}{2} & 1 < t \\ 1 & t > 2 \end{cases}$$
(4.8)

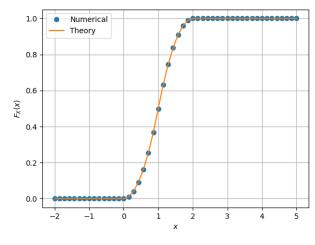


Fig. 4.5: CDF for (4)

4.3 Find the PDF of T.

Solution:

$$P_T(t) = \frac{d(F_T(t))}{dt} \tag{4.9}$$

$$\therefore P_T(t) = \begin{cases} 0 & t < 0 \\ t & 0 \le t \le 1 \\ 2 - t & 0 < t \le 2 \\ 0 & t > 2 \end{cases}$$
 (4.10)

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

Solution:

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

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

4.5 Verify your results through a plot.

Solution: Run the below code to get the cdf

Use the following command in the terminal to run the code

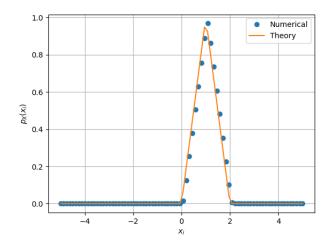


Fig. 4.5: PDF for (4)

Run the below code to get the pdf

https://github.com/Pranavb060504/ Random_numbers/blob/main/4.5/T_pdf. py

Use the following command in the terminal to run the code

python3 T pdf.py

5 Maximul Likelihood

5.1 Generate equiprobable $X \in \{1, -1\}$.

Solution: Run the below code,

https://github.com/Pranavb060504/ Random_numbers/blob/main/5.1/ bernoulli.c Use the below command in the terminal to run the code

cc bernoulli.c -lm ./a.out

5.2 Generate

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

where A = 5 dB, and $N \sim \mathcal{N}(0, 1)$.

Solution: Run the below code for generating samples of Y,

https://github.com/Pranavb060504/ Random_numbers/blob/main/5.2/Ygen.c

Use the below command in the terminal to run the code

cc Ygen.c -lm ./a.out

5.3 Plot Y using a scatter plot.

Solution:

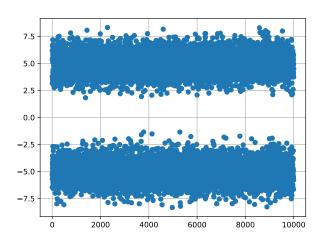


Fig. 5.3: plot for (5.3)

Run the following code to generate the scatter plot

https://github.com/Pranavb060504/ Random_numbers/blob/main/5.3/ Yplot.py

Use the below command to run the code,

python3 Yplot.py

5.4 Guess how to estimate *X* from *Y*.

Solution: if the received signal is greater than 0, then the receiver assumes s_1 was transmitted.

if the received signal is less than or equal to 0, then the receiver assumes s_0 was transmitted, where s_0 and s_1 are cases of X = 1 and X = -1 respectively where threshold 0 is taken to be the decision boundary.

$$y > 0 \implies s_1 \tag{5.2}$$

$$y \le 0 \implies s_0$$
 (5.3)

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1)$$
 (5.4)

and

$$P_{e|1} = \Pr(\hat{X} = 1|X = -1)$$
 (5.5)

Solution: Here s_1 and s_2 are equally probable ie, $p(s_1) = p(s_0) = \frac{1}{2}$

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_{x}^{\infty} e^{\frac{-x^{2}}{2}} dx$$
 (5.6)

$$p(e|s_{1}) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{0} e^{-\frac{(y-5)^{2}}{2}} dy$$
 (5.7)

$$p(e|s_{0}) = \frac{1}{\sqrt{2\pi}} \int_{0}^{\infty} e^{-\frac{(y+5)^{2}}{2}} dy$$
 (5.8)

5.6 Find P_e assuming that X has equiprobable symbols.

Solution: Total probability of bit error:

$$P_e = p(s_1)p(e|s_1) + p(s_0)p(e|s_0)$$
 (5.9)

$$= \frac{1}{2}[Q(5) + Q(5)] \tag{5.10}$$

$$\therefore p(s_1) = p(s_0) = \frac{1}{2}, X \text{ has equiprobable symbols}$$
(5.11)

= Q(5)

5.7 Verify by plotting the theoretical P_e with respect to A from 0 to 10 dB.

Solution:

Run the following code to generate the scatter plot

https://github.com/Pranavb060504/ Random_numbers/blob/main/5.7/ Pplot.py

Use the below command to run the code,

python3 Pplot.py

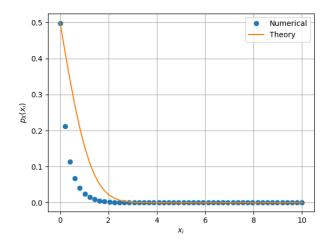


Fig. 5.7: plot for (5.7)

5.8 Now, consider a threshold δ while estimating X from Y. Find the value of δ that minimizes the theoretical P_e .

Solution: Threshold= δ ,

$$y > \delta \implies s_1$$
 (5.12)

$$y \le \delta \implies s_0 \tag{5.13}$$

$$p(e|s_1) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\delta} e^{-\frac{(y-5)^2}{2}} dy$$
 (5.14)

$$p(e|s_0) = \frac{1}{\sqrt{2\pi}} \int_{\delta}^{\infty} e^{-\frac{(y+5)^2}{2}} dy$$

$$P_e = \frac{1}{2\sqrt{2\pi}} \left(\int_{-\infty}^{\delta} e^{-\frac{(y-5)^2}{2}} dy + \int_{\delta}^{\infty} e^{-\frac{(y+5)^2}{2}} dy \right)$$

$$P_e = \frac{Q(\delta + 5) + Q(5 - \delta)}{2}$$
 (5.15)

$$P_e = f(\delta) \tag{5.16}$$

to minimize
$$P_e$$
, $\frac{d(f(\delta))}{d\delta} = 0$ and $f''(\delta) > 0$ (5.17)

$$e^{\frac{-(5-\delta)^2}{2}} - e^{\frac{-(5+\delta)^2}{2}} = 0 (5.18)$$

$$\therefore 5 - \delta = 5 + \delta, \implies \delta = 0 \tag{5.19}$$

$$f'''(\delta) = k((5 - \delta)e^{\frac{-(5 - \delta)^2}{2}} + (5 + \delta)e^{\frac{-(5 + \delta)^2}{2}}) > 0$$
(5.20)

5.9 Repeat the above exercise when

$$p_X(0) = p \tag{5.21}$$

Solution: $p_X(0) = p$

$$\implies p_X(1) = 1 - p$$

$$P_e = pP(e|s_0) + (1-p)P(e|s_1)$$
 (5.22)

$$= pQ(5+\delta) + (1-p)Q(5-\delta)$$
 (5.23)

$$\frac{d(P_e)}{d(\delta)} = 0 \qquad (5.24)$$

$$\implies e^{\frac{(5+\delta)^2-(5-\delta)^2}{2}} = \frac{p}{1-p} \qquad (5.25)$$

$$\therefore \delta = \frac{1}{10} log(\frac{p}{1-p}) \qquad (5.26)$$

$$\frac{d(P_e)}{d(\delta)} \quad at \quad \delta + \epsilon > 0 \qquad (5.27)$$

$$\frac{d(P_e)}{d(\delta)} \quad at \quad \delta - \epsilon < 0$$

$$\therefore \delta = \frac{1}{10} log\left(\frac{p}{1-p}\right) \longrightarrow minima$$

5.10 Repeat the above exercise using the MAP criterion.

Solution:

$$P_{X|Y}(x|y)\Big|_{X=1} = \frac{P(Y=y|X=1)P(X=1)}{P(Y=y)}$$
(5.28)

$$P(Y = y) = P(Y = y|X = 1)P(X = 1) + P(Y = y|X = -1)P(X = -1)$$
(5.29)

$$P(Y = y|X = 1)P(X = 1) = \frac{1}{2}P(Y = 5 + N)$$
(5.30)

$$=\frac{1}{2}\left(\frac{1}{\sqrt{2\pi}}e^{\frac{-(y-5)^2}{2}}\right) \tag{5.31}$$

$$\therefore P_{X|Y}(x|y)\big|_{X=1} = \frac{\frac{1}{2}\left(\frac{1}{\sqrt{2\pi}}e^{\frac{-(y-5)^2}{2}}\right)}{P(Y=y)}$$
 (5.32)

$$P_{X|Y}(x|y)\Big|_{X=-1} = \frac{P(Y=y|X=-1)P(X=-1)}{P(Y=y)}$$
(5.33)

$$P(Y = y|X = -1)P(X = -1) = \frac{1}{2}P(Y = -5 + N)$$

$$=\frac{1}{2}\left(\frac{1}{\sqrt{2\pi}}e^{\frac{-(y+5)^2}{2}}\right) \tag{5.34}$$

$$\therefore P_{X|Y}(x|y)\big|_{X=-1} = \frac{\frac{1}{2}\left(\frac{1}{\sqrt{2\pi}}e^{\frac{-(y+5)^2}{2}}\right)}{P(Y=y)}$$
 (5.35)

Now comparing $a = P_{X|Y}(x|y)|_{X=-1}$ and $b = P_{X|Y}(x|y)|_{X=1}$, if a > b, X = -1 is more likely

a < b, X = 1 is more likely.

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: The sum of squares of k independent standard random normal variables is nothing but a χ^2 distribution with k degrees of freedom.

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

Here k=2,

$$\therefore \chi^2(2) = P_V(v) = \frac{e^{\frac{-x}{2}}}{2}$$
 (6.2)

$$\implies F_V(v) = \int_0^v \frac{e^{\frac{-x}{2}}}{2} dx \tag{6.3}$$

$$=1-e^{\frac{-x}{2}} \tag{6.4}$$

To generate data for V, run the following code,

https://github.com/Pranavb060504/ Random numbers/blob/main/6.1/V g

Random_numbers/blob/main/6.1/V_gen.c

Run the below command in terminal,

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

https://github.com/Pranavb060504/ Random_numbers/blob/main/6.1/chi_pdf. py

Use the following command in the terminal to run the code

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

https://github.com/Pranavb060504/ Random_numbers/blob/main/6.1/chi_cdf. py

Use the following command in the terminal to run the code

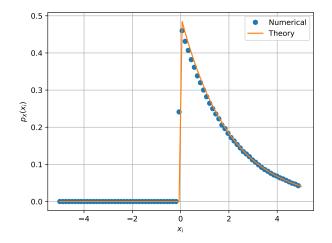


Fig. 6.1: PDF for (6.1)

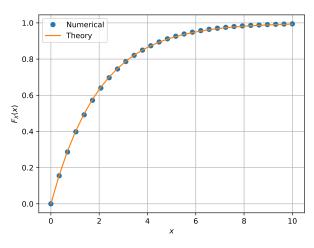


Fig. 6.1: PDF for (6.1)

6.2 If

$$F_V(x) = \begin{cases} 1 - e^{-\alpha x} & x \ge 0\\ 0 & x < 0, \end{cases}$$
 (6.5)

find α .

Solution: From (6.4) $\alpha = 0.5$

6.3 Plot the CDF and PDf of

$$A = \sqrt{V} \tag{6.6}$$

Solution:

$$F_A(a) = P(A < a) = P(V < a^2)$$
 (6.7)

from(6.4), =
$$\begin{cases} 1 - e^{\frac{-a^2}{2}} & a > 0\\ 0 & a <= 0 \end{cases}$$
 (6.8)

$$\implies P_A(a) = \frac{d(F_A(a))}{da}$$
 (6.9)

$$\Rightarrow P_A(a) = \frac{d(F_A(a))}{da}$$
 (6.9)
=\begin{cases} ae^{\frac{-a^2}{2}} & a > 0 \\ 0 & a <= 0 \end{cases} (6.10)

To generate data for A, run the following code,

https://github.com/Pranavb060504/ Random numbers/blob/main/6.3/A gen.

Run the below command in terminal.

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

https://github.com/Pranavb060504/ Random numbers/blob/main/6.3/A pdf. py

Use the following command in the terminal to run the code

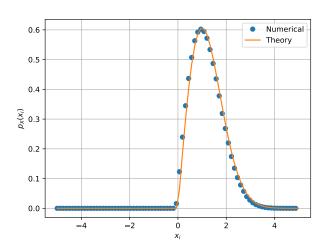


Fig. 6.3: PDF for (6.3)

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

https://github.com/Pranavb060504/ Random numbers/blob/main/6.3/A cdf.

Use the following command in the terminal to run the code

python3 A_cdf.py

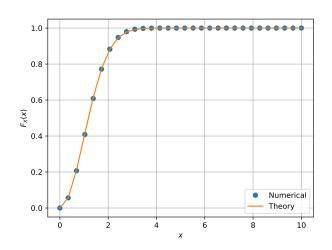


Fig. 6.3: CDF for (6.3)

7 CONDITIONAL PROBABILITY

7.1 Plot

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

for

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

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

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

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

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

7.4 Plot P_e in problems 7.1 and 7.3 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)

$$\mathbf{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix}, n_1, n_2 \sim \mathcal{N}(0, 1).$$
 (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 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.