

Random Numbers

AI1110: Probability and Random Variables

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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/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/1.1.c
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/header.h
```

Compile and run the C program by executing the following

```
cc -lm 1.1.c
./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

$$F_U(x) = \Pr(U \leq x) \quad (1.1)$$

Solution: Download the following Python code that plots Fig. 1.2

```
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/1.2.py
```

Run the code by executing

```
python 1.2.py
```

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

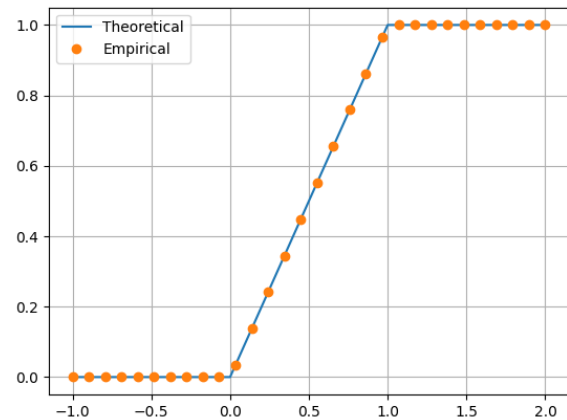


Fig. 1.2. The CDF of U

Solution: The PDF of U is given by

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

The CDF of U is given by

$$F_U(x) = \Pr(U \leq x) = \int_{-\infty}^x p_U(x) dx \quad (1.3)$$

If $x < 0$,

$$\int_{-\infty}^x p_U(x) dx = \int_{-\infty}^x 0 dx = 0 \quad (1.4)$$

If c ,

$$\int_{-\infty}^x p_U(x) dx = \int_{-\infty}^0 0 dx + \int_0^x 1 dx \quad (1.5)$$

$$= 0 + x \quad (1.6)$$

$$= x \quad (1.7)$$

If $x > 1$,

$$\int_{-\infty}^x p_U(x) dx = \int_{-\infty}^0 0 dx + \int_0^1 1 dx + \int_1^x 0 dx \quad (1.8)$$

$$\int_{-\infty}^x p_U(x) dx = 0 + 1 + 0 \quad (1.9)$$

$$= 1 \quad (1.10)$$

Therefore, we obtain the CDF of U as

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq 1 \\ 1 & x > 1 \end{cases} \quad (1.11)$$

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^N U_i \quad (1.12)$$

and its variance as

$$\text{Var}[U] = E[U - E[U]]^2 \quad (1.13)$$

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

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/1.4.c
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/header.h
```

Compile and run the C program by executing the following

```
cc -lm 1.4.c
./a.out
```

The output of the code is

$$\mu_{\text{emp}} = 0.500007 \quad (1.14)$$

$$\mu_{\text{the}} = 0.500000 \quad (1.15)$$

$$\sigma_{\text{emp}}^2 = 0.083301 \quad (1.16)$$

$$\sigma_{\text{the}}^2 = 0.083333 \quad (1.17)$$

1.5 Verify your result theoretically given that

$$E[U^k] = \int_{-\infty}^{\infty} x^k dF_U(x) \quad (1.18)$$

Solution: The mean of U is given by

$$E[U] = \int_{-\infty}^{\infty} x dF_U(x) \quad (1.19)$$

On differentiating the CDF of U , we get

$$dF_U(x) = \begin{cases} 0 & x < 0 \\ dx & 0 \leq x \leq 1 \\ 0 & x > 1 \end{cases} \quad (1.20)$$

$$\therefore E[U] = \int_0^1 x dx = \frac{1}{2} = 0.5 \quad (1.21)$$

Similarly,

$$\therefore E[U^2] = \int_0^1 x^2 dx = \frac{1}{3} \quad (1.22)$$

Now, the variance of U is given by

$$\text{Var}[U] \quad (1.23)$$

$$= E[U - E[U]]^2 \quad (1.24)$$

$$= E[U^2 - 2UE[U] + (E[U])^2] \quad (1.25)$$

By linearity of expectation, we have

$$E[U^2] + E[-2UE[U]] + E[(E[U])^2] \quad (1.26)$$

$$= E[U^2] - 2E[U]E[U] + (E[U])^2 \quad (1.27)$$

$$= E[U^2] - (E[U])^2 \quad (1.28)$$

$$= \frac{1}{3} - \left(\frac{1}{2}\right)^2 \quad (1.29)$$

$$= \frac{1}{12} \approx 0.083333 \quad (1.30)$$

2. CENTRAL LIMIT THEOREM

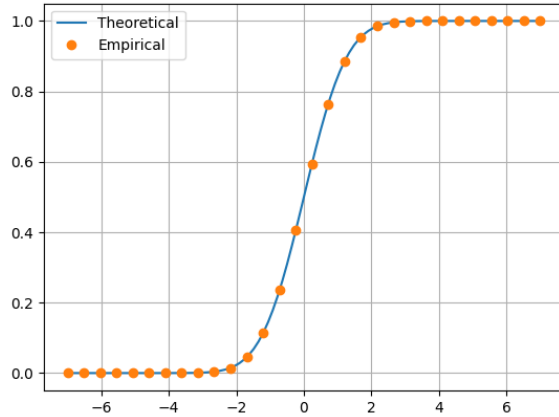
2.1 Generate 10^6 samples of the random variable

$$X = \sum_{i=1}^{12} U_i - 6 \quad (2.1)$$

using a C program, where $U_i, i = 1, 2, \dots, 12$ are a set of independent uniform random variables between 0 and 1 and save in a file called gau.dat

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/2.1.c
wget https://github.com/Ankit-Saha-2003/
  AI1110/raw/main/Random-Numbers/
  codes/header.h
```

Fig. 2.2. The CDF of X

Compile and run the C program by executing the following

```
cc -lm 2.1.c
./a.out
```

- 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 Python code that plots Fig. 2.2

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/2.2.py
```

Run the code by executing

```
python 2.2.py
```

Every CDF is monotone increasing and right-continuous. Furthermore,

$$\lim_{x \rightarrow -\infty} F_X(x) = 0 \quad \lim_{x \rightarrow \infty} F_X(x) = 1 \quad (2.2)$$

Thus, every CDF is bounded between 0 and 1 and hence, convergent.

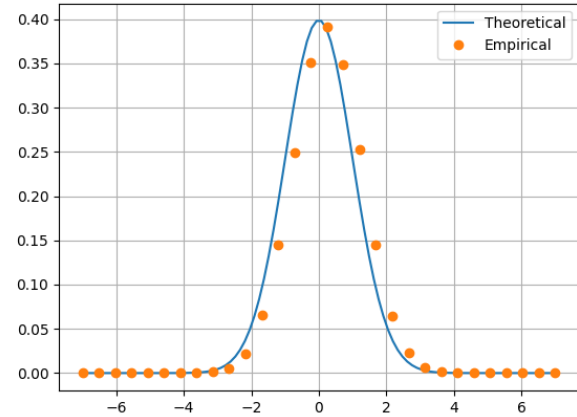
In this case, the CDF is also left-continuous. Therefore, X is a continuous random variable.

- 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

$$p_X(x) = \frac{d}{dx} F_X(x) \quad (2.3)$$

What properties does the PDF have?

Solution: Download the following Python code that plots Fig. 2.2

Fig. 2.3. The PDF of X

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/2.3.py
```

Run the code by executing

```
python 2.3.py
```

Every PDF is bounded between 0 and 1 and

$$\int_{-\infty}^{\infty} p_X(x) dx = 1 \quad (2.4)$$

In this case, the PDF is symmetric about $x = 0$

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

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/2.4.c
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/header.h
```

Compile and run the C program by executing the following

```
cc -lm 2.4.c
./a.out
```

The output of the code is

$$\mu_{\text{emp}} = 0.000294 \quad (2.5)$$

$$\mu_{\text{the}} = 0.000000 \quad (2.6)$$

$$\sigma_{\text{emp}}^2 = 0.999560 \quad (2.7)$$

$$\sigma_{\text{the}}^2 = 1.000000 \quad (2.8)$$

2.5 Given that

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

repeat the above exercise theoretically

Solution: The mean of X is given by

$$E[X] = \int_{-\infty}^{\infty} x p_X(x) dx \quad (2.10)$$

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

Now, let

$$g(x) = \frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \quad (2.12)$$

$$\Rightarrow g(-x) = \frac{-x}{\sqrt{2\pi}} \exp\left(-\frac{(-x)^2}{2}\right) \quad (2.13)$$

$$= -\frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \quad (2.14)$$

$$= -g(x) \quad (2.15)$$

Thus, $g(x)$ is an odd function

$$\therefore E[X] = \int_{-\infty}^{\infty} g(x) dx = 0 \quad (2.16)$$

Now,

$$E[X^2] = \int_{-\infty}^{\infty} x^2 p_X(x) dx \quad (2.17)$$

$$= \int_{-\infty}^{\infty} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \quad (2.18)$$

$$= 2 \int_0^{\infty} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \quad (2.19)$$

since $\frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$ is an even function

Using integration by parts,

$$E[X^2] = \sqrt{\frac{2}{\pi}} \int_0^{\infty} x \cdot x \exp\left(-\frac{x^2}{2}\right) dx \quad (2.20)$$

$$= \sqrt{\frac{2}{\pi}} \left(x \int x \exp\left(-\frac{x^2}{2}\right) dx \right) \Big|_0^{\infty} - \sqrt{\frac{2}{\pi}} \int_0^{\infty} 1 \cdot \int x \exp\left(-\frac{x^2}{2}\right) dx \quad (2.21)$$

Substitute $t = \frac{x^2}{2} \Rightarrow dt = x dx$

$$\int x \exp\left(-\frac{x^2}{2}\right) dx = \int \exp(-t) dt \quad (2.22)$$

$$= -\exp(-t) \quad (2.23)$$

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

Now,

$$-x \exp\left(-\frac{x^2}{2}\right) \Big|_0^{\infty} = 0 - 0 = 0 \quad (2.25)$$

$$\therefore \lim_{x \rightarrow \infty} x \exp\left(-\frac{x^2}{2}\right) = \lim_{x \rightarrow \infty} \frac{x}{\exp\left(\frac{x^2}{2}\right)} = 0 \quad (2.26)$$

as exponential function grows much faster than a polynomial function

Also,

$$\int_0^{\infty} -\exp\left(-\frac{x^2}{2}\right) dx \quad (2.27)$$

$$\xleftrightarrow{x=t\sqrt{2}} \int_0^{\infty} -\exp(-t^2) dt \sqrt{2} \quad (2.28)$$

$$= -\sqrt{2} \int_0^{\infty} \exp(-t^2) dt \quad (2.29)$$

$$= -\sqrt{2} \frac{\sqrt{\pi}}{2} \quad (2.30)$$

$$= -\sqrt{\frac{\pi}{2}} \quad (2.31)$$

Therefore,

$$E[X^2] = 0 - \sqrt{\frac{2}{\pi}} \left(-\sqrt{\frac{\pi}{2}} \right) \quad (2.32)$$

$$= 1 \quad (2.33)$$

$$\therefore \text{Var}[X] = E[X^2] - (E[X])^2 \quad (2.34)$$

$$= 1 - 0 \quad (2.35)$$

$$= 1 \quad (2.36)$$

3. FROM UNIFORM TO OTHER

3.1 Generate samples of

$$V = -2 \ln(1 - U) \quad (3.1)$$

and plot its CDF

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/3.1.c
```

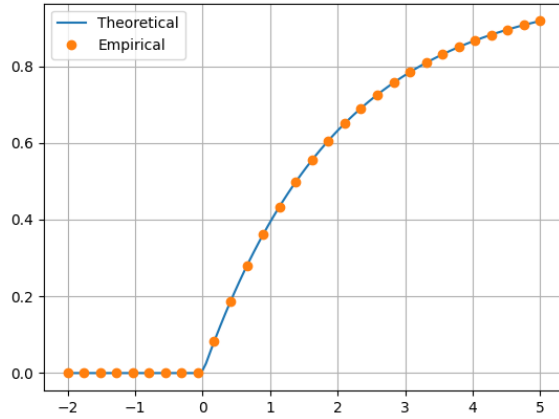


Fig. 3.1. The CDF of V

Compile and run the C program by executing the following

```
cc -lm 3.1.c
./a.out
```

Download the following Python code that plots Fig. 3.1

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/3.1.py
```

Run the code by executing

```
python 3.1.py
```

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

Solution: We have

$$F_V(x) = \Pr(V \leq x) \quad (3.2)$$

$$= \Pr(-2 \ln(1 - U) \leq x) \quad (3.3)$$

$$= \Pr\left(\ln(1 - U) \geq -\frac{x}{2}\right) \quad (3.4)$$

$$= \Pr\left(1 - U \geq \exp\left(-\frac{x}{2}\right)\right) \quad (3.5)$$

$$= \Pr\left(U \leq 1 - \exp\left(-\frac{x}{2}\right)\right) \quad (3.6)$$

$$= F_U\left(1 - \exp\left(-\frac{x}{2}\right)\right) \quad (3.7)$$

Now,

$$0 \leq 1 - \exp\left(-\frac{x}{2}\right) < 1 \quad \text{if } x \geq 0 \quad (3.8)$$

$$1 - \exp\left(-\frac{x}{2}\right) < 0 \quad \text{if } x < 0 \quad (3.9)$$

Therefore,

$$F_V(x) = \begin{cases} 1 - \exp\left(-\frac{x}{2}\right) & x \geq 0 \\ 0 & x < 0 \end{cases} \quad (3.10)$$

4. TRIANGULAR DISTRIBUTION

4.1 Generate

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

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/4.1.c
```

Compile and run the C program by executing the following

```
cc -lm 4.1.c
./a.out
```

4.2 Find the CDF of T

Solution: The CDF of T is given by

$$F_T(t) = \Pr(T \leq t) = \Pr(U_1 + U_2 \leq t) \quad (4.2)$$

Since $U_1, U_2 \in [0, 1] \implies U_1 + U_2 \in [0, 2]$
Therefore, if $t \geq 2$, then $U_1 + U_2 \leq t$ is always true and if $t < 0$, then $U_1 + U_2 \leq t$ is always false.

Now, fix the value of U_1 to be some x

$$x + U_2 \leq t \implies U_2 \leq t - x \quad (4.3)$$

If $0 \leq t \leq 1$, then x can take all values in $[0, t]$

$$F_T(t) = \int_0^t \Pr(U_2 \leq t - x) p_{U_1}(x) dx \quad (4.4)$$

$$= \int_0^t F_{U_2}(t - x) p_{U_1}(x) dx \quad (4.5)$$

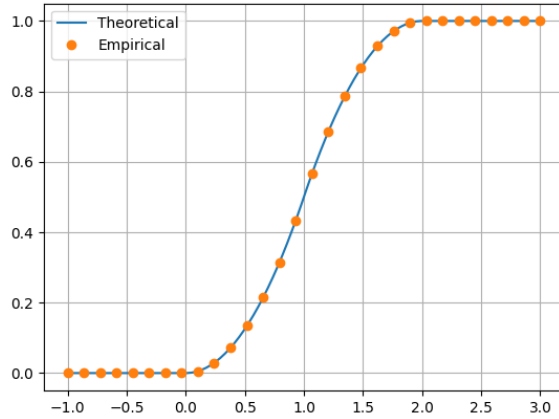
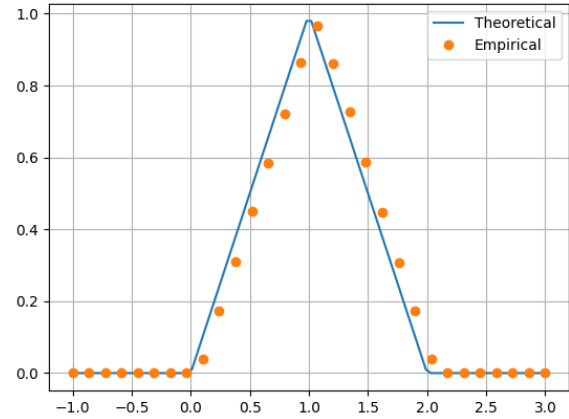
$$0 \leq x \leq t \implies 0 \leq t - x \leq t \leq 1 \quad (4.6)$$

$$\implies F_{U_2}(t - x) = t - x \quad (4.7)$$

$$F_T(t) = \int_0^t (t - x) \cdot 1 \cdot dx \quad (4.8)$$

$$= tx - \frac{x^2}{2} \Big|_0^t \quad (4.9)$$

$$= \frac{t^2}{2} \quad (4.10)$$

Fig. 4.2. The CDF of T Fig. 4.3. The PDF of T

If $1 < t < 2$, x can only take values in $[0, 1]$ as $U_1 \leq 1$

$$F_T(t) = \int_0^1 F_{U_2}(t-x) \cdot 1 \cdot dx \quad (4.11)$$

$$0 \leq x \leq t-1 \implies 1 \leq t-x \leq t \quad (4.12)$$

$$t-1 \leq x \leq 1 \implies 0 < t-1 \leq t-x \leq 1 \quad (4.13)$$

$$F_T(t) = \int_0^{t-1} 1 dx + \int_{t-1}^1 (t-x) dx \quad (4.14)$$

$$= t-1 + t(1-(t-1)) - \frac{1}{2} + \frac{(t-1)^2}{2} \quad (4.15)$$

$$= t-1 + 2t-t^2 - \frac{1}{2} + \frac{t^2}{2} + \frac{1}{2} - t \quad (4.16)$$

$$= -\frac{t^2}{2} + 2t - 1 \quad (4.17)$$

Therefore,

$$F_T(t) = \begin{cases} 0 & t < 0 \\ \frac{t^2}{2} & 0 \leq t \leq 1 \\ 2t - \frac{t^2}{2} - 1 & 1 < t < 2 \\ 1 & t \geq 2 \end{cases} \quad (4.18)$$

4.3 Find the PDF of T

Solution: The PDF of T is given by

$$p_T(t) = \frac{d}{dt} F_T(t) \quad (4.19)$$

$$\therefore p_T(t) = \begin{cases} 0 & t < 0 \\ t & 0 \leq t \leq 1 \\ 2-t & 1 < t < 2 \\ 0 & t \geq 2 \end{cases} \quad (4.20)$$

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

Solution: The theoretical expressions for the CDF and PDF have been found in problems 4.2 and 4.3 respectively

4.5 Verify your results through a plot

Solution: Download the following Python codes that plot Fig. 4.2 and Fig. 4.3

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/4.2.py
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/4.3.py
```

Run the codes by executing

```
python 4.2.py
python 4.3.py
```

5. MAXIMAL LIKELIHOOD

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

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/5.1.c
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/header.h
```

Compile and run the C program by executing the following

```
cc -lm 5.1.c
./a.out
```

5.2 Generate

$$Y = AX + N \quad (5.1)$$

where $A = 5$ dB, $X \in \{-1, 1\}$ is Bernoulli and $N \sim \mathcal{N}(0, 1)$

Solution: Download the C source code by executing the following commands

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/5.2.c
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/header.h
```

Compile and run the C program by executing the following

```
cc -lm 5.2.c
./a.out
```

5.3 Plot Y

Solution: Download the following Python code that plots Fig. 5.3

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/5.3.py
```

Run the code by executing

```
python 5.3.py
```

5.4 Guess how to estimate X from Y

Solution:

$$X = \begin{cases} 1 & Y > 0 \\ -1 & Y < 0 \end{cases} \quad (5.2)$$

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1) \quad (5.3)$$

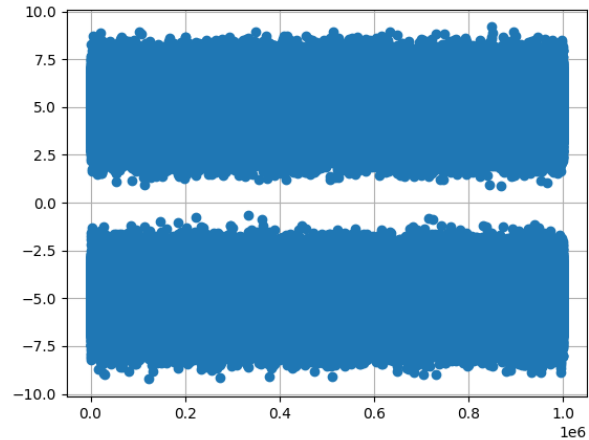


Fig. 5.3. Plot of Y

and

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

Solution:

$$\Pr(\hat{X} = -1|X = 1) = \Pr(Y < 0|X = 1) \quad (5.5)$$

$$= \Pr(A + N < 0) \quad (5.6)$$

$$= \Pr(N < -5) \quad (5.7)$$

$$= 1 - \Pr(N > -5) \quad (5.8)$$

$$= 1 - Q(-5) \quad (5.9)$$

$$= Q(5) \quad (5.10)$$

where $Q(x) = \Pr(N > x)$ is the Q-function

$$Q(x) = 1 - Q(-x) \quad \forall x \in \mathbb{R} \quad (5.11)$$

$$\Pr(\hat{X} = 1|X = -1) = \Pr(Y > 0|X = -1) \quad (5.12)$$

$$= \Pr(-A + N > 0) \quad (5.13)$$

$$= \Pr(N > 5) \quad (5.14)$$

$$= Q(5) \quad (5.15)$$

5.6 Find P_e assuming that X has equiprobable symbols

Solution:

$$P_e = \Pr(X = -1) P_{e|0} + \Pr(X = 1) P_{e|1} \quad (5.16)$$

Since X has equiprobable symbols, $\Pr(X = -1) = \Pr(X = 1) = \frac{1}{2}$

$$P_e = \frac{1}{2} Q(5) + \frac{1}{2} Q(5) \quad (5.17)$$

$$= Q(5) \quad (5.18)$$

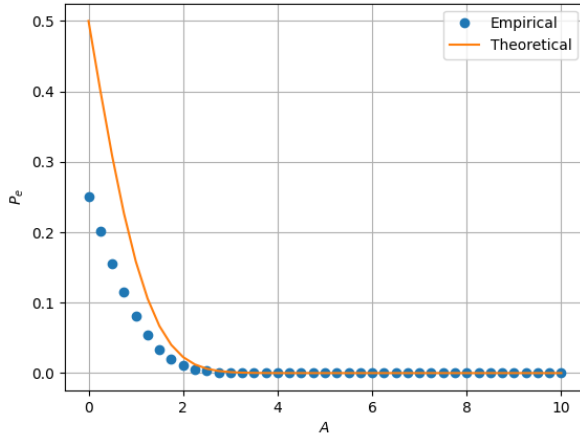


Fig. 5.7. Plot of P_e

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

Solution: Download the following Python code that plots Fig. 5.7

```
wget https://github.com/Ankit-Saha-2003/
AI1110/raw/main/Random-Numbers/
codes/5.7.py
```

Run the code by executing

```
python 5.7.py
```

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

Solution:

$$X = \begin{cases} 1 & Y > \delta \\ -1 & Y < \delta \end{cases} \quad (5.19)$$

$$P_{e|0} = \Pr(\hat{X} = -1 | X = 1) \quad (5.20)$$

$$= \Pr(Y < \delta | X = 1) \quad (5.21)$$

$$= \Pr(A + N < \delta) \quad (5.22)$$

$$= \Pr(N < \delta - 5) \quad (5.23)$$

$$= Q(5 - \delta) \quad (5.24)$$

$$P_{e|1} = \Pr(\hat{X} = 1 | X = -1) \quad (5.25)$$

$$= \Pr(Y > \delta | X = -1) \quad (5.26)$$

$$= \Pr(-A + N > \delta) \quad (5.27)$$

$$= \Pr(N > \delta + 5) \quad (5.28)$$

$$= Q(5 + \delta) \quad (5.29)$$

$$(5.30)$$

Now, P_e is given by

$$P_e = \Pr(X = -1) P_{e|0} + \Pr(X = 1) P_{e|1} \quad (5.31)$$

$$= \frac{1}{2} Q(5 - \delta) + \frac{1}{2} Q(5 + \delta) \quad (5.32)$$

$$= \frac{Q(5 - \delta) + Q(5 + \delta)}{2} \quad (5.33)$$

$$= g(\delta) \quad (5.34)$$

On differentiating g with respect to δ , we get

$$g'(\delta) = \frac{Q'(5 + \delta) - Q'(5 - \delta)}{2} \quad (5.35)$$

Recall the definition of $Q(x)$

$$Q(x) = \int_x^\infty \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u^2}{2}\right) du \quad (5.36)$$

$$\Rightarrow Q'(x) = -\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \quad (5.37)$$

Thus,

$$g'(\delta) = \frac{\exp\left(-\frac{(5-\delta)^2}{2}\right) - \exp\left(-\frac{(5+\delta)^2}{2}\right)}{2\sqrt{2\pi}} \quad (5.38)$$

$$g'(\delta) = 0 \Rightarrow (5 - \delta)^2 = (5 + \delta)^2 \quad (5.39)$$

$$\Rightarrow |5 - \delta| = |5 + \delta| \quad (5.40)$$

$$\Rightarrow \delta = 0 \quad (5.41)$$

$$g''(\delta) = \frac{(5 - \delta)}{2\sqrt{2\pi}} \exp\left(-\frac{(5 - \delta)^2}{2}\right) + \frac{(5 + \delta)}{2\sqrt{2\pi}} \exp\left(-\frac{(5 + \delta)^2}{2}\right) \quad (5.42)$$

$$g''(0) = \frac{5}{\sqrt{2\pi}} \exp\left(-\frac{25}{2}\right) > 0 \quad (5.43)$$

Therefore, $\hat{\delta} = 0$ is a minima and it is what minimizes P_e

5.9 Repeat the above exercise when

$$p_X(-1) = p \quad (5.44)$$

Solution:

$$P_e = p_X(1)P_{e|0} + p_X(-1)P_{e|1} \quad (5.45)$$

$$= (1 - p)Q(5 - \delta) + pQ(5 + \delta) \quad (5.46)$$

$$= g(\delta) \quad (5.47)$$

On differentiating g with respect to δ , we get

$$g'(\delta) = \frac{(1-p)\exp\left(-\frac{(5-\delta)^2}{2}\right) - p\exp\left(-\frac{(5+\delta)^2}{2}\right)}{\sqrt{2\pi}} \quad (5.48)$$

$g'(\delta) = 0$ when

$$(1-p)\exp\left(-\frac{(5-\delta)^2}{2}\right) = p\exp\left(-\frac{(5+\delta)^2}{2}\right) \quad (5.49)$$

$$\Rightarrow \exp\left(\frac{(5+\delta)^2 - (5-\delta)^2}{2}\right) = \frac{p}{1-p} \quad (5.50)$$

$$\Rightarrow \exp(10\delta) = \frac{p}{1-p} \quad (5.51)$$

$$\therefore \hat{\delta} = \frac{1}{10} \ln \frac{p}{1-p} \quad (5.52)$$

5.10 Repeat the above exercise using the MAP criterion

Solution: The PDF of $X|Y$ is given by

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x)p_X(x)}{p_Y(y)} \quad (5.53)$$

Assuming X has equiprobable symbols,
 $p_X(x) = \frac{1}{2}, \quad x = -1, 1$

$$p_Y(y) = p_X(-1)p_{Y|X}(y|-1) + p_X(1)p_{Y|X}(y|1) \quad (5.54)$$

$$= \frac{1}{2} \Pr(-A + N = y) + \frac{1}{2} \Pr(A + N = y) \quad (5.55)$$

$$= \frac{p_N(y+5) + p_N(y-5)}{2} \quad (5.56)$$

$$= \frac{\exp\left(-\frac{(y+5)^2}{2}\right) + \exp\left(-\frac{(y-5)^2}{2}\right)}{2\sqrt{2\pi}} \quad (5.57)$$

$$(5.58)$$

Now,

$$p_{X|Y}(1|y) = \frac{\Pr(A + N = y) p_X(1)}{p_Y(y)} \quad (5.59)$$

$$= \frac{p_N(y-5)p_X(1)}{p_Y(y)} \quad (5.60)$$

$$= \frac{\exp\left(-\frac{(y-5)^2}{2}\right)}{2\left(\exp\left(-\frac{(y+5)^2}{2}\right) + \exp\left(-\frac{(y-5)^2}{2}\right)\right)} \quad (5.61)$$

$$= \frac{1}{2\left(1 + \exp\left(\frac{(y-5)^2 - (y+5)^2}{2}\right)\right)} \quad (5.62)$$

$$= \frac{1}{2(1 + \exp(-10y))} \quad (5.63)$$

Similarly,

$$p_{X|Y}(-1|y) = \frac{1}{2(1 + \exp(10y))} \quad (5.64)$$

Now,

$$p_{X|Y}(1|y) > p_{X|Y}(-1|y) \quad (5.65)$$

$$\Leftrightarrow \frac{1}{2(1 + \exp(-10y))} > \frac{1}{2(1 + \exp(10y))} \quad (5.66)$$

$$\Leftrightarrow \exp(-10y) < \exp(10y) \quad (5.67)$$

$$\Leftrightarrow y > 0 \quad (5.68)$$

And $p_{X|Y}(1|y) < p_{X|Y}(-1|y) \Leftrightarrow y < 0$

Therefore, $X = 1$ is more probable than $X = -1$ when $Y > 0$ and vice versa

Consider now a general Bernoulli random variable X with $p_X(-1) = p, p_X(1) = 1 - p$

$$p_Y(y) = p_X(-1)p_{Y|X}(y|-1) + p_X(1)p_{Y|X}(y|1) \quad (5.69)$$

$$= pp_N(y+5) + (1-p)p_N(y-5) \quad (5.70)$$

$$= \frac{p\exp\left(-\frac{(y+5)^2}{2}\right) + (1-p)\exp\left(-\frac{(y-5)^2}{2}\right)}{\sqrt{2\pi}} \quad (5.71)$$

$$p_{X|Y}(1|y) = \frac{p_{Y|X}(y|1)p_X(1)}{p_Y(y)} \quad (5.72)$$

$$= \frac{(1-p)\exp\left(-\frac{(y-5)^2}{2}\right)}{p\exp\left(-\frac{(y+5)^2}{2}\right) + (1-p)\exp\left(-\frac{(y-5)^2}{2}\right)} \quad (5.73)$$

$$= \frac{1-p}{1-p + p\exp(-10y)} \quad (5.74)$$

Similarly,

$$p_{X|Y}(-1|y) = \frac{p}{p + (1 - p) \exp(10y)} \quad (5.75)$$

Now,

$$p_{X|Y}(1|y) > p_{X|Y}(-1|y) \quad (5.76)$$

$$\Leftrightarrow \frac{1 - p}{1 - p + p \exp(-10y)} > \frac{p}{p + (1 - p) \exp(10y)} \quad (5.77)$$

$$\Leftrightarrow 1 + \frac{p}{1 - p} \exp(-10y) < 1 + \frac{1 - p}{p} \exp(10y) \quad (5.78)$$

$$\Leftrightarrow \exp(20y) > \left(\frac{p}{1 - p} \right)^2 \quad (5.79)$$

$$\Leftrightarrow y > \frac{1}{10} \ln \frac{p}{1 - p} = \hat{\delta} \quad (5.80)$$

and

$$p_{X|Y}(1|y) < p_{X|Y}(-1|y) \quad (5.81)$$

$$\Leftrightarrow \frac{1 - p}{1 - p + p \exp(-10y)} < \frac{p}{p + (1 - p) \exp(10y)} \quad (5.82)$$

$$\Leftrightarrow 1 + \frac{p}{1 - p} \exp(-10y) > 1 + \frac{1 - p}{p} \exp(10y) \quad (5.83)$$

$$\Leftrightarrow \exp(20y) < \left(\frac{p}{1 - p} \right)^2 \quad (5.84)$$

$$\Leftrightarrow y < \frac{1}{10} \ln \frac{p}{1 - p} = \hat{\delta} \quad (5.85)$$

Therefore, $X = 1$ is more probable than $X = -1$ when $Y > \hat{\delta}$ and vice versa