

# Assignment

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**Abstract**—This manual provides solutions to the Assignment on Random Numbers

## 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.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/1.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 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:** The python code for the plot in Fig. 1.2 is given below,

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/1.2CDF.py>

Download the above file and execute the command below to produce Fig.1.2

```
$ python3 1.2CDF.py
```

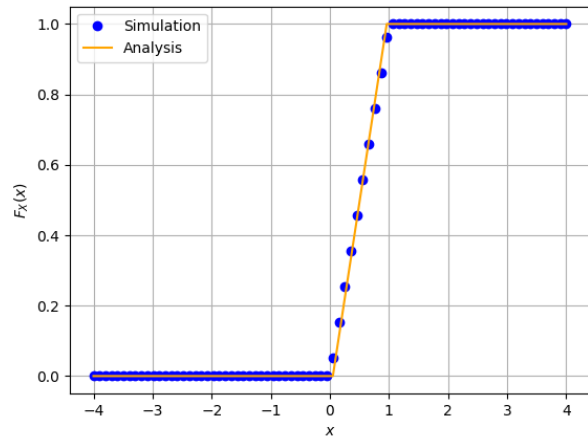


Fig. 1.2: The CDF of  $U$

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

**Solution:** Given  $U$  is a uniform Random Variable between 0 and 1,

$$F_U(x) = \Pr(U \leq x) = \int_{-\infty}^x p_U(u) du \quad (1.2)$$

We have three parts:

- i) For  $x < 0$ ;  $p_X(x) = 0$ , So  $F_U(x) = 0$ .
- ii) For  $0 \leq x < 1$ ;

$$F_U(x) = \int_0^x (1) du = x \quad (1.3)$$

- iii) For  $x \geq 1$ ; CDF is 1 as all the random numbers are between 0 and 1.

Therefore,

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x < 1 \\ 1 & x \geq 1 \end{cases} \quad (1.4)$$

1.4 The mean of  $U$  is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^N U_i \quad (1.5)$$

and its variance as

$$\text{var}[U] = E[U - E[U]]^2 \quad (1.6)$$

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

**Solution:** Download the following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/1.4.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 1.4.c
$ ./a.out
```

The actual analysis values,

$$\text{Mean} = 0.500007 \quad (1.7)$$

$$\text{Variance} = 0.083301 \quad (1.8)$$

1.5 Verify your result theoretically given that

$$E[U^k] = \int_{-\infty}^{\infty} x^k dF_U(x) \quad (1.9)$$

**Solution:** W.K.T,

$$\text{var}[U] = E[U^2] - E[U]^2 \quad (1.10)$$

Where  $E[U]$  is,

$$E[U] = \int_{-\infty}^{\infty} x dF_U(x) \quad (1.11)$$

$$= \int_0^1 x \quad (1.12)$$

$$= \frac{1}{2} = 0.5 \quad (1.13)$$

And  $E[U^2]$  is,

$$E[U^2] = \int_{-\infty}^{\infty} x^2 dF_U(x) \quad (1.14)$$

$$= \int_0^1 x^2 dF_U(x) \quad (1.15)$$

$$= \frac{1}{3} \quad (1.16)$$

$$(1.17)$$

Hence finally,

$$\therefore \text{var}[U] = \frac{1}{12} = 0.0833 \quad (1.18)$$

The theoretically calculated values are,

$$\text{Mean} = 0.5 \quad (1.19)$$

$$\text{Variance} = 0.0833 \quad (1.20)$$

These values matches with the actual analysis values from above,

$$\text{Mean} = 0.500007 \quad (1.21)$$

$$\text{Variance} = 0.083301 \quad (1.22)$$

Hence it verifies the result.

## 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 following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/2.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 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:** The CDF of  $X$  is plotted in Fig. 2.2 using the code below

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/2.2CDF.py>

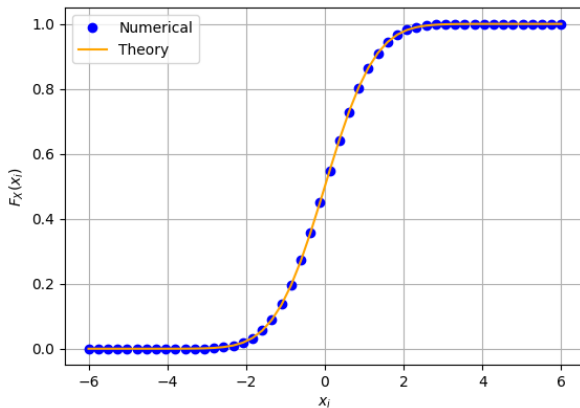


Fig. 2.2: The CDF of  $X$

Download the above file and execute the following command to produce Fig.2.2

\$ python3 2.2CDF.py

Some of the properties of CDF

- $F_X(x)$  is non decreasing function.
- As  $x \rightarrow -\infty$ ,  $F_X(x) \rightarrow 0$  and when  $x \rightarrow \infty$ ,  $F_X(x) \rightarrow 1$
- Graph is linear upto some region around  $x = 0$ .

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

$$p_X(x) = \frac{d}{dx} F_X(x) \quad (2.2)$$

What properties does the PDF have?

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

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/2.3PDF.py>

Download the above files and execute the following commands to produce Fig.2.3

\$ python3 2.3PDF.py

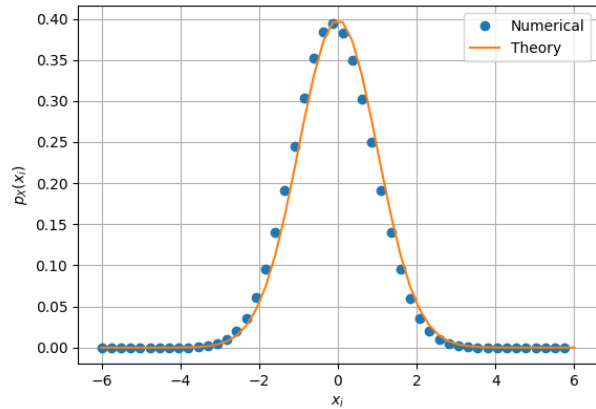


Fig. 2.3: The PDF of  $X$

Some of the properties of the PDF:

- Symmetric about  $x = \mu$
- Area under the PDF graph is unity.
- Increasing function for  $x < \mu$  and decreasing for  $x > \mu$  and attains maximum at  $x = \mu$ .

Let  $x \sim \mathcal{N}(0, 1)$ . The Q-function  $Q(x)$  is defined as:

$$Q(x) = \Pr(X > x) \quad (2.3)$$

$$= 1 - \Pr(X \leq x) \quad (2.4)$$

$$= 1 - F_X(x) \quad (2.5)$$

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

**Solution:** Download the following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/2.4.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

\$ gcc 2.4.c  
\$ ./a.out

The actual analysis values,

$$\text{Mean} = 0.000294 \quad (2.6)$$

$$\text{Variance} = 0.999561 \quad (2.7)$$

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.8)$$

repeat the above exercise theoretically.

**Solution:** CDF is defined as

$$F_X(x) = \int_{-\infty}^{\infty} p_X(x) dx \quad (2.9)$$

$$\text{W.K.T, } \boxed{F_X(x) = 1} \quad (2.10)$$

Mean is given by

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

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

It is a odd function function, So its value is 0.

$$\boxed{E(x) = 0} \quad (2.13)$$

$E(x^2)$  is given by,

$$E(X^2) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} x^2 \exp\left(-\frac{x^2}{2}\right) dx \quad (2.14)$$

On applying by-parts for the above integral we get,

$$\int_{-\infty}^{\infty} x \times x \exp\left(-\frac{x^2}{2}\right) dx = -x \exp\left(-\frac{x^2}{2}\right) \Big|_{-\infty}^{\infty} \quad (2.15)$$

$$+ \int_{-\infty}^{\infty} \exp\left(-\frac{x^2}{2}\right) dx \quad (2.16)$$

Where,  $x \exp\left(-\frac{x^2}{2}\right) \Big|_{-\infty}^{\infty} = 0$ ; because it is converges to zero as x tends to infinity.

$$\therefore E(X^2) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp\left(-\frac{x^2}{2}\right) dx \quad (2.17)$$

$$= \frac{\sqrt{2\pi}}{\sqrt{2\pi}} \quad (2.18)$$

$$= 1 \quad (2.19)$$

Variance is given by

$$\text{var}[U] = E(U^2) - (E(U))^2 \quad (2.20)$$

$$\therefore \boxed{\text{var}[U] = 1} \quad (2.21)$$

### 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 following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/3.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 3.1.c -lm
$ ./a.out
```

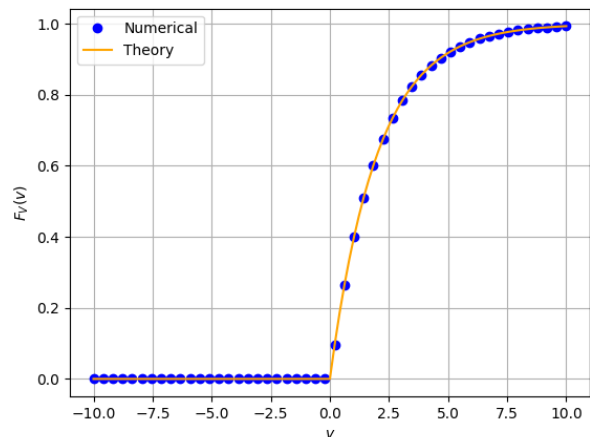


Fig. 3.1: The PDF of X

The CDF of V is plotted in Fig. 3.1 using the code below

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/3.1CDF.py>

Download the above files and execute the following commands to produce plot Fig.3.1

```
$ python3 3.1CDF.py
```

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

**Solution:** If  $Y = g(X)$ ,

W.K.T,

$$F_Y(y) = F_X(g^{-1}(y)) \quad (3.2)$$

$$V = -2 \ln(1 - U) \quad (3.3)$$

$$1 - U = e^{\frac{-V}{2}} \quad (3.4)$$

$$U = 1 - e^{\frac{-V}{2}} \quad (3.5)$$

$$F_V(x) = F_U(1 - e^{\frac{-x}{2}}) \quad (3.6)$$

$$\Rightarrow F_V(x) = \begin{cases} 0 & x < 0 \\ 1 - e^{\frac{-x}{2}} & x \geq 0 \end{cases} \quad (3.7)$$

#### 4 TRIANGULAR DISTRIBUTION

##### 4.1 Generate

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

**Solution:** Download the following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/4.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 4.1.c
$ ./a.out
```

##### 4.2 Find the CDF of $T$ .

**Solution:** The CDF of  $T$  is plotted in Fig. 4.2 using the code below

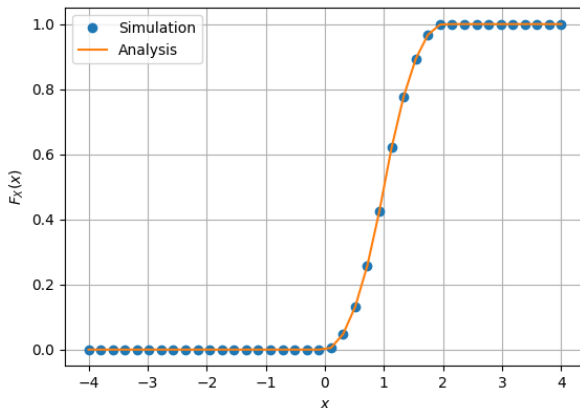


Fig. 4.2: The CDF of  $T$

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/4.2CDF.py>

Download the above files and execute the following commands to produce Fig.4.2

```
$ python3 4.2CDF.py
```

##### 4.3 Find the PDF of $T$ .

**Solution:** The PDF of  $T$  is plotted in Fig. 4.2 using the code below

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/4.3PDF.py>

Download the above files and execute the following commands to produce Fig.4.2

```
$ python3 4.3PDF.py
```

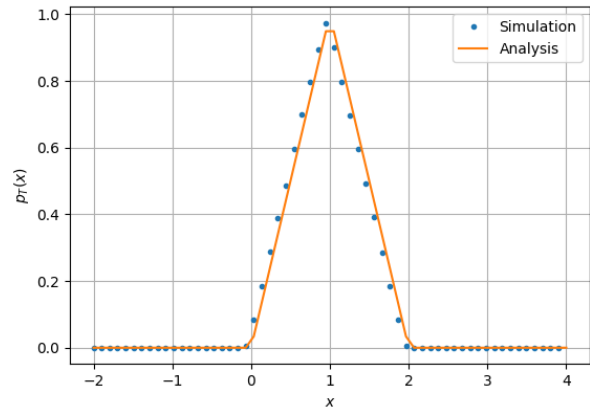


Fig. 4.3: The PDF of  $T$

##### 4.4 Find the Theoretical Expression for the PDF and CDF of $T$

**Solution:**

$$T = U_1 + U_2 \quad (4.2)$$

$$\Rightarrow p_T(t) = \int_{-\infty}^t p_{U1}(x)p_{U2}(y)dx \quad (4.3)$$

$$\text{As, } p_{U1}(x) = p_{U1}(y) = p_U(u) \quad (4.4)$$

$$\Rightarrow p_T(t) = \int_{-\infty}^t p_U(u)p_U(t-u)du \quad (4.5)$$

a) Theoretical PDF

i) For  $0 < t \leq 1$

$$p_T(t) = \int_0^t p_U(t-u)du \quad (4.6)$$

$$\Rightarrow p_T(t) = \int_0^t du = t \quad (4.7)$$

ii) For  $1 < t \leq 2$

$$p_T(t) = \int_0^1 p_U(t-u)du \quad (4.8)$$

$$\Rightarrow p_T(t) = \int_{t-1}^1 du = 2-t \quad (4.9)$$

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

b) Theoretical CDF

$$F_T(t) = \int_{-\infty}^t p_T(u)du \quad (4.10)$$

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

4.5 Verify your results through a plot

**Solution:** The above theoretical results are verified in the plots Fig 4.2 and Fig 4.3

## 5 MAXIMUM LIKELIHOOD

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

**Solution:** Download the following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/4.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 5.1.c
$ ./a.out
```

5.2 Generate

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

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

**Solution:** Download the following files and execute the C program.

<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/4.1.c>  
<https://github.com/bhargav0383/AI1110-Assignments/blob/main/RandomNumbers/codes/functions.h>

Execute the above C program files using the following commands

```
$ gcc 5.2.c
$ ./a.out
```

5.3 Plot  $Y$  using a scatter plot.

**Solution:** The CDF of  $V$  is plotted in Fig. 5.3 using the code below

`wget https://github.com/GovindaRohith/Assignments/blob/main/Randomnum/codes/5.3.py`

Download the above files and execute the following commands to produce Fig.5.3

```
$ python3 5.3.py
```

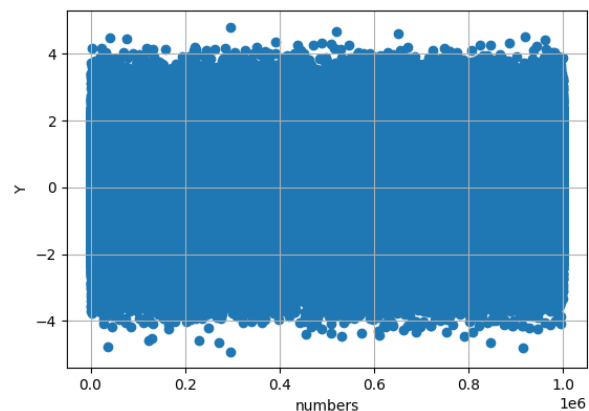


Fig. 5.3: The Scatter Plot

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

**Solution:**

- If  $Y < 0$  then probably  $X = -1$
- If  $Y > 0$  then probably  $X = 1$

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1) \quad (5.2)$$

and

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

**Solution:** Download the following files and execute the C program.

```
wget https://github.com/GovindaRohith/
  Assignments/blob/main/Randomnum/
  codes/5.5.c
wget https://github.com/GovindaRohith/
  Assignments/blob/main/Randomnum/
  codes/source.h
```

Download the above files and execute the following commands to get the result

```
$ gcc 5.5.c
$ ./a.out
```

$$P_{e|0} = 0.499035 \quad (5.4)$$

$$P_{e|1} = 0.500176 \quad (5.5)$$

5.6 Find  $P_e$  assuming that  $X$  has equiprobable symbols.

**Solution:**

$$P_e = P(X = 1)P_{e|0} + P(X = -1)P_{e|1} \quad (5.6)$$

Since  $X$  is equiprobable

$$P(X = 1) = P(X = -1) = 0.5 \quad (5.7)$$

$$\Rightarrow P_e = \frac{P_{e|0} + P_{e|1}}{2} \quad (5.8)$$

$$\Rightarrow P_e = Q_N(A) \quad (5.9)$$

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

**Solution:**

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1) \quad (5.10)$$

$$P_{e|0} = \Pr(AX + N < 0|X = 1) \quad (5.11)$$

$$P_{e|0} = \Pr(N < -A) \quad (5.12)$$

$$P_{e|0} = \int_{-\infty}^{-A} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \quad (5.13)$$

$$P_{e|0} = \int_A^{\infty} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \quad (5.14)$$

$$P_{e|0} = Q_N(A) \quad (5.15)$$

$$\text{Similarly, } P_{e|1} = Q_N(A) \quad (5.16)$$

```
wget https://github.com/GovindaRohith/
  Assignments/blob/main/Randomnum/
  codes/5.6.py
```

Download the above files and execute the following commands to produce Fig.5.7

```
$ python3 5.7.py
```

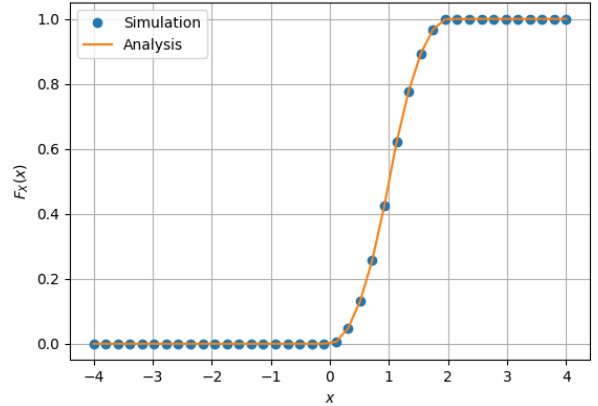


Fig. 5.7:  $P_e(A)$  with semilog-y axis

5.8 Now, consider a threshold  $\delta$  while estimating  $X$  from  $Y$ . Find the value of  $\delta$  that maximizes the theoretical  $P_e$ .

**Solution:** To estimate  $X$  from  $Y$ , we now consider the following:

$$X = \begin{cases} 1, & Y > \delta \\ -1, & Y < \delta \end{cases} \quad (5.17)$$

Therefore,

$$P_{e|0} = \Pr(\hat{X} = -1|X = 1) \quad (5.18)$$

$$= \Pr(AX + N < \delta|X = 1) \quad (5.19)$$

$$\Rightarrow P_{e|0} = \Pr(N < \delta - A) \quad (5.20)$$

$$= \int_{-\infty}^{\delta-A} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \quad (5.21)$$

$$= \int_{A-\delta}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \quad (5.22)$$

$$\Rightarrow P_{e|0} = Q_N(A - \delta) \quad (5.23)$$

Similarly,

$$P_{e|1} = Q_N(A + \delta) \quad (5.24)$$

$$P_e = P_{e|0} \Pr(X = 1) + P_{e|1} \Pr(X = -1) \quad (5.25)$$

$$= \frac{Q_N(A - \delta) + Q_N(A + \delta)}{2} \quad (5.26)$$

Differentiating the above equation wrt  $\delta$ :

$$0 = \frac{d}{d\delta} \left( \frac{Q_N(A - \delta) + Q_N(A + \delta)}{2} \right) \quad (5.27)$$

$$= \frac{1}{2} \left( \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta-A)^2}{2}} - \frac{1}{\sqrt{2\pi}} e^{-\frac{(A+\delta)^2}{2}} \right) \quad (5.28)$$

$$\Rightarrow (\delta - A)^2 = (A + \delta)^2 \quad (5.29)$$

$$\Rightarrow \boxed{\delta = 0} \quad (5.30)$$

5.9 Repeat the above exercise when

$$p_X(0) = p \quad (5.31)$$

**Solution:** Using Eq. (5.25), we have:

$$P_e = P_{e|0}p + P_{e|1}(1 - p) \quad (5.32)$$

$$= pQ_N(A - \delta) + (1 - p)Q_N(A + \delta) \quad (5.33)$$

Differentiating as before, we get:

$$0 = p \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta-A)^2}{2}} - (1 - p) \frac{1}{\sqrt{2\pi}} e^{-\frac{(A+\delta)^2}{2}} \quad (5.34)$$

$$e^{\frac{(\delta+A)^2}{2} - \frac{(\delta-A)^2}{2}} = \frac{1 - p}{p} \quad (5.35)$$

$$\Rightarrow \boxed{\delta = 0} \quad (5.36)$$

5.10 Repeat the above exercise using the MAP criterion.

**Solution:** Assume that

$$\Pr(X = -1) = p \quad (5.37)$$

$$\Pr(X = 1) = (1 - p) \quad (5.38)$$

From Total Probability Theorem, we have:

$$p_Y(y) = p_{Y|X=-1}(y|-1) \Pr(X = -1) + p_{Y|X=1}(y|1) \Pr(X = 1) \quad (5.39)$$

$$p_Y(y) = p \times p_{(-A+N)}(y) \quad (5.40)$$

$$+ (1 - p) \times p_{(A+N)}(y) \quad (5.41)$$

Now,  $p_{(-A+N)}$  is just the pdf of a shifted normal distribution, and therefore:

$$p_Y(y) = p \frac{e^{-\frac{(y+A)^2}{2}}}{\sqrt{2\pi}} + (1 - p) \frac{e^{-\frac{(y-A)^2}{2}}}{\sqrt{2\pi}} \quad (5.42)$$

To use the MAP criterion, we must find  $p_{X|Y}(x|y)$ . To do this, we use the Theorem of Conditional Probability:

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x) \times p_X(x)}{p_Y(y)} \quad (5.43)$$

When  $X = 1$ , we have:

$$p_{X|Y}(1|y) = \frac{p_{Y|X}(y|1) \times p_X(1)}{p_Y(y)} \quad (5.44)$$

$$= \frac{(1 - p) \frac{e^{-\frac{(y-A)^2}{2}}}{\sqrt{2\pi}}}{p \frac{e^{-\frac{(y+A)^2}{2}}}{\sqrt{2\pi}} + (1 - p) \frac{e^{-\frac{(y-A)^2}{2}}}{\sqrt{2\pi}}} \quad (5.45)$$

$$= \frac{(1 - p) e^{2yA}}{p + (1 - p) e^{2yA}} \quad (5.46)$$

Similarly, when  $X = -1$ , we get:

$$p_{X|Y}(-1|y) = \frac{p}{p + (1 - p) e^{2yA}} \quad (5.47)$$

Therefore, when  $p_{X|Y}(1|y) > p_{X|Y}(-1|y)$ , we have:

$$\frac{(1 - p) e^{2yA}}{p + (1 - p) e^{2yA}} > \frac{p}{p + (1 - p) e^{2yA}} \quad (5.48)$$

$$e^{2yA} > \frac{p}{(1 - p)} \quad (5.49)$$

$$y > \frac{1}{2A} \ln \frac{p}{(1 - p)} \quad (5.50)$$

Therefore, when Eq. (5.50), we can assert that  $X = 1$ , and  $X = -1$  otherwise. Now, consider



when  $p = \frac{1}{2}$ . We have:

$$y > \frac{1}{2A} \ln \frac{p}{(1-p)} \quad (5.51)$$

$$= \frac{1}{2A} \ln 1 \quad (5.52)$$

$$= 0 \quad (5.53)$$

Therefore, when  $y > 0$ , we choose  $X = 1$ , and we choose  $X = -1$  otherwise.

## 6 GAUSSIAN TO OTHER

6.1 Let  $X_1 \sim \mathcal{N}(0, 1)$   $X_2 \sim \mathcal{N}(0, 1)$ . Plot the CDF and PDF of

$$V = X_1^2 + X_2^2 \quad (6.1)$$

6.2 If

$$F_V(x) = \begin{cases} 1 - e^{-\alpha x} & x \geq 0 \\ 0 & x < 0, \end{cases} \quad (6.2)$$

find  $\alpha$ .

6.3 Plot the CDF and Pdf of

$$A = \sqrt{V} \quad (6.3)$$