

# Random Numbers

**Abstract**—This manual provides a simple introduction to the generation of 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:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/exrand.c
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

- 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 following code plots Fig. 1.2

```
https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/cdf_plot.py
```

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

**Solution:**

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

$$F_U(x) = \int_{-\infty}^x f_U(x) dx \quad (1.3)$$

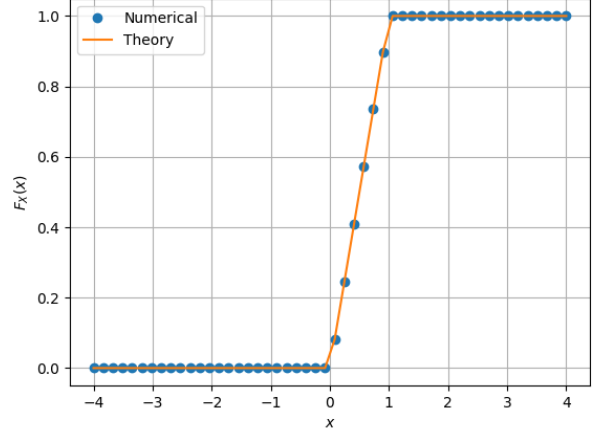


Fig. 1.2: The CDF of  $U$

Hence, If  $x \leq 0$ ,

$$F_U(x) = \int_{-\infty}^x f_U(x) dx \quad (1.4)$$

$$F_U(x) = \int_{-\infty}^x 0 dx \quad (1.5)$$

$$= 0 \quad (1.6)$$

If  $0 < x < 1$ ,

$$F_U(x) = \int_0^x f_U(x) dx \quad (1.7)$$

$$F_U(x) = \int_0^x 1 dx \quad (1.8)$$

$$= x \quad (1.9)$$

If  $x \geq 1$ ,

$$F_U(x) = \int_1^x f_U(x) dx \quad (1.10)$$

$$F_U(x) = \int_1^x 0 dx \quad (1.11)$$

$$= 0 \quad (1.12)$$

Hence,

$$F_U(x) = \begin{cases} x, & x \in (0, 1) \\ 0, & \text{otherwise} \end{cases} \quad (1.13)$$

1.4 The mean of  $U$  is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^N U_i \quad (1.14)$$

and its variance as

$$\text{var}[U] = E[U - E[U]]^2 \quad (1.15)$$

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

**Solution:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/mean_variance.c
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

$$\text{Mean obtained} = 0.500007 \quad (1.16)$$

$$\text{Variance obtained} = 0.083301 \quad (1.17)$$

1.5 Verify your result theoretically given that

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

**Solution:**

$$E[U^k] = 0 + \int_0^1 x^k dF_U(x) + 0 \quad (1.19)$$

$$= \frac{1}{k+1} \quad (1.20)$$

$$\Rightarrow E[U] = \frac{1}{2} \quad (1.21)$$

$$= 0.50 \quad (1.22)$$

$$E[U^2] = \frac{1}{3} \quad (1.23)$$

Hence,

$$\text{variance} = E[U^2] - (E[U])^2 \quad (1.24)$$

$$= \frac{1}{3} - \left(\frac{1}{2}\right)^2 \quad (1.25)$$

$$= \frac{1}{12} \quad (1.26)$$

$$= 0.0833 \quad (1.27)$$

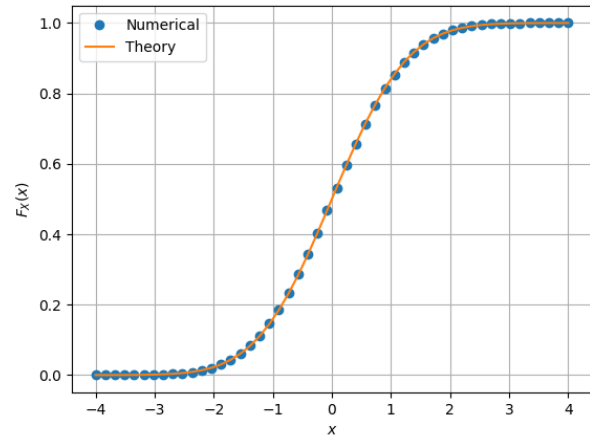


Fig. 2.2: The CDF of  $X$

## 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:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/exrand.c
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

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 code

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/cdf_plot.py
```

The CDF of a random variable  $U$  has the following properties:

- a)  $F_U(x)$  is a non decreasing function of  $x$  where  $-\infty < x < \infty$

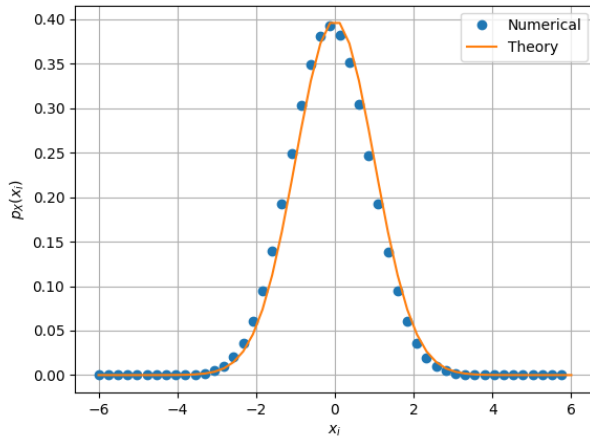


Fig. 2.3: The PDF of  $X$

b)  $F_U(x)$  ranges from 0 to 1

c)  $F_U(x) = 0$  as  $x \rightarrow -\infty$

d)  $F_U(x) = 1$  as  $x \rightarrow \infty$

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

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/pdf_plot.py
```

The PDF of a random variable  $X$  has the following properties:

a) The probability density function is non-negative for all the possible values.

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

c)  $f(x) = 0$  as  $x \rightarrow -\infty$

d)  $f(x) = 0$  as  $x \rightarrow \infty$

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

**Solution:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/mean_variance.c
```

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

$$\text{Mean obtained} = 0.000326 \quad (2.3)$$

$$\text{Variance obtained} = 1.000907 \quad (2.4)$$

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.5)$$

repeat the above exercise theoretically.

**Solution:** By definition,

$$p_X(x) dx = dF_U(x) \quad (2.6)$$

Hence,

$$E[U] = \int_{-\infty}^{\infty} x p_X(x) dx \quad (2.7)$$

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

$$= 0 \quad (2.9)$$

Also,

$$\text{variance} = E[U^2] - (E[U])^2 \quad (2.10)$$

$$E[U^2] = \int_{-\infty}^{\infty} x^2 p_X(x) dx \quad (2.11)$$

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

Using integration by parts, (2.13)

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

$$= \left[ -x \frac{1}{\sqrt{2\pi}} e^{\left(-\frac{x^2}{2}\right)} + \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \right] \quad (2.15)$$

$$= 0 + \frac{1}{\sqrt{2\pi}} \sqrt{2\pi} \quad (2.16)$$

$$= 1 \implies \text{variance} = 1 \quad (2.17)$$

Also,

$$F_U(x) = \int_{-\infty}^x p_X(x) dx \quad (2.18)$$

$$= \int_{-\infty}^x e^{-\frac{x^2}{2}} dx \quad (2.19)$$

### 3 FROM UNIFORM TO OTHER

#### 3.1 Generate samples of

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

and plot its CDF.

**Solution:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/exrand.c
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

The CDF of  $X$  is plotted in Fig. 3.2 using the code below

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/cdf_plot.py
```

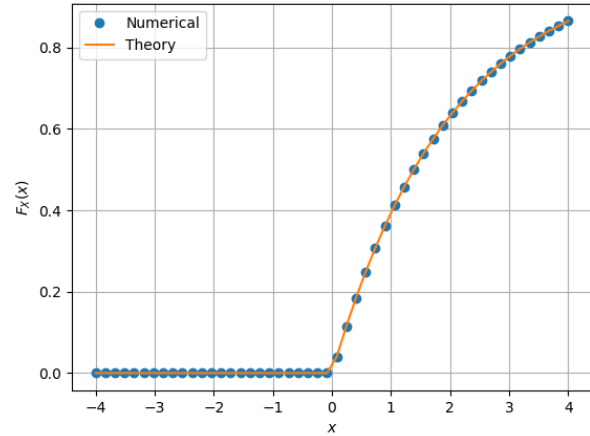


Fig. 3.2: CDF of  $V$

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/coeffs.h
```

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

**Solution:** Let,  $V = g(U)$

$$V = -2 \ln(1 - U) \quad (3.2)$$

$$\Rightarrow U = 1 - e^{-\frac{V}{2}} \quad (3.3)$$

Now,

$$F_V(x) = P(g(U) \leq x) \quad (3.4)$$

$$= P(X < g^{-1}(V)) \quad (3.5)$$

$$= F_U(g^{-1}(V)) \quad (3.6)$$

$$= F_U(1 - e^{-\frac{V}{2}}) \quad (3.7)$$

$$F_U(1 - e^{-\frac{V}{2}}) = \begin{cases} 1 - e^{-\frac{V}{2}}, & V \in (0, \infty) \\ 0, & \text{otherwise} \end{cases} \quad (3.8)$$

### 4 TRIANGULAR DISTRIBUTION

#### 4.1 Generate

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

**Solution:**

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/exrand.c
```

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

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

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

**Solution:**

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

$$\Rightarrow f_T(T) = f_{U_1+U_2}(T) \quad (4.3)$$

$$= f_{U_1}(T) * f_{U_2}(T) \quad (4.4)$$

$$= \int_{-\infty}^{\infty} f_{U_1}(T - U_2) f_{U_2}(U_2) dU_2 \quad (4.5)$$

since  $U_1$  and  $U_2$  are independent Also,

$$f_{U_1}(U_1) = f_{U_2}(U_2) = \begin{cases} 1, & x \in (0, 1) \\ 0, & \text{otherwise} \end{cases} \quad (4.6)$$

Hence for  $0 \leq U_2 \leq 1$ , Eq:(4.5) becomes,

$$f_T(T) = \int_0^1 f_{U_1}(T - U_2) dU_2 \quad (4.7)$$

If  $0 \leq T \leq 1$ ,

$$f_T(T) = \int_0^T dU_2 \quad (4.8)$$

$$= T \quad (4.9)$$

If  $1 < T \leq 2$ ,

$$f_T(T) = \int_{T-1}^1 dU_2 \quad (4.10)$$

$$= 2 - T \quad (4.11)$$

$$f_T(T) = \begin{cases} T, & 0 \leq T \leq 1 \\ 2 - T, & 1 < T \leq 2 \\ 0, & \text{otherwise} \end{cases} \quad (4.12)$$

Now,

$$F_T(T) = \int_{-\infty}^T f_T(T) dT \quad (4.13)$$

Hence, if  $T < 0$ ,

$$F_T(T) = \int_{-\infty}^T f_T(T) dT \quad (4.14)$$

$$= \int_{-\infty}^T 0 dT \quad (4.15)$$

$$= 0 \quad (4.16)$$

If  $0 \leq T \leq 1$ ,

$$F_T(T) = \int_0^T f_T(T) dT \quad (4.17)$$

$$= \int_0^T T dT \quad (4.18)$$

$$= \frac{T^2}{2} \quad (4.19)$$

If  $1 < T \leq 2$ ,

$$F_T(T) = \int_0^1 f_T(T) dT + \int_1^T f_T(T) dT \quad (4.20)$$

$$= \int_0^1 T dT + \int_1^T (2 - T) dT \quad (4.21)$$

$$= \frac{1}{2} + \frac{4T - T^2 - 3}{2} \quad (4.22)$$

$$= 1 - \frac{(2 - T)^2}{2} \quad (4.23)$$

If  $T > 2$ ,

$$F_T(T) = \int_2^T f_T(T) dT \quad (4.24)$$

$$= \int_2^T 0 dT \quad (4.25)$$

$$= 0 \quad (4.26)$$

$$F_T(T) = \begin{cases} \frac{T^2}{2}, & 0 \leq T \leq 1 \\ 1 - \frac{(2-T)^2}{2}, & 1 < T \leq 2 \\ 0, & \text{otherwise} \end{cases} \quad (4.27)$$

4.5 Verify your results through a plot.

**Solution:** The CDF of T is plotted in Fig. 4.5 below

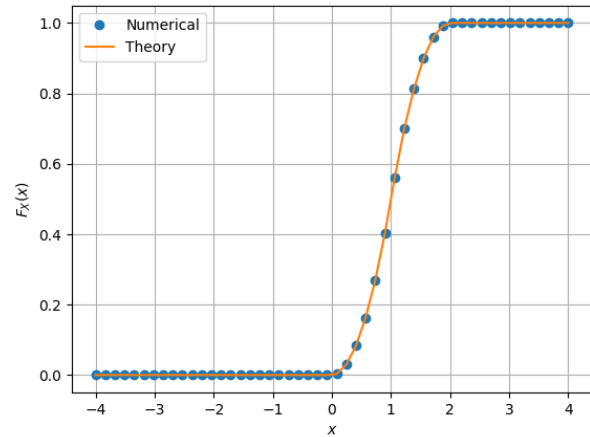


Fig. 4.5: CDF of T

The PDF of T is plotted in Fig. 4.5 below

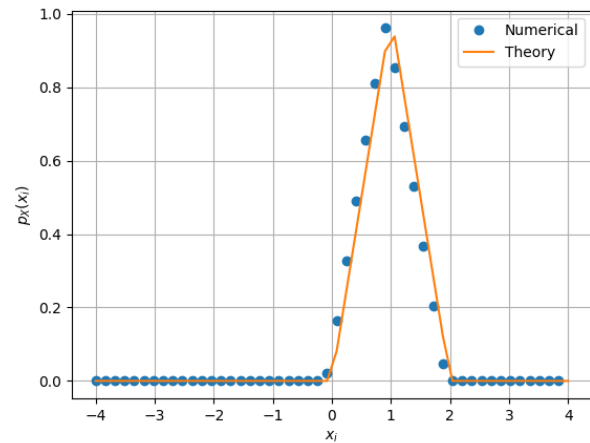


Fig. 4.5: PDF of T

Using codes

```
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
Random%20Variables%20Assignment/
codes/cdf_plot.py
wget https://github.com/tejalkul/AI1110-
Assignments/blob/main/AI1110%20
```

Random%20Variables%20Assignment/  
codes/pdf\_plot.py

## 5 MAXIMUM LIKELIHOOD

### 5.1 Generate

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

where  $A = 5 \text{ dB}$ ,  $X_1 \in \{1, -1\}$ , is Bernoulli and  $N \sim \mathcal{N}(0, 1)$ .

### 5.2 Plot $Y$ .

### 5.3 Guess how to estimate $X$ from $Y$ .

### 5.4 Find

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

and

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

### 5.5 Find $P_e$ .

### 5.6 Verify by plotting the theoretical $P_e$ .

## 6 GAUSSIAN TO OTHER

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

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

### 6.2 If

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

find  $\alpha$ .

**Solution:**  $X_1$  and  $X_2$  are i.i.d. Let,

$$X_1 = R \cos \theta \quad (6.3)$$

$$X_2 = R \sin \theta \quad (6.4)$$

Using Jacobian transform we have,

$$f_{X_1 X_2}(X_1, X_2) = \frac{1}{|J(R, \theta)|} f_{R\theta}(R, \theta) \quad (6.5)$$

Now, Jacobian matrix is given as follows,

$$J = \begin{pmatrix} \frac{\partial x_1}{\partial R} & \frac{\partial x_1}{\partial \theta} \\ \frac{\partial x_2}{\partial R} & \frac{\partial x_2}{\partial \theta} \end{pmatrix} \quad (6.6)$$

$$J = \begin{pmatrix} \cos \theta & -R \sin \theta \\ \sin \theta & R \cos \theta \end{pmatrix} \quad (6.7)$$

$$|J| = R \quad (6.8)$$

$$\Rightarrow f_{X_1 X_2}(X_1, X_2) = \frac{1}{R} f_{R\theta}(R, \theta) \quad (6.9)$$

Now,

$$f_{X_1 X_2}(X_1, X_2) = f_{X_1}(X_1) f_{X_2}(X_2) \quad (6.10)$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{x_1^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x_2^2}{2}} \quad (6.11)$$

$$= \frac{1}{2\pi} e^{-\frac{(x_1^2 + x_2^2)}{2}} \quad (6.12)$$

$$= \frac{1}{2\pi} e^{-\frac{R^2}{2}} \quad (6.13)$$

Hence,

$$f_{R\theta}(R, \theta) = \frac{R}{2\pi} e^{-\frac{R^2}{2}} \quad (6.14)$$

$$\Rightarrow f_R(R) = \int_0^{2\pi} \frac{R}{2\pi} e^{-\frac{R^2}{2}} d\theta \quad (6.15)$$

$$= R e^{-\frac{R^2}{2}} \quad (6.16)$$

Hence,

$$f_R(R) = \begin{cases} R e^{-\frac{R^2}{2}} & x \geq 0 \\ 0 & x < 0, \end{cases} \quad (6.17)$$

$$F_R(R) = \begin{cases} 1 - e^{-\frac{R^2}{2}} & x \geq 0 \\ 0 & x < 0, \end{cases} \quad (6.18)$$

Now,

$$F_V(x) = P(V \leq x) \quad (6.19)$$

$$= P(R^2 \leq x) \quad (6.20)$$

$$= P(R \leq \sqrt{x}) \quad (6.21)$$

$$= F_R(\sqrt{x}) \quad (6.22)$$

Hence,

$$F_V(x) = \begin{cases} 1 - e^{-\frac{x}{2}} & x \geq 0 \\ 0 & x < 0, \end{cases} \quad (6.23)$$

$$\Rightarrow \alpha = \frac{1}{2} \quad (6.24)$$

### 6.3 Plot the CDF and PDF of

$$A = \sqrt{V} \quad (6.25)$$

## 7 CONDITIONAL PROBABILITY

### 7.1

### 7.2 Plot

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

for

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

where  $A$  is Rayleigh with  $E[A^2] = \gamma$ ,  $N \sim 01$ ,  $X \in (-1, 1)$  for  $0 \leq \gamma \leq 10$  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 7.2 and 7.4 on the same graph w.r.t  $\gamma$ . Comment.

## 8 TWO DIMENSIONS

Let

$$\mathbf{y} = A\mathbf{x} + \mathbf{n}, \quad (8.1)$$

where

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

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

8.1 Plot

$$\mathbf{y}|\mathbf{s}_0 \text{ and } \mathbf{y}|\mathbf{s}_1 \quad (8.4)$$

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

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

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.