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## Assignment

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**Uniform Random Numbers** 

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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

$$F_U(x) = \Pr\left(U \le x\right) \tag{1.1}$$

**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

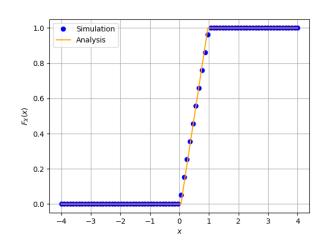


Fig. 1.2: The CDF of U

1.3 Find a theoretical expression for  $F_U(x)$ . Solution: Given U is a uniform Random Variable between 0 and 1,

$$F_U(x) = \Pr\left(U \le x\right) = \int_{-\infty}^x p_U(u) du \qquad (1.2)$$

We have three parts:

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

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

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

Therefore,

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \le x < 1 \\ 1 & x \ge 1 \end{cases}$$
 (1.4)

1.4 The mean of U is defined as

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

and its variance as

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

The actual analysis values,

$$Mean = 0.500007 \tag{1.7}$$

Variance = 
$$0.083301$$
 (1.8)

1.5 Verify your result theoretically given that

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

Solution: W.K.T,

$$var[U] = E[U^2] - E[U]^2$$
 (1.10)

Where E[U] is,

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

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

$$=\frac{1}{2}=0.5\tag{1.13}$$

And  $E[U^2]$  is,

$$E\left[U^{2}\right] = \int_{-\infty}^{\infty} x^{2} dF_{U}(x) \tag{1.14}$$

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

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

(1.17)

Hence finally,

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

The theoretically calculated values are,

Mean = 
$$0.5$$
 (1.19)

Variance = 
$$0.0833$$
 (1.20)

These values matches with the actual analysis values from above,

$$Mean = 0.500007 \tag{1.21}$$

Variance = 
$$0.083301$$
 (1.22)

Hence it verifies the result.

#### 2 Central Limit Theorem

2.1 Generate 10<sup>6</sup> 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:** 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

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

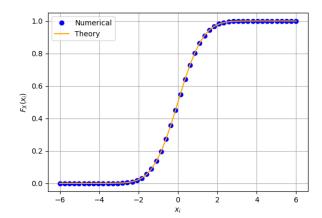


Fig. 2.2: The CDF of X

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

### \$ python3 2.2CDF.py

Some of the properties of CDF

- a)  $F_X(x)$  is non decreasing function.
- b) As  $x \to -\infty$ ,  $F_X(x) \to 0$  and when  $x \to \infty$ ,  $F_X(x) \to 1$
- c) 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

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

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

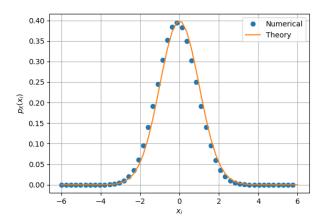


Fig. 2.3: The PDF of X

Some of the properties of the PDF:

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

Let  $x \sim \aleph(0, 1)$ . The Q-function Q(x) is defined as:

$$Q(x) = Pr(X > x) \tag{2.3}$$

$$= 1 - Pr(X \le x) \tag{2.4}$$

$$= 1 - F_X(x) \tag{2.5}$$

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,

$$Mean = 0.000294 \tag{2.6}$$

Variance = 
$$0.999561$$
 (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

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

W.K.T, 
$$F_X(x) = 1$$
 (2.10)

Mean is given by

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

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

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

$$E(x) = 0 \tag{2.13}$$

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

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

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

$$\int_{-\infty}^{\infty} x \times x \exp\left(-\frac{x^2}{2}\right) dx = -x \exp\left(-\frac{x^2}{2}\right) \Big|_{-\infty}^{\infty} (2.15)$$
$$+ \int_{-\infty}^{\infty} \exp\left(-\frac{x^2}{2}\right) dx (2.16)$$

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

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

$$=\frac{\sqrt{2\pi}}{\sqrt{2\pi}}\tag{2.18}$$

$$=1 \tag{2.19}$$

Variance is given by

$$var[U] = E(U^2) - (E(U))^2$$
 (2.20)

$$\therefore \text{ var } [U] = 1 \tag{2.21}$$

3 From Uniform to Other

3.1 Generate samples of

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

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

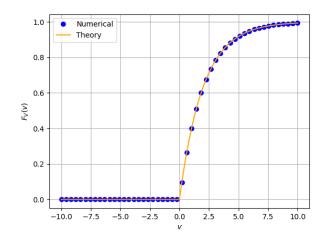


Fig. 3.1: The PDF of *X* 

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,

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

$$V = -2\ln(1 - U) \tag{3.3}$$

$$1 - U = e^{\frac{-V}{2}} \tag{3.4}$$

$$U = 1 - e^{\frac{-V}{2}} \tag{3.5}$$

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

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

#### 4 Triangular Distribution

#### 4.1 Generate

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

**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

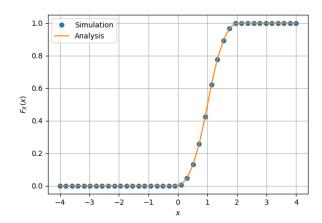


Fig. 4.2: The CDF of T

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

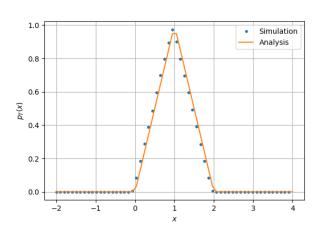


Fig. 4.3: The PDF of T

## 4.4 Find the Theoreotical Expression for the PDF and CDF of *T*

**Solution:** 

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

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

$$As, p_{U1}(x) = p_{U1}(y) = p_{U}(u)$$
 (4.4)

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

a) Theoretical PDF

i) For  $0 < t \le 1$ 

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

$$\implies p_T(t) = \int_0^t du = t \tag{4.7}$$

ii) For  $1 < t \le 2$ 

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

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

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

b) Theoretical CDF

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

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

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/5.1.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

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

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

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

https://github.com/bhargav0383/AI1110— Assignments/blob/main/RandomNumbers/codes/5.2.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

https://github.com/bhargav0383/AI1110— Assignments/blob/main/RandomNumbers/ codes/5.3scatt.py

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

\$ python3 5.3scatt.py

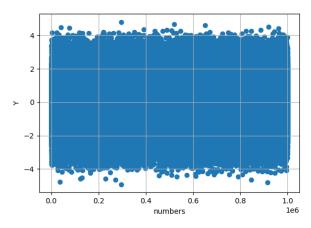


Fig. 5.3: The Scatter Plot

- 5.4 Guess how to estimate *X* from *Y*. **Solution:** 
  - a) If Y < 0 then probably X = -1
  - b) If Y > 0 then probably X = 1

5.5 Find

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

and

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

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

https://github.com/bhargav0383/AI1110— Assignments/blob/main/RandomNumbers/codes/5.5.c

https://github.com/bhargav0383/AI1110— Assignments/blob/main/RandomNumbers/codes/functions.h

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

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

$$P_{e|0} = 0.499035 \tag{5.4}$$

$$P_{e|1} = 0.500176 \tag{5.5}$$

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

#### **Solution:**

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

Since X is equiprobable

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

$$\implies P_e = \frac{P_{e|0} + P_{e|1}}{2} \tag{5.8}$$

$$\Longrightarrow \boxed{P_e = Q_N(A)} \tag{5.9}$$

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

**Solution:** 

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

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

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

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

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

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

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

https://github.com/bhargav0383/AI1110— Assignments/blob/main/RandomNumbers/ codes/5.7.py

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

\$ python3 5.7.py

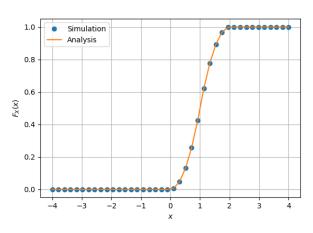


Fig. 5.7:  $P_e(A)$  with semilog-y axis

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:

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

Therefore,

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

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

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

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

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

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

Similarly,

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

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

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

Differentiating the above equation wrt  $\delta$ :

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

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

$$\implies (\delta - A)^2 = (\delta + A)^2 \tag{5.29}$$

$$\implies \boxed{\delta = 0} \tag{5.30}$$

5.9 Repeat the above exercise when

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

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

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

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

Differentiating as before, we get:

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

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

$$\implies \boxed{\delta = 0} \tag{5.36}$$

5.10 Repeat the above exercise using the MAP criterion.

**Solution:** Assume that

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

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

From Total Probability Theorem, we have:

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

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

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

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

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

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

When X = 1, we have:

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

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

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

Similarly, when X = -1, we get:

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

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

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

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

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

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

$$= \frac{1}{2A} \ln 1$$
 (5.52)  
= 0 (5.53)

$$=0 (5.53)$$

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

### 6 Gaussian to Other

6.1 Let  $X_1 \sim \aleph(0,1) \ X_2 \sim \aleph(0,1)$ . Plot the CDF and PDF of

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

6.2 If

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

find  $\alpha$ .

6.3 Plot the CDF and PDf of

$$A = \sqrt{V} \tag{6.3}$$