# Random Numbers

# Anshul Sangrame CS21BTECH11004

# **CONTENTS**

1 (	Jniform	Random	Numbers	1

- **2** Central Limit Theorem 2
- 3 From Uniform to Other 4
- 4 Triangular Distribution 4
- 5 Maximum Likelihood 6
- 6 Gaussian to Other 9

#### 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,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/1/ coeffs.h

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /1/1 1.c

Execute the above files using code,

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:**

Download the following files

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /1/1\_2.py

Execute the code using command

python3 1 2.py

Plot 1.1 is obtained,

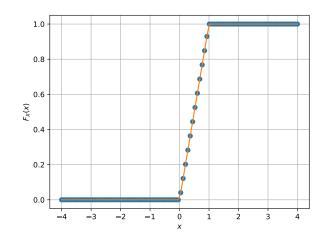


Fig. 1.1

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

#### **Solution:**

Pdf of Uniform distribution between [0,1] is given by,

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

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

Case-1: x < 0,

$$F_U(x) = \int_{-\infty}^x 0 dx \tag{1.4}$$

$$=0 (1.5)$$

Case-2:  $x \in [0,1]$ ,

$$F_U(x) = \int_{-\infty}^0 0 dx + \int_0^x 1 dx$$
 (1.6)

$$= x \tag{1.7}$$

Case-3: x > 1,

$$F_U(x) = \int_{-\infty}^0 0 dx + \int_0^1 1 dx + \int_1^x 0 dx \quad (1.8)$$
  
= 1 (1.9)

Hence,

$$F_U(x) = \begin{cases} 0, & x < 0 \\ x, & x \in [0, 1] \\ 1, & x > 1 \end{cases}$$
 (1.10)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.11)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.12)

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

# **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/1/ coeffs.h

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /1/1 4.c

Execute the above files using code,

The following result is obtained

$$E[U] = 0.500007$$
  
var $[U] = 0.083301$ 

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) dx \qquad (1.13)$$

**Solution:** 

 $F_U(x)$  for uniform distribution,

$$F_U(x) = \begin{cases} 0, & x < 0 \\ x, & x \in [0, 1] \\ 1, & x > 1 \end{cases}$$
 (1.14)

$$E[U] = \int_{-\infty}^{\infty} x dF_U(x)$$
 (1.15)

$$= \int_{-\infty}^{0} 0 dx + \int_{0}^{1} x dx + \int_{1}^{\infty} 0 dx \qquad (1.16)$$

$$=\frac{1}{2}$$
 (1.17)

$$E[U^2] = \int_{-\infty}^{\infty} x^2 dF_U(x) \tag{1.18}$$

$$= \int_{-\infty}^{0} 0 dx + \int_{0}^{1} x^{2} dx + \int_{1}^{\infty} 0 dx \quad (1.19)$$

$$=\frac{1}{3}$$
 (1.20)

$$E[U - E[U]]^{2} = E[U^{2}] - [E[U]]^{2}$$
 (1.21)

From Equation (1.17) and (1.20),

$$E[U - E[U]]^2 = \frac{1}{3} - \left(\frac{1}{2}\right)^2$$
 (1.22)

$$=\frac{1}{3}-\frac{1}{4}\tag{1.23}$$

$$= \frac{1}{12} \approx 0.083 \tag{1.24}$$

# 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,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/1/ coeffs.h

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /2/2 1.c Execute the above files using code,

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 files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /2/2\_2.py

Execute the above file using code,

Plot 2.1 obtained is symmetric about (0,0.5)

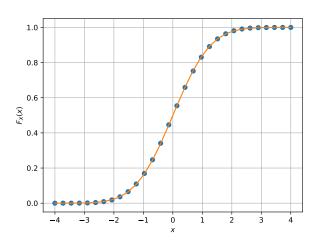


Fig. 2.1

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:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /2/2\_3.py

Execute the above file using code,

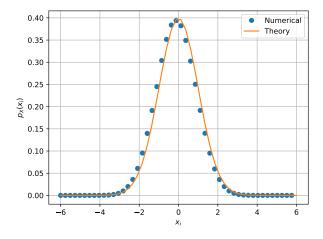


Fig. 2.2

Plot 2.2 obtained is symmetric about y-axis 2.4 Find the mean and variance of *X* by writing a C program.

### **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/soltion/1/ coeffs.h

wget https://github.com/Anshul-Sangrame/AI1110/blob/main/Assignment/solution/2/2\_4.c

Execute the above files using code,

The result is,

$$E[U] = 0.000294$$
 (2.3)

$$var[U] = 0.999561$$
 (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:**

$$E[U] = \int_{-\infty}^{\infty} u p_X(u) du \qquad (2.6)$$

 $up_X(u)$  is a odd function.

$$E[U] = 0 \tag{2.8}$$

(2.9)

$$E\left[U^{2}\right] = \int_{-\infty}^{\infty} \frac{u^{2}}{\sqrt{2\pi}} \exp\left(-\frac{u^{2}}{2}\right) du \qquad (2.10)$$
$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} u \left(u \exp\left(-\frac{u^{2}}{2}\right)\right) du \quad (2.11)$$

Using integration by parts,

$$E\left[U^{2}\right] = -u\frac{1}{\sqrt{2\pi}}\exp\left(-\frac{u^{2}}{2}\right)\Big|_{-\infty}^{\infty} + \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}}\exp\left(-\frac{u^{2}}{2}\right)$$
(2.12)

$$= 0 + \frac{1}{\sqrt{2\pi}} \sqrt{2\pi} \tag{2.13}$$

$$= 1 \tag{2.14}$$

Hence,

$$var[U] = E[U^{2}] - [E[U]]^{2}$$
 (2.15)  
= 1 (2.16)

#### 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,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/ coeffs.h

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /3/3\_1.c

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /3/3 1.py

Execute the above files using code,

# Plot 3.1 is obtained

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

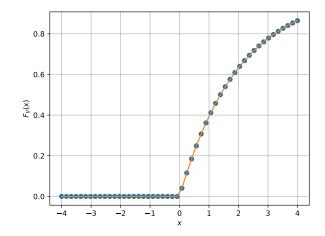


Fig. 3.1

# **Solution:**

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

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

$$= Pr\left(U \le 1 - e^{-\frac{x}{2}}\right) \tag{3.4}$$

$$F_V(x) = F_U \left( 1 - e^{-\frac{x}{2}} \right) \tag{3.5}$$

$$= \begin{cases} 0, & 1 - e^{-\frac{x}{2}} < 0\\ 1 - e^{-\frac{x}{2}}, & 1 - e^{-\frac{x}{2}} \in [0, 1]\\ 1, & 1 - e^{-\frac{x}{2}} > 1 \end{cases}$$
 (3.6)

Now,

$$1 - e^{-\frac{x}{2}} < 0 \tag{3.7}$$

$$\implies x < 0$$
 (3.8)

$$0 \le 1 - e^{-\frac{x}{2}} \le 1 \tag{3.9}$$

$$\implies x \ge 0$$
 (3.10)

$$1 < 1 - e^{-\frac{x}{2}} \tag{3.11}$$

$$\implies x \in \phi$$
 (3.12)

Hence,

$$F_V(x) = \begin{cases} 0, & x < 0 \\ 1 - e^{-\frac{x}{2}}, & x \ge 0 \end{cases}$$
 (3.13)

# 4 Triangular Distribution

#### 4.1 Generate

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

# **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/ coeffs.h

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /4/4\_1.c

Execute the above files using code,

4.2 Find the CDF of T.

# **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /4/4 2.py

Execute the above files using code,

Plot 4.1 is obtained

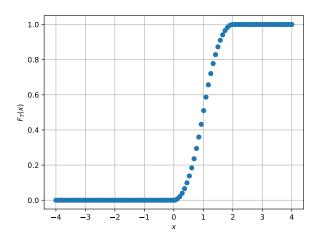


Fig. 4.1

#### 4.3 Find the PDF of T.

# **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /4/4\_3.py

Execute the above files using code,

# python3 4\_3.py

Plot 4.2 is obtained

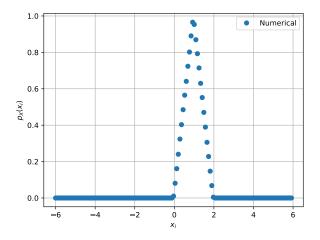


Fig. 4.2

# 4.4 Find the theoretical PDF and CDF of T.

# **Solution:**

The CDF of T is given by

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

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

Now, fix the value of  $U_1$  to be some x

$$x + U_2 \le t \implies U_2 \le t - x \tag{4.3}$$

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

$$F_T(t) = \int_0^t \Pr(U_2 \le t - x) \, p_{U_1}(x) \mathrm{d}x \qquad (4.4)$$

$$= \int_0^t F_{U_2}(t-x)p_{U_1}(x)\mathrm{d}x \tag{4.5}$$

$$0 \le x \le t \implies 0 \le t - x \le t \le 1$$
 (4.6)

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

$$F_T(t) = \int_0^t (t - x) \cdot 1 \cdot \mathrm{d}x \tag{4.8}$$

$$= tx - \frac{x^2}{2} \bigg|_0^t \tag{4.9}$$

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

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

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

$$0 \le x \le t - 1 \implies 1 \le t - x \le t$$

$$t - 1 \le x \le 1 \implies 0 < t - 1 \le t - x \le 1$$

$$(4.12)$$

$$(4.13)$$

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

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

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

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

Therefore,

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

The PDF of T is given by

$$p_T(t) = \frac{\mathrm{d}}{\mathrm{d}t} F_T(t) \tag{4.19}$$

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

4.5 Verify your results through a plot.

#### **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /4/4\_5\_1.py

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /4/4 5 2.py

Execute the above files using code,

python3 4 5 2.py

Plot 4.3 is obtained is pdf Plot 4.4 is obtained

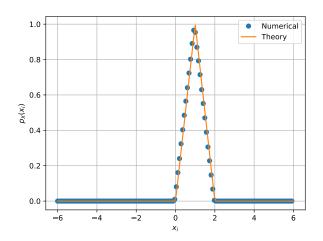


Fig. 4.3

is cdf

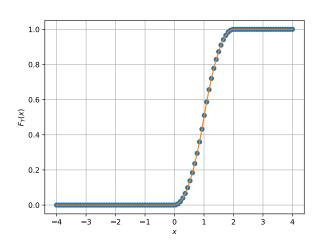


Fig. 4.4

#### 5 Maximum Likelihood

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

#### **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.1.c

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/ coeffs.h Execute the above files using code,

#### 5.2 Generate

$$Y = AX + N \tag{5.1}$$

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

# **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.2.c

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution/ coeffs.h

Execute the above files using code,

5.3 Plot Y using a scatter plot.

#### **Solution:**

Download the following files,

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.3.py

Execute the above files using code,

python3 5.3.py

Plot 5.1 is obtained

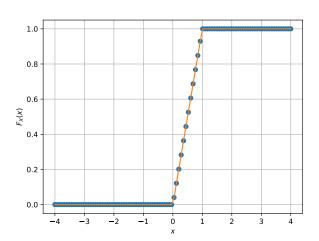


Fig. 5.1

5.4 Guess how to estimate X from Y.

### **Solution:**

From the plot of Y, we see that the estimate model can be written as

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

**5.5** Find

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

and

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

# **Solution:**

Letting X = 1 and X = -1 respectively, we see the number of mismatched data points to compute the error probabilities. The simulation is coded in

\$ wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.5.py

and can be run by typing

\$ python3 5.5.py

The results are

$$P_{e|0} = 5.02 \times 10^{-4} \tag{5.5}$$

$$P_{e|1} = 5.24 \times 10^{-4} \tag{5.6}$$

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

# **Solution:**

Here, Pr(X = 1) = Pr(X = -1) = 0.5. Thus,

$$P_e = \Pr(X = 1) P_{e|1} + \Pr(X = -1) P_{e|0}$$
 (5.7)

$$= \frac{1}{2} \left( P_{e|0} + P_{e|1} \right) = 5.13 \times 10^{-4} \tag{5.8}$$

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

### **Solution:**

$$P_{e|0} = \Pr(\hat{X} = 1|X = -1)$$
 (5.9)

$$= \Pr(Y > 0 | X = -1) \tag{5.10}$$

$$= \Pr(AX + N > 0 | X = -1) \tag{5.11}$$

$$= \Pr(N > A) = Q(A)$$
 (5.12)

since X and N are independent. Writing a

similar expression for  $P_{e|1}$  and noting that

$$Pr(N < -A) = Pr(N > A) = Q(A)$$
 (5.13)

it follows that  $P_e = Q(A)$ . This is the idea used to plot the theoretical  $P_e$ . The plot is coded both in the rectangular axes and the semilog-y axes. Download the relevant codes using

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.7.py

wget https://github.com/Anshul-Sangrame/ AI1110/blob/main/Assignment/solution /5/5.7\_semilog.py

Execute the above files using code,

Plot 5.2 obtained in rectangular axis

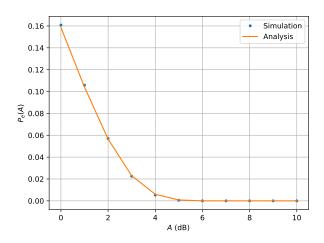


Fig. 5.2

Plot 5.3 obtained in 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:

$$X = \begin{cases} 1, & Y > \delta \\ -1, & Y < \delta \end{cases} \tag{5.14}$$

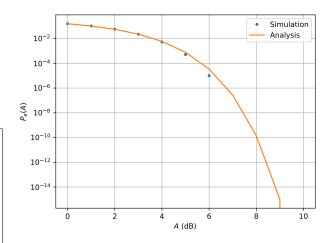


Fig. 5.3

Therefore,

$$P_{e} = \Pr(X = -1) Q(A + \delta) + \Pr(X = 1) Q(A - \delta)$$

$$= \frac{1}{2} (Q(A + \delta) + Q(A - \delta))$$
(5.15)

To minimise  $P_e$ , we differentiate the above equation wrt  $\delta$ :

$$0 = \frac{d}{d\delta} \left( \frac{1}{2} (Q_N(A - \delta) + Q_N(A + \delta)) \right)$$

$$= \frac{1}{2} \left( \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta - A)^2}{2}} - \frac{1}{\sqrt{2\pi}} e^{-\frac{(A + \delta)^2}{2}} \right)$$
(5.18)

Therefore,

$$(\delta - A)^2 = (\delta + A)^2$$

$$\implies \delta = 0$$
(5.19)
$$(5.20)$$

5.9 Repeat the above exercise when

$$p_{x}(0) = p$$

**Solution:** 

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

$$= pQ_N(A-\delta) + (1-p)Q_N(A+\delta)$$
 (5.22)

Differentiating as before, we get:

$$0 = p \frac{1}{\sqrt{2\pi}} e^{-\frac{(\delta - A)^2}{2}} - (1 - p) \frac{1}{\sqrt{2\pi}} e^{-\frac{(A + \delta)^2}{2}}$$
(5.23)

Taking In on both sides we have:

$$\ln p - \frac{(\delta - A)^2}{2} = \ln 1 - p - \frac{(\delta + A)^2}{2} \quad (5.24)$$

$$\implies 2\delta A = \ln \frac{1-p}{p} \tag{5.25}$$

$$\implies \delta = \frac{1}{2A} \ln \frac{1-p}{p} \tag{5.26}$$

5.10 Repeat the above exercise using the MAP criterion.

# **Solution:**

Using theorem of conditional probabilities,

$$p_{X|Y}(x|y) = \frac{p_{Y|X}(y|x) \times p_X(x)}{p_Y(y)}$$
 (5.27)

$$p_{X|Y}(x|y) = \frac{p_X(x)Pr(Ax + N = y)}{p_Y(y)}$$

$$p_{X|Y}(x|y) = \frac{p_X(x)Pr(N = y - Ax)}{p_Y(y)}$$

$$p_{X|Y}(x|y) = \frac{p_X(x)Pr(N = y - Ax)}{p_Y(y)}$$

$$p_{X|Y}(x|y) = \frac{p_X(x)p_N(y - Ax)}{p_Y(y)}$$
(5.29)

$$p_{X|Y}(x|y) = \frac{p_X(x)Pr(N = y - Ax)}{p_Y(y)}$$
 (5.29)

$$p_{X|Y}(x|y) = \frac{p_X(x)p_N(y - Ax)}{p_Y(y)}$$
 (5.30)

$$p_{X|Y}(x|y) = \frac{\frac{1}{\sqrt{2\pi}} p_X(x) e^{-\frac{(y-Ax)^2}{2}}}{p_Y(y)}$$
 (5.31)

(5.32)

When X = 1, we have:

$$p_{X|Y}(1|y) = \frac{p_{Y|X}(y|1) \times p_X(1)}{p_Y(y)}$$
 (5.33)

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

$$(5.34)$$

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

Similarly when X = -1,

$$p_{X|Y}(-1|y) = \frac{p}{p + (1-p)e^{2yA}}$$
 (5.36)

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

have:

$$\frac{(1-p)e^{2yA}}{p+(1-p)e^{2yA}} > \frac{p}{p+(1-p)e^{2yA}}$$
 (5.37)

$$e^{2yA} > \frac{p}{(1-p)} \tag{5.38}$$

$$y > \frac{1}{2A} \ln \frac{p}{(1-p)}$$
 (5.39)

Therefore, when Eq. (5.39), we can assert that X = 1, and X = -1 otherwise.

# 6 Gaussian to Other

6.1 gaytam singh