

# Probability

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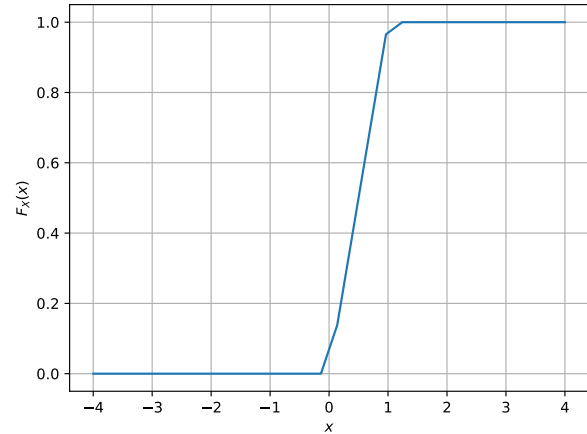


Fig. 1.2. The CDF of  $U$

## 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.

```
codes / coeffs . h
codes / uni _ var _ mean . c
```

- 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

$$F_U(x) = \Pr(U \leq x) \quad (1.2.1)$$

**Solution:** The following code plots Fig. 1.2

```
codes / uni _ cdf _ plot . py
```

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

**Solution:** Probability density function is given by:

$$\begin{cases} 1 & \text{for } U \in (0, 1) \\ 0 & \text{otherwise} \end{cases}$$

CDF of random variable  $U$  in range  $\in [0,1)$

$$F_U(x) = \int_0^U f(x)dx \quad (1.3.1)$$

$$= \int_0^U \frac{1}{1-0}dx \quad (1.3.2)$$

$$= \frac{[U]_0^U}{1-0} \quad (1.3.3)$$

$$= \frac{U}{1} \quad (1.3.4)$$

Cumulative density function:

$$\begin{cases} 0 & \text{for } U < 0 \\ U & \text{for } U \in [0, 1) \\ 1 & \text{for } U \geq b \end{cases}$$

- 1.4 The mean of  $U$  is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^N U_i \quad (1.4.1)$$

and its variance as

$$\text{var}[U] = E[U - E[U]]^2 \quad (1.4.2)$$

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

**Solution:**

codes/uni\_var\_mean.c

1.5 Verify your result theoretically given that

$$E[U^k] = \int_{-\infty}^{\infty} x^k dF_U(x) \quad (1.5.1)$$

**Solution:**

$$\begin{aligned} &= E[U - E[U]]^2 \\ &= E[U^2 - 2UE[U] + E[U]^2] \\ &= E[U^2] - E[U]^2 \end{aligned}$$

$$\text{Given, } E[U]^k = \int_{-\infty}^{\infty} x^k dF_U(x)$$

For  $E[U]$ ,  $k=1$  and limits are from 0-1.

$$\begin{aligned} E[U] &= \int_0^1 x dF_U(x) \\ &= \int_0^1 x dx \end{aligned}$$

$dF_U(x)$  for uniformly distributed function is  $\frac{1}{b-a}$  where  $b=1$  and  $a=0$ .

$$\begin{aligned} &= \frac{1}{1-0} \int_0^1 x dx \\ &= \frac{1}{1-0} * \left[ \frac{x^2}{2} \right]_0^1 \\ &= \frac{1}{2} = 0.5. \end{aligned}$$

$$\begin{aligned} E[U^2] &= \frac{1}{1-0} \int_0^1 x^2 dx \\ &= \frac{1}{1-0} * \left[ \frac{x^3}{3} \right]_0^1 \\ &= \frac{1}{3}. \end{aligned}$$

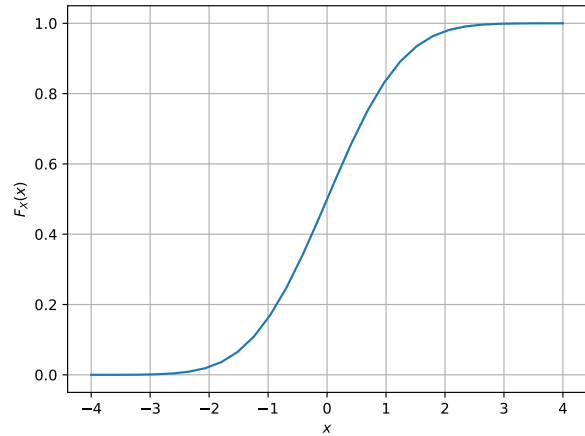


Fig. 2.2. The CDF of  $X$

Variance will be

$$\begin{aligned} \text{var}[U] &= E[U^2] - E[U]^2 \\ &= \frac{1}{3} - \left[ \frac{1}{2} \right]^2 \\ &= \frac{1}{12} \\ &= 0.0833 \end{aligned}$$

## 2 CENTRAL LIMIT THEOREM

2.1 Generate  $10^6$  samples of the random variable

$$X = \sum_{i=1}^{12} U_i - 6 \quad (2.1.1)$$

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:**

codes/coeffs.h  
codes/gauss\_var\_mean.c

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

codes/gauss\_cdf\_plot.py

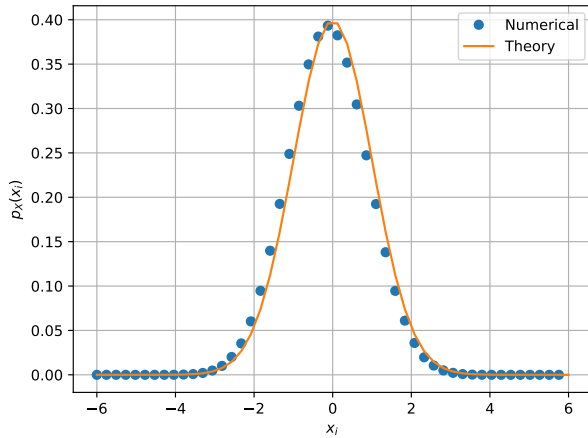


Fig. 2.3. The PDF of  $X$

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

$$p_X(x) = \frac{d}{dx} F_X(x) \quad (2.3.1)$$

What properties does the PDF have?

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

```
codes / gauss_pdf_plot . py
```

Properties of PDF is

1.  $P(x) = \int_a^b f(x)dx$
2.  $\int_{-\infty}^{\infty} f(x)dx = 1$
3.  $f(x)$  is non-negative for all possible values.

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

**Solution:** The code give below gives the mean and variance of  $X$

```
codes / gauss_var_mean . c
```

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.5.1)$$

repeat the above exercise theoretically.

**Solution:** We have

### 3 FROM UNIFORM TO OTHER

3.1 Generate samples of

$$V = -2 \ln(1 - U) \quad (3.1.1)$$

and plot its CDF.

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

### 4 TRIANGULAR DISTRIBUTION

4.1 Generate

$$T = U_1 + U_2 \quad (4.1.1)$$

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  $T$ .

4.5 Verify your results through a plot.

### 5 MAXIMUM LIKELIHOOD

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

5.2 Generate

$$Y = AX + N, \quad (5.2.1)$$

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

5.3 Plot  $Y$  using a scatter plot.

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

5.5 Find

$$P_{e|0} = \Pr(\hat{X} = -1 | X = 1) \quad (5.5.1)$$

and

$$P_{e|1} = \Pr(\hat{X} = 1 | X = -1) \quad (5.5.2)$$

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

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

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

5.9 Repeat the above exercise when

$$p_X(0) = p \quad (5.9.1)$$

5.10 Repeat the above exercise using the MAP criterion.

## 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

$$V = X_1^2 + X_2^2 \quad (6.1.1)$$

- 6.2 If

$$F_V(x) = \begin{cases} 1 - e^{-\alpha x} & x \geq 0 \\ 0 & x < 0, \end{cases} \quad (6.2.1)$$

find  $\alpha$ .

- 6.3 Plot the CDF and PDF of

$$A = \sqrt{V} \quad (6.3.1)$$

## 7 CONDITIONAL PROBABILITY

- 7.1 Plot

$$P_e = \Pr(\hat{X} = -1 | X = 1) \quad (7.1.1)$$

for

$$Y = AX + N, \quad (7.1.2)$$

where  $A$  is Rayleigh with  $E[A^2] = \gamma$ ,  $N \sim \mathcal{N}(0, 1)$ ,  $X \in (-1, 1)$  for  $0 \leq \gamma \leq 10$  dB.

- 7.2 Assuming that  $N$  is a constant, find an expression for  $P_e$ . Call this  $P_e(N)$

- 7.3 For a function  $g$ ,

$$E[g(X)] = \int_{-\infty}^{\infty} g(x)p_X(x)dx \quad (7.3.1)$$

Find  $P_e = E[P_e(N)]$ .

- 7.4 Plot  $P_e$  in problems 7.1 and 7.3 on the same graph w.r.t  $\gamma$ . Comment.

## 8 TWO DIMENSIONS

Let

$$\mathbf{y} = A\mathbf{x} + \mathbf{n}, \quad (4.1)$$

where

$$\mathbf{x} \in (\mathbf{s}_0, \mathbf{s}_1), \mathbf{s}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{s}_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad (4.2)$$

$$\mathbf{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix}, n_1, n_2 \sim \mathcal{N}(0, 1). \quad (4.3)$$

- 8.1 Plot

$$\mathbf{y}|\mathbf{s}_0 \text{ and } \mathbf{y}|\mathbf{s}_1 \quad (8.1.1)$$

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

$$P_e = \Pr(\hat{\mathbf{x}} = \mathbf{s}_1 | \mathbf{x} = \mathbf{s}_0) \quad (8.3.1)$$

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.