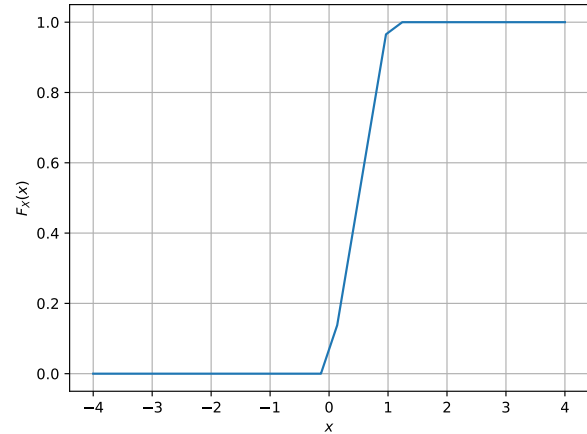# Probability

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Contents

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.

```
wget https://github.com/
    gadepall/probability/raw/
    master/manual/codes/exrand.c
wget https://github.com/
    gadepall/probability/raw/
    master/manual/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

$$F_U(x) = \Pr(U \le x) \quad (1.0.1)$$

Solution: The following code plots Fig. 1.2

```
wget https://github.com/
    gadepall/probability/raw/
    master/manual/codes/cdf_plot
    .py
```

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

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i \quad (1.0.2)$$

and its variance as

$$\text{var}[U] = E[U - E[U]]^2 \quad (1.0.3)$$

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

1.5 Verify your result theoretically given that

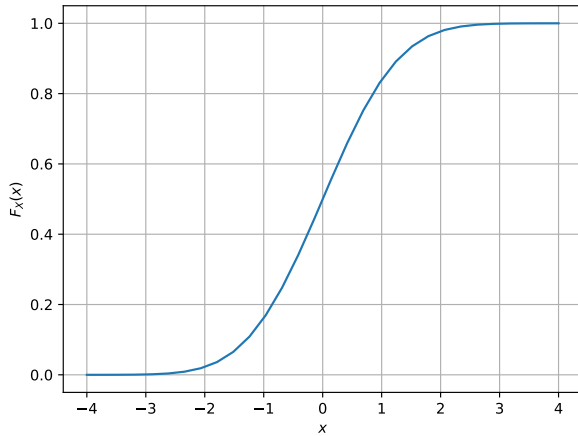$$E[U^k] = \int_{-\infty}^{\infty} x^k dF_U(x) \quad (1.0.4)$$

2 Central Limit Theorem

2.1 Generate 10^6 samples of the random variable

$$X = \sum_{i=1}^{12} U_i - 6 \quad (2.0.1)$$

using a C program, where $U_i, i = 1, 2, \ldots, 12$ are a set of independent uniform random variables between 0 and 1 and save in a file called gau.dat

2.2 Load gau.dat in python and plot the empirical CDF of X using the samples in gau.dat. What

Fig. 2.2. The CDF of X

Fig. 2.3. The PDF of X

properties does a CDF have?

Solution: The CDF of X is plotted in Fig. 2.2

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

$$p_X(x) = \frac{d}{dx} F_X(x) \qquad (2.0.2)$$

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/
   gadepall/probability/raw/
   master/manual/codes/pdf_plot
   .py
```

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

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty,$$
$$(2.0.3)$$

repeat the above exercise theoretically.
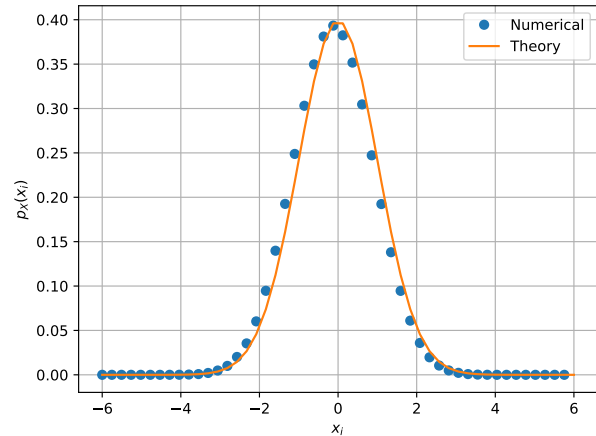
3 From Uniform to Other

3.1 Generate samples of

$$V = -2\ln(1 - U) \qquad (3.0.1)$$

and plot its CDF.

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

4 Triangular Distribution

4.1 Generate

$$T = U_1 + U_2 \qquad (4.0.1)$$

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

$$Y = AX + N, \qquad (5.0.1)$$

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

$$P_{e|0} = \Pr\left(\hat{X} = -1 | X = 1\right) \qquad (5.0.2)$$

and

$$P_{e|1} = \Pr\left(\hat{X} = 1 | X = -1\right) \qquad (5.0.3)$$

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 δ while estimating X from Y. Find the value of δ that maximizes the theoretical P_e.

5.9 Repeat the above exercise when

$$p_X(0) = p \qquad (5.0.4)$$

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

$$V = X_1^2 + X_2^2 \qquad (6.0.1)$$

6.2 If

$$F_V(x) = \begin{cases} 1 - e^{-\alpha x} & x \geq 0 \\ 0 & x < 0, \end{cases} \qquad (6.0.2)$$

find α.

6.3 Plot the CDF and PDf of

$$A = \sqrt{V} \qquad (6.0.3)$$

7 CONDITIONAL PROBABILITY

7.1 Plot

$$P_e = \Pr\left(\hat{X} = -1 | X = 1\right) \qquad (7.0.1)$$

for

$$Y = AX + N, \qquad (7.0.2)$$

where A is Raleigh with $E\left[A^2\right] = \gamma, N \sim \mathcal{N}(0,1), X \in (-1,1)$ 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,

$$E\left[g(X)\right] = \int_{-\infty}^{\infty} g(x)p_X(x)\,dx \qquad (7.0.3)$$

Find $P_e = E\left[P_e(N)\right]$.

7.4 Plot P_e in problems 7.1 and 7.3 on the same graph w.r.t γ. Comment.

8 TWO DIMENSIONS

Let

$$\mathbf{y} = A\mathbf{x} + \mathbf{n}, \qquad (8.0.1)$$

where

$$x \in (\mathbf{s}_0, \mathbf{s}_1), \mathbf{s}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{s}_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \qquad (8.0.2)$$

$$\mathbf{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix}, n_1, n_2 \sim \mathcal{N}(0,1). \qquad (8.0.3)$$

8.1 Plot

$$\mathbf{y}|\mathbf{s}_0 \text{ and } \mathbf{y}|\mathbf{s}_1 \qquad (8.0.4)$$

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

$$P_e = \Pr(\hat{\mathbf{x}} = \mathbf{s}_1 | \mathbf{x} = \mathbf{s}_0) \qquad (8.0.5)$$

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.