

# Random Numbers

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## 0.1 Uniform Random Numbers

Let  $U$  be a uniform random variable between 0 and 1.

0.1.1 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/include/coeffs.h
codes/src/uni_gen_stat.c
```

0.1.2 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

$$F_U(x) = \Pr(U \leq x) \quad (0.1.2.1)$$

**Solution:** The following code plots Fig. 0.1.2.1

```
codes/src/cdf_plot_uni.py
```

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

**Solution:**

$$F_U(x) = \int_{-\infty}^x f_U(x) dx \quad (0.1.3.1)$$

For the uniform random variable  $U$ ,  $f_U(x)$  is given by

$$f_U(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{elsewhere} \end{cases} \quad (0.1.3.2)$$

Substituting (0.1.3.2) in (0.1.3.1),  $F_U(x)$  is found to be

$$F_U(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq 1 \\ 1 & x > 1 \end{cases} \quad (0.1.3.3)$$

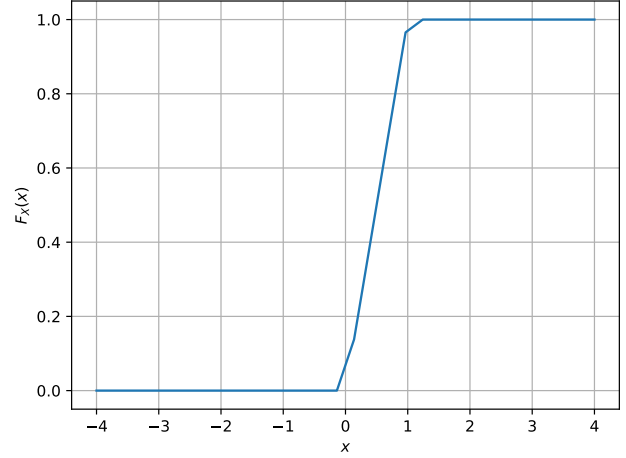


Figure 0.1.2.1: The CDF of  $U$

0.1.4 The mean of  $U$  is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^N U_i \quad (0.1.4.1)$$

and its variance as

$$\text{var}[U] = E[U - E[U]]^2 \quad (0.1.4.2)$$

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

**Solution:** The following code prints the mean and variance of  $U$

```
codes/src/uni_gen_stat.c
```

0.1.5 Verify your result theoretically given that

$$E[U^k] = \int_{-\infty}^{\infty} x^k dF_U(x) \quad (0.1.5.1)$$

**Solution:** For a random variable  $X$ , the mean  $\mu_X$  and variance  $\sigma_X^2$  are given by

$$\mu_X = E[X] = \int_{-\infty}^{\infty} x dF_U(x) \quad (0.1.5.2)$$

$$\sigma_X^2 = E[X^2] - \mu_X^2 = \int_{-\infty}^{\infty} x^2 dF_U(x) - \mu_X^2 \quad (0.1.5.3)$$

Substituting the CDF of  $U$  from (0.1.3.3) in (0.1.5.2) and

(0.1.5.3), we get

$$\mu_U = \frac{1}{2} \quad (0.1.5.4)$$

$$\sigma_U^2 = \frac{1}{12} \quad (0.1.5.5)$$

which match with the values printed in problem 0.1.4

## 0.2 Central Limit Theorem

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

$$X = \sum_{i=1}^{12} U_i - 6 \quad (0.2.1.1)$$

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

```
codes/include/coeffs.h
codes/src/gau_gen_stat.c
```

0.2.2 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. 0.2.2.1

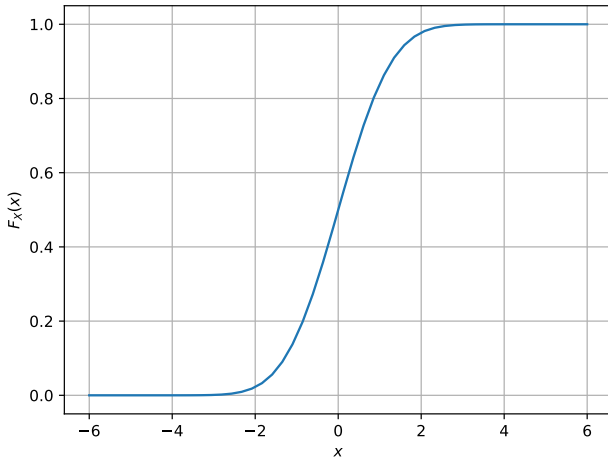


Figure 0.2.2.1: The CDF of  $X$

The properties of a CDF are

$$F_X(-\infty) = 0 \quad (0.2.2.1)$$

$$F_X(\infty) = 1 \quad (0.2.2.2)$$

$$\frac{dF_X(x)}{dx} \geq 0 \quad (0.2.2.3)$$

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

$$p_X(x) = \frac{d}{dx} F_X(x) \quad (0.2.3.1)$$

What properties does the PDF have?

**Solution:** The PDF of  $X$  is plotted in Fig. 0.2.3.1 using the code below

```
codes/src/cdf_pdf_plot_gau.py
```

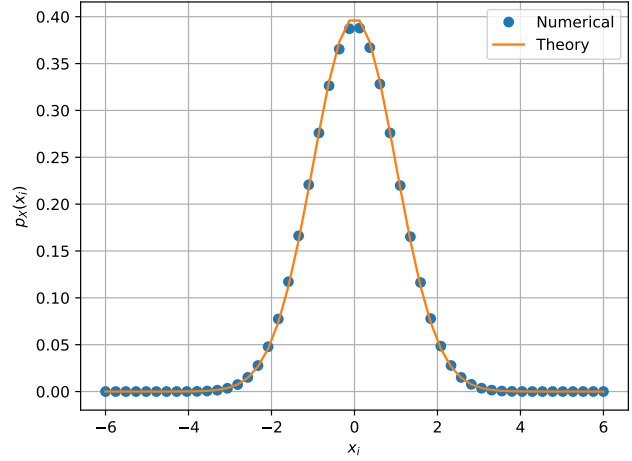


Figure 0.2.3.1: The PDF of  $X$

The properties of PDF are

$$f_X(x) \geq 0 \quad (0.2.3.2)$$

$$\int_{-\infty}^{\infty} f_X(x) dx = 1 \quad (0.2.3.3)$$

0.2.4 Find the mean and variance of  $X$  by writing a C program. **Solution:** The following code prints the mean and variance of  $X$

```
codes/src/gau_gen_stat.c
```

0.2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (0.2.5.1)$$

repeat the above exercise theoretically.

**Solution:** Substituting the PDF from (0.2.5.1) in (0.1.5.2),

$$\mu_X = \int_{-\infty}^{\infty} \frac{x}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \quad (0.2.5.2)$$

Using

$$(0.2.5.3)$$

$$\int x \cdot \exp(-ax^2) dx = -\frac{1}{2a} \cdot \exp(-ax^2) \quad (0.2.5.4)$$

$$\mu_X = \frac{1}{\sqrt{2\pi}} \left[ -\exp\left(-\frac{x^2}{2}\right) \right]_{-\infty}^{\infty} \quad (0.2.5.5)$$

$$\mu_X = 0 \quad (0.2.5.6)$$

Substituting  $\mu_X$  and the PDF in (0.1.5.3) to compute 0.3.2 Find a theoretical expression for  $F_V(x)$ .  
variance,

$$\sigma_X^2 = \int_{-\infty}^{\infty} \frac{x^2}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \quad (0.2.5.7)$$

Substituting

$$t = \frac{x^2}{2}, \quad (0.2.5.8)$$

$$\begin{aligned} \sigma_X^2 &= \frac{2}{\sqrt{\pi}} \int_0^{\infty} t^{\frac{1}{2}} \exp(-t) dt \\ &= \frac{2}{\sqrt{\pi}} \int_0^{\infty} t^{\frac{3}{2}-1} \exp(-t) dt \end{aligned} \quad (0.2.5.9)$$

Using the gamma function

$$\Gamma(x) = \int_0^{\infty} z^{x-1} \cdot e^{-z} dz \quad (0.2.5.10)$$

$$\begin{aligned} \sigma_X^2 &= \frac{2}{\sqrt{\pi}} \Gamma\left(\frac{3}{2}\right) \\ &= \frac{2}{\sqrt{\pi}} \frac{\sqrt{\pi}}{2} \\ &= 1 \end{aligned} \quad (0.2.5.11)$$

## 0.3 From Uniform to Other

### 0.3.1 Generate samples of

$$V = -2 \ln(1 - U) \quad (0.3.1.1)$$

and plot its CDF.

**Solution:** The samples for  $U$  are loaded from uni.dat file generated in problem 0.1.4. The CDF of  $V$  is plotted in Fig. 0.3.1.1 using the code below,

```
codes/src/function_1.py
```

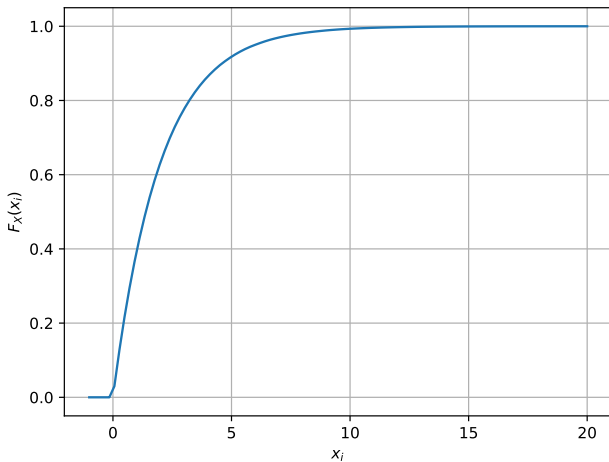


Figure 0.3.1.1: The CDF of  $V$

$$F_V(x) = P(V < x) \quad (0.3.2.1)$$

$$= P(-2 \ln(1 - U) < x) \quad (0.3.2.2)$$

$$= P(U < 1 - e^{-\frac{x}{2}}) \quad (0.3.2.3)$$

$$= F_U(1 - e^{-\frac{x}{2}}) \quad (0.3.2.4)$$

Using  $F_U(x)$  defined in (0.1.3.3),

$$F_V(x) = \begin{cases} 0 & x < 0 \\ 1 - e^{-\frac{x}{2}} & x \geq 0 \end{cases} \quad (0.3.2.5)$$

## 0.4 Triangular Distribution

### 0.4.1 Generate

$$T = U_1 + U_2 \quad (0.4.1.1)$$

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

```
codes/include/coeffs.h
codes/src/two_uni_gen.c
```

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

**Solution:** Loading the samples from uni1.dat and uni2.dat in python, the CDF is plotted in Fig. 0.4.2.1

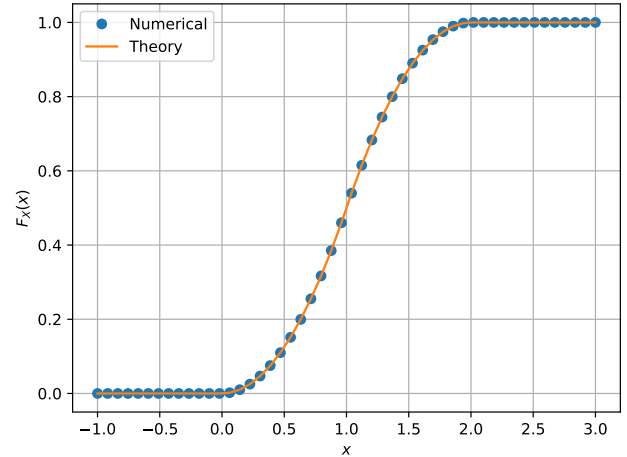


Figure 0.4.2.1: The CDF of  $T$

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

**Solution:** The PDF of  $T$  is plotted in Fig. 0.4.3.1 using the code below

```
codes/src/function_2.py
```

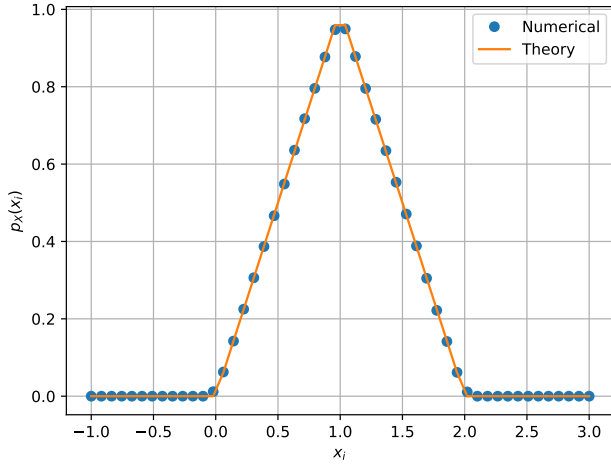


Figure 0.4.3.1: The PDF of  $T$

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

**Solution:** Since  $T$  is the sum of two independent random variables  $U_1$  and  $U_2$ , the PDF of  $T$  is given by

$$p_T(x) = p_{U_1}(x) * p_{U_2}(x) \quad (0.4.4.1)$$

Using the PDF of  $U$  from (0.1.3.2), the convolution results in

$$p_T(x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x \leq 1 \\ 2 - x & 1 \leq x \leq 2 \\ 0 & x > 2 \end{cases} \quad (0.4.4.2)$$

The CDF of  $T$  is found using (0.1.3.1) by replacing  $U$  with  $T$ . Evaluating the integral for the piecewise function  $p_T(x)$ ,

$$F_T(x) = \begin{cases} 0 & x < 0 \\ \frac{x^2}{2} & 0 \leq x \leq 1 \\ 2x - \frac{x^2}{2} - 1 & 1 \leq x \leq 2 \\ 1 & x > 2 \end{cases} \quad (0.4.4.3)$$

0.4.5 Verify your results through a plot.

**Solution:** The theoretical and numerical plots for the CDF and PDF of  $T$  closely match in Fig. 0.4.2.1 and Fig. 0.4.3.1