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# Assignment

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2

#### **CONTENTS**

1	T Ingiformer	Dandam	Numbers	
	Unitorm	Kandom	Numbers	

### 2 Central Limit Theorem

### **3** From Uniform to Other

Abstract—This manual provides a simple introduction to the generation of 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

\$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/

R N Assignment/Codes/exrand.c

\$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/

R\_N\_Assignment/Codes/coeffs.h

and compile and execute the C program using

- \$ gcc exrand.c -lm
- \$ ./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 following code plots Fig. 1.2

\$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/ R N Assignment/Codes/cdf plot.py

It is executed with

\$ python3 cdf plot.py

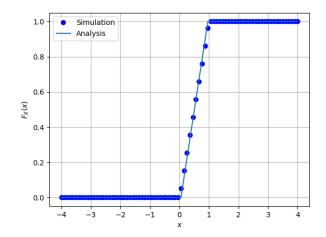


Fig. 1.2: The CDF of U

1.3 Find a theoretical expression for  $F_U(x)$ . **Solution:** The CDF of U is given by

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

We now have three cases:

- a) x < 0:  $p_X(x) = 0$ , and hence  $F_U(x) = 0$ .
- b)  $0 \le x < 1$ : Here,

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

c)  $x \ge 1$ : Put x = 1 in (1.3) as U is uniform in [0, 1] to get  $F_U(x) = 1$ .

Therefore,

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

This is verified in Figure (1.2)

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:** The C program can be downloaded using

\$ wget https://github.com/Prabhav11004/
AI1110-ASSIGNMENTS/blob/main/
R\_N\_Assignment/Codes/mean\_var.c
\$ wget https://github.com/Prabhav11004/
AI1110-ASSIGNMENTS/blob/main/
R N Assignment/Codes/mean var.h

#### and compiled and executed with

\$ gcc mean\_var.c -lm \$ ./a.out

The calculated mean is 0.500007 and the calculated variance is 0.083301.

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) dx \tag{1.7}$$

**Solution:** We write

$$E\left[U^{2}\right] = \int_{-\infty}^{\infty} x^{2} dF_{U}(x) \tag{1.8}$$

$$= \int_{-\infty}^{\infty} x^2 p_U(x) dx \tag{1.9}$$

$$= \int_0^1 x^2 dx = \frac{1}{3}$$
 (1.10)

and

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

$$= \int_{-\infty}^{\infty} x p_U(x) dx \tag{1.12}$$

$$= \int_0^1 x dx = \frac{1}{2}$$
 (1.13)

which checks out with the empirical mean on 0.500007. Now, using linearity of expectation,

$$var[U] = E[U - E[U]]^{2}$$
 (1.14)

$$= E\left[U^2 - 2UE[U] + (E[U])^2\right]$$
 (1.15)

$$= E\left[U^{2}\right] - 2\left(E\left[U\right]\right)^{2} + \left(E\left[U\right]\right)^{2} \qquad (1.16)$$

$$= E[U^2] - (E[U])^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12} \quad (1.17)$$

(1.18)

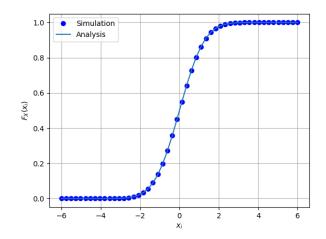


Fig. 2.2: The CDF of X

and this checks out with the empirical variance 0.083301 of the sample data.

#### 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.

\$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/ R\_N\_Assignment/Codes/exrand.c \$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/ R N Assignment/Codes/coeffs.h

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 The required python file can be downloaded using

\$ wget https://github.com/Prabhav11004/ AI1110-ASSIGNMENTS/blob/main/ R\_N\_Assignment/Codes/q2.2 \_cdf\_gau\_plot.py

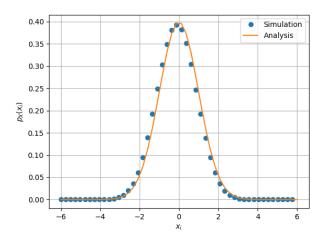


Fig. 2.3: The PDF of X

and executed using

\$ python3 q2.2 cdf gauss plot.py

- a) The CDF is non-decreasing
- b) It is right-continuous.
- c)  $\lim_{x\to-\infty} F_X(x) = 0$
- d)  $\lim_{x\to\infty} F_X(x) = 1$
- 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:** The PDF of X is plotted in Fig. 2.3 using the code below

\$ wget https://raw.githubusercontent.com/ goats-9/ai1110-assignments/master/ manual/codes/pdf plot.py

The figure is generated using

The properties of a PDF  $p_X(x)$  are as follows:

- a)  $\forall x \in \mathbb{R}, p_X(x) \ge 0$ b)  $\int_{-\infty}^{\infty} p_X(x) dx = 1$ c) For  $a < b, a, b \in \mathbb{R}$

$$\Pr(a < X < b) = \Pr(a \le X \le b)$$
 (2.3)

$$= \int_a^b p_X(x)dx \tag{2.4}$$

If we take a = b, then we get Pr(X = a) = 0.

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

**Solution:** The mean and variance have been calculated using (1.5) and (1.6) respectively. The C program can be downloaded using

\$ wget https://raw.githubusercontent.com/ goats-9/ai1110-assignments/master/ manual/codes/mean var gau.c

and compiled and executed with the following commands

The calculated mean is 0.000326 and the calculated variance is 1.000906.

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:** The mean is given by

$$E[X] = \int_{-\infty}^{\infty} x \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) = 0 \qquad (2.6)$$

as the integrand is odd. This checks out with the empirical mean of 0.000326. The variance is given by

$$var[X] = E[X^2] - (E[X])^2$$
 (2.7)

$$= \int_{-\infty}^{\infty} x^2 \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx \qquad (2.8)$$

$$= \int_0^\infty \frac{2}{\sqrt{2\pi}} \sqrt{2t} e^{-t} dt \tag{2.9}$$

$$=\frac{2}{\sqrt{\pi}}\Gamma\left(\frac{3}{2}\right) \tag{2.10}$$

$$= \frac{1}{\sqrt{\pi}} \Gamma\left(\frac{1}{2}\right) = 1 \tag{2.11}$$

where we have used  $t = \frac{x^2}{2}$  and so dt = xdx. We have also used the gamma function given as

$$\Gamma(n) = \int_{-\infty}^{\infty} x^{n-1} e^{-x} dx \qquad (2.12)$$

$$\Gamma(n) = (n-1)\Gamma(n-1) \text{ for } n > 1$$
 (2.13)

and the fact that  $\Gamma(1/2) = \sqrt{\pi}$ . This agrees with the empirical variance of 1.000906.

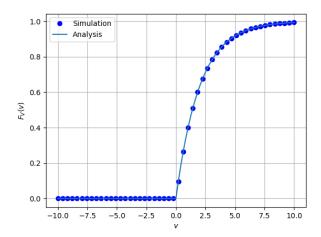


Fig. 3.1: The CDF of V

#### 3 From Uniform to Other

#### 3.1 Generate samples of

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

and plot its CDF.

Solution: The relevant python code is at

\$ wget https://raw.githubusercontent.com/ goats-9/ai1110-probability/master/ manual/codes/cdf exp plot.py

and can be executed with

and the CDF is plotted in Figure (3.1).

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

Solution: Note that the function

$$v = f(u) = -2\ln(1 - u) \tag{3.2}$$

is monotonically increasing in [0, 1] and  $v \in$  $\mathbb{R}^+$ . Hence, it is invertible and the inverse function is given by

$$u = f^{-1}(v) = 1 - \exp\left(-\frac{v}{2}\right)$$
 (3.3)

Therefore, from the monotonicity of v, and using (1.4),

$$F_V(v) = F_U \left( 1 - \exp\left(-\frac{v}{2}\right) \right) \tag{3.4}$$

$$F_V(v) = F_U \left( 1 - \exp\left(-\frac{v}{2}\right) \right)$$
 (3.4)  

$$\implies F_V(v) = \begin{cases} 0 & v < 0 \\ 1 - \exp\left(-\frac{v}{2}\right) & v \ge 0 \end{cases}$$
 (3.5)