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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: : Link to the code : C 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: The following code plots Fig: Python code

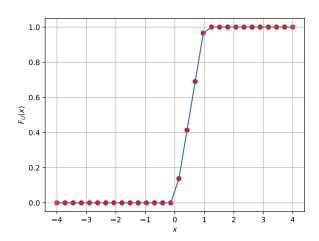


Fig. 1: The CDF of U

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

Solution: As U is uniformly distributed ran- Variance(σ^2) = $E(U^2) - E(U)^2$ dom variable in the interval (0,1) and

$$p_U(x) = \begin{cases} 1, & x \in (0,1) \\ 0, & otherwise \end{cases}$$
 (1.2)

Hence,

$$F_U(x) = \begin{cases} 0, & x \in (-\infty, 0) \\ x, & x \in (0, 1) \\ 1, & x \in (1, \infty) \end{cases}$$
 (1.3)

1.4 The mean of U is defined as

$$E[U] = \frac{1}{N} \sum_{i=1}^{N} U_i$$
 (1.4)

and its variance as

$$var[U] = E[U - E[U]]^2$$
 (1.5)

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

Solution: Link to the code : C code

aniket@aniket-HP:~/Desktop/PR\$ gcc code/code1 4.c aniket@aniket-HP:~/Desktop/PR\$./a.out Mean is 0.500007 Variance is 0.083301 aniket@aniket-HP:~/Desktop/PR\$

Fig. 1: Output

1.5 Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.6}$$

Solution: we know that,

$$dF_U(x) = p_U(x)dx (1.7)$$

also mean (μ) is E(U): Hence,

$$\mu = \int_{-\infty}^{\infty} x p_U(x) dx$$
$$= \int_{0}^{1} x dx$$
$$= \frac{x^2}{2} \Big|_{0}^{1}$$
$$= \frac{1}{2}$$

$$E(U^{2}) = \int_{-\infty}^{\infty} x^{2} p_{U}(x) dx$$
$$= \int_{0}^{1} x^{2} dx$$
$$= \frac{x^{3}}{3} \Big|_{0}^{1}$$
$$= \frac{1}{3}$$

$$\sigma^2 = \frac{1}{3} - \frac{1}{4}$$
$$= \frac{1}{12}$$

2 Central Limit Theorem

2.1 Generate 10⁶ 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: link to the code : C 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: Link to Python code: Python code

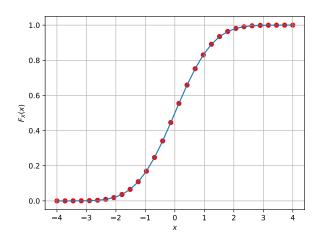


Fig. 2: The CDF of X

Properties:

- (1) Graph is symmetric about a single point
- (2) The $F_X(x)$ is non-decreasing function
- (3) $\lim_{x\to\infty} F_X(x)$
- 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

$$p_X(x) = \frac{d}{dx} F_X(x) \tag{2.2}$$

What properties does the PDF have? **Solution:** Link to the code :Python code

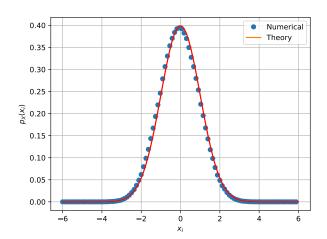


Fig. 2: The PDF of X

Properties:

- (1) Area under the curve is One.
- (2) Symmetric about line $x = \mu$.
- (3) Increasing in first half and decreasing in other half.
- 2.4 Find the mean and variance of *X* by writing a C program.

Solution: Link to the code : C code

```
aniket@aniket-HP:~/Desktop/PR$ gcc code/code2_4.c
aniket@aniket-HP:~/Desktop/PR$ ./a.out
Mean is 0.000294
Variance is 0.999560
aniket@aniket-HP:~/Desktop/PR$
```

Fig. 2: Output

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty, \quad (2.3)$$

repeat the above exercise theoretically. **Solution:**

(1) CDF is

$$F_X(x) = \int_{-\infty}^{\infty} p_X(x) dx$$
$$= 1$$

(2) Mean is

$$\mu = E(x) = \int_{-\infty}^{\infty} x p_X(x) dx$$

from equation (1.3),

$$F_U(x) = \begin{cases} 0, & x \in (-\infty, 0) \\ (1 - exp\frac{-x}{2}), & x \in (0, \infty) \end{cases}$$

Due to symmetry clearly, $\mu = 0$

(3) Variance is

$$var[X] = E(X^{2}) - E(X)^{2}$$
$$= \int_{-\infty}^{\infty} x^{2} p_{X}(x) dx - 0$$
$$= 1$$

3 From Uniform to Other

3.1 Generate samples of

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

and plot its CDF.

Solution: Link to the code: Python code

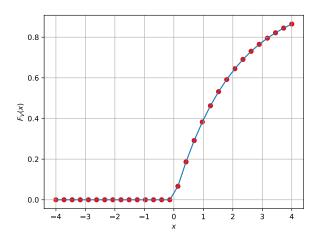


Fig. 3: CDF of V

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

$$\begin{split} F_V(X) &= Pr(V \leq x) \\ &= Pr(-2ln(1-U) \leq x) \\ &= Pr((1-U) \geq exp(\frac{-x}{2})) \\ &= Pr(U \leq (1-exp(\frac{-x}{2}))) \\ &= F_U(1-exp\frac{-x}{2}) \end{split}$$