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Random Numners

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Uniform Random Numbers

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1 Uniform Random Numbers

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

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ coeffs.h wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ exrand.c gcc exrand.c ./a.out

1.2 **Solution:** The following code plots Fig. 1.1

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ cdf_plot.py python cdf_plot.py

1.3 Find a theoretical expression for $F_U(x)$. **Solution:** Since U is a uniform distribution

$$F_U(x) = x \tag{1.1}$$

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

```
wget https://github.com/gunjitmittal/
Random_Numbers/blob/main/codes/
mean.c
gcc mean.c
./a.out
```

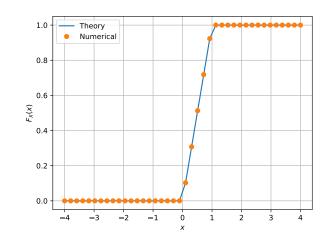


Fig. 1.1. The CDF of U

Mean: 0.500004 Variance: 0.083269

1.5

Verify your result theoretically given that

$$E\left[U^{k}\right] = \int_{-\infty}^{\infty} x^{k} dF_{U}(x) \tag{1.2}$$

Solution:

$$E[U] = \int_{-\infty}^{\infty} x dF_U(x)$$
 (1.3)

$$= \int_0^1 x dx \tag{1.4}$$

$$= \left[\frac{x^2}{2}\right]_0^1 \tag{1.5}$$

$$=\frac{1}{2}\tag{1.6}$$

$$E[U - E[U]]^2 = E[U^2] - E[U]^2$$
 (1.7)

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

$$= \int_0^1 x^2 dx \tag{1.9}$$

$$= \left[\frac{x^3}{3}\right]_0^1 \tag{1.10}$$

$$=\frac{1}{3}$$
 (1.11)

$$E[U - E[U]]^2 = \frac{1}{3} - \left(\frac{1}{2}\right)^2$$
 (1.12)

$$=\frac{1}{3} - \frac{1}{4} = \frac{1}{12} \tag{1.13}$$

2 CENTRAL LIMIT THEOREM

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

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ coeffs.h wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ exrand2.c

gcc exrand2.c

2.2 What properties does a CDF have?

Solution: The following code plots Fig. 2.1

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ cdf_plot2.py python cdf_plot2.py

- CDF reaches 0.5 as it reaches the mean(0)
- It reaches 1 as the graph reaches the end
- Differentiating the graph we obtain the graph of the PDF.
- 2.3 What properties does the PDF have?

Solution: The following code plots Fig. 2.2

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ pdf_plot.py python pdf_plot.py

• PDF is symmetric about x = mean(0)

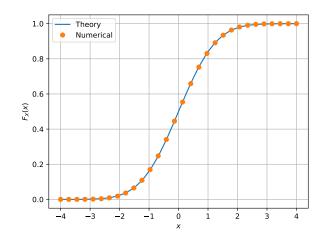


Fig. 2.1. The CDF of X

• Graph is shpaed like a bell

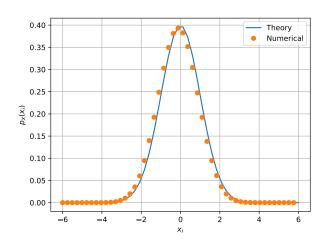


Fig. 2.2. The PDF of X

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

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/codes/ mean2.c gcc mean2.c ./a.out

> Mean: 0.000326 Variance: 1.000467

2.5 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty,$$
(2.1)

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

$$F_X(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$
 (2.2)

$$E[X] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} x e^{-\frac{x^2}{2}} dx$$
 (2.3)

Taking
$$\frac{x^2}{2} = t \implies xdx = dt$$
 (2.4)

$$E[X] = \frac{1}{\sqrt{2\pi}} \int_{\infty}^{\infty} e^{-t} dt = 0$$
 (2.5)

$$E[X^{2}] = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} x^{2} e^{-\frac{x^{2}}{2}} dx$$
 (2.6)

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} x(xe^{-\frac{x^2}{2}}) dx$$
 (2.7)

$$= \frac{1}{\sqrt{2\pi}} \left[-xe^{-\frac{-x^2}{2}} + \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx \right]_{-\infty}^{\infty}$$
(2.8)

$$=1 \tag{2.9}$$

Variance =
$$E[X^2] - E[X]^2 = 1 - 0 = 1$$
 (2.10)

3 From Uniform to Other

3.1 **Solution:** The following code plots Fig. 3.1

wget https://github.com/gunjitmittal/ Random Numbers/blob/main/codes/ cdf_plot3.py python cdf_plot3.py

3.2 Given that

$$p_X(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right), -\infty < x < \infty,$$
(3.1)

repeat the above exercise theoretically.

$$F_V(x) = P(V \le x) \tag{3.2}$$

$$= P(-2\ln(1-U) < x) \tag{3.3}$$

$$= P(U < 1 - e^{\frac{-x}{2}}) \tag{3.4}$$

$$P(U < x) = \int_{0}^{x} dx = x$$
 (3.5)

$$\therefore P(U \le 1 - e^{\frac{-x}{2}}) = 1 - e^{\frac{-x}{2}} \tag{3.6}$$

$$\implies F_V(x) = 1 - e^{\frac{-x}{2}}$$
 (3.7)

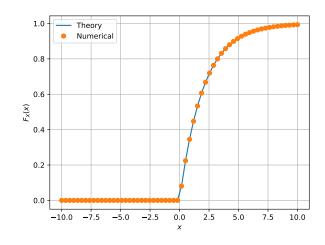


Fig. 3.1. The CDF of V

4 TRIANGULAR DISTRIBUTION

4.1 Generate

$$T = U_1 + U_2 (4.1)$$

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

> wget https://github.com/gunjitmittal/ Random Numbers/blob/main/ codes/coeffs.h wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/ codes/triang.c gcc triang.c ./a.out

4.2 Find the CDF of T

Solution:

$$P_T(x) = P(U_1 + U_2 = x)$$

$$= \int_{-\infty}^{\infty} P_{U_1}(a) P_{U_2}(x - a) da$$
(4.2)

As $P_{U_1}(a) = 1$ for 0 < a < 1 and 0 otherwise

$$\implies \int_0^1 P_{U_2}(x-a)da \qquad (4.4)$$

$$P_{U_2}(x-a) = 1$$
 for $0 < x-a < 1 \implies x-1 < a < x$
If $x < 1, 0 < a < x$

$$= \int_0^x 1da = x \tag{4.5}$$

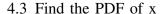
If
$$x > 1, x - 1 < a < 1$$

$$= \int_{x-1}^{1} 1da = 2 - x \tag{4.6}$$

$$\therefore P_T(x) = \begin{cases} 0, & x < 0 \\ x, & 0 < x < 1 \\ 2 - x, & 1 < x < 2 \\ 0, & x > 2 \end{cases}$$
 (4.7)

Integrating $P_T(x)$ we obtain $F_T(x)$

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



$$P_T(x) = \begin{cases} 0, & x < 0 \\ x, & 0 < x < 1 \\ 2 - x, & 1 < x < 2 \\ 0, & x > 2 \end{cases}$$
 (4.9)

4.4 Find the theoretical expressions for the PDF and CDF of T.

$$P_T(x) = \begin{cases} 0, & x < 0 \\ x, & 0 < x < 1 \\ 2 - x, & 1 < x < 2 \\ 0, & x > 2 \end{cases}$$
 (4.10)

$$F_T(x) = \begin{cases} 0, & x < 0\\ \frac{x^2}{2}, & 0 < x < 1\\ 1 - \frac{(2-x)^2}{2}, & 1 < x < 2\\ 1, & x > 2 \end{cases}$$
(4.11)

4.5 Verify your results through a plot.

5 MAXIMUL LIKELIHOOD

5.1 solution Download the following files and execute the C program.

wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/ codes/coeffs.h wget https://github.com/gunjitmittal/ Random_Numbers/blob/main/ codes/generate5.c gcc generate5.c ./a.out

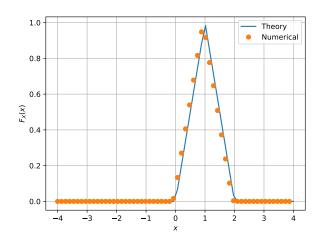


Fig. 4.1. The PDF of ${\cal T}$

(4.8)

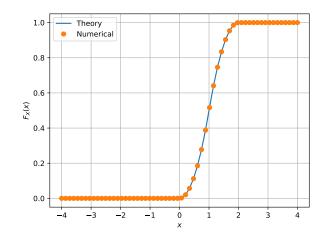


Fig. 4.2. The CDF of T

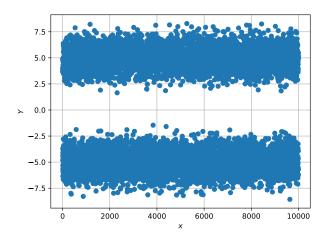


Fig. 5.1. The plot of Y

- 5.2 Plot Y.
- 5.3 Guess how to estimate X from Y. Since X lies in $\{-1,1\}$ and is multiplied by 5, while the normal distribution is concentrated mostly in the [-1,1] region so if X is 1 then Y lies mostly in the region [4,6] and if X is -1 then Y lies mostly in the region [-6,-4] so we can take a threshold δ and say that if $Y > \delta$ then X is most probably 1 and if $Y < \delta$ then X is mostly -1. Since it is symmetric about X axis let $\delta = 0$ for now.

5.4 Find

$$P_{e|0} = \Pr\left(\hat{X} = -1|X = 1\right)$$
 (5.1)

and

$$P_{e|1} = \Pr\left(\hat{X} = 1|X = -1\right)$$
 (5.2)

Solution:

$$P_{e|0} = \Pr\left(\hat{X} = -1|X = 1\right)$$
 (5.3)

$$= \Pr(Y < 0 | X = 1) \tag{5.4}$$

$$= \Pr\left(A + N < 0\right) \tag{5.5}$$

$$= \Pr\left(N < -A\right) \tag{5.6}$$

$$= F_N(-A) = 1 - Q_N(-A)$$
 (5.7)

$$P_{e|1} = \Pr\left(\hat{X} = 1|X = -1\right)$$
 (5.8)

$$= \Pr(Y > 0 | X = -1) \tag{5.9}$$

$$= \Pr(-A + N > 0) \tag{5.10}$$

$$= \Pr\left(N > A\right) \tag{5.11}$$

$$=Q_N(A) (5.12)$$

5.5 Find P_e assuming that X has equiprobable symbols.

Solution:

$$P_e = \frac{1}{2}P_{e|0} + \frac{1}{2}P_{e|1} \tag{5.13}$$

$$P_e = \frac{1}{2} \left(Q_N(A) + Q_N(A) \right) \tag{5.14}$$

$$=Q_N(A) \tag{5.15}$$