

# Assignment

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Question : Let  $\phi(\cdot)$  denote the cumulative distribution function of a standard normal random variable. If the random variable  $X$  has the cumulative distribution function

$$F(x) = \begin{cases} \phi(x), & x < -1 \\ \phi(x+1), & x \geq -1 \end{cases} \quad (1)$$

then which one of the following statements is true?

- A)  $P(X \leq -1) = \frac{1}{2}$
- B)  $P(X = -1) = \frac{1}{2}$
- C)  $P(X < -1) = \frac{1}{2}$
- D)  $P(X \leq 0) = \frac{1}{2}$

**Solution: Gaussian**

Q function is defined

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\frac{u^2}{2}} du \quad (2)$$

From (1) and (2);

$$F_X(x) = \begin{cases} Q(-x), & x < -1 \\ 1 - Q(x+1), & x \geq -1 \end{cases} \quad (3)$$

From (6);

A)

$$\begin{aligned} \Pr(X \leq -1) &= F_X(-1) = 1 - Q(0) \\ &= 0.5 \end{aligned} \quad \begin{matrix} (4) \\ (5) \end{matrix}$$

So Option A i.e.,  $P(X < -1) = \frac{1}{2}$  is correct

B) The pdf of  $X$  can be defined in terms of cdf as

$$\Pr(X = b) = F_X(b) - \lim_{x \rightarrow b^-} F_X(x) \quad (6)$$

From (6);

$$\Pr(X = -1) = F_X(-1) - \lim_{x \rightarrow -1^-} F_X(x) \quad (7)$$

$$= 1 - Q(0) - Q(-(-1)) \quad (8)$$

$$= 0.341 \quad (9)$$

So Option B i.e.,  $P(X = -1) = \frac{1}{2}$  is incorrect

C)

$$\Pr(X < -1) = \lim_{x \rightarrow -1^-} F_X(x) = F_X(-1) \quad (10)$$

$$= Q(-(-1)) \quad (11)$$

$$= 0.159 \quad (12)$$

So Option C i.e.,  $P(X < -1) = \frac{1}{2}$  is incorrect

D)

$$\Pr(X \leq 0) = F_X(0) = 1 - Q(1) \quad (13)$$

$$= 0.8413 \quad (14)$$

So Option D i.e.,  $P(X \leq 0) = \frac{1}{2}$  is incorrect

Gaussian CDF plot of  $X$  is given in fig1

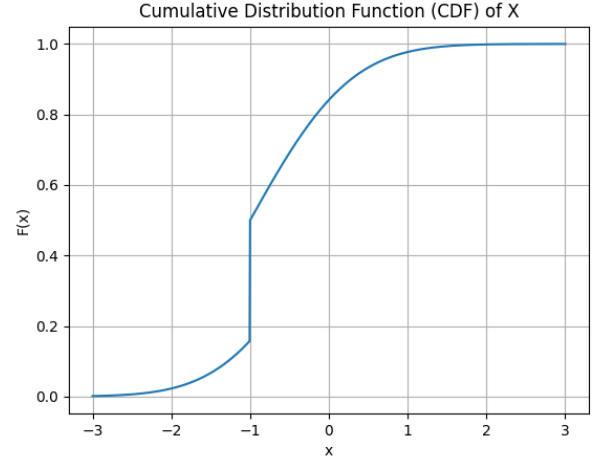


Fig. 1.

**Box-Muller**

**STATEMENT:** Suppose  $U_1$  and  $U_2$  are independent samples chosen from the uniform distribution on the unit interval  $(0, 1)$ . Let

$$Z_0 = R \cos(\Theta) = \sqrt{-2 \ln U_1} \cos(2\pi U_2) \quad (15)$$

and

$$Z_1 = R \sin(\Theta) = \sqrt{-2 \ln U_1} \sin(2\pi U_2) \quad (16)$$

Then  $Z_0$  and  $Z_1$  are independent random variables with a standard normal distribution. So, now we will generate  $Z_0$  using C Code as given below

**PROOF:** Let  $X$  and  $Y$  be independent standard normal variables

$$X, Y \sim \mathcal{N}(0, 1) \quad \text{and} \quad X \perp Y \quad (17)$$

The joint pdf of  $X$  and  $Y$  is given by

$$f_{XY}(X, Y) = f(x)f(y) \quad (18)$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \quad (19)$$

$$= \frac{1}{2\pi} e^{-\frac{(x^2+y^2)}{2}} \quad (20)$$

The relationship between Cartesian coordinates  $(x, y)$  and polar coordinates  $(r, \theta)$  is as follows

$$x = r \cos \theta \quad (21)$$

$$y = r \sin \theta \quad (22)$$

Change  $f_{XY}(x, y)$  to polar coordinates :

$$f_{XY}(x, y) dx dy = f_{R\theta}(r, \theta) dr d\theta \quad (23)$$

i.e.,

$$f_{R\theta}(r, \theta) = f_{XY}(x, y) \frac{dx dy}{dr d\theta} = f_{XY}(x, y) \left| \frac{\partial(x, y)}{\partial(r, \theta)} \right| \quad (24)$$

where  $J$  is the Jacobian

$$J = \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} \quad (25)$$

$$= \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r \quad (26)$$

For  $r \geq 0$  and  $\theta \in [0, 2\pi)$ , we have

$$f_{R\theta}(r, \theta) dr d\theta = \frac{1}{2\pi} e^{-\frac{r^2}{2}} r dr d\theta \quad (27)$$

Now we change the variable from  $(r, \theta)$  to  $(r^2, \theta)$ ;

$$r dr = \frac{1}{2} dr^2 \quad (28)$$

Now it can be written as,

$$f_{R\theta}(r, \theta) dr d\theta = f_{R^2\theta}(r^2, \theta) dr^2 d\theta \quad (29)$$

$$= \frac{1}{2\pi} e^{-\frac{r^2}{2}} \frac{1}{2} dr^2 d\theta \quad (30)$$

$$= \left( \frac{1}{2} e^{-\frac{r^2}{2}} dr^2 \right) \left( \frac{1}{2} d\theta \right) \quad (31)$$

$$= f_{R^2}(r^2) dr^2 f_{\theta}(\theta) d\theta \quad (32)$$

from the above equation, we can say that

$$R^2 \perp \theta \quad (\text{i.e., } R^2 \text{ and } \theta \text{ are independent}) \quad (33)$$

Generate,  $\theta \sim \text{Unif}(0, 2\pi)$  Generate,  $V \sim \text{Exp}(\lambda = \frac{1}{2})$  (i.e.,  $V = R^2$ ) and compute

$$R = \sqrt{V} \quad (34)$$

Compute

$$X = R \cos \theta \quad (35)$$

$$Y = R \sin \theta \quad (36)$$

$$(37)$$

where,  $X, Y$  are i.i.d in  $\sim \mathcal{N}(0, 1)$  Then

$$\theta = 2\pi U_1 \quad (38)$$

$$V = -2\log(U_2) \quad (39)$$

We can prove  $V = -2\log(U_2)$  using the c.d.f definition of the exponential distribution

Let  $X$  be a random variable following an exponential distribution with rate parameter  $\lambda = \frac{1}{2}$ , denoted as  $X \sim \text{Exp}(\frac{1}{2})$ . The cumulative distribution function (CDF) of the exponential distribution is given by:

$$F(x) = 1 - e^{-\frac{x}{2}} \quad (40)$$

Now, suppose we have a random variable  $U$  following a uniform distribution in the interval  $[0, 1]$ , denoted as  $U \sim \text{U}(0, 1)$ .

The CDF of the uniform distribution is simply:

$$F_U(u) = u, \text{ for } 0 \leq u \leq 1 \quad (41)$$

We can use the probability integral transform to express the exponential random variable  $X$  in terms of the uniform random variable  $U$ :

$$F(x) = F_U(u) \quad (42)$$

$$1 - e^{-\frac{x}{2}} = u \quad (43)$$

Now, solve for  $x$ :

$$e^{-\frac{x}{2}} = 1 - u \quad (44)$$

$$-\frac{x}{2} = \ln(1 - u) \quad (45)$$

$$x = -2 \ln(1 - u) \quad (46)$$

So,  $X \sim -2 \ln(1 - U)$  for  $U \sim \text{U}(0, 1)$ . This expression represents the exponential random variable  $X$  in terms of a uniform random variable  $U$ .

C and Python codes are given below :

```
(28)1 #include <stdio.h>
      2 #include <stdlib.h>
      3 #include <math.h>
      4 #include <time.h>
      5
      6 double sn() {
      7     double u1, u2;
      8     while (1) {
      9         u1 = ((double)rand() / RAND_MAX);
      10        if (u1 > 0 && u1 < 1)
```

```

11         break;};
12     while (1) {
13         u2 = ((double)rand() / RAND_MAX);
14         if (u2 > 0 && u2 < 1)
15             break;};
16     double z1 = sqrt(-2 * log(u1)) * cos
17         (2 * M_PI * u2);
18     return z1;
19 }
20 int main() {
21     // Seed the random number generator
22     srand(time(0));
23
24     int numSamples = 100000; // You can
25     // change this to the desired number
26     // of samples
27     FILE *file = fopen("uni.dat", "w");
28
29     if (file == NULL) {
30         printf("Unable to open file for
31         writing.\n");
32         return 1;
33     }
34
35     for (int i = 0; i < numSamples; i++)
36     {
37         double sample = sn();
38         fprintf(file, "%lf\n", sample);
39     }
40
41     fclose(file);
42
43     return 0;
44 }

```

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 # Read data from the "uni.dat" file
5 x = np.genfromtxt("uni.dat")
6
7 # Create a histogram plot of the PDF
8 plt.hist(x, bins=50, density=True, alpha
9         =0.6, color='b', label='PDF (Histogram
10         )')
11 plt.xlabel('x (Standard Normal Random
12         Variable)')
13 plt.ylabel('Frequency')
14 plt.title('PDF of Standard Normal
15         Distribution (Histogram)')
16 plt.savefig('/home/sayyam/KHP/figs/
17         figure1.png')
18 plt.show()

```

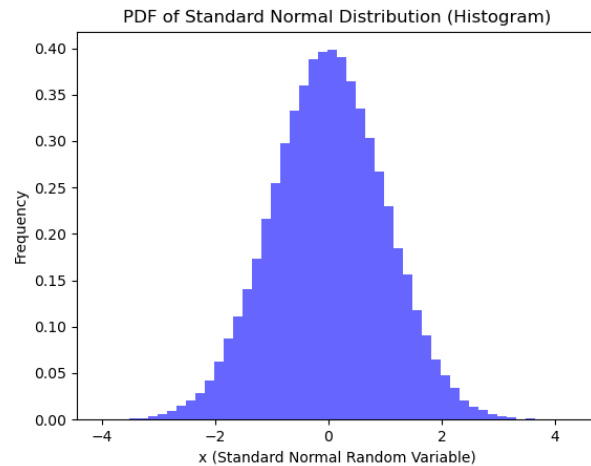


Fig. 2. Histogram plot of density of  $Z_0$