

MA-323 MONTE CARLO SIMULATION LAB-2

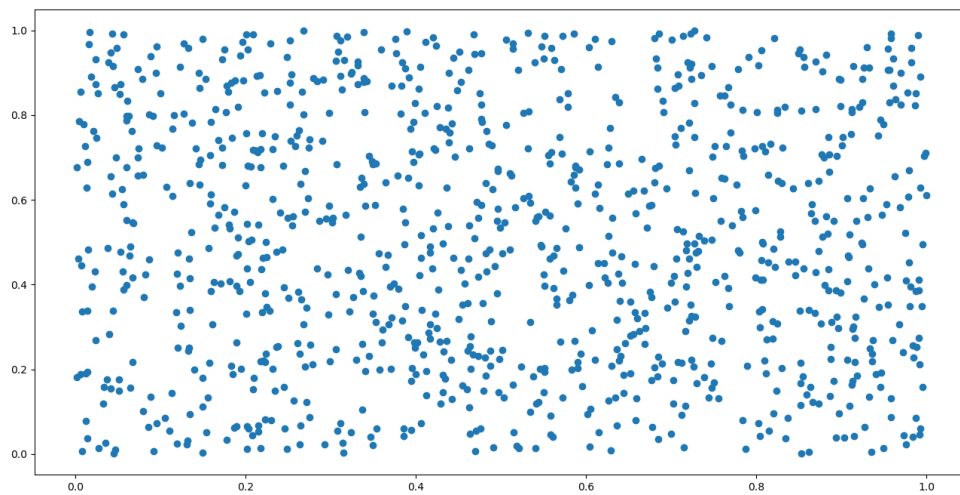
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210123083

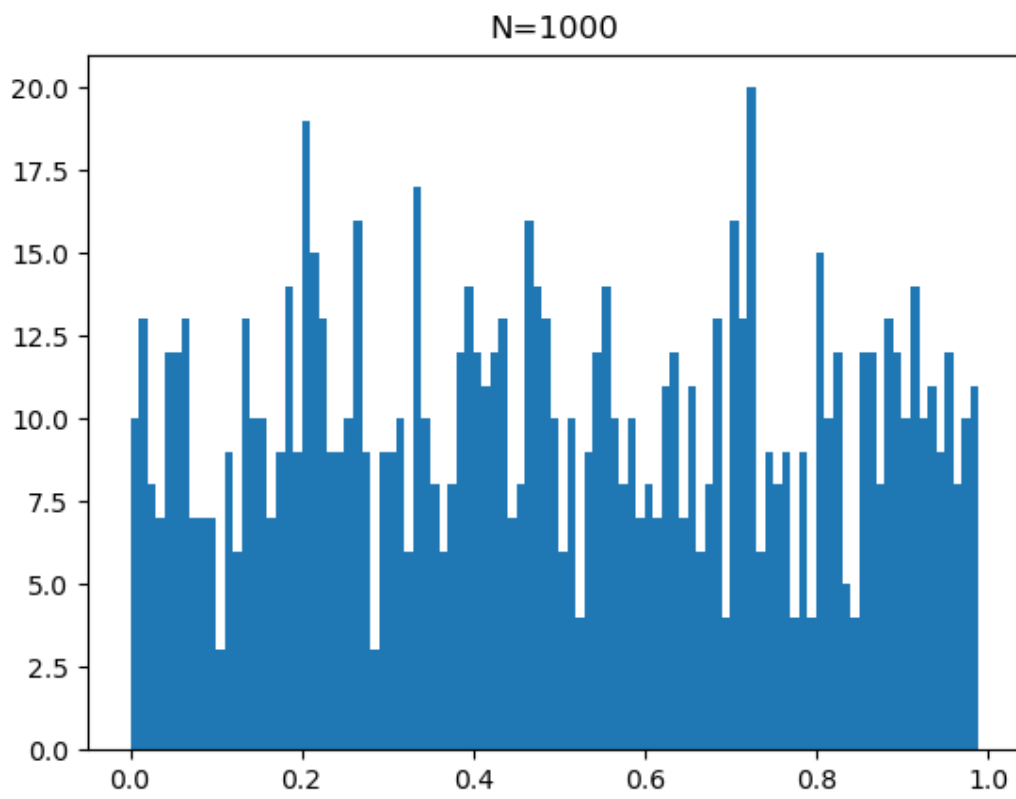
Question 1

Linear Congruence Generator

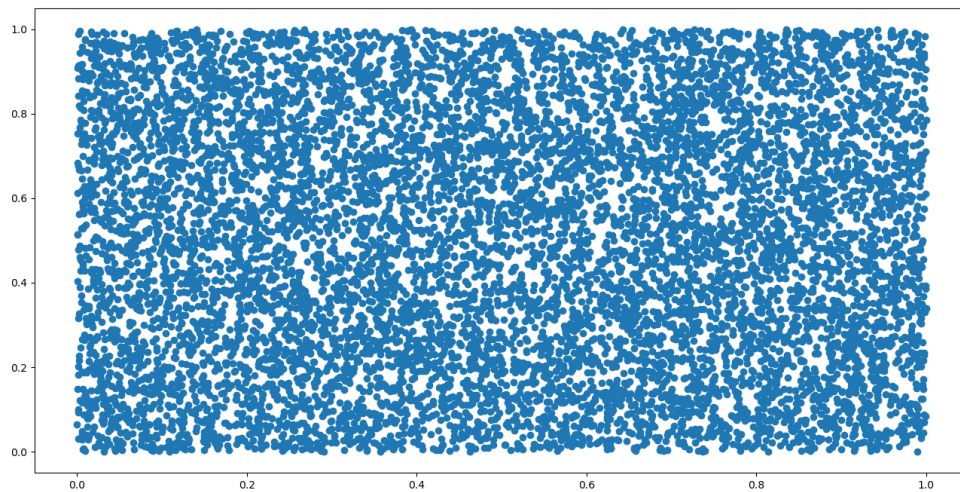
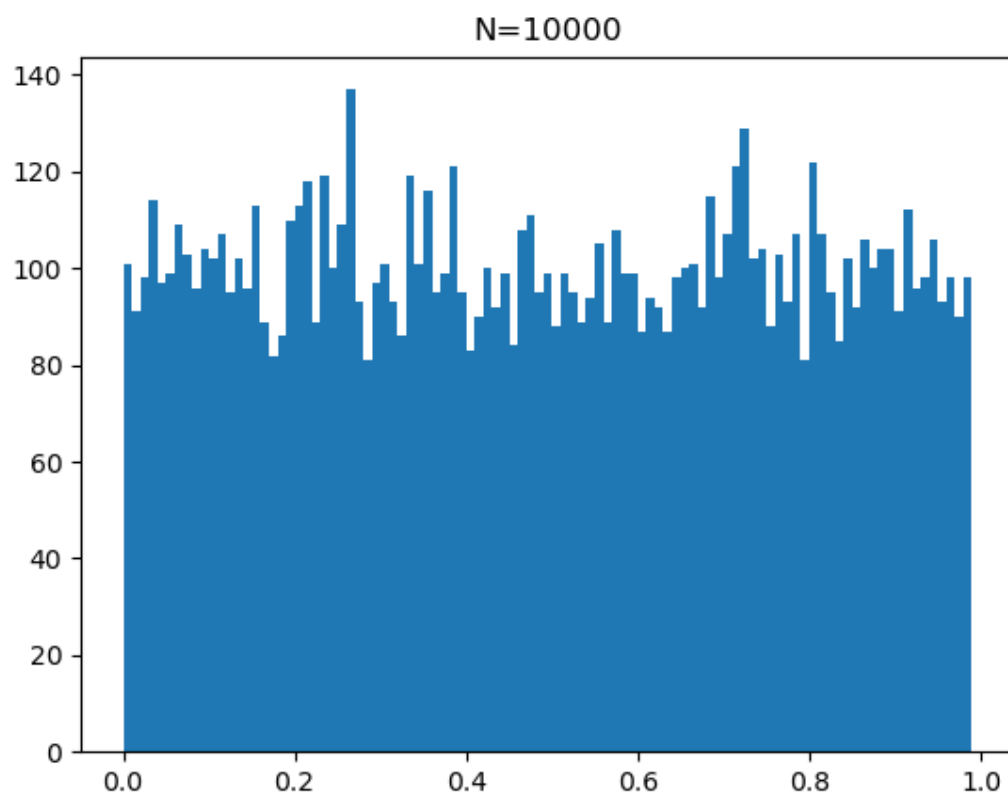
$x_0 = 1$, $a(\text{multiplier}) = 1229$, $m(\text{modulus}) = 2048$ and $b = 1$

- **Plot for $n=1000$**

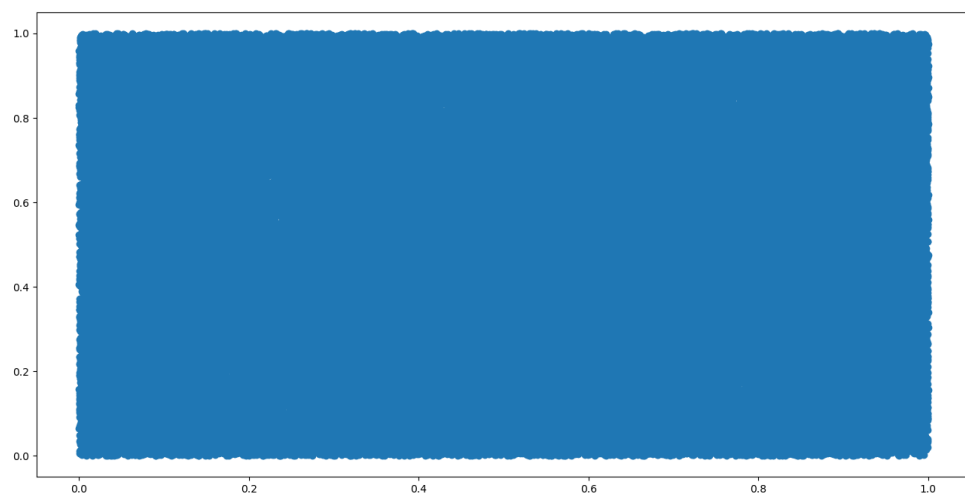
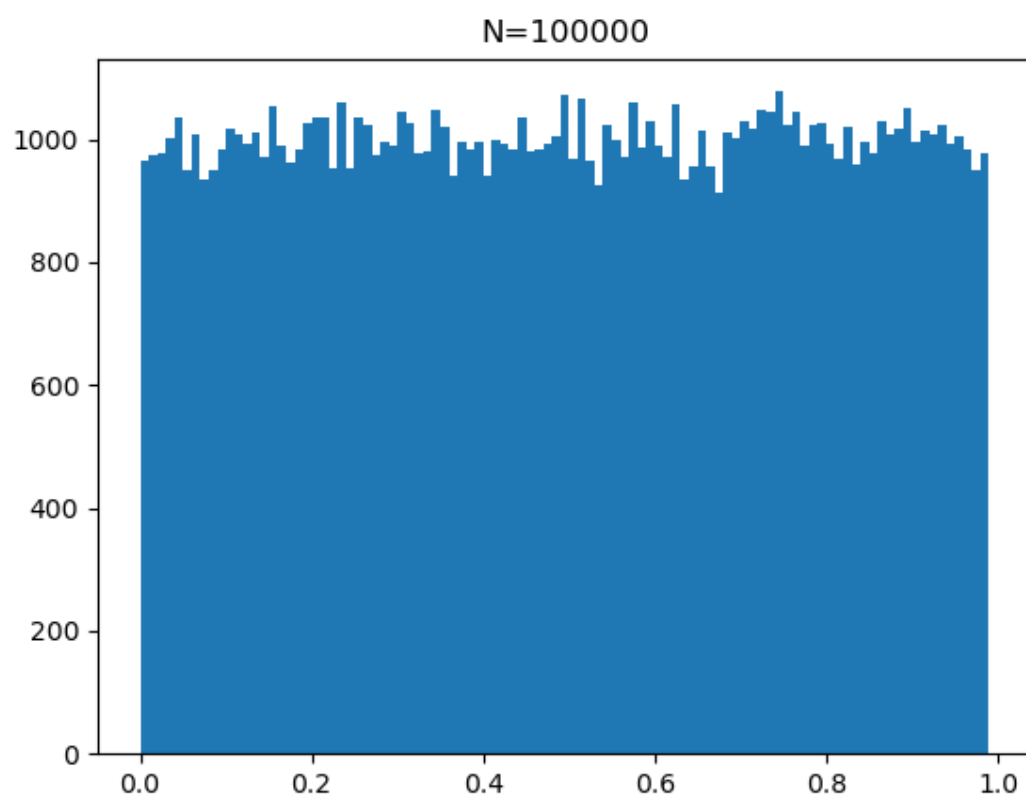




- Plot for $n=10000$



- Plot for n= 100000



CONCLUSIONS

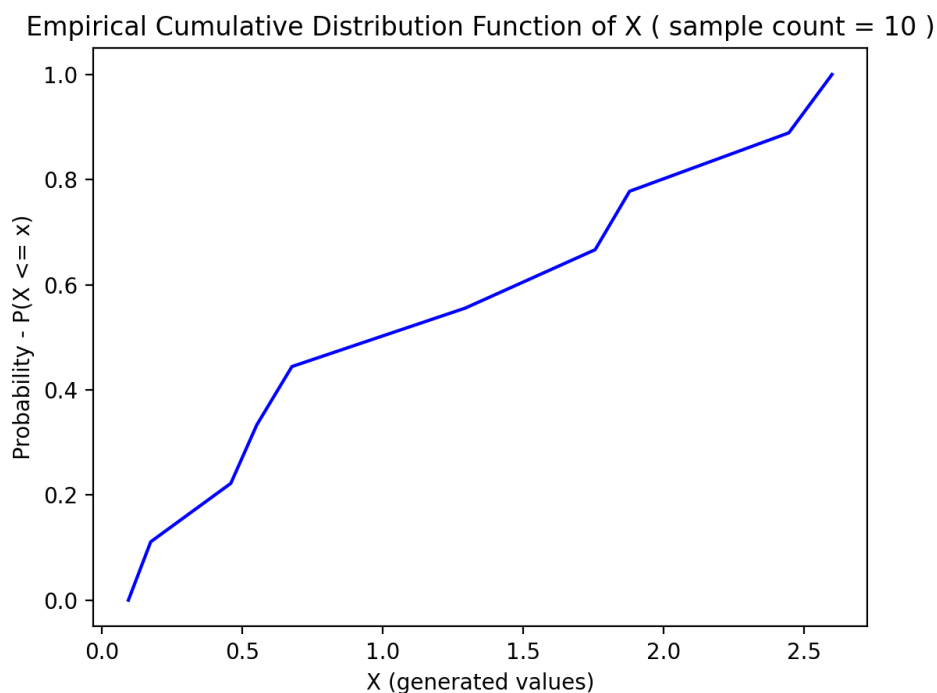
- 1) This scatter plot seems random. There does not seem to be a definite relationship between U_i and U_{i+1} . U_i 's do not follow a particular pattern and thus are random.
- 2) The histogram and Kde plot suggest that the lagged Fibonacci generator closely follows what an ideal generator should follow i.e. all the values sandwiched between 0 and 1 and All U_i to be mutually independent.
- 3) Also as we increase the value of n (or size you can say) we can see all histogram bars getting at the same level(One can also see the KDE plot line getting straighter as n increases).
When $n=1000$ we can histogram bars height differing considerably but as n increases say to 100000 height differences reduces. It pertains to having uniform distribution.

Question 2

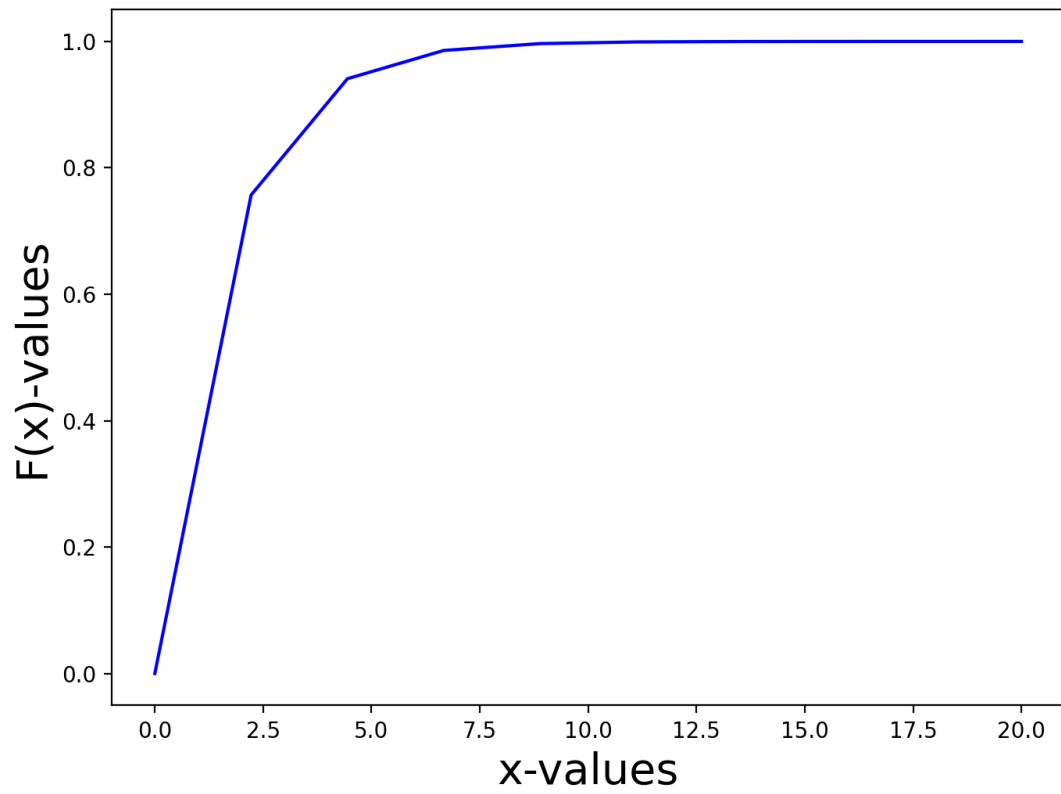
Assumed mean = $\pi/2(1.57)$

Below are graphs for different values of n :-

Plot for $n=10$



Actual CDF of $F(x)$
Number of rounds in Simulation = 10



Plot for n=100



Figure 1

Empirical Cumulative Distribution Function of X (sample count = 100)

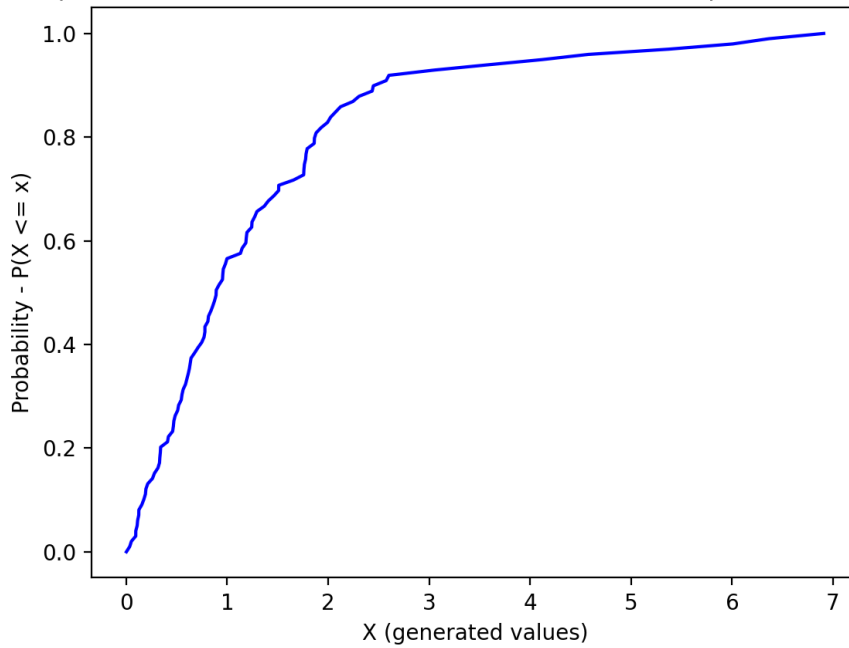
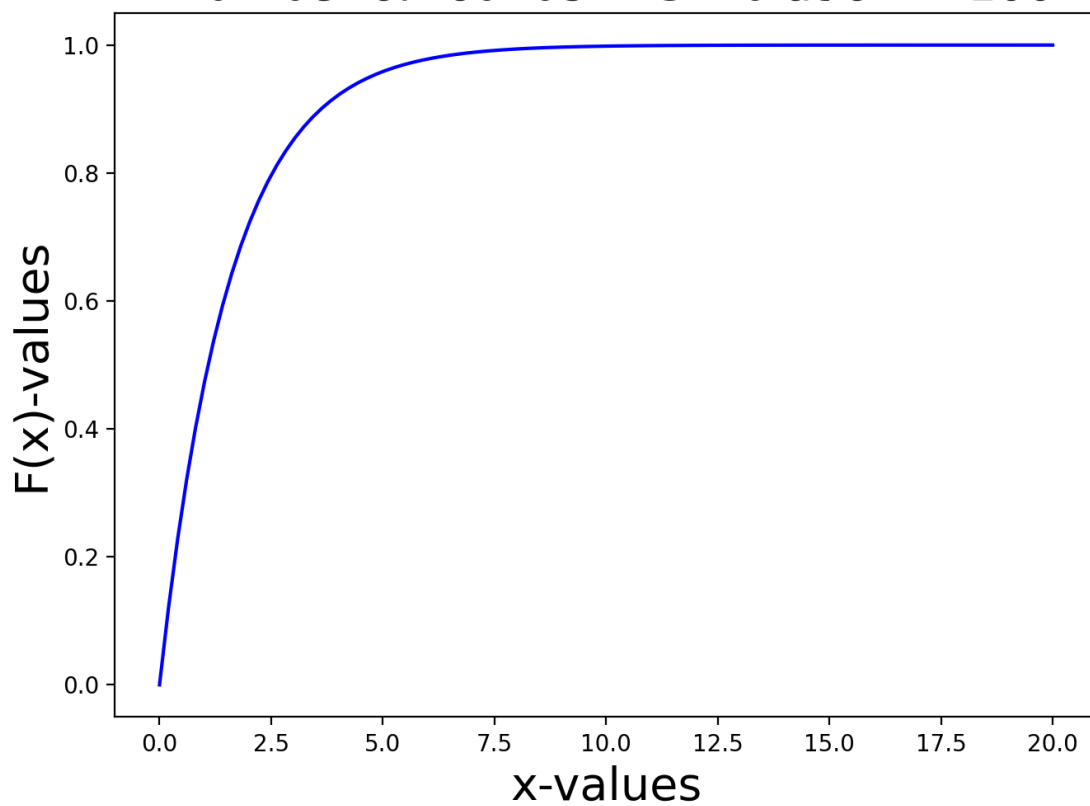


Figure 1

Actual CDF of F(x)

Number of rounds in Simulation = 100



Plot for n=1000

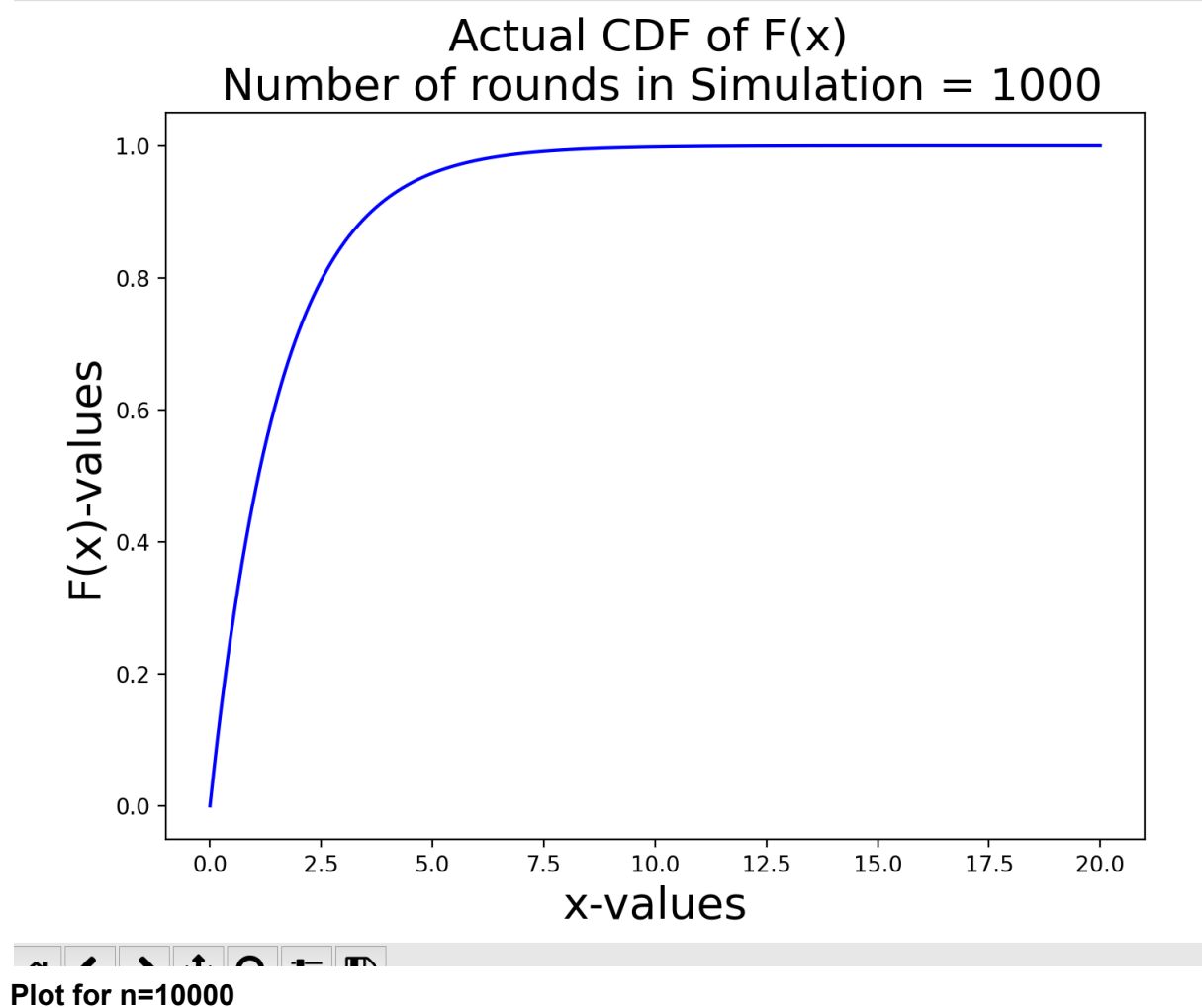
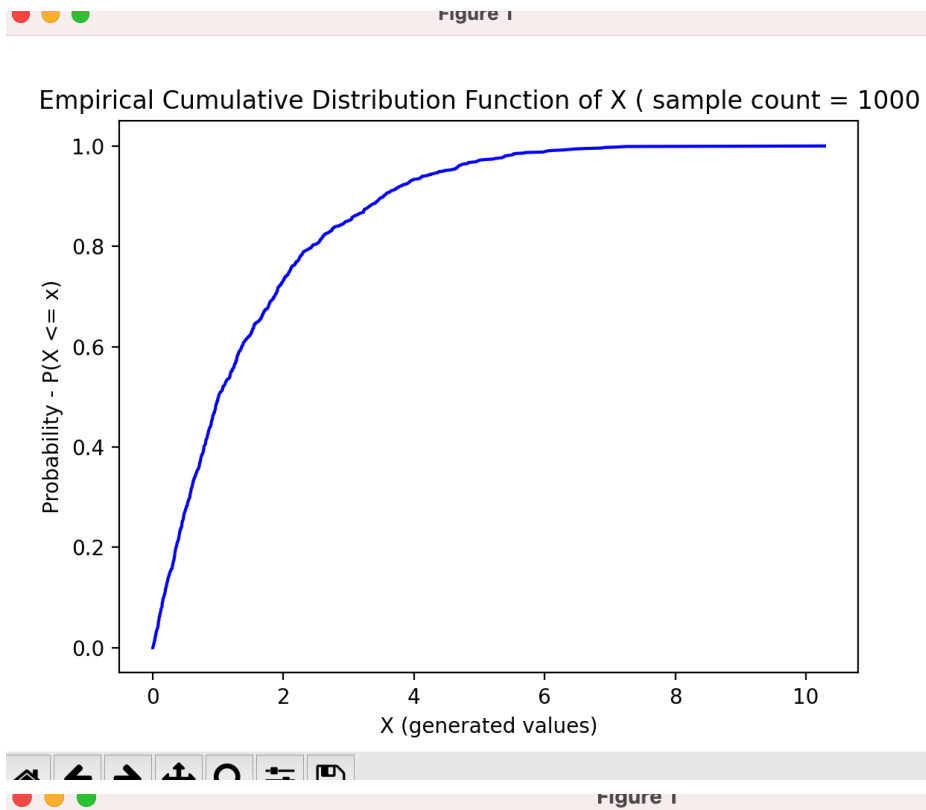


Figure 1

Empirical Cumulative Distribution Function of X (sample count = 10000)

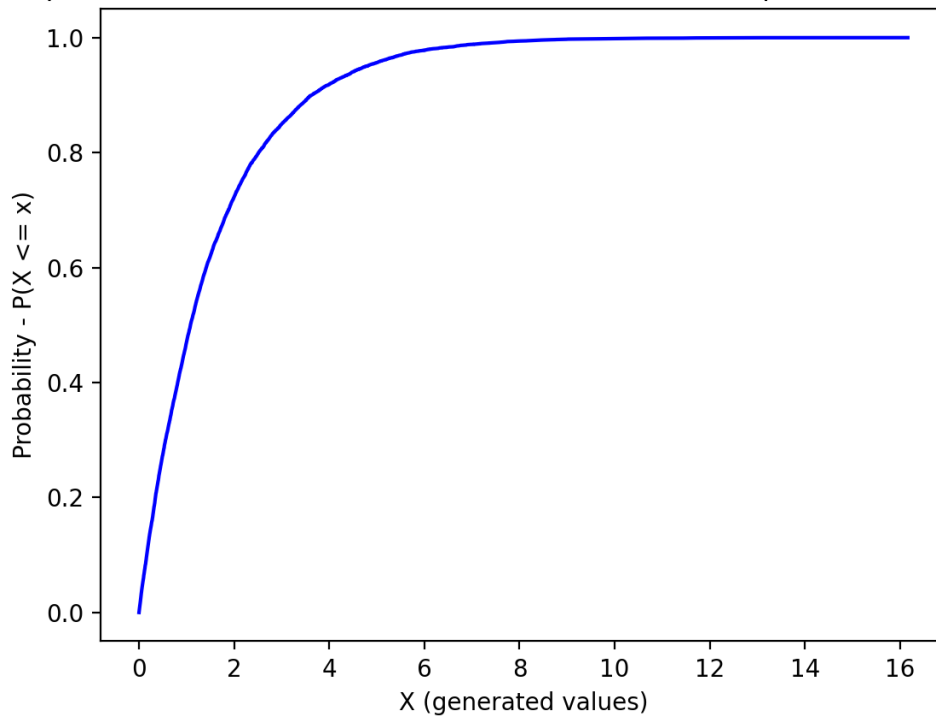
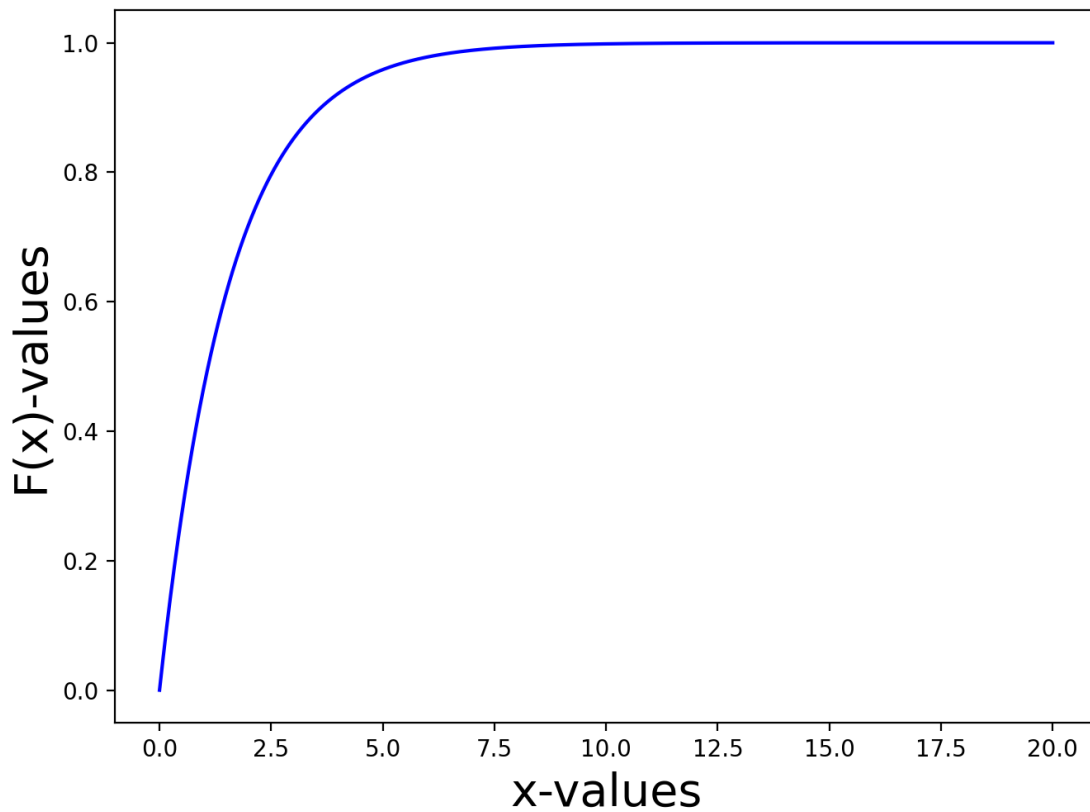


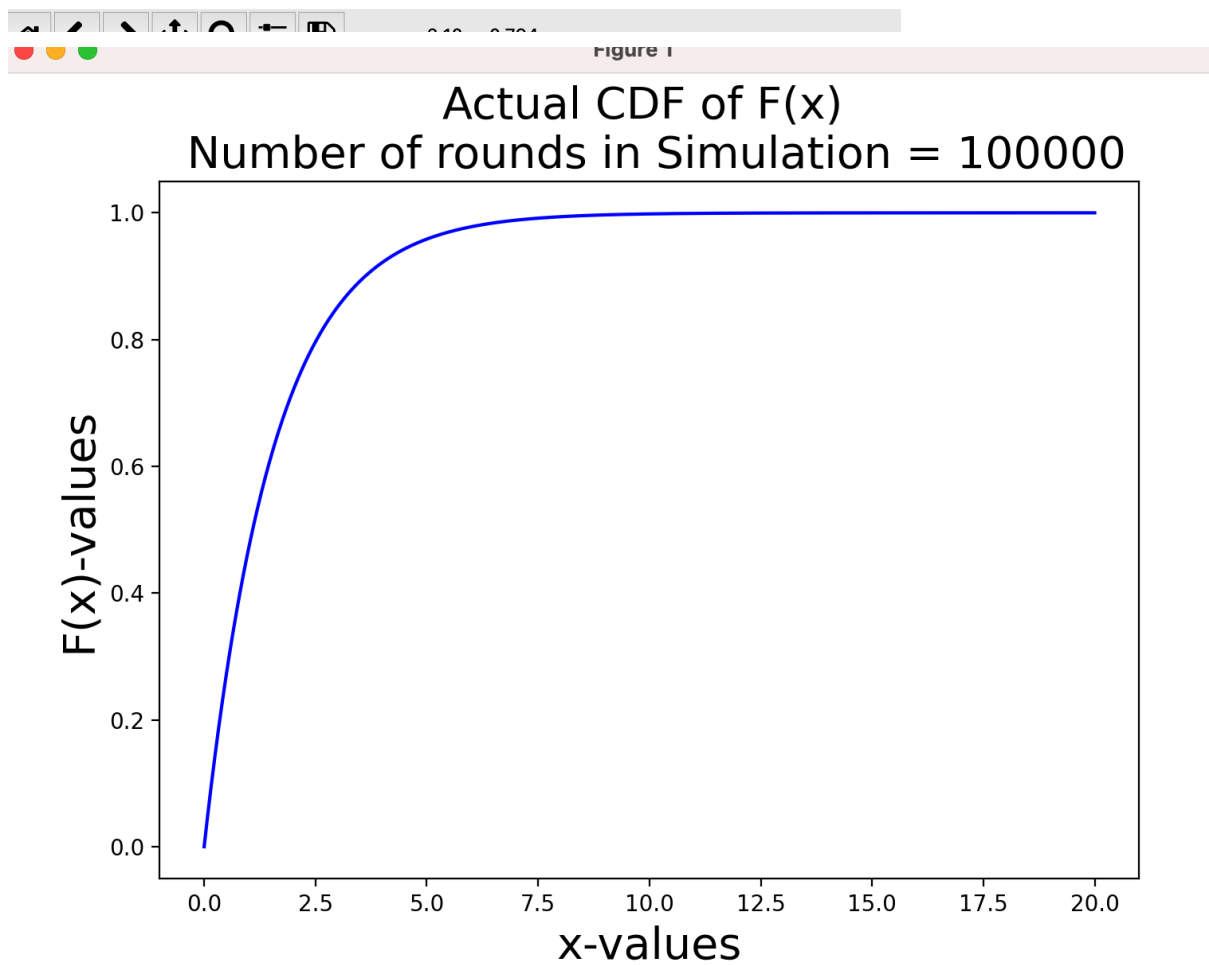
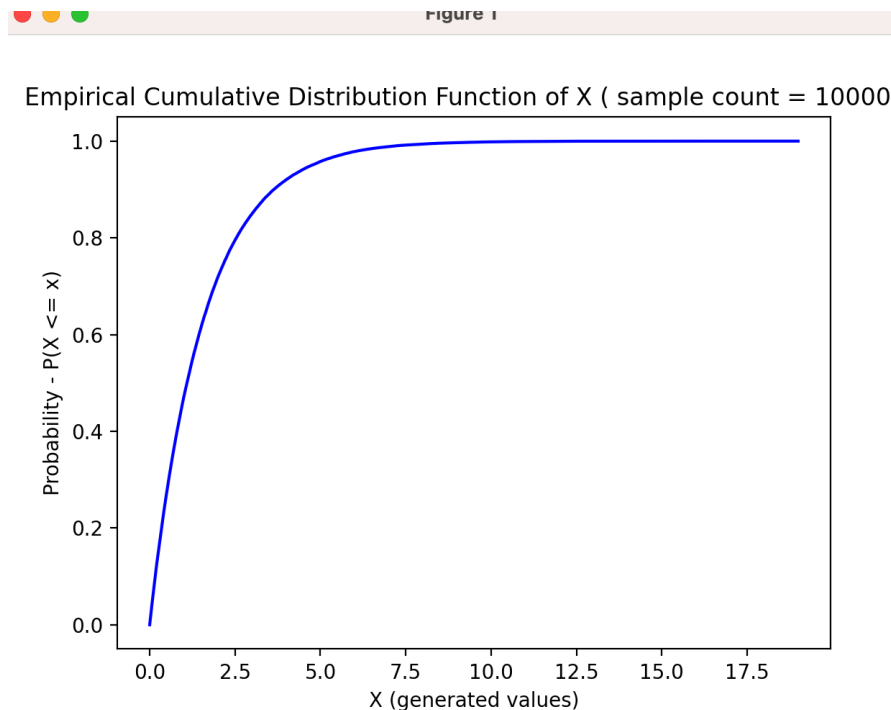
Figure 1

Actual CDF of F(x)

Number of rounds in Simulation = 10000



Plot for n=100000



Conclusion:

1) Actual mean is approx 1.570796

2) Actual variance is approx 2.467401

Here variance = mean*mean and mean is $\pi/2$.

Sample count = 10

Mean = 1.1936962912966962

Variance = 0.7814990069541273

Sample count = 100

Mean = 1.2994460101344962

Variance = 1.762657457409761

Sample count = 1000

Mean = 1.4965421325216637

Variance = 1.9869188361903516

Sample count = 10000

Mean = 1.5701159836132759

Variance = 2.488296189302775

Sample count = 100000

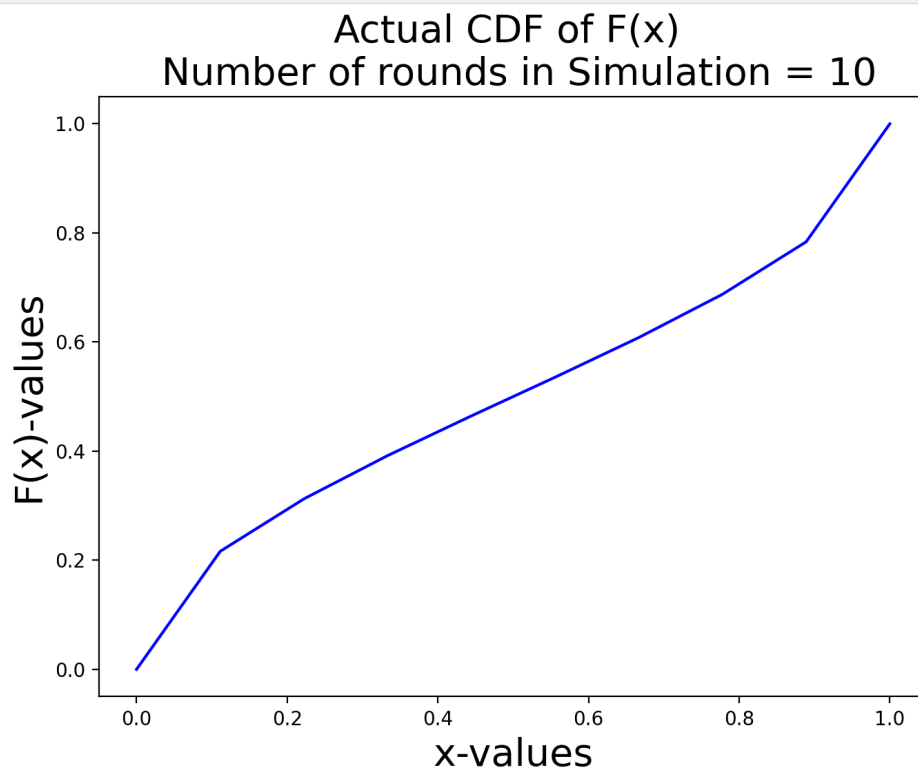
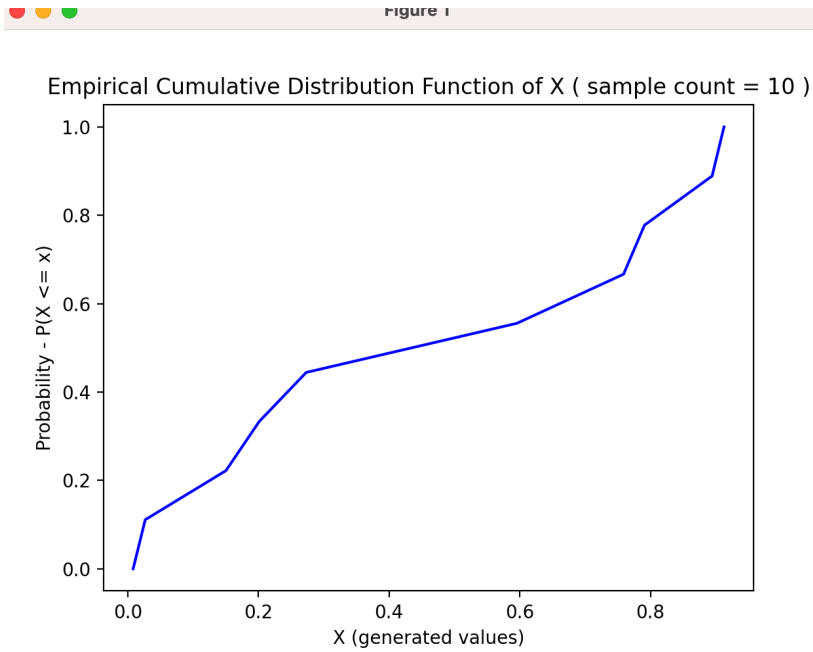
Mean = 1.572536789079683

Variance = 2.461267751081261

As we increase the value of n we can see sample mean and sample variance coming closer to actual mean and actual variance.

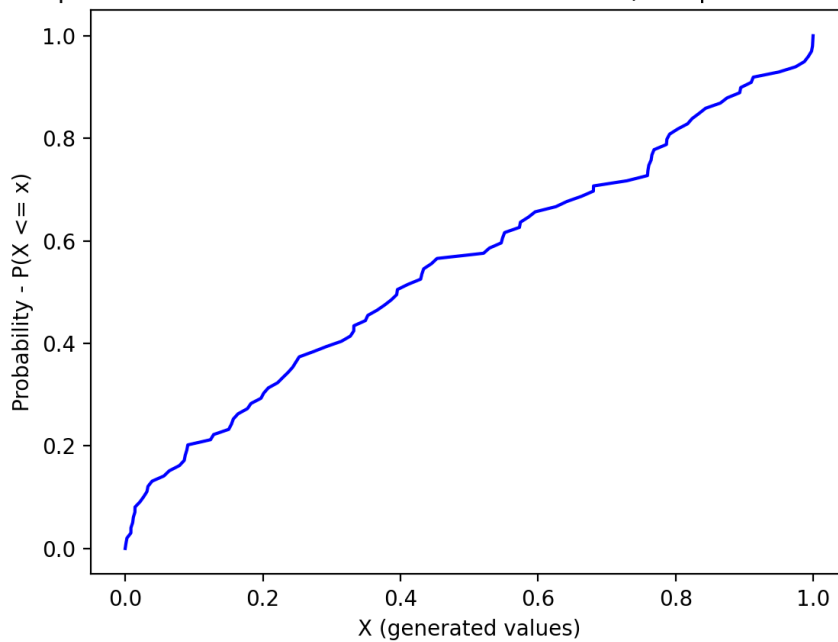
Question 3

Plot for $n=10$

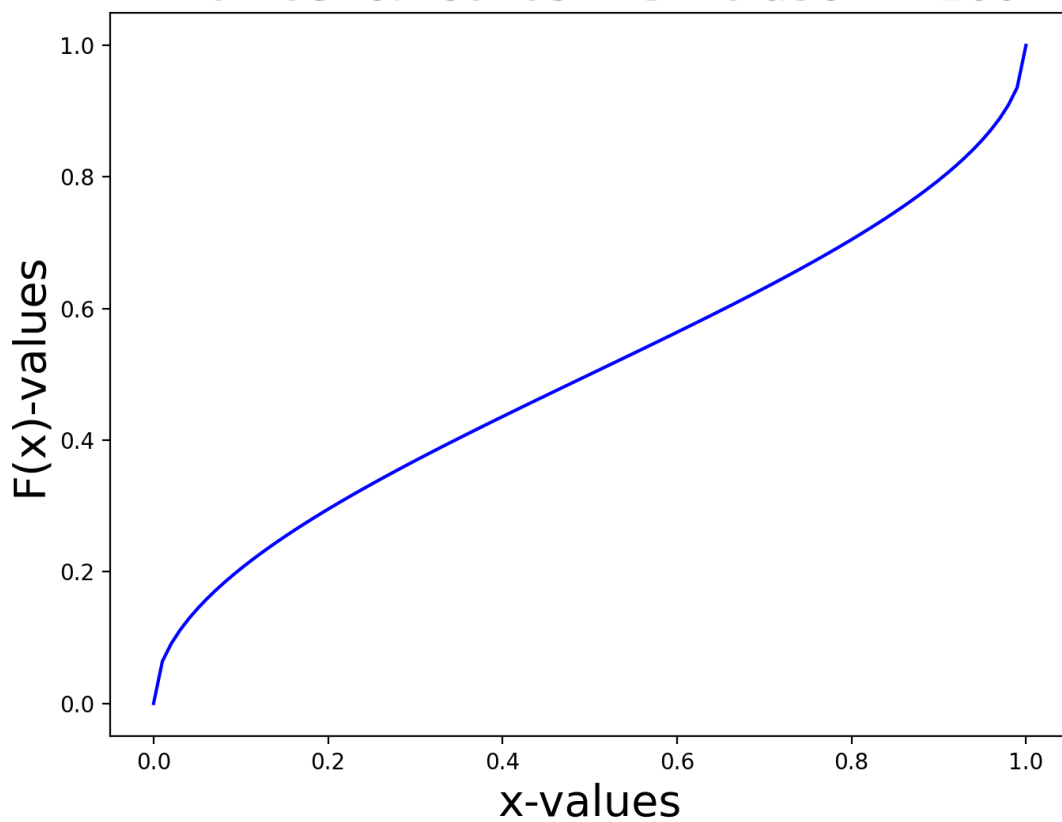


Plot for n=100

Empirical Cumulative Distribution Function of X (sample count = 100)

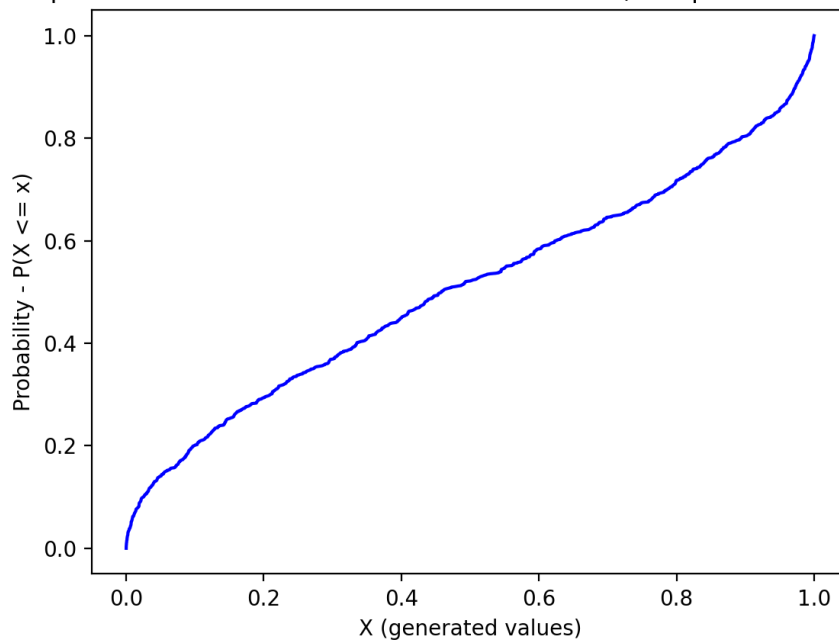


Actual CDF of F(x)
Number of rounds in Simulation = 100

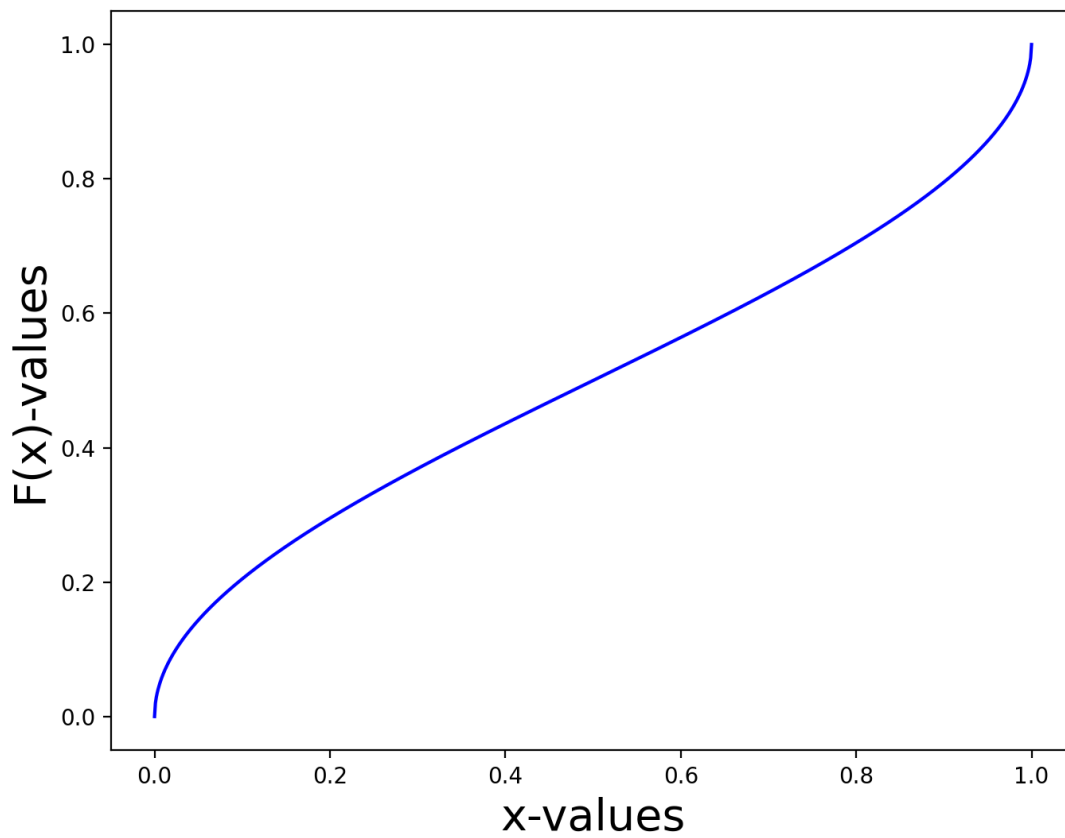


Plot for n=1000

Empirical Cumulative Distribution Function of X (sample count = 1000)

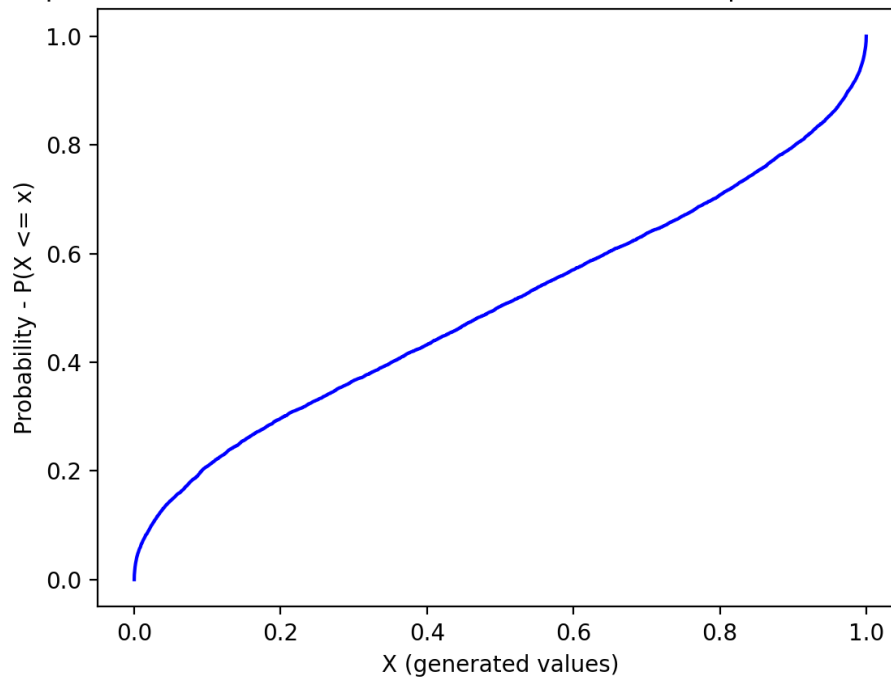


Actual CDF of F(x)
Number of rounds in Simulation = 1000

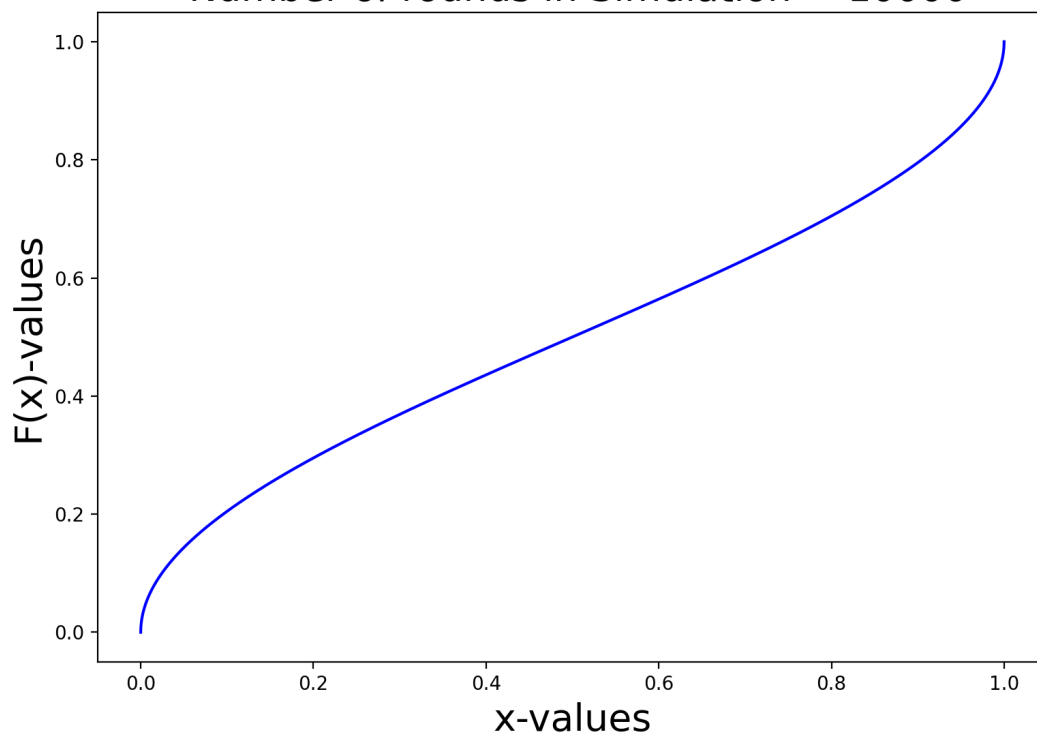


Plot for $n=10000$

Empirical Cumulative Distribution Function of X (sample count = 10000)

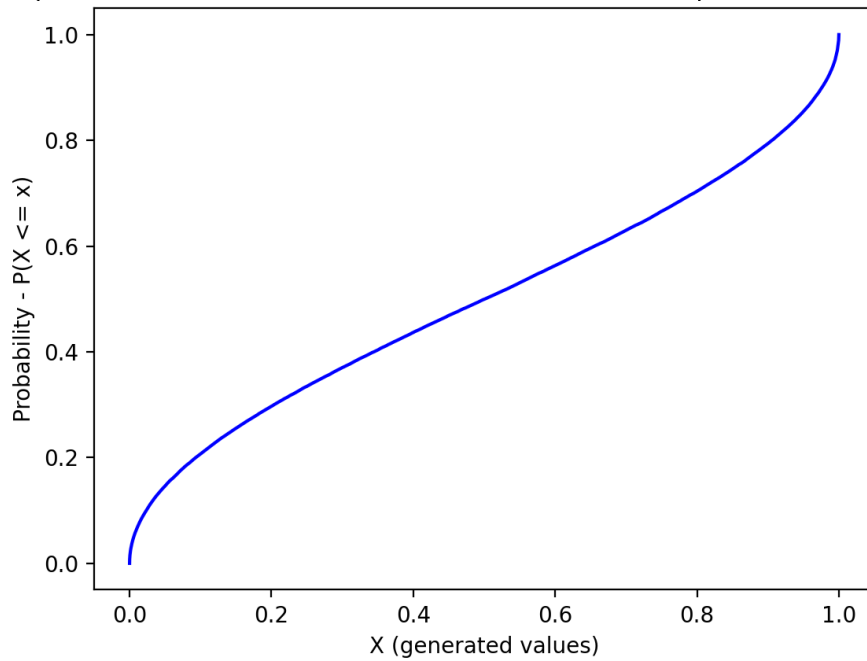


Actual CDF of F(x)
Number of rounds in Simulation = 10000



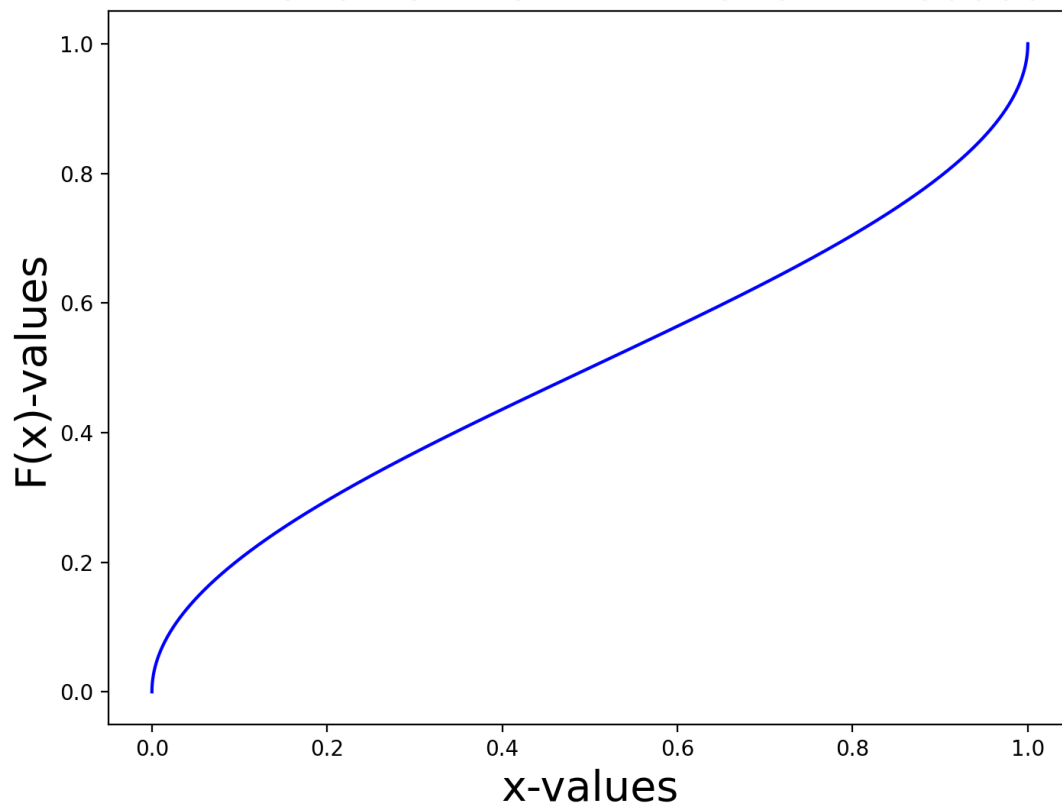
Plot for n= 100000

Empirical Cumulative Distribution Function of X (sample count = 100000)



Actual CDF of F(x)

Number of rounds in Simulation = 100000



CONCLUSION:

- 1) Actual mean of the distribution is 0.5.
- 2) Actual variance of distribution is 0.125.

Sample count = 10
Mean = 0.4613610544911812
Variance = 0.12002030136126798

Sample count = 100
Mean = 0.4451046366945898
Variance = 0.10441125911733588

Sample count = 1000
Mean = 0.49173183519814806
Variance = 0.12251431095826601

Sample count = 10000
Mean = 0.4985638361801597
Variance = 0.12464348755138321

Sample count = 100000
Mean = 0.4998199454583099
Variance = 0.12574974575513317

As we increase the value of n we can see sample mean and sample variance coming closer to actual mean and actual variance.

The reason the distribution of the random variable X matches the cumulative distribution function (CDF) $F(x)$ is because $F(x)$ is a smooth and always increasing function. When we use a uniform distribution U on the interval $[0, 1]$ and apply the inverse of F , denoted as $F^{-1}(U)$, the result represents a sample from the distribution of F . This relationship illustrates the concept of the Inverse Transform Method.

Question 4 :-

Discrete uniform distribution on $\{1, 3, 5, \dots, 9999\}$ for sample size 100000 is as follows :-

