Probabilistic Numerical Methods

Homework-1

I) 1. Random Number Generator

```
In [1]:
         # Simulating some realizations of random generator
         # Generating 20 random numbers between 10 and 30
         import random as rnd
         random_sample_1 = []
         for i in range(20):
           n = rnd.uniform(10,30)
           random_sample_1.append(n)
         random sample 1
Out[1]: [21.906249882596285,
         20.50614376571922,
         10.30812841208671,
         25.358483490923447,
         25.466759320694788,
         24.318258861177448,
         17.26787971853736,
         25.147708532945963,
         10.619646059484896,
         20.97706545356424,
         11.78356251570115,
         15.014885056099029,
         18.582971534106175,
         15.660003934227314,
         15.277905583402704,
         16.698574994762804,
         17.245228442399302,
         12.463812329904837,
         28.81244534090522,
         23.104011336247062]
In [2]:
         # Again simulating different set of random numbers using random generator
         # Generating 20 random numbers between 10 and 30
         random sample 2 = []
         for i in range(20):
           n = rnd.uniform(10,30)
           random sample 2.append(n)
         random_sample_2
Out[2]: [21.518853324986324,
         11.473937046401161,
         19.508119866657847,
         29.488707503223715,
         10.594972581309413,
         15.254661499851164,
         19.854759470710732,
         17.923050904945562,
         12.59708104005796,
         23.088790414631255,
         11.260005038799765,
         23.52018344488799,
```

```
19.581200649812544,

12.652498431353402,

15.834714826259972,

22.90710165543807,

10.902486970418996,

12.57367387376744,

18.510001955443183,

28.80827181946009]
```

COMMENT- It is evident that there is no co-relation between above generated two sample sets (random_sample_1, random_sample_2) and they appear to be completely random.

```
In [3]:
         # Simulating some realizations of random generator by fixing the seed value as 1000
         # Generating 20 random numbers between 10 and 30
         random sample 3 = []
         rnd.seed(1000)
         for i in range(20):
           n = rnd.uniform(10,30)
           random sample 3.append(n)
         random sample 3
Out[3]: [25.54713285401128,
         23.396511191184995,
         11.98279207849634,
         17.059410223802907,
         19.35815485801684,
         20.69367482941755,
         29.566181218247948,
         12.60630700317318,
         23.424869364605325,
         17.284588318947513,
         19.776714143239715,
         14.060244214681074,
         23.323967511427185,
         14.553260624138641,
         19.161281165935264,
         10.81444795109915,
         29.48579590755657,
         19.749521485378132,
         19.232277272747194,
         24.282943116164006]
In [4]:
         # Simulating again some realizations of random generator by fixing the seed value as 10
         # Generating 20 random numbers between 10 and 30
         random sample 4 = []
         rnd.seed(1000)
         for i in range(20):
           n = rnd.uniform(10,30)
           random sample 4.append(n)
         random_sample_4
Out[4]: [25.54713285401128,
         23.396511191184995,
         11.98279207849634,
         17.059410223802907,
         19.35815485801684,
         20.69367482941755,
         29.566181218247948,
```

```
12.60630700317318,
         23.424869364605325,
         17.284588318947513,
         19.776714143239715,
         14.060244214681074,
         23.323967511427185,
         14.553260624138641,
         19.161281165935264,
         10.81444795109915,
         29.48579590755657,
         19.749521485378132,
         19.232277272747194,
         24.282943116164006]
In [5]:
         # Simulating some realizations of random generator by fixing the seed value as 2000
         # Generating 20 random numbers between 10 and 30
         random sample 5 = []
         rnd.seed(2000)
         for i in range(20):
           n = rnd.uniform(10,30)
           random_sample_5.append(n)
         random sample 5
Out[5]: [18.96914035821057,
         28.94327428468175,
         19.489466540151703,
         25.36465955298137,
         25.81804569999036,
         28.682023115766356,
         15.081154660056557,
         25.918416854846306,
         21.49232680153613,
         12.779539939714255,
         25.107094854101213,
         13.757990423017867,
         26.39688069394593,
         12.920816024311923,
         23.156436177640764,
         24.2715193832855,
         16.876339096231685,
         18.470473413811263,
         29.308699523064156,
         14.127877463713634]
```

COMMENT- From above three samples it is evident that Random Number Generator generates the same set of random numbers if we use same seed value and different set of random numbers for different seed values.

I) 2. Quality of Random Number Generator

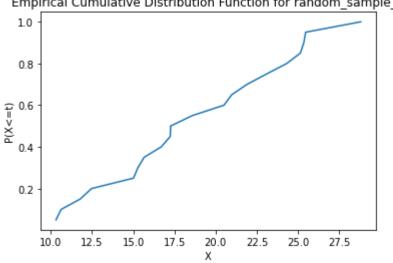
i) Plotting the Empirical Cumulative Distribution Function for random_sample_1

```
import numpy as np
from statsmodels.distributions.empirical_distribution import ECDF
from matplotlib import pyplot
ecdf = ECDF(random_sample_1)
# Printing intermediate cdf's
print('P(x<=15): %.3f' % ecdf(15))</pre>
```

```
print('P(x<=20): %.3f' % ecdf(20))</pre>
print('P(x<=25): %.3f' % ecdf(25))</pre>
print('P(x<=28): %.3f' % ecdf(28))</pre>
pyplot.plot(ecdf.x, ecdf.y)
pyplot.xlabel('X')
pyplot.ylabel('P(X<=t)')</pre>
pyplot.title("Empirical Cumulative Distribution Function for random sample 1")
pyplot.show()
```

P(x<=15): 0.200P(x <= 20): 0.550P(x <= 25): 0.800P(x <= 28): 0.950

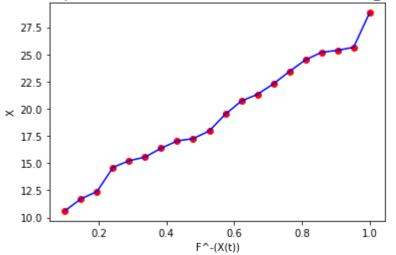
Empirical Cumulative Distribution Function for random sample 1



ii) Plotting the Inverse Empirical Cumulative Distribution Function for random sample 1

```
In [7]:
         from scipy.interpolate import interp1d
         # Ordering the Sample
         Ordered Sample = sorted(set(random sample 1))
         # Assigning Cumulative Probability Distribution to the Ordered Sample (P<=X)
         Ecdf Value = [ ecdf(item) for item in Ordered Sample]
         inverted edf = interp1d(Ecdf Value, Ordered Sample)
         x = np.linspace(0.1, 1, num=20)
         y = inverted \ edf(x)
         # Finding the respective value of random number for given percentile of cdf for compari
         print ('30 percentile:', inverted_edf(0.3))
         print ('60 percentile:', inverted_edf(0.6))
         print ('75 percentile:', inverted_edf(0.75))
         print ('80 percentile:', inverted_edf(0.8))
         pyplot.plot(x, y, 'ro', x, y, 'b-')
         pyplot.xlabel("F^-(X(t))")
         pyplot.ylabel('X')
         pyplot.title("Inverse Empirical Cumulative Distribution Function for random_sample_1")
        30 percentile: 15.277905583402704
        60 percentile: 20.50614376571922
        75 percentile: 23.104011336247062
        80 percentile: 24.31825886117745
Out[7]: Text(0.5, 1.0, 'Inverse Empirical Cumulative Distribution Function for random_sample_1')
```

Inverse Empirical Cumulative Distribution Function for random_sample_1

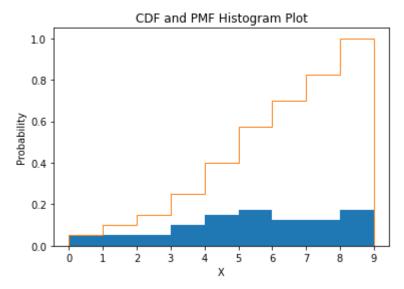


COMMENT It appears from the plots that both the functions are satisfying the properties of a CDF and Inverse CDF functions like Ranging from 0 to 1, Non-Decreasing and Right Continuous. Also the corresponding values "P(x <= 15): 0.150, P(x <= 20): 0.600, P(x <= 25): 0.850, P(x <= 28): 1.000" seems to match with there inverse "30 percentile: 16.881795822402125, 60 percentile: 19.974238562412452, 75 percentile: 21.660666306034074, 80 percentile: 22.890655546445117" approximately.

II) Simulating Discrete Random Variable

Here we take a random sample of a Discrete Integer Random Variable ranging from 0 to 9 and plot it's PMF and CDF

```
In [8]:
    data = [3,4,2,3,4,5,4,7,8,5,4,6,2,1,0,9,7,6,6,5,4,1,3,5,7,6,8,9,0,5,7,8,5,6,4,3,7,8,5,9
    pyplot.hist(data,bins=9, density=True, cumulative=True, histtype='step')
    pyplot.xlabel("X")
    pyplot.ylabel("Probability")
    pyplot.xticks(np.arange(0,10))
    pyplot.title("CDF and PMF Histogram Plot")
    pyplot.show()
```



COMMENT Like mentioned previously the CDF maintains all the properties and the Histogram Plot is like a step function which is typical for a Discrete Random Variable