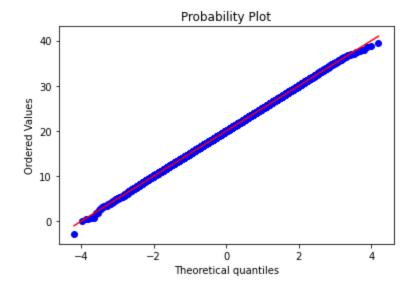
```
import numpy as np
import pylab
import scipy.stats as stats
import matplotlib.pyplot as plt
import seaborn as sns
import random
from sklearn import datasets
from sklearn.utils import resample
from sklearn.metrics import accuracy_score
```

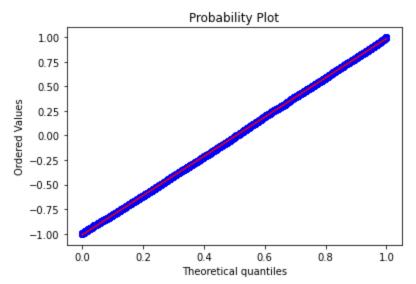
## ▼ QQ Plot

```
# generate 100 samples from N(20,5)
measurements = np.random.normal(loc = 20, scale = 5, size=50000)
#try size=1000
stats.probplot(measurements, dist="norm", plot=pylab)
pylab.show()
```



```
# generate 100 sanples from N(20,5)
measurements = np.random.uniform(low=-1, high=1, size=10000)
#try size=1000
stats.probplot(measurements, dist="norm", plot=pylab)
pylab.show()
```

```
# generate 100 samples from N(20,5)
measurements = np.random.uniform(low=-1, high=1, size=10000)
#try size=1000
stats.probplot(measurements, dist="uniform", plot=pylab)
pylab.show()
```



## ▼ Random Numbers Generator (random distribution)

```
print(random.random()) #0 to 1, uniform distribution
rands = [random.random() for i in range(1000)]
plt.scatter(np.linspace(0,1, num=1000),rands)
plt.show()

#CDF
cnt, edges = np.histogram(rands, bins=20)
plt.plot(edges[1:], np.cumsum(cnt/sum(cnt)))
```

```
0.3209522535255618
      1.0
      0.8
      0.6
      0.4
      0.2
      0.0
          0.0
                  0.2
                          0.4
                                  0.6
                                          0.8
                                                  1.0
     [<matplotlib.lines.Line2D at 0x7f7b713f8550>]
      1.0
#n - datapoints, sample them of size m uniformly
df = datasets.load iris().data;
n = df.shape[0]
m = 30
p = m/n
for j in range(1,5):
  sampled_data = []
  for i in range(n):
    if random.random() <= p:</pre>
      sampled_data.append(i)
  print(len(sampled data)) #need not be 30 always, roughly around 30
     28
     33
     27
```

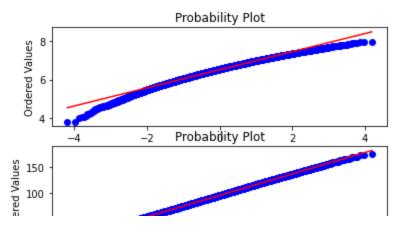
#### ▼ Box Cox Transformation

33

```
#power law distribution
fig, axes = plt.subplots(nrows = 2, ncols=1)
ax1 = axes[0]
X = stats.loggamma.rvs(5, size=50000) + 5
stats.probplot(X, dist="norm", plot=ax1)

Y, _ = stats.boxcox(X)
stats.probplot(Y, dist="norm", plot=axes[1])

plt.show()
```

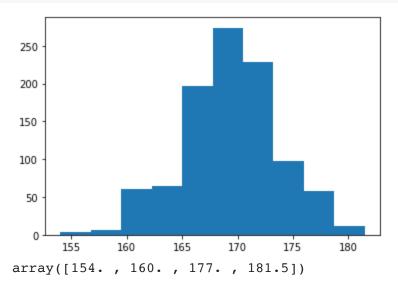


# ▼ Empirical bootstrap for CI

```
X = np.array([180,162,158,172,168,150,171,183,165,176])
k = 1000
m = int((len(X)*0.8))
md = []
for i in range(k):
    sample = resample(X, n_samples = m)
    md.append(np.median(sample))

plt.hist(md)
plt.show()

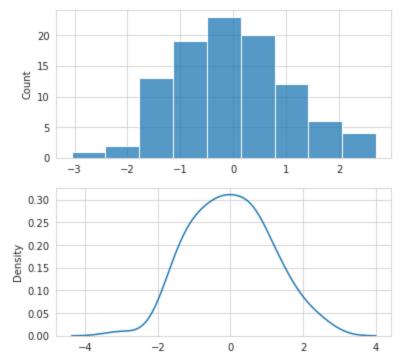
alpha = 0.95 #95% CI
range_n = np.array([0, ((1-alpha)/2)*100, 100*((1+alpha)/2),100])
md = np.sort(md)
np.percentile(md, range_n)
```



### → KS Test

```
X = stats.norm.rvs(size=100)
fig, axes = plt.subplots(nrows = 2, ncols=1, figsize=(6,6))
sns.set_style('whitegrid')
#sns.scatterplot(x=range(0,1000), y=X)
```

```
sns.histplot(X, ax=axes[0])
sns.kdeplot(X, ax=axes[1])
plt.show()
print(stats.kstest(X, 'norm'))
```



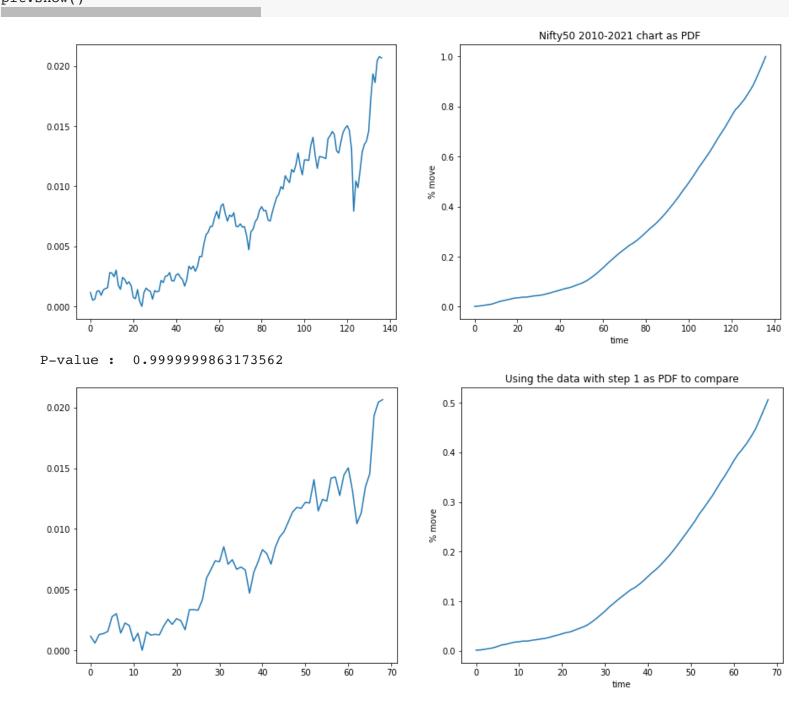
KstestResult(statistic=0.07403083345749373, pvalue=0.6368378097402114)

```
X = stats.uniform.rvs(size=1000)
print(X[0:50])
fig, axes = plt.subplots(nrows=2, ncols=1, figsize=(6,6))
sns.set_style('whitegrid')
#sns.scatterplot(data=X, ax=axes[0])
sns.histplot(X, ax=axes[0])
sns.kdeplot(X, bw=0.5, ax=axes[1])
plt.show()
print(stats.kstest(X, 'norm'))
```

```
0.78349851 0.25667239 0.11282044 0.26346924 0.81493582 0.45306526
      0.37891529 0.31140773 0.4583943 0.17486253 0.83452177 0.1480231
      0.03743587 0.86591515 0.81731302 0.69751853 0.90149015 0.40924313
      0.32390345 0.65339424 0.57019138 0.3665048 0.36414264 0.89174567
      0.27038761 0.42762324 0.24452063 0.78719723 0.0614169 0.47512458
      0.00561059 \ 0.22037126 \ 0.41811627 \ 0.68009393 \ 0.20992509 \ 0.00196789
      0.36169091 0.29999564 0.18642837 0.947758
                                                 0.66751821 0.08403089
      0.05087222 0.88002339]
     /usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:1657: Futur
       warnings.warn(msq, FutureWarning)
       100
X = stats.norm.rvs(size=100)
Y = stats.uniform.rvs(size=1000)
Z = stats.norm.rvs(size=1000)
print(stats.ks_2samp(X, Y))
print(stats.ks 2samp(X, Z))
    Ks 2sampResult(statistic=0.55, pvalue=2.5991628500150062e-24)
    Ks_2sampResult(statistic=0.101, pvalue=0.2955254321543853)
#Thinking the stock market graph as PDF and comparing it with another
percentages = np.array([-6.13, 0.82, 6.64, 0.55, -3.63, 4.45, 1.04, 0.65, 11.62, -0.20, -2.58, 4.64, -10.2]
X = []
prev = 100
for i in range(len(percentages)):
  pp = percentages[i]/100
  if(i==0):
    X.append(100)
  val = prev*(1+(pp))
  X.append(val)
  prev = val
X_{norm} = ((X - np.min(X))/(np.max(X) - np.min(X)))
X norm = X norm/np.sum(X norm)
fig, axes = plt.subplots(nrows = 1, ncols=2, figsize=(15,6))
axes[0].plot(range(len(X norm)), X norm)
axes[1].plot(range(len(X norm)), np.cumsum(X norm))
plt.xlabel('time')
plt.ylabel('% move')
plt.title("Nifty50 2010-2021 chart as PDF")
plt.show()
#print(stats.kstest(X, lambda X: cdf norm))
step = 1
test = X norm[::2]
d,p_val = stats.ks_2samp(X_norm, test)
print("P-value : ", p val)
fig, axes = plt.subplots(nrows = 1, ncols=2, figsize=(15,6))
axes[0].plot(range(len(test)), test)
axes[1].plot(range(len(test)), np.cumsum(test), label='Test CDF')
plt.xlabel('time')
plt.ylabel('% move')
```

[0.31906831 0.81668202 0.46445003 0.60371156 0.96736522 0.8039904

plt.title("Using the data with step {} as PDF to compare".format(step))
plt.show()



×