# **Statistical Functions in Python**

```
In [25]: import numpy as np
import pandas as pd

In [2]: import statistics as stat
```

# **Central Tendency**

## **Mean Function**

# Mean is the sum of the value of each observations in data set divided by the number of observations

```
In [42]: from IPython.display import Image
Image(filename='E:/pyimages/mean0.png')

Out[42]:

Mean of Grouped Data:

\[ \overline{x} = \frac{\sum fx}{n} \]

where: \[ \vec{x} = mean \\ f = frequency of each class \\ x = mid-interval value of each class \\ n = total frequency \\ \sum fx = sum of the producst of \\ mid - interval values and \\ their corresponding frequency \end{array}
```

```
from IPython.display import Image
In [37]:
          Image(filename='E:/pyimages/mean.png')
Out[37]:
                                Mean
            1
           0.8
           0.6
                              σ
           0.4
           0.2
            0
                            3
                            х
In [38]: np.mean(n)
Out[38]: 7.75
In [3]: stat.mean([1,2,3,4])
Out[3]: 2.5
 In [4]: stat.mean([1,2,3,4,50])
Out[4]: 12
```

# Median

```
In [47]: from IPython.display import Image
          Image(filename='E:/pyimages/median.png')
Out[47]:
           50 % of the data ← → 50 % of the data
In [48]: np.median(n)
Out[48]: 8.0
In [5]: stat.median({1,2,3,4,50})
Out[5]: 3
In [6]: | stat.median([1,3,50,4,2])
Out[6]: 3
 In [7]: stat.median([1,2,30,50,60,55,40,50]) # in this case finds avg of two middle nu
Out[7]: 45.0
In [8]: | stat.median_low([1,3,50,4,2])
Out[8]: 3
In [9]: stat.median_high([1,6,3,50,4,2])
Out[9]: 4
```

### Mode

```
In [49]: from IPython.display import Image
          Image(filename='E:/pyimages/mode0.png')
Out[49]:
            6, 3, 9, 6, 6, 5, 9, 3
                               8
In [50]:
          from IPython.display import Image
          Image(filename='E:/pyimages/mode.png')
Out[50]:
                    mode = 3
                                               mode = \{2, 6\}
                  3 4 5 6
                          7 8
                                             2 3 4 5 6
In [10]: stat.mode([1,2,1,3,3,4,4,4,5])
Out[10]: 4
In [11]: stat.mode([1,2,3,4])
          StatisticsError
                                                     Traceback (most recent call last)
          <ipython-input-11-b9b6425c424e> in <module>
          ----> 1 stat.mode([1,2,3,4])
         C:\anaconda3\lib\statistics.py in mode(data)
                      elif table:
              504
              505
                          raise StatisticsError(
                                   'no unique mode; found %d equally common values' % le
          --> 506
          n(table)
              507
                                   )
              508
                      else:
         StatisticsError: no unique mode; found 4 equally common values
```

#### **Harmonic Mean**

```
In [ ]: stat.harmonic_mean([1,2,3,4])
```

#### **Standard Deviation Function**

[1,2,3,4,5] # consicutive difference is 1 no lot of disperssion among number

[1,2,500,600,-2] # dispession varies alot in number

```
In [21]: stat.stdev([1,2,3])
Out[21]: 1.0
In [22]: stat.stdev([1,2,500,600,-2])
Out[22]: 303.13726263856114
In [23]: stat.variance([1,2,3,4,5,600])
Out[23]: 59403.5
```

## **Creation of Population dataset**

```
population = np.random.randint(10,20,100)
In [56]:
         population
Out[56]: array([18, 11, 12, 18, 15, 13, 19, 18, 11, 19, 10, 19, 16, 11, 13, 15, 15,
                11, 14, 16, 18, 11, 19, 16, 19, 17, 13, 17, 12, 19, 11, 12, 17, 17,
                14, 14, 17, 19, 17, 16, 13, 10, 19, 14, 16, 16, 12, 18, 18, 16, 13,
                15, 14, 12, 19, 16, 14, 17, 18, 19, 10, 19, 11, 11, 15, 19, 11, 17,
                12, 15, 16, 16, 19, 19, 15, 10, 16, 11, 14, 18, 11, 15, 17, 14, 13,
                10, 10, 14, 11, 14, 12, 11, 12, 12, 14, 18, 12, 13, 14, 13])
In [57]: | np.mean(population)
Out[57]: 14.72
In [59]: np.median(population)
Out[59]: 15.0
In [61]: from statistics import mode
         mode(population)
Out[61]: 19
In [62]:
         sample=np.random.choice(population,20)
         sample
Out[62]: array([15, 13, 11, 16, 12, 17, 16, 17, 13, 19, 15, 10, 16, 19, 10, 12, 14,
                14, 11, 12])
```

```
In [63]: np.mean(sample)
Out[63]: 14.1
In [64]: np.median(sample)
Out[64]: 14.0
In [68]: sample1=np.random.choice(population,10)
         sample1
Out[68]: array([19, 11, 17, 14, 19, 18, 19, 19, 14, 11])
In [69]: np.mean(sample1)
Out[69]: 16.1
In [71]: np.median(sample1)
Out[71]: 17.5
In [72]:
         sample=np.random.choice(population, 20)
         sample1=np.random.choice(population,15)
         sample2=np.random.choice(population,10)
         sample3=np.random.choice(population,5)
In [75]: | all_samples=[sample,sample1,sample2,sample3]
         sample mean=[]
         for i in all samples:
             sample_mean.append(np.mean(1))
         sample_mean
Out[75]: [1.0, 1.0, 1.0, 1.0]
In [76]: np.mean(sample_mean)
Out[76]: 1.0
```

## Range

```
import numpy as np
In [7]:
         import pandas as pd
         from matplotlib import pyplot as plt
In [6]: n=np.random.randn(9)
Out[6]: array([ 0.64835067, -0.39925531, 0.64320203, 0.24283355,
                                                                    0.78986282,
                -0.96110071, 0.99804129, 0.22722202, -0.77016285])
In [8]:
        # Range:
         np.max(n)-np.min(n)
Out[8]: 1.959141997817723
         n=np.random.randn(30)
In [9]:
Out[9]: array([ 0.30660136, 0.18955531, -0.41204737, 0.4298422 ,
                                                                    0.63080927,
                -1.64107963, 0.2620718, -0.4127817, -1.82698773,
                                                                    0.0291693,
                 0.69688716, -0.30101659, -0.7317028, 0.61273998,
                                                                    0.411687
                 0.14236308, -0.90975577, -1.4865773 , 0.30066822,
                                                                    0.38880768,
                -1.48479099, -1.36498011, -0.8690589 , 0.46676093,
                                                                    0.88125612,
                 1.64162044, 0.39106158, 1.39715284, 0.17389935, 0.30019409])
In [10]: | np.max(n)-np.min(n)
Out[10]: 3.4686081689554884
```

# Interquatile Range

#### Interquartile range

## **IQR= q3-q1**

```
In [19]: IQR= q3-q1
In [20]: IQR
Out[20]: 0.0
```

## variance

In [25]: from IPython.display import Image Image(filename='E:/pyimages/meanvariance.png')

Out[25]: Population Sample

# of subjects 
$$N$$
  $n$ 

Mean  $\mu = \frac{\sum_{i=1}^{N} x_i}{N}$   $\bar{x} = \frac{\sum_{i=1}^{n} x_i}{n}$ 

Variance  $\sigma^2 = \frac{\sum_{i=1}^{N} (x_i - \mu)^2}{N}$   $S^2 = \frac{\sum_{i=1}^{n} (x_i - \bar{x})^2}{n-1}$ 

Note:  $S^2$  is the formula for unbiased sample variance, since we're dividing by n-1.

Note: Finding S by taking  $\sqrt{S^2}$  reintroduces bias.

```
population = np.random.randn(100)
In [23]:
         population
Out[23]: array([ 1.13364281, 0.20813156, -0.49039347, 0.47607173,
                                                                       1.57715744,
                 1.14729139, -0.51960049,
                                           1.13747426, -1.28885488,
                                                                       1.85517791,
                 0.11698179, -0.86450702, -1.06594138, -1.79239126,
                                                                       0.20973238,
                 0.05388449, 0.58854362, 1.55263027, 0.68690154, -0.90941053,
                              1.77251892, -0.36496575, 0.76970442, -0.38617487,
                 -1.88125526,
                 0.99580853,
                              0.87559861, 0.91986788, 1.46623713, -1.41431437,
                 0.52688283, -0.22368626, -0.32982899, -0.82980222,
                                                                       0.14615214,
                 -0.41513706, -0.13472317, -0.47941308, -0.29690663, -0.34841352,
                 0.12709846, 0.09934264, -1.56349637, -2.39193288, -0.69650291,
                 0.33057995, 0.54481002, -1.07170811, -0.24867409, 0.13860025,
                 0.1149794 ,
                               0.75312523, 0.01454702, 1.15784119,
                                                                       0.41790696,
                              1.935609 ,
                                            0.3724933 , -1.26240029, -1.1671613 ,
                 1.69612267,
                                            0.97325808, -0.14645279, -0.51579546,
                 -0.59061535,
                              0.74263151,
                 -1.26982565, -1.03846874,
                                            0.9101527 , -2.07570431, -2.06042914,
                 -0.53444461, 0.62360466, 1.2091826, 0.44100514, -0.37055556,
                 -1.27923513,
                              0.77171813, -0.47292255, -0.76284864, 0.84026693,
                 0.64853318, -1.88610845, 0.71826119, -0.33981284,
                                                                      1.09014468,
                 -0.6925029 , -0.72124843, -1.79760064, -0.27191069, -1.12522377,
                              1.00079606, 0.29802669, 1.9686116, 0.23098387,
                 0.74962558,
                              0.80799811, -0.18477792, -0.55540335, 0.23820809])
                 0.39512942,
In [24]: | np.var(population)
Out[24]: 0.9826764141230292
In [26]: | np.std(population)
Out[26]: 0.9913003652390274
         from IPython.display import Image
In [84]:
         Image(filename='E:/pyimages/typesstat.png')
Out[84]:
                           Statistics
                          Transition from
            Descriptive
                                          Inferential
                           Descriptive to
                            Inferential
                                          Statistics
             Statistics
                            Statistics
```

```
In [86]: from IPython.display import Image
Image(filename='E:/pyimages/typestat1.png')

Out[86]:

Descriptive vs. Inferential
Statistics

Descriptive
Methods for summarizing data
Summaries usually consist of graphs and numerical
summaries of the data
Inferential
Methods of making decisions or predictions about a populations based on sample information.
```

# **Descriptive statistics:**

Deals with presentation of collections of data

```
In [36]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          %matplotlib inline
In [41]: | df1 =pd.DataFrame(dict(id=range(6), age=np.random.randint(18,31,size=6)))
          df1
Out[41]:
             id
                age
          0
             0
                 27
             1
                 25
             2
          2
                 20
             3
          3
                 27
                 18
             5
                 19
In [46]:
         df1.age.mean()
          df1.mean()
Out[46]: id
                  2.500000
                 22.666667
          age
         dtype: float64
In [68]: df1["age"].mean()
Out[68]: 22.6666666666668
```

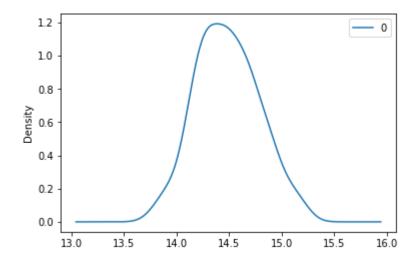
```
In [69]: df1.age.median()
Out[69]: 22.5
In [70]: df1.age.mode()
Out[70]: 0
         dtype: int32
In [71]: df1.age.var()
Out[71]: 17.06666666666667
In [72]: df1.age.std()
Out[72]: 4.1311822359545785
In [73]: df1.age.max()
Out[73]: 27
In [74]: df1.age.min()
Out[74]: 18
In [75]: # Range
         df1.age.max()-df1.age.min()
Out[75]: 9
In [76]: df1.boxplot(column='age',return_type='axes')
Out[76]: <matplotlib.axes._subplots.AxesSubplot at 0x1f1f367e7f0>
          26
          24
          22
          20
          18
                                  age
```

## **Skewness and Kurtosis**

#### **Inferential Statistics**

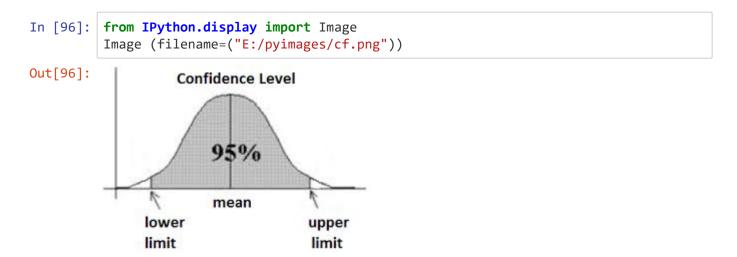
```
In [90]: pd.DataFrame(estimates).plot(kind="density")
```

Out[90]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1f1f3a3a898>



## **Point of Estimations**

#### **Confidence Interval**



## Margin of error

```
In [97]: from IPython.display import Image
            Image (filename=("E:/pyimages/mirro.png"))
 Out[97]:
             Formula
              MOE = Z \frac{\sigma}{\sqrt{n}}
                       critical value x standard error
                    critical value

    standard error

                     → standard deviation
                 n → sample size
               MOE → margin of error
                                               metcalc ...
 In [99]: # to find Critical value
            import scipy.stats as stats
            z critical = stats.norm.ppf(q=0.975) ##percentage point function
In [100]: t_critical = stats.t.ppf(q=0.975,df=24) ## df is degree of freedom (sample six
            e minus 1)
In [101]: margin_of_error = z_critical = (np.std(estimates)/np.sqrt(200))
In [103]: ## Lower : sample_mean_of_erroe
           np.mean(estimates)- margin_of_error
Out[103]: 14.464465800050748
In [104]: ## upper : sample_mean_of_erroe
           np.mean(estimates) + margin_of_error
Out[104]: 14.506634199949252
  In [ ]:
```