```
In [141...
           # importing the libraries
           import numpy as np
           import pandas as pd
In [143...
           # loading the csv file
           dataset=pd.read_csv(r"C:\Users\user\Documents\Salary_Data.csv")
          dataset.head()
In [145...
Out[145...
              YearsExperience Salary
           0
                          1.1
                               39343
           1
                          1.3 46205
           2
                          1.5 37731
           3
                          2.0 43525
           4
                          2.2 39891
```

#### Mean

In [148	<pre>dataset.mean() # this gives the entire dataframe mean</pre>			
Out[148	YearsExperience 5.313333 Salary 76003.000000 dtype: float64			
In [150	dataset['Salary'].mean() # this gives that particular column mean			
Out[150	76003.0			

# Meadian

#### Mode

In [158... dataset.mode() # this gives the mode of that entire dataframe

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	YearsExperience	Salary
0	3.2	37731
1	4.0	39343
2	NaN	39891
3	NaN	43525
4	NaN	46205
5	NaN	54445
6	NaN	55794
7	NaN	56642
8	NaN	56957
9	NaN	57081
10	NaN	57189
11	NaN	60150
12	NaN	61111
13	NaN	63218
14	NaN	64445
15	NaN	66029
16	NaN	67938
17	NaN	81363
18	NaN	83088
19	NaN	91738
20	NaN	93940
21	NaN	98273
22	NaN	101302
23	NaN	105582
24	NaN	109431
25	NaN	112635
26	NaN	113812
27	NaN	116969
28	NaN	121872
29	NaN	122391

In [160... dataset['Salary'].mode() # gives mode of that particular column

```
Out[160...
                   37731
           1
                   39343
           2
                   39891
           3
                   43525
           4
                   46205
           5
                   54445
           6
                   55794
           7
                   56642
           8
                   56957
           9
                   57081
           10
                   57189
           11
                   60150
           12
                   61111
           13
                   63218
           14
                   64445
           15
                   66029
           16
                   67938
           17
                   81363
                   83088
           18
           19
                   91738
                   93940
           20
           21
                   98273
           22
                  101302
                  105582
           23
           24
                  109431
           25
                  112635
           26
                  113812
           27
                  116969
           28
                  121872
                  122391
           29
           Name: Salary, dtype: int64
```

#### **Variance**

```
In [163...
           dataset.var()
Out[163...
           YearsExperience
                                8.053609e+00
           Salary
                                7.515510e+08
           dtype: float64
           dataset['Salary'].var()
In [165...
Out[165...
           751550960.4137931
```

#### **Standard Deviation**

```
In [168...
           dataset.std()
Out[168...
           YearsExperience
                                    2.837888
           Salary
                                27414.429785
           dtype: float64
In [170...
           dataset['Salary'].std()
Out[170...
           27414.4297845823
```

## **Coefficient of Variance**

## Correlation

Out[178		YearsExperience	Salary
	YearsExperience	1.000000	0.978242
	Salary	0.978242	1.000000

```
In [180... dataset['Salary'].corr(dataset['YearsExperience']) # gives correlation of the pa
Out[180... 0.9782416184887598
```

#### **Skewness**

In [183	dataset.skew()				
Out[183	YearsExperience Salary dtype: float64	0.37956 0.35412			
In [185	<pre>dataset['Salary'].skew()</pre>				
Out[185	0.354119679229591	.53			

#### Standard error

In [188	dataset.sem() # gives standard error of the entire dataset				
Out[188	YearsExperience 0.518125 Salary 5005.167198 dtype: float64				
In [190	dataset['Salary'].sem() # gives the standard error of the particular column				
Out[190	5005.167198052405				

#### **Zscore**

```
In [193... # for calculating the zscore we need to import the library import scipy.stats as stats
```

dataset.apply(stats.zscore)

0		$\Gamma =$		7	
U	uч	-	LЭ	5	

0-1.510053-1.3601131-1.438373-1.1055272-1.366693-1.4199193-1.187494-1.2049574-1.115814-1.3397815-0.864935-0.7183076-0.829096-0.5881587-0.757416-0.7998178-0.757416-0.4288109-0.578216-0.69801310-0.506537-0.47433311-0.470697-0.74976912-0.470697-0.70662013-0.434857-0.70202014-0.291498-0.55250415-0.148138-0.29921716-0.076458-0.37004317-0.0047790.262859180.2102610.198860190.2461000.665476200.5328190.583780210.6403390.826233220.9270580.938611231.0345771.402741241.2137771.240203251.3212961.097402261.5004961.519868271.5363361.359074281.7872151.721028291.8588941.701773		YearsExperience	Salary
2 -1.366693 -1.419919 3 -1.187494 -1.204957 4 -1.115814 -1.339781 5 -0.864935 -0.718307 6 -0.829096 -0.588158 7 -0.757416 -0.799817 8 -0.757416 -0.428810 9 -0.578216 -0.698013 10 -0.506537 -0.474333 11 -0.470697 -0.749769 12 -0.470697 -0.706620 13 -0.434857 -0.702020 14 -0.291498 -0.552504 15 -0.148138 -0.299217 16 -0.076458 -0.370043 17 -0.004779 0.262859 18 0.210261 0.198860 19 0.246100 0.665476 20 0.532819 0.583780 21 0.640339 0.826233 22 0.927058 0.938611 23 1.034577 1.402741 24 1.213777 1.240203 25 1.321296 1.097402 26 1.500496 1.519868 27 1.536336 1.359074 28 1.787215 1.721028	0	-1.510053	-1.360113
3 -1.187494 -1.204957 4 -1.115814 -1.339781 5 -0.864935 -0.718307 6 -0.829096 -0.588158 7 -0.757416 -0.799817 8 -0.757416 -0.428810 9 -0.578216 -0.698013 10 -0.506537 -0.474333 11 -0.470697 -0.749769 12 -0.470697 -0.706620 13 -0.434857 -0.702020 14 -0.291498 -0.552504 15 -0.148138 -0.299217 16 -0.076458 -0.370043 17 -0.004779 0.262859 18 0.210261 0.198860 19 0.246100 0.665476 20 0.532819 0.583780 21 0.640339 0.826233 22 0.927058 0.938611 23 1.034577 1.402741 24 1.213777 1.240203 25 1.321296 1.097402 26 1.500496 1.519868 27 1.536336 1.359074 28 1.787215 1.721028	1	-1.438373	-1.105527
4-1.115814-1.3397815-0.864935-0.7183076-0.829096-0.5881587-0.757416-0.7998178-0.757416-0.4288109-0.578216-0.69801310-0.506537-0.47433311-0.470697-0.74976912-0.470697-0.70662013-0.434857-0.70202014-0.291498-0.55250415-0.148138-0.29921716-0.076458-0.37004317-0.0047790.262859180.2102610.198860190.2461000.665476200.5328190.583780210.6403390.826233220.9270580.938611231.0345771.402741241.2137771.240203251.3212961.097402261.5004961.519868271.5363361.359074281.7872151.721028	2	-1.366693	-1.419919
5       -0.864935       -0.718307         6       -0.829096       -0.588158         7       -0.757416       -0.799817         8       -0.757416       -0.428810         9       -0.578216       -0.698013         10       -0.506537       -0.474333         11       -0.470697       -0.749769         12       -0.470697       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	3	-1.187494	-1.204957
6	4	-1.115814	-1.339781
7       -0.757416       -0.799817         8       -0.757416       -0.428810         9       -0.578216       -0.698013         10       -0.506537       -0.474333         11       -0.470697       -0.7049769         12       -0.470697       -0.702020         13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	5	-0.864935	-0.718307
8       -0.757416       -0.428810         9       -0.578216       -0.698013         10       -0.506537       -0.474333         11       -0.470697       -0.749769         12       -0.470697       -0.702020         13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	6	-0.829096	-0.588158
9-0.578216-0.69801310-0.506537-0.47433311-0.470697-0.74976912-0.470697-0.70662013-0.434857-0.70202014-0.291498-0.55250415-0.148138-0.29921716-0.076458-0.37004317-0.0047790.262859180.2102610.198860190.2461000.665476200.5328190.583780210.6403390.826233220.9270580.938611231.0345771.402741241.2137771.240203251.3212961.097402261.5004961.519868271.5363361.359074281.7872151.721028	7	-0.757416	-0.799817
10       -0.506537       -0.474333         11       -0.470697       -0.7049769         12       -0.470697       -0.706620         13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	8	-0.757416	-0.428810
11       -0.470697       -0.749769         12       -0.470697       -0.706620         13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	9	-0.578216	-0.698013
12       -0.470697       -0.706620         13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	10	-0.506537	-0.474333
13       -0.434857       -0.702020         14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	11	-0.470697	-0.749769
14       -0.291498       -0.552504         15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	12	-0.470697	-0.706620
15       -0.148138       -0.299217         16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	13	-0.434857	-0.702020
16       -0.076458       -0.370043         17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	14	-0.291498	-0.552504
17       -0.004779       0.262859         18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	15	-0.148138	-0.299217
18       0.210261       0.198860         19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	16	-0.076458	-0.370043
19       0.246100       0.665476         20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	17	-0.004779	0.262859
20       0.532819       0.583780         21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	18	0.210261	0.198860
21       0.640339       0.826233         22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	19	0.246100	0.665476
22       0.927058       0.938611         23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	20	0.532819	0.583780
23       1.034577       1.402741         24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	21	0.640339	0.826233
24       1.213777       1.240203         25       1.321296       1.097402         26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	22	0.927058	0.938611
25 1.321296 1.097402 26 1.500496 1.519868 27 1.536336 1.359074 28 1.787215 1.721028	23	1.034577	1.402741
26       1.500496       1.519868         27       1.536336       1.359074         28       1.787215       1.721028	24	1.213777	1.240203
<b>27</b> 1.536336 1.359074 <b>28</b> 1.787215 1.721028	25	1.321296	1.097402
<b>28</b> 1.787215 1.721028	26	1.500496	1.519868
	27	1.536336	1.359074
<b>29</b> 1.858894 1.701773	28	1.787215	1.721028
	29	1.858894	1.701773

In [195... stats.zscore(dataset['Salary'])

```
Out[195...
             -1.360113
          1
               -1.105527
          2
               -1.419919
          3
               -1.204957
          4
               -1.339781
          5
               -0.718307
               -0.588158
          6
               -0.799817
          7
          8
               -0.428810
               -0.698013
          9
          10
               -0.474333
          11
               -0.749769
          12
               -0.706620
          13
                -0.702020
          14
               -0.552504
          15
               -0.299217
          16
               -0.370043
          17
                0.262859
          18
                0.198860
          19
                0.665476
                0.583780
          20
          21
                0.826233
                0.938611
          22
          23
                1.402741
          24
                1.240203
          25
                1.097402
          26
              1.519868
                1.359074
          27
          28
                 1.721028
          29
                1.701773
          Name: Salary, dtype: float64
```

# **Degree of Freedom**

```
In [198... a=dataset.shape[0]
    b=dataset.shape[1]

degree_of_freedom=a-b
    print(degree_of_freedom)
```

# Sum Of Squares Of Regression(SSR)

```
In [206... # first we have to seperate the data into dependent and independent variables
    x=dataset.iloc[:,:-1].values # independent variable
    y=dataset.iloc[:,1].values # dependen variables
    y_mean=np.mean(y)
    from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.20,random_state=0)

from sklearn.linear_model import LinearRegression
    reg=LinearRegression()
    reg.fit(x_train,y_train)
    y_predict=reg.predict(x_test) # predicting the model
    y=y[0:6]
    SSR=np.sum((y_predict-y_mean)**2)
    print(SSR)
```

28

6263152884.284127

# Sum Of Squares Of Error(SSE)

```
In [220... # first we have to seperate the data into dependent and independent variables
    x=dataset.iloc[:,:-1].values # independent variable
    y=dataset.iloc[:,1].values # dependen variables
    from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.20,random_state=0)

from sklearn.linear_model import LinearRegression
    reg=LinearRegression()
    reg.fit(x_train,y_train)
    y_predict=reg.predict(x_test) # predicting the model
    y=y[0:6]
    SSE=np.sum((y-y_predict)**2)
    print(SSE)
```

15274062883.9432

# **Sum Of Squares Total**

```
In [225... mean_total=np.mean(dataset.values)
    SST=np.sum((dataset.values-mean_total)**2)
    print(SST)
```

108429703765.82735

### **R-Square**

```
In [228... r_square=SSR/SST r_square

Out[228... 0.05776233510524465

In []:
```