```
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
In [2]:
          wcat=pd.read csv("C:/Users/USER/Desktop/Datasets/wc.at.csv")
          wcat
                        ΑT
Out[2]:
               Waist
           0
               74.75
                      25.72
               72.60
           1
                      25.89
           2
               81.80
                      42.60
           3
               83.95
                      42.80
               74.65
                      29.84
         104 100.10 124.00
         105
               93.30
                      62.20
         106 101.80 133.00
         107 107.90 208.00
         108 108.50 208.00
        109 rows × 2 columns
In [3]:
          wcat.describe()
Out[3]:
                    Waist
                                  AT
         count 109.000000 109.000000
         mean
                 91.901835 101.894037
                 13.559116
           std
                            57.294763
           min
                 63.500000
                            11.440000
          25%
                 80.000000
                            50.880000
          50%
                 90.800000
                            96.540000
                104.000000 137.000000
          75%
          max 121.000000 253.000000
In [4]:
          wcat.info()
         <class 'pandas.core.frame.DataFrame'>
```

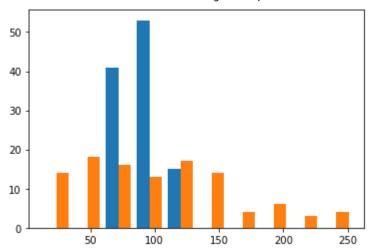
```
RangeIndex: 109 entries, 0 to 108
          Data columns (total 2 columns):
               Column Non-Null Count Dtype
                                        float64
           0
               Waist
                       109 non-null
           1
               ΑT
                       109 non-null
                                        float64
          dtypes: float64(2)
          memory usage: 1.8 KB
 In [5]:
           wcat.median()
                   90.80
          Waist
 Out[5]:
                   96.54
          dtype: float64
 In [6]:
           wcat.mode()
 Out[6]:
             Waist
                     AT
              94.5 121.0
             106.0 123.0
             108.5
                    NaN
 In [7]:
           wcat.isnull().sum()
          Waist
                   0
 Out[7]:
          ΑT
                   0
          dtype: int64
 In [8]:
           wcat.std()
          Waist
                   13.559116
 Out[8]:
                   57.294763
          dtype: float64
 In [9]:
           wcat.var()
          Waist
                    183.849626
 Out[9]:
                   3282.689835
          dtype: float64
In [10]:
           range_AT=wcat["AT"].max()-wcat["AT"].min()
           range_AT
          241.56
Out[10]:
In [11]:
           range_Waist=wcat["Waist"].max()-wcat["Waist"].min()
           range_Waist
          57.5
Out[11]:
```

```
wcat.kurtosis()
In [12]:
          Waist
                   -1.102667
Out[12]:
                   -0.285576
          dtype: float64
In [13]:
           wcat.skew()
          Waist
                    0.134056
Out[13]:
                    0.584869
          dtype: float64
In [14]:
           wcat.hist()
          array([[<AxesSubplot:title={'center':'Waist'}>,
Out[14]:
                   <AxesSubplot:title={'center':'AT'}>]], dtype=object)
                        Waist
                                                      AΤ
                                        17.5
           20.0
          17.5
                                        15.0
          15.0
                                        12.5
          12.5
                                        10.0
           10.0
                                         7.5
            7.5
                                         5.0
            5.0
                                         2.5
            2.5
           0.0
                                         0.0
                     80
                            100
                                   120
                                                   100
                                                            200
In [15]:
           boxpot=wcat.boxplot()
           250
           200
          150
          100
            50
                          Waist
In [16]:
           plt.plot(wcat)
          [<matplotlib.lines.Line2D at 0x233cad3e460>,
Out[16]:
           <matplotlib.lines.Line2D at 0x233cad3e490>]
```

```
250 -
200 -
150 -
100 -
50 -
0 20 40 60 80 100
```

```
In [17]: plt.hist(wcat)
```

Out[17]: (array([[0., 0., 41., 53., 15., 0., 0., 0., 0., 0., 0.], [14., 18., 16., 13., 17., 14., 4., 6., 3., 4.]]), array([11.44 , 35.596, 59.752, 83.908, 108.064, 132.22 , 156.376, 180.532, 204.688, 228.844, 253.]), <a list of 2 BarContainer objects>)



```
In [18]: plt.scatter(x=wcat["Waist"],y=wcat["AT"])
```

Out[18]: <matplotlib.collections.PathCollection at 0x233cae2eeb0>

```
250 -
200 -
150 -
100 -
50 -
70 80 90 100 110 120
```

```
In [19]:
            np.corrcoef(x=wcat["Waist"],y=wcat['AT'])
                                , 0.81855781],
           array([[1.
Out[19]:
                   [0.81855781, 1.
                                             ]])
In [20]:
            import statsmodels.formula.api as smf
           mdl=smf.ols('AT~Waist',data=wcat).fit()
           mdl.params
           Intercept
                         -215.981488
Out[20]:
           Waist
                            3.458859
           dtype: float64
In [21]:
           mdl.summary()
                               OLS Regression Results
Out[21]:
               Dep. Variable:
                                          ΑT
                                                    R-squared:
                                                                   0.670
                     Model:
                                         OLS
                                                Adj. R-squared:
                                                                   0.667
                   Method:
                                 Least Squares
                                                     F-statistic:
                                                                   217.3
                             Mon, 28 Feb 2022
                                              Prob (F-statistic):
                                                                1.62e-27
                      Time:
                                      14:52:36
                                                Log-Likelihood:
                                                                 -534.99
           No. Observations:
                                         109
                                                           AIC:
                                                                   1074.
                                                           BIC:
               Df Residuals:
                                         107
                                                                   1079.
                  Df Model:
                                            1
            Covariance Type:
                                    nonrobust
```

Omnibus: 3.960 Durbin-Watson: 1.560

Prob(Omnibus): 0.138 Jarque-Bera (JB): 4.596

t P>|t|

-9.909 0.000

0.235 14.740 0.000

[0.025

-259.190

2.994

0.975]

3.924

-172.773

coef std err

21.796

3.4589

Intercept -215.9815

Waist

 Skew:
 0.104
 Prob(JB):
 0.100

 Kurtosis:
 3.984
 Cond. No.
 639.

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [22]:
            mdl.conf_int(0.05)
Out[22]:
                               0
                                            1
           Intercept
                     -259.190053 -172.772923
              Waist
                        2.993689
                                     3.924030
In [23]:
            mdl 1 = smf.ols('AT~np.log(Waist)',data = wcat).fit()
            mdl 1.params
           Intercept
                             -1328.341989
Out[23]:
           np.log(Waist)
                                317.135564
           dtype: float64
In [24]:
            mdl 1.summary()
                                OLS Regression Results
Out[24]:
               Dep. Variable:
                                           ΑT
                                                     R-squared:
                                                                    0.675
                     Model:
                                          OLS
                                                 Adj. R-squared:
                                                                    0.672
                    Method:
                                 Least Squares
                                                     F-statistic:
                                                                    222.6
                       Date: Mon, 28 Feb 2022 Prob (F-statistic): 6.80e-28
                      Time:
                                      14:52:45
                                                 Log-Likelihood:
                                                                  -534.11
           No. Observations:
                                          109
                                                           AIC:
                                                                    1072.
                Df Residuals:
                                          107
                                                            BIC:
                                                                    1078.
                  Df Model:
            Covariance Type:
                                    nonrobust
                                                                          0.975]
                               coef std err
                                                   t P>|t|
                                                               [0.025
               Intercept -1328.3420
                                     95.923 -13.848 0.000 -1518.498
                                                                       -1138.186
                                              14.918 0.000
           np.log(Waist)
                            317.1356 21.258
                                                              274.994
                                                                         359.277
                 Omnibus: 3.317
                                    Durbin-Watson: 1.599
           Prob(Omnibus): 0.190
                                  Jarque-Bera (JB): 2.908
                    Skew: 0.235
                                          Prob(JB): 0.234
```

Kurtosis: 3.647 **Cond. No.** 145.

Notes:

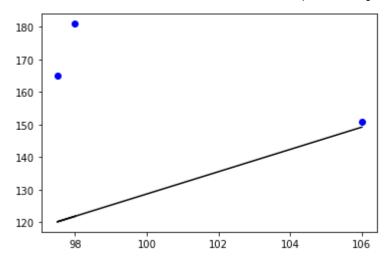
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [25]:
           print(mdl_1.conf_int(0.05))
                                      0
                                                    1
          Intercept
                          -1518.498012 -1138.185966
          np.log(Waist)
                            274.993649
                                          359.277480
In [26]:
           x=wcat.iloc[:,:1]
Out[26]:
                Waist
                74.75
                72.60
             1
            2
                81.80
            3
                83.95
                74.65
          104 100.10
          105
                93.30
          106 101.80
          107 107.90
          108 108.50
          109 rows × 1 columns
In [27]:
           y = wcat.iloc[:,1:2]
Out[27]:
                  AT
            0
                25.72
                25.89
             2
                42.60
                42.80
```

29.84

AT

```
104 124.00
          105
               62.20
          106 133.00
          107 208.00
          108 208.00
         109 rows × 1 columns
In [28]:
          from sklearn.model selection import train test split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size = 0.02,random_state = 1)
 In [ ]:
          from sklearn.preprocessing import MinMaxScaler
          scale=MinMaxScaler()
          S x=scale.fit(x)
          S_x=scale.transform(x)
          S_x=pd.DataFrame(S_x, columns=names)
          S_x
In [54]:
          from sklearn.linear model import LinearRegression
          from sklearn import metrics
          LR= LinearRegression()
          LR.fit(x train,y train)
         LinearRegression()
Out[54]:
In [55]:
          y pred = LR.predict(x test)
          y_pred
         array([[149.23424574],
Out[55]:
                 [120.10667673],
                 [121.82006314]])
In [49]:
          print('mean_square_error',metrics.mean_squared_error(ytest,ypred))
         mean square error 1840.2644296349736
In [50]:
          print('rmse',np.sqrt(metrics.mean squared error(ytest,ypred)))
         rmse 42.89830334214832
In [64]:
          plt.scatter(x_test, y_test, color='b')
          plt.plot(x_test, y_pred, color='k')
          plt.show()
```



In [65]:

N_wcat=wcat.iloc[0:50,:]

ΑT

25.72

In [66]:

Out[66]:

N_wcat

Waist

74.75

1	72.60	25.89
2	81.80	42.60
3	83.95	42.80
4	74.65	29.84
5	71.85	21.68
6	80.90	29.08
7	83.40	32.98
8	63.50	11.44
9	73.20	32.22
10	71.90	28.32
11	75.00	43.86
12	73.10	38.21

13

14

15

16

17

18

19

79.00

77.00

68.85

75.95

74.15

73.80

75.90

42.48

30.96

55.78

43.78

33.41

43.35

29.31

	Waist	AT
20	76.85	36.60
21	80.90	40.25
22	79.90	35.43
23	89.20	60.09
24	82.00	45.84
25	92.00	70.40
26	86.60	83.45
27	80.50	84.30
28	86.00	78.89
29	82.50	64.75
30	83.50	72.56
31	88.10	89.31
32	90.80	78.94
33	89.40	83.55
34	102.00	127.00
35	94.50	121.00
36	91.00	107.00
37	103.00	129.00
38	80.00	74.02
39	79.00	55.48
40	83.50	73.13
41	76.00	50.50
42	80.50	50.88
43	86.50	140.00
44	83.00	96.54
45	107.10	118.00
46	94.30	107.00
47	94.50	123.00
48	79.70	65.92
49	79.30	81.29

```
input=np.array(N_wcat["Waist"]).reshape(-1,1)
output=np.array(N_wcat["AT"]).reshape(-1,1)
```

```
X_train, X_test, Y_train, Y_test=train_test_split(input, output, test_size=0.25)
In [68]:
In [71]:
           regr=LinearRegression()
           regr.fit(X_train, Y_train)
           print(regr.score(X_test, Y_test))
          0.6673727245502035
In [73]:
           Y_pred = regr.predict(X_test)
           Y_pred
          array([[34.23556367],
Out[73]:
                 [51.43187331],
                 [39.57234942],
                 [82.26663542],
                 [51.43187331],
                 [97.38752839],
                 [60.32651623],
                 [42.53723039],
                 [42.38898634],
                 [ 5.47621823],
                 [72.18604012],
                 [32.45663508],
                 [73.6684806]])
In [74]:
           plt.scatter(X_test, Y_test, color='b')
           plt.plot(X_test, Y_pred, color='k')
           plt.show()
          140
          120
          100
           80
           60
           40
           20
                                75
                                       80
                                              85
                         70
                                                      90
                                                             95
 In [ ]:
           #done :)
```