```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.metrics import r2_score
from sklearn.linear_model import LinearRegression
import statsmodels.formula.api as smf
```

```
In [2]: data=pd.read_csv('wc_at.csv')
   data.head()
```

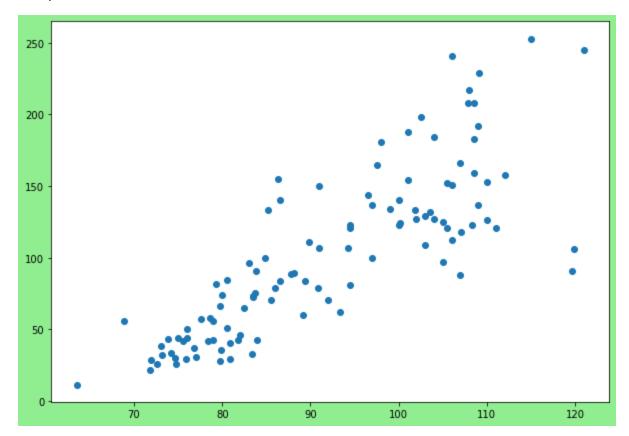
Out[2]:

	Waist	AT
0	74.75	25.72
1	72.60	25.89
2	81.80	42.60
2	92 NE	12 00

4 74.65 29.84

```
In [3]: plt.figure(figsize=(10,7),facecolor="lightgreen")
plt.scatter(data.Waist,data.AT)
```

Out[3]: <matplotlib.collections.PathCollection at 0x19861a45bb0>



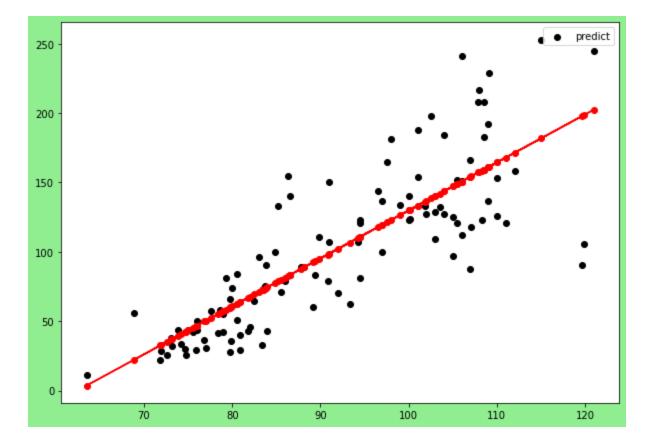
1.Method1

[-215.98148796]

```
In [4]: lm=LinearRegression()
In [5]: x=data['Waist'].values.reshape(-1,1)
    y=data['AT'].values.reshape(-1,1)
In [6]: model=lm.fit(x,y)
In [7]: y_pred=model.predict(x)
In [8]: print(model.coef_)
    print(model.intercept_)
    [[3.45885939]]
```

```
In [9]: plt.figure(figsize=(10,7),facecolor="lightgreen")
    plt.scatter(data.Waist,data.AT,color="black",label="predict")
    plt.plot(data.Waist,y_pred,color="red")
    plt.scatter(data.Waist,y_pred,color="red")
    plt.legend(loc="best")
```

Out[9]: <matplotlib.legend.Legend at 0x19861c1dfa0>



```
In [10]: r2_score(data['AT'],y_pred)
```

Out[10]: 0.6700368930528429

2.Method 2

```
In [11]: data['Waist_sq']=data['Waist']**2
```

```
In [12]: model=smf.ols("np.log(AT)~Waist+Waist_sq",data=data).fit()
```

In [13]: model.summary()

Out[13]: OLS Regression Results

Dep. \	Variable:		np.log(AT)		R-squared:		0.779	
Model:		:	OLS Ad		j. R-squared:		0.775	
Method:		։ Լ	Least Squares		F-statistic:		186.8	
	Date	: Wed	, 29	Jun 2022	2 Prol	(F-statis	tic):	1.80e-35
	Time:		18:38:06 Lo		g-Likelihood:		-24.779	
No. Observations:		:	109			AIC:		55.56
Df Residuals:		:	106			BIC:		63.63
Df Model:		:	2					
Covarian	:	nonrobust						
	coe	ef std	err	t	P> t	[0.025	0.97	5]
Intercept	-7.824	1 1.4	73	-5.312	0.000	-10.744	-4.90)4
Waist	0.228	9 0.0	32	7.107	0.000	0.165	0.29	93
Waist_sq	-0.001	0.0	000	-5.871	0.000	-0.001	-0.00)1
Omi	nibus:	0.325	D	ourbin-W	/atson:	1.46	4	
Prob(Omnibus): 0		0.850	50 Jarque-Bera (JB):		0.271			
Skew: 0		0.119	Prob(JB):			0.873		

Notes:

Kurtosis: 2.949

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

Cond. No. 4.49e+05

[2] The condition number is large, 4.49e+05. This might indicate that there are strong multicollinearity or other numerical problems.