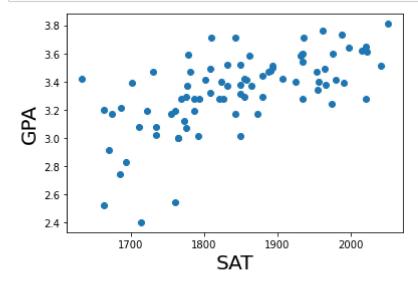
```
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import statsmodels.api as sm
In [7]: data = pd.read_csv('1.01. Simple linear regression.csv')
In [8]: data
Out[8]:
               SAT GPA
           0 1714
                   2.40
              1664
                   2.52
             1760 2.54
              1685
                   2.74
              1693
                    2.83
           79
              1936
                    3.71
           80
              1810
                   3.71
              1987
           81
                   3.73
           82
             1962
                    3.76
           83 2050
                   3.81
          84 rows × 2 columns
In [11]: #provides statics data from each column
         data.describe()
Out[11]:
```

	SAT	GPA
count	84.000000	84.000000
mean	1845.273810	3.330238
std	104.530661	0.271617
min	1634.000000	2.400000
25%	1772.000000	3.190000
50%	1846.000000	3.380000
75%	1934.000000	3.502500
max	2050.000000	3.810000

```
In [12]: #We are going to create a linear regression which predits GPA based on the SAT so
y = data['GPA']
x1 = data['SAT']
```

```
In [13]: plt.scatter(x1,y)
    plt.xlabel('SAT',fontsize=20)
    plt.ylabel('GPA',fontsize=20)
    plt.show()
```



```
In [15]: x = sm.add_constant(x1)
results = sm.OLS(y,x).fit() #fit will apply a specific estimation technique (OLS
results.summary()
```

## Out[15]:

## **OLS Regression Results**

Dep. Variable:	GPA	R-squared:	0.406
Model:	OLS	Adj. R-squared:	0.399
Method:	Least Squares	F-statistic:	56.05
Date:	Tue, 24 Aug 2021	Prob (F-statistic):	7.20e-11
Time:	21:11:40	Log-Likelihood:	12.672
No. Observations:	84	AIC:	-21.34
Df Residuals:	82	BIC:	-16.48
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.2750	0.409	0.673	0.503	-0.538	1.088
SAT	0.0017	0.000	7.487	0.000	0.001	0.002

Omnibus: 12.839 Durbin-Watson: 0.950

Prob(Omnibus): 0.002 Jarque-Bera (JB): 16.155

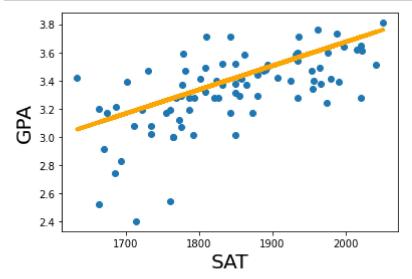
 Skew:
 -0.722
 Prob(JB):
 0.000310

 Kurtosis:
 4.590
 Cond. No.
 3.29e+04

## Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 3.29e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [18]: plt.scatter(x1,y)
    yhat = 0.0017*x1 + 0.275
    fig = plt.plot(x1, yhat, lw=4, c='orange', label='regression line')
    plt.xlabel('SAT', fontsize=20)
    plt.ylabel('GPA', fontsize=20)
    plt.show()
```



```
In [ ]:
```