

```
In [1]: import numpy as np
import pandas as pd
import scipy
import statsmodels.api as sm
import matplotlib.pyplot as plt
import seaborn as sns
sns.set()
```

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

```
In [3]: data
```

```
Out[3]:
```

	SAT	GPA
0	1714	2.40
1	1664	2.52
2	1760	2.54
3	1685	2.74
4	1693	2.83
...
79	1936	3.71
80	1810	3.71
81	1987	3.73
82	1962	3.76
83	2050	3.81

84 rows × 2 columns

```
In [4]: data.describe()
```

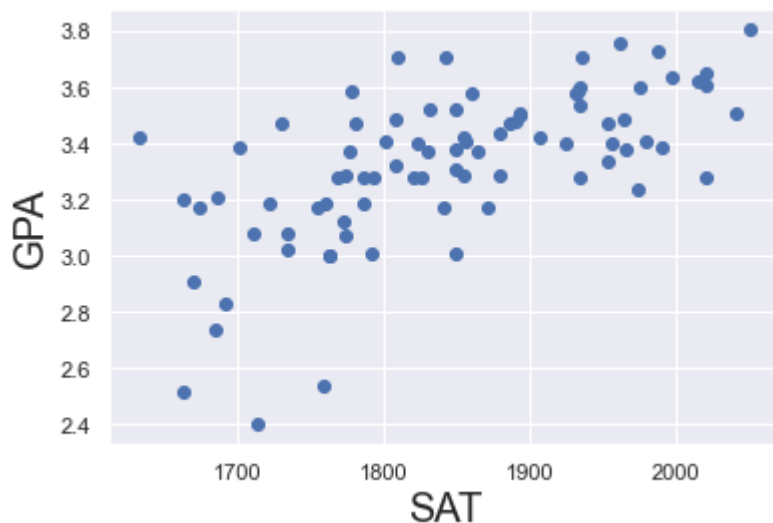
```
Out[4]:
```

	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

	SAT	GPA
max	2050.000000	3.810000

```
In [5]: y = data['GPA']
        x1 = data['SAT']
```

```
In [6]: plt.scatter(x1, y)
        plt.xlabel('SAT', fontsize = 20)
        plt.ylabel('GPA', fontsize = 20)
        plt.show()
```



```
In [7]: x = sm.add_constant(x1)
        results = sm.OLS(y, x).fit()
        results.summary()
```

```
Out[7]:
```

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:	Thu, 21 Dec 2023	Prob (F-statistic):	7.20e-11		
Time:	10:07:15	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

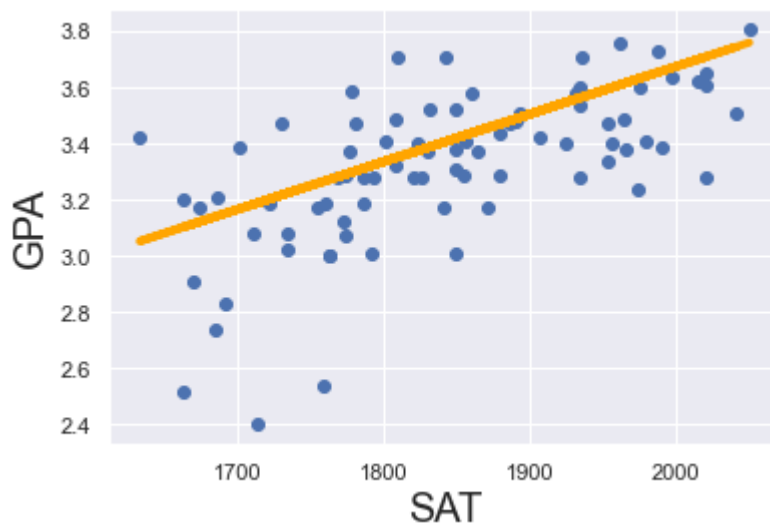
Notes:

[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 [8]:

```
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 [9]:

```
print(results.summary())
```

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:	Thu, 21 Dec 2023	Prob (F-statistic):	7.20e-11			
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	coef	std err	t	P> t	[0.025	0.975]

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