

# Simple linear regression - exercise

You are given a real estate dataset.

Real estate is one of those examples that every regression course goes through as it is extremely easy to understand and there is a (almost always) certain causal relationship to be found.

The data is located in the file: 'real\_estate\_price\_size.csv'.

You are expected to create a simple linear regression (similar to the one in the lecture), using the new data.

In this exercise, the dependent variable is 'price', while the independent variables is 'size'.

Good luck!

## Import the relevant libraries

```
In [3]: import numpy as np #multidimensional arrays
import pandas as pd #format data into columns and rows
import matplotlib.pyplot as plt #2d visualization
import statsmodels.api as sm #summaries
import seaborn #nice graphs
seaborn.set()
```

## Load the data

```
In [4]: data = pd.read_csv('real_estate_price_size.csv')
```

In [22]: data

Out[22]:

	price	size
0	234314.144	643.09
1	228581.528	656.22
2	281626.336	487.29
3	401255.608	1504.75
4	458674.256	1275.46
...	...	...
95	252460.400	549.80
96	310522.592	1037.44
97	383635.568	1504.75
98	225145.248	648.29
99	274922.856	705.29

100 rows × 2 columns

In [20]: data.head()

Out[20]:

	price	size
0	234314.144	643.09
1	228581.528	656.22
2	281626.336	487.29
3	401255.608	1504.75
4	458674.256	1275.46

```
In [21]: data.describe()
```

```
Out[21]:
```

	price	size
count	100.000000	100.000000
mean	292289.470160	853.024200
std	77051.727525	297.941951
min	154282.128000	479.750000
25%	234280.148000	643.330000
50%	280590.716000	696.405000
75%	335723.696000	1029.322500
max	500681.128000	1842.510000

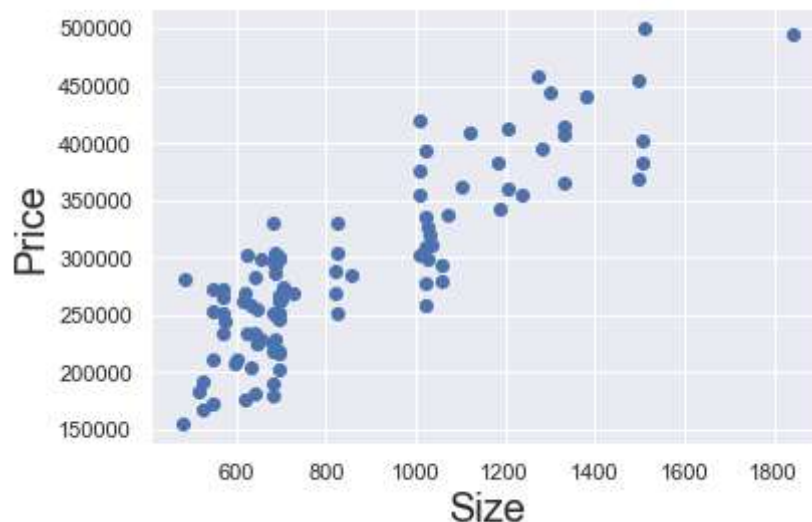
## Create the regression

### Declare the dependent and the independent variables

```
In [8]: y = data['price']  
x1 = data['size']
```

### Explore the data

```
In [9]: plt.scatter(x1,y)  
plt.xlabel('Size',fontsize=20)  
plt.ylabel('Price',fontsize=20)  
plt.show()
```



## Regression itself

```
In [10]: x = sm.add_constant(x1)
results = sm.OLS(y,x).fit()
results.summary()
```

Out[10]: OLS Regression Results

<b>Dep. Variable:</b>	price	<b>R-squared:</b>	0.745
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.742
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	285.9
<b>Date:</b>	Wed, 25 Aug 2021	<b>Prob (F-statistic):</b>	8.13e-31
<b>Time:</b>	21:36:09	<b>Log-Likelihood:</b>	-1198.3
<b>No. Observations:</b>	100	<b>AIC:</b>	2401.
<b>Df Residuals:</b>	98	<b>BIC:</b>	2406.
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
<b>const</b>	1.019e+05	1.19e+04	8.550	0.000	7.83e+04	1.26e+05
<b>size</b>	223.1787	13.199	16.909	0.000	196.986	249.371

<b>Omnibus:</b>	6.262	<b>Durbin-Watson:</b>	2.267
<b>Prob(Omnibus):</b>	0.044	<b>Jarque-Bera (JB):</b>	2.938
<b>Skew:</b>	0.117	<b>Prob(JB):</b>	0.230
<b>Kurtosis:</b>	2.194	<b>Cond. No.</b>	2.75e+03

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.75e+03. This might indicate that there are strong multicollinearity or other numerical problems.

## Plot the regression line on the initial scatter

```
In [12]: plt.scatter(x1,y)
yhat = 223.1787*x1 + 101900
fig = plt.plot(x1, yhat, lw=4, c='red', label='regression line')
plt.xlabel('Size', fontsize=20)
plt.ylabel('Price', fontsize=20)
plt.show()
```

