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# Machine Learning - Multiple Regression

< Previous

Next >

## Multiple Regression

Multiple regression is like linear regression, but with more than one independent value, meaning that we try to predict a value based on **two or more** variables.

Take a look at the data set below, it contains some information about cars.

Car	Model	Volume	Weight	CO2
Toyota	Aygo	1000	790	99
Mitsubishi	Space Star	1200	1160	95
Skoda	Citigo	1000	929	95
Fiat	500	900	865	90
Mini	Cooper	1500	1140	105
VW	Up!	1000	929	105
Skoda	Fabia	1400	1109	90
Mercedes	A-Class	1500	1365	92
Ford	Fiesta	1500	1112	98

Audi

A1

1600

1150

99



We can predict the CO2 emission of a car based on the size of the engine, but with multiple regression we can throw in more variables, like the weight of the car, to make the prediction more accurate.

## How Does it Work?

In Python we have modules that will do the work for us. Start by importing the Pandas module.

```
import pandas
```

Learn about the Pandas module in our [Pandas Tutorial](#).

The Pandas module allows us to read csv files and return a DataFrame object.

The file is meant for testing purposes only, you can download it here: [cars.csv](#)

```
df = pandas.read_csv("cars.csv")
```

Then make a list of the independent values and call this variable `X`.

Put the dependent values in a variable called `y`.

```
X = df[['Weight', 'Volume']]  
y = df['CO2']
```

**Tip:** It is common to name the list of independent values with a upper case X, and the list of dependent values with a lower case y.

We will use some methods from the sklearn module, so we will have to import that module as well:

```
from sklearn import linear_model
```

From the sklearn module we will use the `LinearRegression()` method to create a linear regression object.

This object has a method called `fit()` that takes the independent and dependent values as parameters and fills the regression object with data that describes the relationship:

```
regr = linear_model.LinearRegression()  
regr.fit(X, y)
```

Now we have a regression object that are ready to predict CO2 values based on a car's weight and volume:

```
#predict the CO2 emission of a car where the weight is 2300kg, and the volume  
is 1300cm3:  
predictedCO2 = regr.predict([[2300, 1300]])
```

## Example

See the whole example in action:

```
import pandas  
from sklearn import linear_model  
  
df = pandas.read_csv("cars.csv")  
  
X = df[['Weight', 'Volume']]  
y = df['CO2']  
  
regr = linear_model.LinearRegression()  
regr.fit(X, y)  
  
#predict the CO2 emission of a car where the weight is 2300kg, and the  
volume is 1300cm3:  
predictedCO2 = regr.predict([[2300, 1300]])
```

```
print(predictedCO2)
```

## Result:

```
[107.2087328]
```

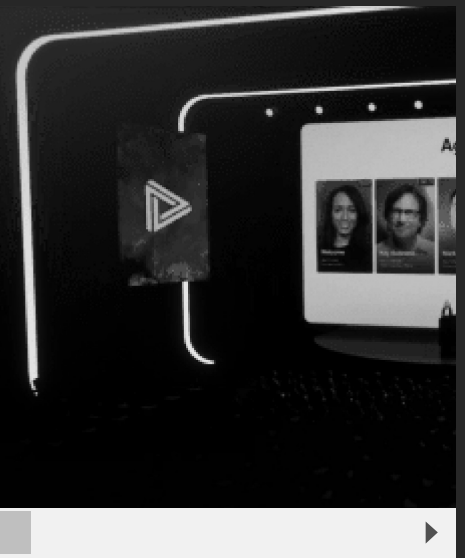
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We have predicted that a car with 1.3 liter engine, and a weight of 2300 kg, will release approximately 107 grams of CO2 for every kilometer it drives.

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## Coefficient

The coefficient is a factor that describes the relationship with an unknown variable.

Example: if  $x$  is a variable, then  $2x$  is  $x$  two times.  $x$  is the unknown variable, and the number  $2$  is the coefficient.

In this case, we can ask for the coefficient value of weight against CO2, and for volume against CO2. The answer(s) we get tells us what would happen if we increase, or decrease, one of the independent values.

## Example

Print the coefficient values of the regression object:

```
import pandas
from sklearn import linear_model

df = pandas.read_csv("cars.csv")

X = df[['Weight', 'Volume']]
y = df['CO2']

regr = linear_model.LinearRegression()
regr.fit(X, y)

print(regr.coef_)
```

Result:

```
[0.00755095 0.00780526]
```

[Run example »](#)

## Result Explained

The result array represents the coefficient values of weight and volume.

Weight: 0.00755095

Volume: 0.00780526

These values tell us that if the weight increase by 1kg, the CO2 emission increases by 0.00755095g.

And if the engine size (Volume) increases by 1 cm<sup>3</sup>, the CO2 emission increases by 0.00780526 g.

I think that is a fair guess, but let test it!

We have already predicted that if a car with a 1300cm<sup>3</sup> engine weighs 2300kg, the CO2 emission will be approximately 107g.

What if we increase the weight with 1000kg?

## Example

Copy the example from before, but change the weight from 2300 to 3300:

```
import pandas
from sklearn import linear_model

df = pandas.read_csv("cars.csv")

X = df[['Weight', 'Volume']]
y = df['CO2']

regr = linear_model.LinearRegression()
regr.fit(X, y)

predictedCO2 = regr.predict([[3300, 1300]])

print(predictedCO2)
```

## Result:

```
[114.75968007]
```

[Run example »](#)


We have predicted that a car with 1.3 liter engine, and a weight of 3300 kg, will release approximately 115 grams of CO2 for every kilometer it drives.

Which shows that the coefficient of 0.00755095 is correct:

$$107.2087328 + (1000 * 0.00755095) = 114.75968$$


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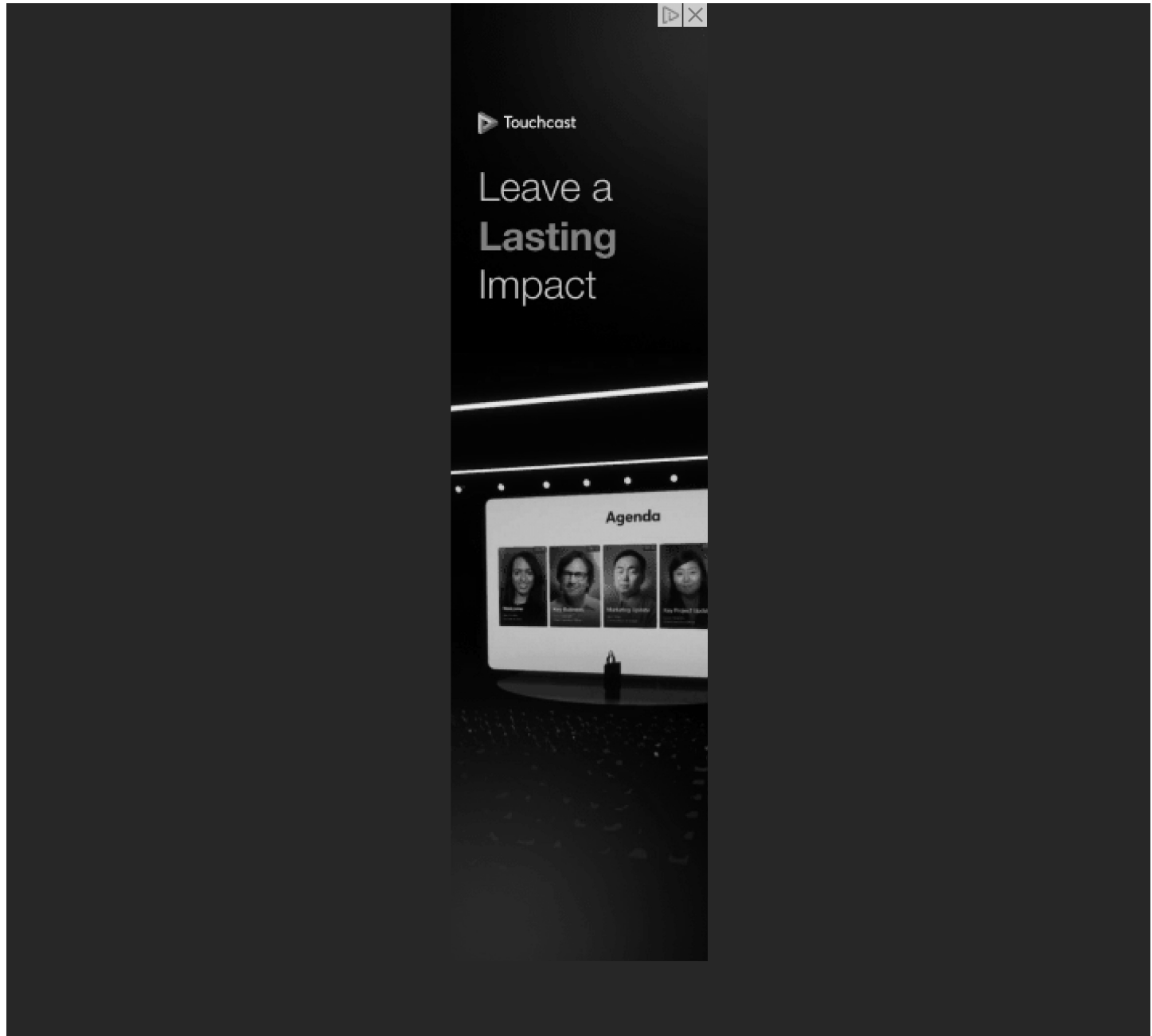
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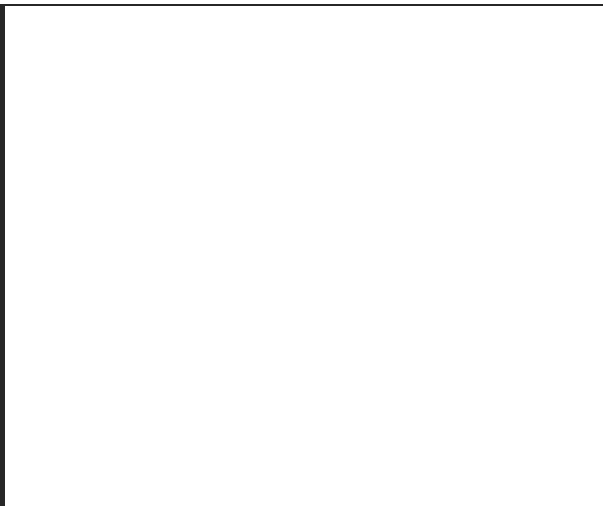




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