

Recursos de escala

Quando seus dados têm valores diferentes e até unidades de medida diferentes, pode ser difícil compará-los. O que é quilogramas comparado a metros? Ou altitude em relação ao tempo?

A resposta para este problema é a escala. Podemos dimensionar os dados em novos valores que são mais fáceis de comparar.

Dê uma olhada na tabela abaixo, é o mesmo conjunto de dados que usamos no capítulo de regressão múltipla, mas desta vez a coluna de **volume** contém valores em *litros* em vez de cm^3 (1,0 em vez de 1000).

O arquivo é apenas para fins de teste, você pode baixá-lo aqui: cars2.csv

Carro	Modelo	Volume	Peso	CO2
Toyota	Aygo	1,0	790	99
Mitsubishi	Estrela do Espaço	1.2	1160	95
Skoda	Citigo	1,0	929	95
Fiat	500	0,9	865	90
Mini	Tanoeiro	1,5	1140	105
VW	Acima!	1,0	929	105

Skoda	Fábia	1,4	1109	90
Mercedes	Uma aula	1,5	1365	92
Ford	Festa	1,5	1112	98
Audi	A1	1,6	1150	99

It can be difficult to compare the volume 1.0 with the weight 790, but if we scale them both into comparable values, we can easily see how much one value is compared to the other.

There are different methods for scaling data, in this tutorial we will use a method called standardization.

The standardization method uses this formula:

$$z = (x - u) / s$$

Where z is the new value, x is the original value, u is the mean and s is the standard deviation.

If you take the **weight** column from the data set above, the first value is 790, and the scaled value will be:

$$(790 - 1292.23) / 238.74 = -2.1$$

If you take the **volume** column from the data set above, the first value is 1.0, and the scaled value will be:

$$(1.0 - 1.61) / 0.38 = -1.59$$

Now you can compare -2.1 with -1.59 instead of comparing 790 with 1.0.

You do not have to do this manually, the Python sklearn module has a method called StandardScaler() which returns a Scaler object with methods for transforming data sets.

Example

Scale all values in the Weight and Volume columns:

```
import pandas
from sklearn import linear_model
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
```

```
df = pandas.read_csv("cars2.csv")

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

scaledX = scale.fit_transform(X)

print(scaledX)
```

Result:

Note that the first two values are -2.1 and -1.59, which corresponds to our calculations:

```
[[-2.10389253 -1.59336644]
[-0.55407235 -1.07190106]
[-1.52166278 -1.59336644]
[-1.78973979 -1.85409913]
[-0.63784641 -0.28970299]
[-1.52166278 -1.59336644]
[-0.76769621 -0.55043568]
[ 0.3046118 -0.28970299]
[-0.7551301 -0.28970299]
[-0.59595938 -0.0289703 ]
[-1.30803892 -1.33263375]
[-1.26615189 -0.81116837]
[-0.7551301 -1.59336644]
[-0.16871166 -0.0289703 ]
[ 0.14125238 -0.0289703 ]
[ 0.15800719 -0.0289703 ]
[ 0.3046118 -0.0289703 ]
[-0.05142797 1.53542584]
[-0.72580918 -0.0289703 ]
[ 1.2219378 -0.0289703 ]
[ 0.51404696 -0.0289703 ]
[ 0.72348212 -0.28970299]
[ 0.96642691 -0.0289703 ]
```

Run example »



Predict CO2 Values

The task in the <u>Multiple Regression chapter</u> was to predict the CO2 emission from a car when you only knew its weight and volume.

When the data set is scaled, you will have to use the scale when you predict values:

Example

Preveja a emissão de CO2 de um carro de 1,3 litro que pesa 2300 kg:

```
import pandas
from sklearn import linear_model
from sklearn.preprocessing import StandardScaler
scale = StandardScaler()

df = pandas.read_csv("cars2.csv")

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

```
y = df['CO2']
scaledX = scale.fit_transform(X)

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

scaled = scale.transform([[2300, 1.3]])

predictedCO2 = regr.predict([scaled[0]])
print(predictedCO2)

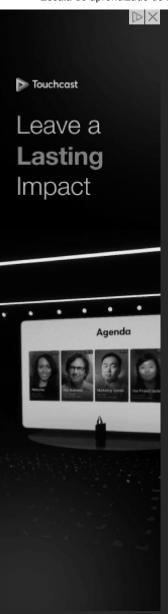
Resultado:

[107.2087328]

Executar exemplo »
```

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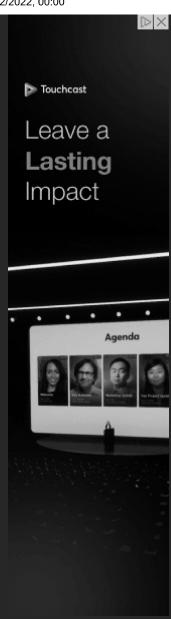
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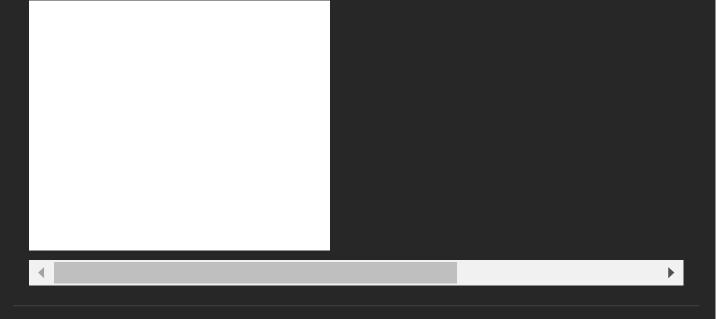


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