

8_Feature_Scaling

January 14, 2025

```
[ ]: '''  
    Feature Scaling -->  
  
    Feature scaling is a preprocessing technique used in machine learning to  
    ↪standardize  
    the range of independent variables (features) in the data. It ensures that  
    ↪no feature  
    dominates others purely due to differences in scale, which can negatively  
    ↪impact certain  
    models (like gradient-based algorithms).  
    '''
```

```
[17]: # Common types of feature scaling -->  
  
# Normalization (Min-Max Scaling) : Rescales the values to a fixed range,  
    ↪usually [0, 1]  
# Standardization (Z-score Scaling) : Centers the data around the mean and  
    ↪scales it according  
# to the standard deviation, giving it a mean of 0 and a standard deviation  
    ↪of 1
```

```
[ ]: # Standardization -->  
  
#  $z = (x - \mu) / \sigma$   
  
# Where,  
  
#  $x$  = Original value  
#  $\mu$  = Mean of dataset  
#  $\sigma$  = Standard deviation of dataset
```

```
[ ]: # Normalization -->  
  
#  $x = (x - \min(x)) / (\max(x) - \min(x))$   
  
# Where,
```

```
# x = Original value
# min(x) = minimum value in dataset
# max(x) = maximum value in dataset
```

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[18]: # Feature scaling is particularly important for algorithms like -->

# K-nearest neighbors (KNN)
# Support Vector Machines (SVM)
# Gradient Descent-based models (e.g., logistic regression, neural networks)

# Some algorithms, like decision trees and random forests, are less sensitive
↳ to feature scaling
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[19]: # Done After Splitting !
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[ ]: # Let's Do Pre-Processing From Scratch !
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```
[20]: import pandas as pd
import numpy as np
from sklearn.impute import SimpleImputer
from sklearn.model_selection import train_test_split
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import OneHotEncoder
from sklearn.preprocessing import StandardScaler
```

```
[21]: data = pd.read_csv('Data/Data.csv')
data
```

```
[21]:
```

	Country	Age	Salary	Purchased
0	France	44.0	72000.0	No
1	Spain	27.0	48000.0	Yes
2	Germany	30.0	54000.0	No
3	Spain	38.0	61000.0	No
4	Germany	40.0	NaN	Yes
5	France	35.0	58000.0	Yes
6	Spain	NaN	52000.0	No
7	France	48.0	79000.0	Yes
8	Germany	50.0	83000.0	No
9	France	37.0	67000.0	Yes

```
[22]: x_data = data.iloc[:, :-1].values
y_data = data.iloc[:, -1].values
```

```
[23]: impute = SimpleImputer(missing_values = np.nan, strategy = 'mean')
impute.fit(x_data[:, 1:3])
x_data[:, 1:3] = impute.transform(x_data[:, 1:3])
```

```
x_data
```

```
[23]: array([[ 'France', 44.0, 72000.0],
             [ 'Spain', 27.0, 48000.0],
             [ 'Germany', 30.0, 54000.0],
             [ 'Spain', 38.0, 61000.0],
             [ 'Germany', 40.0, 63777.77777777778],
             [ 'France', 35.0, 58000.0],
             [ 'Spain', 38.77777777777778, 52000.0],
             [ 'France', 48.0, 79000.0],
             [ 'Germany', 50.0, 83000.0],
             [ 'France', 37.0, 67000.0]], dtype=object)
```

```
[24]: clt = ColumnTransformer(transformers = [('encoder', OneHotEncoder(), [0]),
      ↪ remainder = 'passthrough')
x_data = np.array(clt.fit_transform(x_data))
x_data
```

```
[24]: array([[1.0, 0.0, 0.0, 44.0, 72000.0],
             [0.0, 0.0, 1.0, 27.0, 48000.0],
             [0.0, 1.0, 0.0, 30.0, 54000.0],
             [0.0, 0.0, 1.0, 38.0, 61000.0],
             [0.0, 1.0, 0.0, 40.0, 63777.77777777778],
             [1.0, 0.0, 0.0, 35.0, 58000.0],
             [0.0, 0.0, 1.0, 38.77777777777778, 52000.0],
             [1.0, 0.0, 0.0, 48.0, 79000.0],
             [0.0, 1.0, 0.0, 50.0, 83000.0],
             [1.0, 0.0, 0.0, 37.0, 67000.0]], dtype=object)
```

```
[25]: encode = LabelEncoder()
y_data = encode.fit_transform(y_data)
y_data
```

```
[25]: array([0, 1, 0, 0, 1, 1, 0, 1, 0, 1])
```

```
[26]: x_train, x_test, y_train, y_test = train_test_split(x_data, y_data, test_size =
      ↪ 0.2, random_state = 1)
print(x_train)
```

```
[[0.0 0.0 1.0 38.77777777777778 52000.0]
 [0.0 1.0 0.0 40.0 63777.77777777778]
 [1.0 0.0 0.0 44.0 72000.0]
 [0.0 0.0 1.0 38.0 61000.0]
 [0.0 0.0 1.0 27.0 48000.0]
 [1.0 0.0 0.0 48.0 79000.0]
 [0.0 1.0 0.0 50.0 83000.0]
 [1.0 0.0 0.0 35.0 58000.0]]
```

```
[27]: sc = StandardScaler()  
x_train[:, 3:] = sc.fit_transform(x_train[:, 3:])  
x_test[:, 3:] = sc.transform(x_test[:, 3:])
```

```
[28]: x_train
```

```
[28]: array([[0.0, 0.0, 1.0, -0.19159184384578545, -1.0781259408412425],  
          [0.0, 1.0, 0.0, -0.014117293757057777, -0.07013167641635372],  
          [1.0, 0.0, 0.0, 0.566708506533324, 0.633562432710455],  
          [0.0, 0.0, 1.0, -0.30453019390224867, -0.30786617274297867],  
          [0.0, 0.0, 1.0, -1.9018011447007988, -1.420463615551582],  
          [1.0, 0.0, 0.0, 1.1475343068237058, 1.232653363453549],  
          [0.0, 1.0, 0.0, 1.4379472069688968, 1.5749910381638885],  
          [1.0, 0.0, 0.0, -0.7401495441200351, -0.5646194287757332]],  
          dtype=object)
```

```
[29]: x_test
```

```
[29]: array([[0.0, 1.0, 0.0, -1.4661817944830124, -0.9069571034860727],  
          [1.0, 0.0, 0.0, -0.44973664397484414, 0.2056403393225306]],  
          dtype=object)
```