

- * all about Standard Scaler () -> feature Scalling and when we have to perform it.

 A and the difference between Scalled and non-scalled data.
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 feature Scaling, Such Standardization (2-score scaling)

 er min-max scaling, is intended for Contineous numerical
 - Scaler. fit transform and Scaler. transform are two methods used with Scalling objects from scikit-learn library

Scolar. fit_transform: Scaler. transform

- used to both fit the Scaling weed to transform a dataset

parameters to your training using scaling parameters that

data and transform it in were previously fitted using

single step ('fit' or fit-transform') on

another dataset.

- Standard deviation of your ers, it directly applies the training data and then Scalling based on the parameters the data based on eters already determined during thes computed parameters. The fitting process.
- The training data.

 The training data to ensure that the Scaling is applied as on training data.

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an test data set use only transform().

- it use the paramet scaling parameter from the

Training set to scale the test set.

This is crimial to avoid date leakage and to

maintain the sepration between the training and

test sets.

```
from sklearn.preprocessing import StandardScaler

# Create a toy dataset
data = [[1, 2], [2, 3], [3, 4]]

# Create a StandardScaler
scaler = StandardScaler()

# Fit and transform the data in one step (common for training data)
scaled_data_train = scaler.fit_transform(data)

# New data (e.g., test data)
new_data = [[4, 5], [5, 6]]

# Use the previously fitted scaler to transform the new data (common for test data)
scaled_new_data = scaler.transform(new_data)

print("Scaled Training Data:")
print(scaled_data_train)

print("Scaled_New_Data:")
print(scaled_new_data)
```

Output:-



