LISBON
SCHOOL OF
ECONOMICS &
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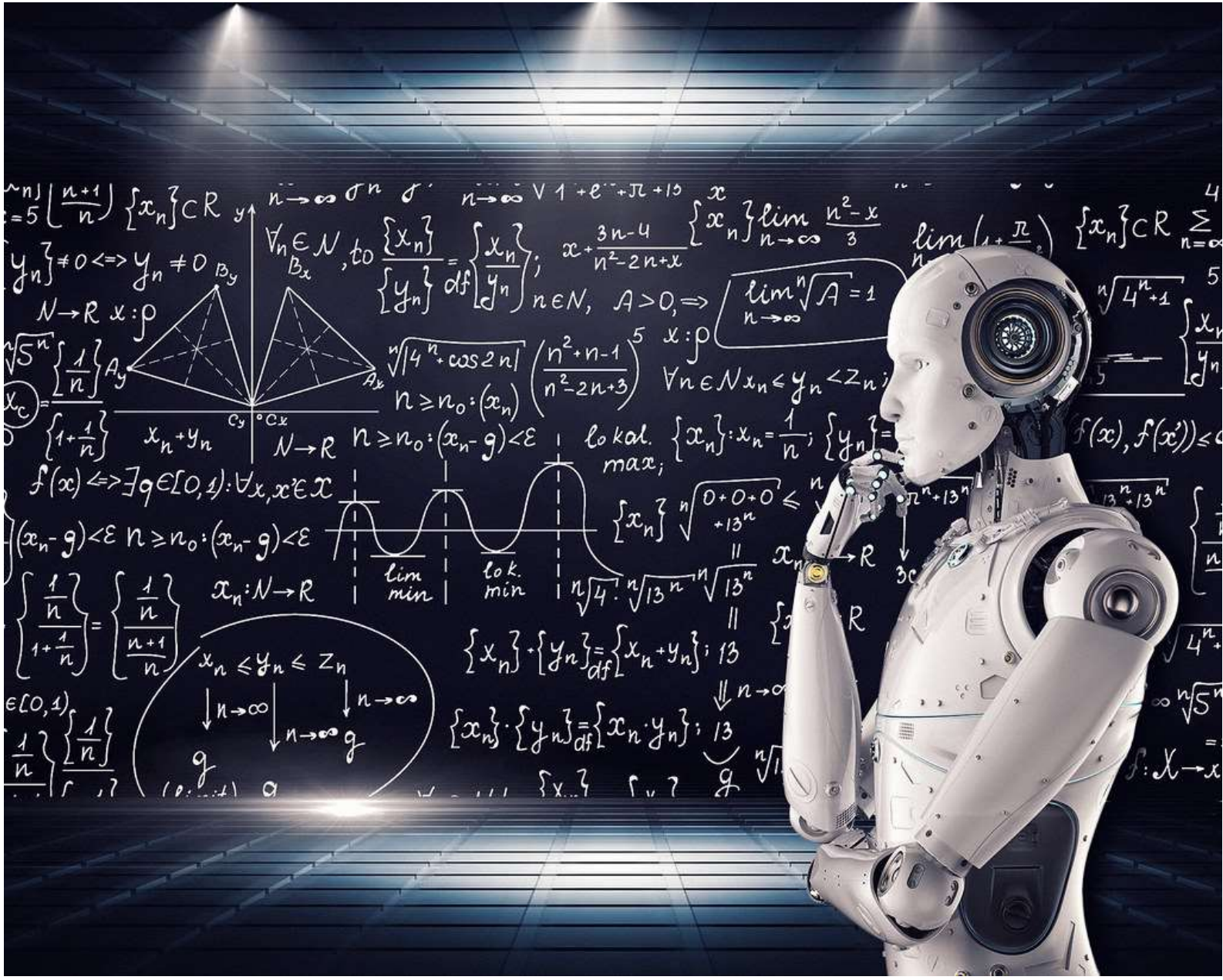
SCIKIT-LEARN

Learning Goals

- Distinguish between supervised and unsupervised learning
- Pre-processing data using scikit-learn
- Creating and fitting model using scikit-learn
- Understand how to use pipeline in scikit-learn

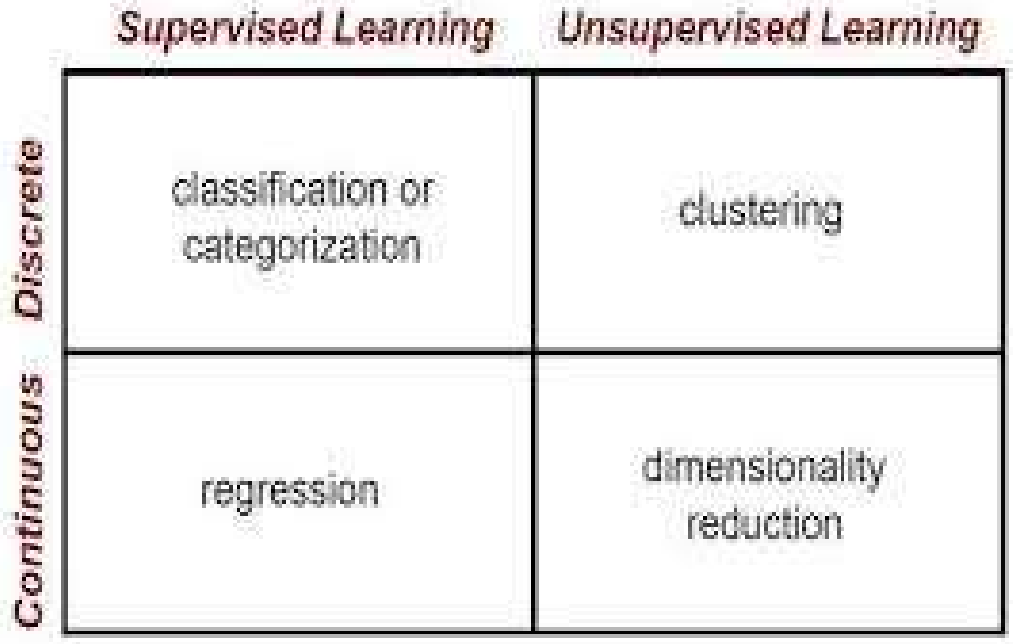
It is as a subset of artificial intelligence that enable systems to learn patterns from data and subsequently improve from experience.



Machine Lerning

Learning problems fall into a few categories:

- supervised learning
- unsupervised learning



supervised learning

Classification

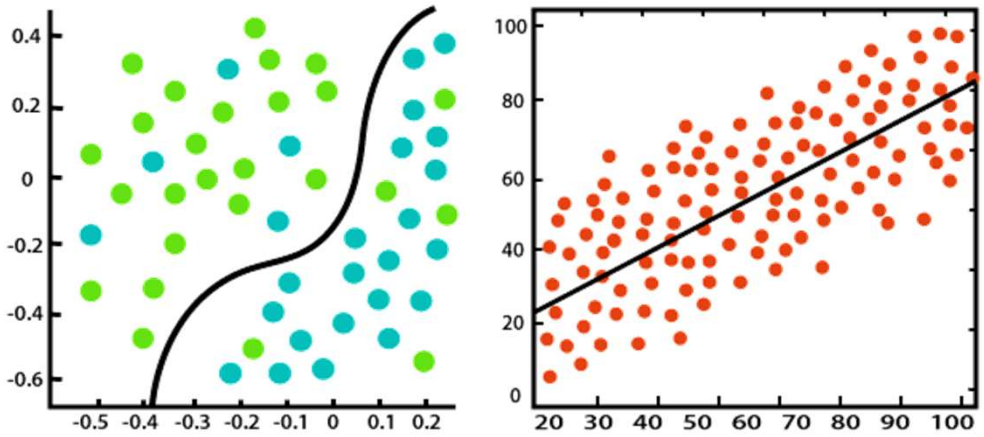
Identifying which category an object belongs to.

Applications: Spam detection, image recognition.
Algorithms: SVM, nearest neighbors, random forest, and more...

Regression

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices.
Algorithms: SVR, nearest neighbors, random forest, and more...



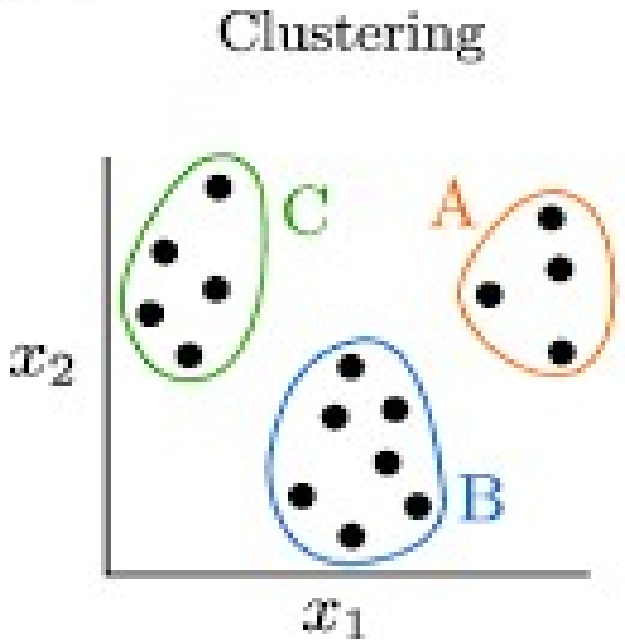
Classification Regression

unsupervised learning

Clustering

Automatic grouping of similar objects into sets.

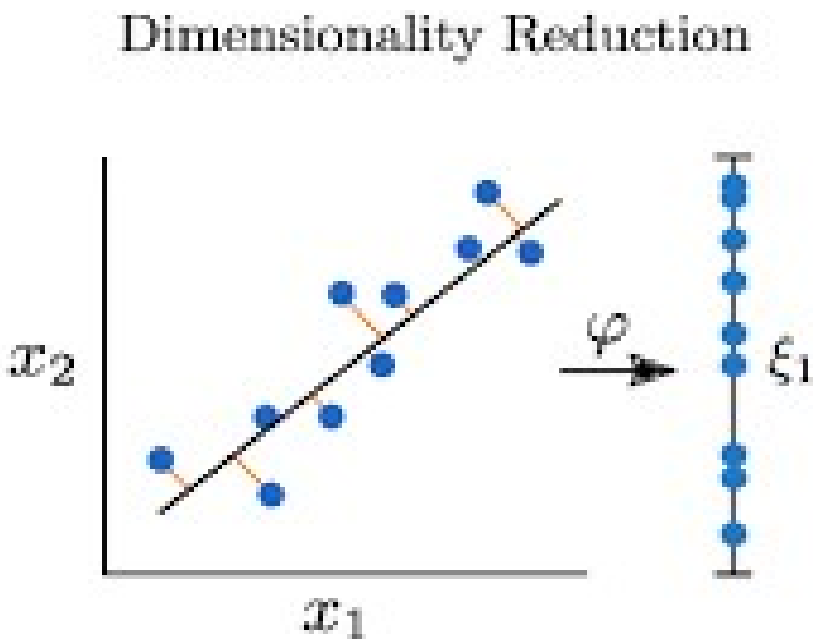
Applications: Customer segmentation, Grouping experiment outcomes
Algorithms: k-Means, spectral clustering, mean-shift, and more...



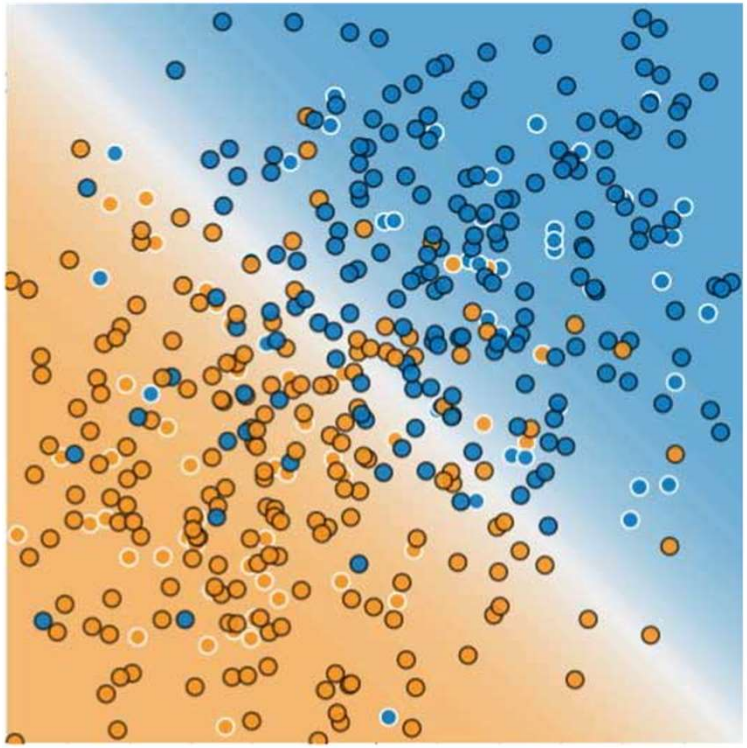
Dimensionality reduction

Reducing the number of random variables to consider.

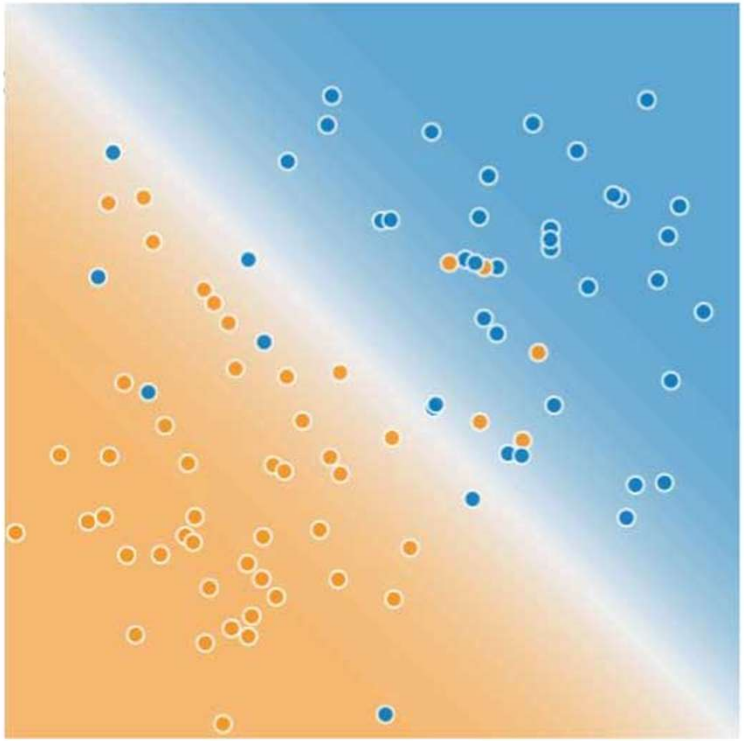
Applications: Visualization, Increased efficiency
Algorithms: k-Means, feature selection, non-negative matrix factorization, and more...



Training set and testing set



Training Data



Test Data

Learning and predicting

```
import sklearn.linear_model as lm
import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
df=pd.DataFrame(data, columns=['x','y'])
```

```
# Create linear regression object
reg = lm.LinearRegression()
# Train the model
results = reg.fit(df[['x']], df['y'])
```

```
results.score(df[['x']], df['y'])
```

```
0.7714285714285715
```

```
results.coef_
```

```
array([0.9])
```

```
results.intercept_
```

```
0.5
```

```
results.predict([[3]])
```

```
array([3.2])
```

Learning and predicting

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```
results.intercept_
```

```
0.5
```

```
results.predict([[3]])
```

```
array([3.2])
```

```
df['x'].shape
```

```
(16,)
```

```
df[['x']].shape
```

```
(16, 1)
```

Learning and predicting

```
import sklearn.linear_model as lm
import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
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```

```
resul
```

```
0.77
```

```
resul
```

```
array
```

```
resul
```

```
0.5
```

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reg = lm.LinearRegression()
# Train the model
results = reg.fit(df[['x']], df['y'])
```

```
results.predict([[3]])
```

```
array([3.2])
```

Learning and predicting

```
import sklearn.linear_model as lm
import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
df=pd.DataFrame(data, columns=['x','y'])

# Create linear regression object
reg = lm.Linea
# Train the mo
results = reg.

results.score(

0.771428571428

results.coef_

array([0.9])

results.interc

0.5

results.predict([[3]])

array([3.2])
```

```
results.coef_
```

```
array([0.9])
```

```
results.intercept_
```

```
0.5
```

Learning and predicting

```
import sklearn.linear_model as lm

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data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
df=pd.DataFrame(data, columns=['x','y'])
```

```
# Split the data into training/testing sets
x_train = x[:-6]
x_test = x[-6:]
# Split the targets into training/testing sets
y_train = y[:-6]
y_test = y[-6:]
```

```
 [[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],   [2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
```

```
# Create linear regression object
regr = lm.LinearRegression()
# Train the model using the training sets
regr.fit(x_train, y_train)
# Make predictions using the testing set
y_pred = regr.predict(x_test)
# The coefficients
print('Coefficients: \n', regr.coef_)
```

```
Coefficients:
 [0.69354839]
```

Learning and predicting

```
import sklearn.linear_model as lm

import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
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y_test = y[-6:]
```

```
[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],
```

```
[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
```

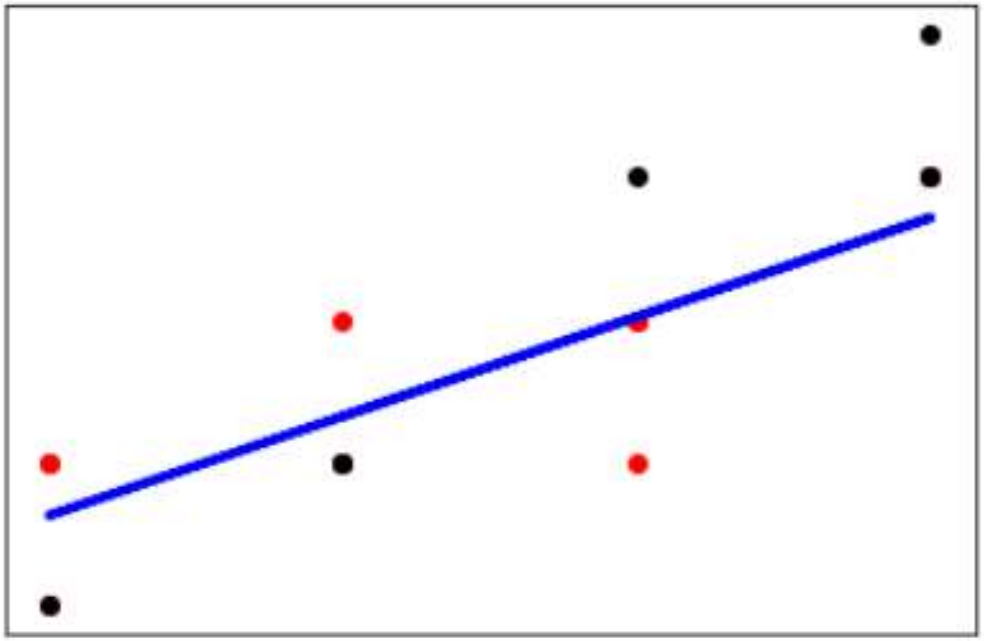
Train

Test

```
# Create linear regression object
regr = lm.LinearRegression()
# Train the model using the training sets
regr.fit(x_train, y_train)
# Make predictions using the testing set
y_pred = regr.predict(x_test)
# The coefficients
print('Coefficients: \n', regr.coef_)
```

```
Coefficients:
 [0.69354839]
```

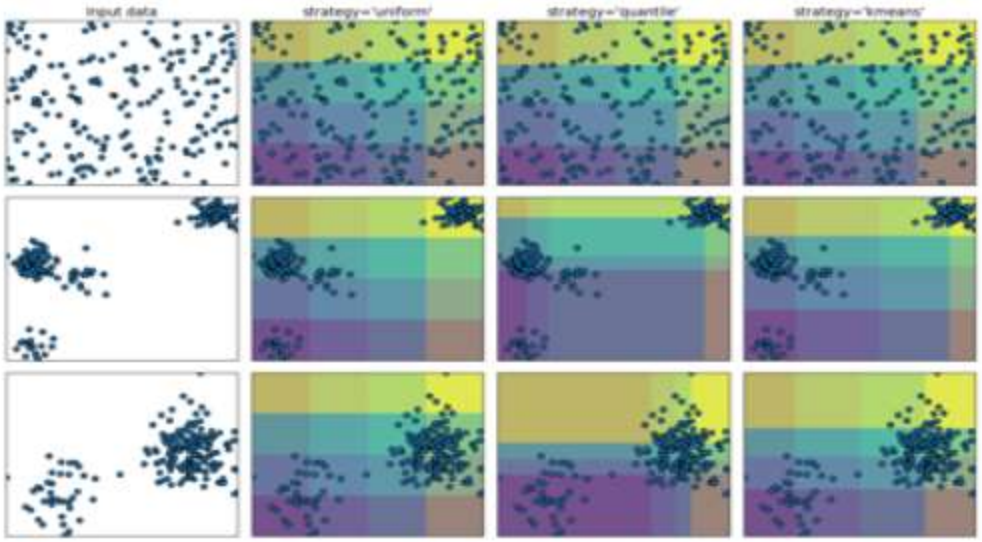
```
import matplotlib.pyplot as plt
# Plot outputs
plt.scatter(x_train, y_train,  color='red')
plt.scatter(x_test, y_test,  color='black')
plt.plot(x_test, y_pred, color='blue', linewidth=3)
plt.xticks(())
plt.yticks(())
plt.show()
```



Preprocessing

Feature extraction and normalization.

Applications: Transforming input data such as text for use with machine learning algorithms.
Algorithms: preprocessing, feature extraction, and more...



Inconsistent preprocessing

```
import sklearn.linear_model as lm
import sklearn.model_selection as ms
import sklearn.metrics as metrics
import sklearn.preprocessing as prep

import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
df=pd.DataFrame(data, columns=['x','y'])
X=df[['x']]
y=df['y']
```

```
X_train, X_test, y_train, y_test = ms.train_test_split(X, y, test_size=0.4, random_state=50)
```

```
scaler = prep.StandardScaler()
X_train_transformed = scaler.fit_transform(X_train)
model = lm.LinearRegression().fit(X_train_transformed, y_train)
metrics.r2_score(y_test, model.predict(X_test))
```

```
-2.1041861708478207
```

```
scaler = prep.StandardScaler()
X_train_transformed = scaler.fit_transform(X_train)
model = lm.LinearRegression().fit(X_train_transformed, y_train)
X_test_transformed = scaler.transform(X_test)
metrics.r2_score(y_test, model.predict(X_test_transformed))
```

```
0.8150675457605236
```

Inconsistent preprocessing

```
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X_test_transformed = scaler.transform(X_test)
metrics.r2_score(y_test, model.predict(X_test_transformed))
```

```
0.8150675457605236
```

Inconsistent preprocessing

```
import sklearn.pipeline as pipeline
import sklearn.linear_model as lm
import sklearn.model_selection as ms
import sklearn.metrics as metrics
import sklearn.preprocessing as prep
import pandas as pd
data=[[3,2],[1,2],[2,2],[4,4],[3,3],[1,2],[2,3],[4,4],[3,3],[1,1],[2,2],[4,5],[3,4],[1,1],[2,2],[4,4]]
df=pd.DataFrame(data, columns=['x','y'])
X=df[['x']]
y=df['y']

X_train, X_test, y_train, y_test = ms.train_test_split(X, y, test_size=0.4, random_state=50)

model = pipeline.make_pipeline(prep.StandardScaler(), lm.LinearRegression())
model.fit(X_train, y_train)

metrics.r2_score(y_test, model.predict(X_test))
```

```
0.8150675457605236
```

Pre-processing

- Standardization
- Non-linear transformation
- Normalization
- Encoding categorical features
- Discretization
- Imputation of missing values
- Generating polynomial features

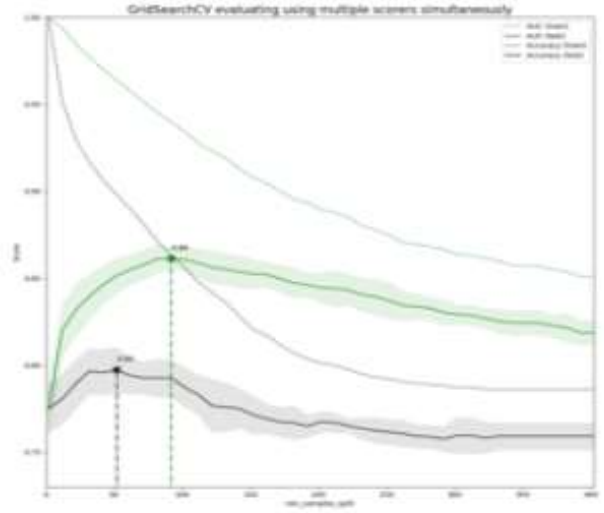
Source: https://scikit-learn.org/stable/modules/preprocessing.html

Model selection

Comparing, validating and choosing parameters and models.

Applications: Improved accuracy via parameter tuning

Algorithms: grid search, cross validation, metrics, and more...



Conclusions

- Supervised and Unsupervised learning
- Main characteristics of scikit-learn