```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
# Load the Iris dataset (you can also download it from various sources including scikit-learn)
from sklearn.datasets import load_iris
iris = load_iris()
# Create a DataFrame
data = pd.DataFrame(data= np.c_[iris['data'], iris['target']], columns= iris['feature_names'] + ['target'])
# Explore the dataset
print(data.head())
# Split the dataset into features (X) and target labels (y)
X = data.drop('target', axis=1)
y = data['target']
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Standardize the features (scaling)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X test = scaler.transform(X test)
# Create a Random Forest Classifier
clf = RandomForestClassifier(n_estimators=100, random_state=42)
# Train the classifier
clf.fit(X_train, y_train)
# Predict on the test set
y_pred = clf.predict(X_test)
# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred)
print(f"Accuracy: {accuracy}")
print(f"Confusion Matrix:\n{conf_matrix}")
print(f"Classification Report:\n{class_report}")
[]
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 sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) \
0
         5.1
                    3.5
                                1.4
                                          0.2
         4.9
                    3.0
                                1.4
                                           0.2
1
2
         4.7
                                          0.2
                    3.2
                                1.3
3
                                1.5
                                           0.2
         4.6
                    3.1
4
         5.0
                    3.6
                                1.4
                                           0.2
 target
0
   0.0
   0.0
1
2
   0.0
3
   0.0
   0.0
Accuracy: 1.0
Confusion Matrix:
[[10 \ 0 \ 0]]
[0 \ 9 \ 0]
[0 0 11]]
Classification Report:
       precision recall f1-score support
    0.0
           1.00
                   1.00
                          1.00
                                   10
    1.0
           1.00
                  1.00
                          1.00
                                   9
    2.0
           1.00
                  1.00
                          1.00
                                   11
                         1.00
                                 30
 accuracy
 macro avg
               1.00
                      1.00
                              1.00
                                       30
weighted avg
                1.00
                       1.00
                                1.00
                                        30
```