```
import pandas as pd
from sklearn.model_selection import train_test_split
from \ sklearn.preprocessing \ import \ StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, classification_report
from sklearn.tree import DecisionTreeClassifier
# Load the dataset
iris_data = pd.read_csv('IRIS.csv')
# Data Exploration
iris_data.head()
         sepal_length sepal_width petal_length petal_width
                                                                  species
                   5.1
                                3.5
                                               1.4
                                                            0.2 Iris-setosa
      1
                   4.9
                                3.0
                                               1.4
                                                            0.2 Iris-setosa
      2
                   4.7
                                3.2
                                               1.3
                                                            0.2 Iris-setosa
      3
                   4.6
                                3.1
                                               1.5
                                                            0.2 Iris-setosa
                   5.0
                                 3.6
                                               1.4
                                                            0.2 Iris-setosa
              View recommended plots
 Next steps:
# Data Exploration
iris_data.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 150 entries, 0 to 149
     Data columns (total 5 columns):
      # Column
                         Non-Null Count Dtype
      0
          sepal_length 150 non-null
                                          float64
      1
          sepal_width
                        150 non-null
                                          float64
          petal_length 150 non-null petal_width 150 non-null
                                          float64
                                          float64
          species
                         150 non-null
                                          object
     dtypes: float64(4), object(1)
     memory usage: 6.0+ KB
# Data Exploration
iris_data.isnull().sum()
     sepal_length
     sepal_width
                      0
     petal_length
                      0
     petal_width
     species
                      0
     dtype: int64
# Data Exploration
iris_data.describe()
```

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000



```
# Data Preprocessing
# Separate features (sepal length, sepal width, petal length, petal width)
X = iris_data[['sepal_length', 'sepal_width', 'petal_length', 'petal_width']]
# Target variable (species)
y = iris_data['species']
# Feature Scaling
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)
# Model Training
model = DecisionTreeClassifier()
model.fit(X_train, y_train)
# Model Evaluation
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print(classification_report(y_test, y_pred))
     Accuracy: 1.0
                                   recall f1-score
                      precision
                                                      support
         Iris-setosa
                           1.00
                                     1.00
                                                1.00
                                                            10
     Iris-versicolor
                           1.00
                                     1.00
                                               1.00
                                                            9
     Iris-virginica
                           1.00
                                     1.00
                                                1.00
                                                            11
            accuracy
                                                1.00
                                                            30
                                     1.00
           macro avg
                           1.00
                                               1.00
                                                            30
        weighted avg
                           1.00
                                     1.00
                                                1.00
                                                            30
# Model Training
model = KNeighborsClassifier(n_neighbors=3) # K-nearest neighbors classifier with k=3
model.fit(X_train, y_train)
# Model Evaluation
y_pred = model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
print(classification_report(y_test, y_pred))
     Accuracy: 1.0
                      precision
                                   recall f1-score
                           1.00
                                     1.00
                                                1.00
         Iris-setosa
                                                            10
     Iris-versicolor
                           1.00
                                     1.00
                                                1.00
                                                             9
      Iris-virginica
                           1.00
                                     1.00
                                                1.00
                                                            11
                                                            30
            accuracy
                                                1.00
           macro avg
                           1.00
                                     1.00
                                                1.00
                                                            30
        weighted avg
                                     1.00
                                                1.00
                                                            30
                           1.00
```



