INTERN:-Bhagyashri Sharad Pisal

Intern ID:- CT4MOTH

Domain:- Machine Learning

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Company:- CODETECH IT SOLUTIONS

Mentor:- Neela Santhosh Kumar

TASK 1:DECISION TREE IMPLEMENTATION

BUILD AND VISUALIZE A DECISION TREE MODEL USING SCIKIT-LEARN TO CLASSIFY OR PREDICT OUTCOMES ON A CHOSEN DATASET. DELIVERABLE: A NOTEBOOK WITH MODEL VISUALIZATION AND ANALYSIS.

```
import numpy as np
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from \ sklearn.tree \ import \ DecisionTreeClassifier, \ export\_text, \ plot\_tree
import matplotlib.pyplot as plt
import seaborn as sns
# Loading the Iris dataset
iris = load iris()
data = pd.DataFrame(iris.data, columns=iris.feature_names)
data['target'] = iris.target
# Displaying the first few rows of the dataset
print("Dataset preview:")
print(data.head())
₹
    Dataset preview:
        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
                                                                              0.2
     1
                                                            1.4
     2
                      4.7
                                         3.2
                                                            1.3
                                                                              0.2
     3
                      4.6
                                        3.1
                                                            1.5
                                                                              0.2
     4
                      5.0
                                         3.6
                                                            1.4
                                                                              0.2
        target
     0
             0
             0
     1
     2
             0
     3
             0
     4
             0
```

#getting the top 10 values
data.head(10)

₹		sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target
	0	5.1	3.5	1.4	0.2	0
	1	4.9	3.0	1.4	0.2	0
	2	4.7	3.2	1.3	0.2	0
	3	4.6	3.1	1.5	0.2	0
	4	5.0	3.6	1.4	0.2	0
	5	5.4	3.9	1.7	0.4	0
	6	4.6	3.4	1.4	0.3	0
	7	5.0	3.4	1.5	0.2	0
	8	4.4	2.9	1.4	0.2	0
	9	4.9	3.1	1.5	0.1	0

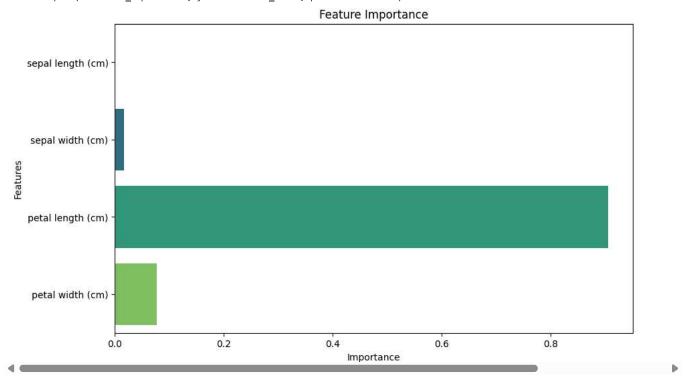
geeting the 10 bottom values
data.tail(10)

```
₹
           sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target
     140
                          6.7
                                              3.1
                                                                                      2.4
                                                                                      2.3
                                                                                                 2
     141
                          6.9
                                              3.1
                                                                   5.1
     142
                                              2.7
                                                                   5.1
                                                                                      1.9
                                                                                                 2
                          5.8
     143
                          6.8
                                              3.2
                                                                   5.9
                                                                                      2.3
                                                                                                 2
     144
                          6.7
                                              3.3
                                                                   5.7
                                                                                      2.5
                                                                                                 2
     145
                           6.7
                                              3.0
                                                                   5.2
                                                                                      2.3
                                                                                                 2
                                              2.5
                                                                                                 2
     146
                          6.3
                                                                   5.0
                                                                                      1.9
     147
                           6.5
                                              3.0
                                                                   5.2
                                                                                      2.0
                                                                                                 2
     148
                                                                                                 2
                           6.2
                                              3.4
                                                                   5.4
                                                                                      2.3
     149
                           5.9
                                              3.0
                                                                   5.1
                                                                                      1.8
                                                                                                 2
```

```
# Spliting the dataset into training and testing sets
X = data[iris.feature_names]
y = data['target']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Initialize the Decision Tree model
dt_model = DecisionTreeClassifier(random_state=42)
# Train the model on the training data
dt_model.fit(X_train, y_train)
₹
            {\tt DecisionTreeClassifier}
     DecisionTreeClassifier(random_state=42)
# Evaluate the model on the test data
accuracy = dt_model.score(X_test, y_test)
print(f"Model accuracy: {accuracy:.2f}")
→ Model accuracy: 1.00
# Visualize feature importance
feature_importances = dt_model.feature_importances_
plt.figure(figsize=(10, 6))
sns.barplot(x=feature_importances, y=iris.feature_names, palette="viridis")
plt.title("Feature Importance")
plt.xlabel("Importance")
plt.ylabel("Features")
plt.show()
```

<ipython-input-9-327d3bdff328>:4: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `le sns.barplot(x=feature_importances, y=iris.feature_names, palette="viridis")



```
# Visualize the Decision Tree
plt.figure(figsize=(16, 10))
plot_tree(dt_model, feature_names=iris.feature_names, class_names=iris.target_names.tolist(), filled=True)
plt.title("Decision Tree Visualization ")
plt.show()
```



plt.show()

Decision Tree Visualization

plt.suptitle("Pairplot of Iris Dataset", y=1.02)

|--- class: 2

petal length (cm) <= 2.45

```
gini = 0.5
samples = 8
value = [0, 4, 4]
Decision Dece Bules:
              voltea ! [PeAg.th] (cm) <
                                                                      class = versicolor
              petal length (cm) > 2.45
                                                                                petal width (cm) <= 1.55
gini = 0.444
samples = 6
value = [0, 2, 4]
class = virginica
                                                                                                                                         sepal width (cm) <= 3.1
gini = 0.444
samples = 3
value = [0, 1, 2]
class = virginica
                -- petal length (cm) <= 4.75
                   --- petal width (cm) <= 1.65amples = 2
                      |--- class: 1
                      - petal width (cm) >
                       |--- class: 2
                  petal length (cm) > 4.75
                                                                                              petal length (cm) <= 5.45
gini = 0.444
samples = 3
                  --- petal width (cm) <= 1.75
                                                                                                                                                            samples = 1
value = [0, 1, 0]
class = versicolor
                         --- petal length (cm) <= 4.95
                                                                                                 value = [0, 2, 1]
class = versicolor
                             |--- class: 1
                              petal length (cm) > 4.95
                               --- petal width (cm) <= 1.55
                                                                                     gini = 2
samples = 2
value = [0, 2, 0]
lass = versicolo
                                  |--- class: 2
                                   petal width (cm) > 1.55
                                     --- petal length (cm) <= 5.45
                                       |--- class: 1
                                    --- petal length (cm) > 5.45
                                       |--- class: 2
                        petal width (cm) > 1.75
                          --- petal length (cm) <= 4.85
                               --- sepal width (cm) <= 3.10
                                  |--- class: 2
                                    sepal width (cm) > 3.10
                                   |--- class: 1
                              petal length (cm) > 4.85
```

