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Domain:- Machine Learning

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Company:- CODETECH IT SOLUTIONS

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TASK ONE: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) \
                     5.1
                                       3.5
                                                          1.4
                                                                            0.2
    1
                     4.9
                                       3.0
                                                          1.4
                                                                            0.2
     2
                     4.7
                                       3.2
                                                          1.3
                                                                            0.2
     3
                     4.6
                                       3.1
                                                          1.5
                                                                            0.2
     4
                                                                            0.2
                     5.0
                                       3.6
       target
    0
            0
     2
            0
     3
            0
     4
#getting the top 10 values
```

data.head(10)

New interactive sheet

⊋ *	sepal	length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	
	0	5.1	3.5	1.4	0.2	0	il.
	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	

View recommended plots

geeting the 10 bottom values

Generate code with data

Next steps:

data.tail(10)

}	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	target	=
140	6.7	3.1	5.6	2.4	2	ıl.
141	6.9	3.1	5.1	2.3	2	
142	5.8	2.7	5.1	1.9	2	
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	
146	6.3	2.5	5.0	1.9	2	
147	6.5	3.0	5.2	2.0	2	
148	6.2	3.4	5.4	2.3	2	
149	5.9	3.0	5.1	1.8	2	

```
# Spliting the dataset 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)

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)

DecisionTreeClassifier ① ?

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")

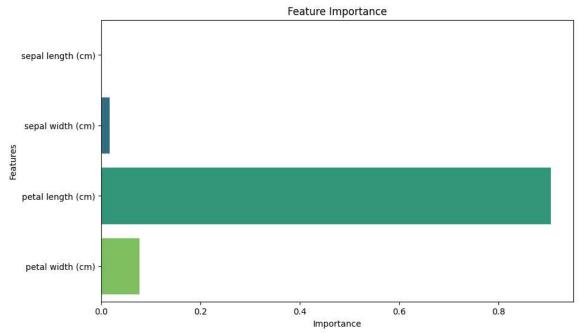
X = data[iris.feature_names]

y = data['target']

```
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 · sns.barplot(x=feature_importances, y=iris.feature_names, palette="viridis")



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



Decision Tree Visualization | petal length (cm) <= 2.45 | gmi = 0.667 | samples = 120 | value = [40, 41, 39] | class = versicolor | samples = 40 | value = [40, 0.0] | class = setosa | value = [40, 0.0] | class = setosa | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] | class = versicolor | value = [40, 0.0] |

```
→ Decision Tree Rules:
     |--- petal length (cm) <= 2.45
        |--- class: 0
     --- petal length (cm) > 2.45
        |--- petal length (cm) <= 4.75
            |--- petal width (cm) <= 1.65
               |--- class: 1
            |--- petal width (cm) > 1.65
              |--- class: 2
         --- petal length (cm) > 4.75
                petal width (cm) <= 1.75
                |--- petal length (cm) <= 4.95
                  |--- class: 1
                   - petal length (cm) > 4.95
                     --- petal width (cm) <= 1.55
                       |--- 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
                    |--- class: 2
```

##