

Ex. No.: 7

## **A PYTHON PROGRAM TO IMPLEMENT DECISION TREE**

### **Aim:**

To implement a decision tree using a python program for the given dataset and plot the trained decision tree.

### **Algorithm:**

Step 1: Import the Iris Dataset

- Import ``load_iris`` from ``sklearn.datasets``.

Step 2: Import Necessary Libraries

- Import numpy as np.
- Import matplotlib.pyplot as plt.
- Import ``DecisionTreeClassifier`` from ``sklearn.tree``.

Step 3: Declare and Initialize Parameters

- Declare and initialize ``n_classes = 3``.
- Declare and initialize ``plot_colors = "ryb"``.
- Declare and initialize ``plot_step = 0.02``.

Step 4: Prepare Data for Model Training

- Load the iris dataset using ``load_iris()``.
- Assign the dataset's data to variable ``X``.
- Assign the dataset's target to variable ``Y``.

Step 5: Train the Model

- Create an instance of ``DecisionTreeClassifier``.
- Fit the classifier using ``clf.fit(X, Y)``.

Step 6: Initialize Pair Index and Plot Graph

- Loop through each pair of features using ``for pairidx, pair in enumerate(combinations(range(X.shape[1]), 2)):``

- Inside the loop, assign `X` with the selected pair of features (e.g., `X = iris.data[:, pair]`).
- Assign `Y` with the target list (e.g., `Y = iris.target`).

#### Step 7: Assign Axis Limits

1. Inside the loop, assign `x\_min` with the minimum value of the selected feature minus 1 (e.g., `x\_min, x\_max = X[:, 0].min() - 1, X[:, 0].max() + 1`).
2. Assign `x\_max` with the maximum value of the selected feature plus 1.
3. Assign `y\_min` with the minimum value of the second selected feature minus 1 (e.g., `y\_min, y\_max = X[:, 1].min() - 1, X[:, 1].max() + 1`).
4. Assign `y\_max` with the maximum value of the second selected feature plus 1.

#### Step 8: Create Meshgrid

1. Use `np.meshgrid` to create a grid of values from `x\_min` to `x\_max` and `y\_min` to `y\_max` with steps of `plot\_step`.
2. Assign the results to variables `xx` and `yy`.

#### Step 9: Plot Graph with Tight Layout

1. Use `plt.tight\_layout()` to adjust the layout of the plots.
2. Set `h\_pad=0.5`, `w\_pad=0.5`, and `pad=2.5`.

#### Step 10: Predict and Reshape

1. Use the classifier to predict on the meshgrid (e.g., `Z = clf.predict(np.c\_[xx.ravel(), yy.ravel()])`).
2. Reshape `Z` to the shape of `xx`.

#### Step 11: Plot Decision Boundary

1. Use `plt.contourf(xx, yy, Z, cmap=plt.cm.RdYlBu)` to plot the decision boundary with the "RdYlBu" color scheme.

#### Step 12: Plot Feature Pairs

1. Inside the loop, label the x-axis and y-axis with the feature names (e.g., `plt.xlabel(iris.feature\_names[pair[0]])` and `plt.ylabel(iris.feature\_names[pair[1]])`).

#### Step 13: Plot Training Points

1. Use `plt.scatter(X[:, 0], X[:, 1], c=Y, cmap=plt.cm.RdYlBu, edgecolor='k', s=15)` to plot the training points with the "RdYlBu" color scheme, black edge color,

and size 15.

#### Step 14: Plot Final Decision Tree

1. Set the title of the plot to "Decision tree trained on all the iris features" (e.g.,

`plt.title("Decision tree trained on all the iris features")`).

2. Display the plot using `plt.show()`.

#### PROGRAM:

```
from sklearn.datasets import load_iris
iris = load_iris()
import numpy as np
import matplotlib.pyplot as plt
from sklearn.tree import DecisionTreeClassifier

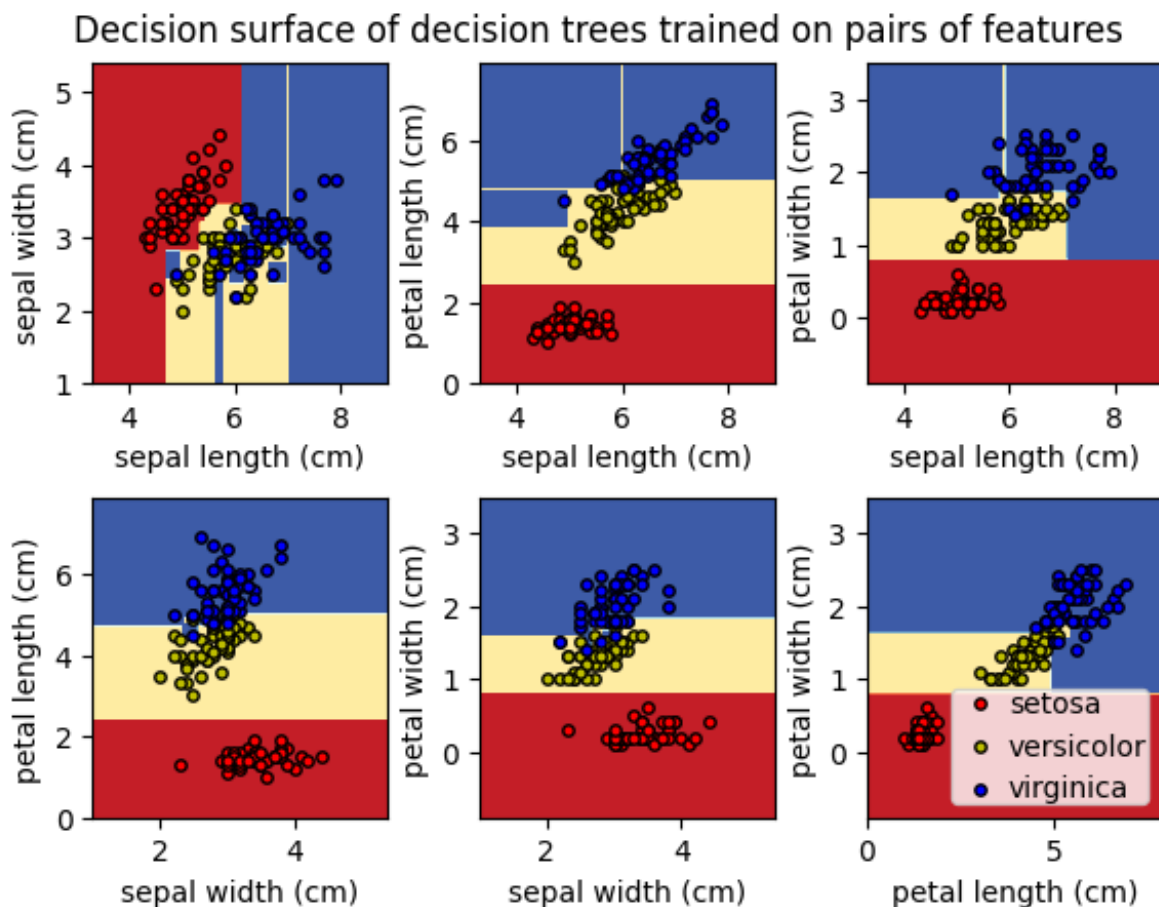
n_classes = 3
plot_colors = "ryb"
plot_step = 0.02

for pairidx, pair in enumerate([[0, 1], [0, 2], [0, 3], [1, 2], [1, 3],
[2, 3]]):
    X = iris.data[:, pair]
    y = iris.target
    clf = DecisionTreeClassifier().fit(X, y)
    plt.subplot(2, 3, pairidx + 1)
    x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
    y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
    xx, yy = np.meshgrid(
        np.arange(x_min, x_max, plot_step), np.arange(y_min, y_max,
plot_step)
    )
    plt.tight_layout(h_pad=0.5, w_pad=0.5, pad=2.5)
    Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])
    Z = Z.reshape(xx.shape)
    cs = plt.contourf(xx, yy, Z, cmap=plt.cm.RdYlBu)
    plt.xlabel(iris.feature_names[pair[0]])
    plt.ylabel(iris.feature_names[pair[1]])
    for i, color in zip(range(n_classes), plot_colors):
        idx = np.where(y == i)
        plt.scatter(
            X[idx, 0],
            X[idx, 1], c=color,
            label=iris.target_names[i], edgecolor="black",
```

```
s=15)
```

```
plt.suptitle("Decision surface of decision trees trained on pairs of  
features")  
plt.legend(loc="lower right", borderpad=0, handletextpad=0)  
plt.axis("tight")  
  
plt.figure()  
clf = DecisionTreeClassifier().fit(iris.data, iris.target)  
from sklearn.tree import plot_tree  
plot_tree(clf, filled=True)  
plt.title("Decision tree trained on all the iris features")  
plt.show()
```

## OUTPUT :



## Decision tree trained on all the iris features



### RESULT:

Thus the python program to implement Decision Tree for the given dataset has been successfully implemented and the results have been verified and analyzed