

## 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()` .