## Python Program to implement Decision Tree.

```
In [1]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         %matplotlib inline
         from sklearn.preprocessing import LabelEncoder
         from sklearn.model selection import train test split
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.metrics import classification report, confusion matrix
         from sklearn.tree import plot tree
         # Input: Dataset
         iris = sns.load_dataset('iris')
         target = iris['species']
         iris1 = iris.copy()
         iris1 = iris1.drop('species', axis =1)
         # Defining the attributes
         X = iris1
         le = LabelEncoder()
                                           # Label Encoding
         target = le.fit_transform(target)
         Y = target
         # Dividing into test and training sets
         X_train, X_test, Y_train, Y_test = train_test_split(X , Y, test_size = 0.3)
         # Create Decision Tree Classifier object
         dtree = DecisionTreeClassifier(criterion = "entropy", max depth = 5)
         # Train the model using the training sets
         dtree.fit(X_train,Y train)
         # Make predictions using the testing set
         Y pred = dtree.predict(X test)
```

```
Exercise-5
         Output for the Decision Tree.
In [3]: # Output: The Classification Report, Confusion Matrix and Decision Tree
          print("Classification Report - \n", classification report(Y test, Y pred))
          cm = confusion matrix(Y test, Y pred)
          plt.figure(figsize=(5,5))
          sns.heatmap(data=cm,linewidths=.5, annot=True,square = True, cmap = 'Greens')
          plt.ylabel('Actual label')
          plt.xlabel('Predicted label')
          all sample title = 'Accuracy Score: {0:.2f}%'.format(dtree.score(X test, Y test)*100)
          plt.title(all sample title, size = 10)
          plt.figure(figsize = (20,20))
          dec tree = plot tree(decision tree=dtree, feature names = iris.columns,
                                   class names =["setosa", "vercicolor", "verginica"] ,
                                   filled = True , precision = 4, rounded = True)
          Classification Report -
                           precision recall f1-score support

      1.00
      1.00
      1.00

      1.00
      0.95
      0.97

      0.93
      1.00
      0.97

                      0
                                                                     11
                                                                     20
                       1
              accuracy
                                                       0.98
                                                                     45
                              0.98
             macro avg
                                                                     45
                                          0.98
                                                      0.98
          weighted avg
                      Accuracy Score: 97.78%
                                                      15.0
                                                     - 12.5
          Actual labe
                                                      10.0
                    0
                               19
                                          1
                                                      7.5
                                                      5.0
                               0
                                                     - 2.5
                                           2
                    0
                               1
                          Predicted label
                                                     - 0.0
                                          petal_width <= 0.8
                                          entropy = 1.5766
                                           samples = 105
                                         value = [39, 30, 36]
                                           class = setosa
                                                   petal length <= 4.85
                                  entropy = 0.0
                                                     entropy = 0.994
                                  samples = 39
                                                      samples = 66
                                value = [39, 0, 0]
                                                    value = [0, 30, 36]
class = verginica
                                 class = setosa
                     petal_width <= 1.65
                                                                                  petal_width <= 1.7
                      entropy = 0.3534
                                                                                   entropy = 0.3095
                       samples = 30
                                                                                    samples = 36
                                                                                   value = [0, 2, 34]
                      value = [0, 28, 2]
                      class = vercicolor
                                                                                   class = verginica
```

