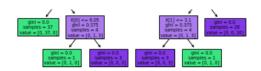
```
In [ ]:
          import pandas as pd
           import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
           df = sns.load_dataset("iris")
           df.head()
            sepal_length sepal_width petal_length petal_width species
         0
                     5.1
                                 3.5
                                              1.4
                                                          0.2 setosa
                     4.9
                                 3.0
         1
                                              1.4
                                                          0.2 setosa
         2
                     4.7
                                 3.2
                                              1.3
                                                               setosa
         3
                     4.6
                                 3.1
                                              1.5
                                                          0.2
                                                               setosa
                     5.0
                                 3.6
                                                          0.2 setosa
          X= df.iloc[ : ,:-1]
          y= df.iloc[ : ,-1:]
In [ ]:
          X.head()
            sepal\_length \quad sepal\_width \quad petal\_length \quad petal\_width
Out[ ]:
         0
                     5.1
                                 3.5
                                              1.4
                                                          0.2
                     4.9
                                                          0.2
         2
                     47
                                 3.2
                                              1.3
                                                          0.2
                     4.6
                                 3.1
                                              1.5
         4
                     5.0
                                                          0.2
                                 3.6
                                              1.4
In [ ]:
          y.head()
Out[ ]: species
         0 setosa
         1 setosa
         2 setosa
             setosa
         4 setosa
In [ ]: from sklearn.tree import DecisionTreeClassifier
           from sklearn.tree import plot_tree
           from sklearn.model_selection import train_test_split
           from sklearn import metrics
          \textbf{from} \  \, \text{sklearn.metrics} \  \, \textbf{import} \  \, \text{confusion\_matrix}
          from sklearn.metrics import accuracy_score
In [ ]: # Splitting the dataset into train and test
           X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3, random_state = 100)
In [ ]:
          \ensuremath{\textit{\#}} Decision tree classifier , fitting the training data
           model = DecisionTreeClassifier(random_state = 100).fit(X_train, y_train)
          plot_tree(model, filled = True)
           plt.title("Decision tree trained model")
          plt.show()
                       Decision tree trained model
```



```
In [ ]:
                            \begin{tabular}{ll} \#plt.savefig("tiff_compressed.tiff", dpi=600, format = "tiff", facecolor = 'white', edgecolor = 'none', \#pil_kwargs = \{"compression": "tiff_lzw"\} \end{tabular} \label{table:plt:savefig} 
 In [ ]:
                           \# prediction on X_{test} (testing data )
                           y_pred = model.predict(X_test)
                           y_pred
Out[]: array(['virginica', 'setosa', 'virginica', 'setosa', 'virginica', 'virginica', 'setosa', 'setosa', 'virginica', 'setosa', 'setosa', 'virginica', 'versicolor', 'versicolor', 'virginica', 'virginica', 'virginica', 'virginica', 'virginica', 'virginica',
                                            'setosa', 'virginica', 'setosa', 'versicolor', 'virginica', 'versicolor', 'setosa', 'versicolor', 'virginica', 'versicolor', 'versicolor', 'setosa', 'setosa', 'versicolor', 'setosa', 'versicolor', 'virginica', 'virginica', 'setosa', 'setosa', 'setosa', 'versicolor', 'virginica', 'virginica', 'setosa', 'versicolor', 'virginica', 'setosa', 'versicolor', 'virginica', 'setosa', 'versicolor', 'virginica', 'virginica', 'setosa', 'versicolor', 'virginica', 'virginica', 'setosa', 'versicolor', 'virginica', 'virginica', 'setosa', 'versicolor', 'virginica', 'virginica', 'versicolor', 'setosa', 'versicolor', 'virginica', 'versicolor', 
                                             'versicolor', 'virginica', 'virginica', 'setosa'], dtype=object)
 In [ ]: #acuuracy score with 10/90
                           score = model.score(X_test, y_test)
print("Acuuracy score is : " ,score)
                         Acuuracy score is : 1.0
 In [ ]: #acuuracy score with 20/80
                            score = model.score(X_test, y_test)
                            print("Acuuracy score is : " ,score)
                         Acuuracy score is : 0.966666666666667
 In [ ]: #acuuracy score with 30/70
                           score = model.score(X_test, y_test)
                           print("Acuuracy score is : " ,score)
                         Acuuracy score is: 0.95555555555556
 In [ ]: #Accuray of the model
                            print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
                           cm= metrics.confusion_matrix(y_test, y_pred)
                           cm
                         Accuracy: 0.95555555555556
                         array([[16, 0, 0],
 Out[ ]:
                                             [ 0, 10, 1],
                                            [ 0, 1, 17]], dtype=int64)
                           X_train.head(10)
Out[ ]:
                                     sepal_length sepal_width petal_length petal_width
                               6
                                                            4.6
                                                                                           3.4
                                                                                                                            1.4
                                                                                                                                                          0.3
                            25
                                                            5.0
                                                                                           3.0
                                                                                                                            1.6
                                                                                                                                                          0.2
                            21
                                                            5.1
                                                                                           3.7
                                                                                                                            1.5
                                                                                                                                                          0.4
                                                            5.8
                                                                                                                           4.0
                            92
                                                                                           2.6
                                                                                                                                                          1.2
                               9
                                                            4.9
                                                                                           3.1
                                                                                                                            1.5
                                                                                                                                                          0.1
                            23
                                                                                           3.3
                                                                                                                            1.7
                                                                                                                                                          0.5
                                                            5.1
                            35
                                                            5.0
                                                                                           3.2
                                                                                                                            1.2
                                                                                                                                                          0.2
                            54
                                                            6.5
                                                                                           2.8
                                                                                                                            4.6
                                                                                                                                                          1.5
                          131
                                                            7.9
                                                                                           3.8
                                                                                                                            6.4
                                                                                                                                                          2.0
                                                                                           3.0
                          127
                                                            6.1
                                                                                                                            4.9
                                                                                                                                                          1.8
 In [ ]:
                            #Sample Value Prediction:
```

print(model.predict([[5.1, 3.2, 1.9, 0.7], [5.1,3.5,1.4,0.2],[4.8,3.9,3.4,0.4],[4.1,3.1,2.6,0.7],[5.2, 3.6,2.9,2.3]]))
['setosa' 'setosa' 'versicolor' 'versicolor' 'versicolor']