# Prediction using Decision Tree

### Done by

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```
In [1]:
          import pandas as pd
          import numpy as np
          from sklearn.datasets import load_iris
          from sklearn.model_selection import train_test_split
          from sklearn.tree import DecisionTreeClassifier
          from sklearn import tree
          from matplotlib import pyplot as plt
          from sklearn.metrics import accuracy_score
          import matplotlib.pyplot as plt
          import seaborn as sns
In [7]:
          iris=load_iris()
          df=pd.DataFrame(iris.data,columns=iris.feature_names)
          df.head()
            sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
Out[7]:
          0
                        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
                        5.0
                                      3.6
                                                     1.4
                                                                    0.2
         Shape
In [8]:
          df.shape
Out[8]: (150, 4)
         Preparing Data for creating a model
In [9]:
          x=iris.data
          y=iris.target
          x_train, x_test, y_train, y_test=train_test_split(x, y, test_size=0.2)
          print("x train data", x_train.shape)
          print("x test data", x_test.shape)
          print("y train data", y_train.shape)
          print("y test data", y_test.shape)
          x train data (120, 4)
          x test data (30, 4)
          y train data (120,)
          y test data (30,)
In [10]:
          model=DecisionTreeClassifier()
          model.fit(x_train,y_train)
          print("model is trained")
          model is trained
         visualizing
          plt.figure(figsize=(20,10))
          tree.plot_tree(model, filled=True)
Out[11]: [Text(558.0, 498.3, 'X[3] <= 0.8\ngini = 0.664\nsamples = 120\nvalue = [41, 35, 44]'),
          Text(472.15384615384613, 407.70000000000005, 'gini = 0.0\nsamples = 41\nvalue = [41, 0, 0]'),

Text(643.8461538461538, 407.700000000005, 'X[3] <= 1.75\ngini = 0.494\nsamples = 79\nvalue = [0, 35, 44]'),

Text(343.38461538461536, 317.1, 'X[2] <= 4.95\ngini = 0.224\nsamples = 39\nvalue = [0, 34, 5]'),
          Text(171.69230769230768, 226.5, 'X[3] \le 1.65 \cdot ngini = 0.059 \cdot nsamples = 33 \cdot nvalue = [0, 32, 1]'),
          Text(429.23076923076917, 135.8999999999999, 'gini = 0.0\nsamples = 3\nvalue = [0, 0, 3]'),

Text(600.9230769230769, 135.899999999999, 'X[0] <= 6.95\ngini = 0.444\nsamples = 3\nvalue = [0, 0, 3]'),

Text(515.0769230769231, 45.2999999999955, 'gini = 0.0\nsamples = 2\nvalue = [0, 2, 0]'),

Text(686.7692307692307, 45.2999999999955, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]'),
          Text(944.3076923076923, 317.1, 'X[2] \le 4.85 \cdot ngini = 0.049 \cdot nsamples = 40 \cdot nvalue = [0, 1, 39]'),
          Text(858.4615384615383, 226.5, 'X[1] \le 3.0 \cdot samples = 2 \cdot value = [0, 1, 1]'),
          Text(772.6153846153845, 135.899999999999, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]'),
          Text(944.3076923076923, 135.8999999999999, 'gini = 0.0\nsamples = 1\nvalue = [0, 1, 0]'),
          Text(1030.1538461538462, 226.5, 'gini = 0.0 \nsamples = 38 \nvalue = [0, 0, 38]')]
                                                                                X[3] \le 0.8
                                                                                gini = 0.664
                                                                               samples = 120
                                                                            value = [41, 35, 44]
                                                                                            X[3] \le 1.75
                                                                      gini = 0.0
                                                                                            gini = 0.494
                                                                    samples = 41
                                                                                            samples = 79
                                                                  value = [41, 0, 0]
                                                                                         value = [0, 35, 44]
                                                  X[2] \le 4.95
                                                                                                                                      X[2] <= 4.85
                                                  gini = 0.224
                                                                                                                                      gini = 0.049
                                                  samples = 39
                                                                                                                                     samples = 40
                                                value = [0, 34, 5]
                                                                                                                                   value = [0, 1, 39]
                                                                                                                          X[1] \le 3.0
                           X[3] <= 1.65
                                                                          X[3] <= 1.55
                                                                                                                                                   gini = 0.0
                           gini = 0.059
                                                                          gini = 0.444
                                                                                                                           gini = 0.5
                                                                                                                                                 samples = 38
                                                                          samples = 6
                          samples = 33
                                                                                                                          samples = 2
                                                                                                                                               value = [0, 0, 38]
                        value = [0, 32, 1]
                                                                         value = [0, 2, 4]
                                                                                                                        value = [0, 1, 1]
                                                                                      X[0] <= 6.95
                 gini = 0.0
                                                                gini = 0.0
                                                                                                                gini = 0.0
                                        gini = 0.0
                                                                                                                                       gini = 0.0
                                                                                      gini = 0.444
              samples = 32
                                                              samples = 3
                                                                                                              samples = 1
                                                                                                                                      samples = 1
                                       samples = 1
                                                                                      samples = 3
            value = [0, 32, 0]
                                     value = [0, 0, 1]
                                                             value = [0, 0, 3]
                                                                                                            value = [0, 0, 1]
                                                                                                                                    value = [0, 1, 0]
                                                                                     value = [0, 2, 1]
                                                                            gini = 0.0
                                                                                                   gini = 0.0
                                                                          samples = 2
                                                                                                  samples = 1
```

## Testing the model created

```
prediction=model.predict(x_test)
prediction
df=pd.DataFrame({'predicted value':prediction, 'actual value':y_test})
df['label']=df['predicted value'].replace(dict(enumerate(iris.target_names)))
df.head()
```

value = [0, 0, 1]

value = [0, 2, 0]

label	actual value	predicted value		Out[12]:
virginica	2	2	0	
setosa	0	0	1	
virginica	2	2	2	
versicolor	1	1	3	
versicolor	1	1	4	

### Accuracy

```
score=accuracy_score(prediction, y_test)
print(f"Accuracy score is {score} i.e. {score*100}%")
```

Accuracy score is 1.0 i.e. 100.0%