```
In [1]: # importing some important libraries
           import pandas as pd
           import numpy as np
           from sklearn.ensemble import BaggingClassifier
           from sklearn.model selection import train test split, GridSearchCV
           from sklearn.tree import DecisionTreeClassifier
           from sklearn import tree
           import matplotlib.pyplot as plt
           %matplotlib inline
 In [3]: # reading winequality csv file
           df = pd.read csv("https://raw.githubusercontent.com/aniruddhachoudhury/Red-Wine-Q
In [21]: df
Out[21]:
                                                              free
                                                                      total
                         volatile
                                 citric
                   fixed
                                        residual
                                                 chlorides
                                                             sulfur
                                                                     sulfur
                                                                            density
                                                                                         sulphates alcoh
                  acidity
                          acidity
                                  acid
                                          sugar
                                                           dioxide
                                                                   dioxide
               0
                           0.700
                                  0.00
                                                     0.076
                                                              11.0
                                                                       34.0 0.99780 3.51
                     7.4
                                             1.9
                                                                                               0.56
                                                                                                         9
               1
                     7.8
                           0.880
                                  0.00
                                             2.6
                                                     0.098
                                                              25.0
                                                                       67.0 0.99680
                                                                                    3.20
                                                                                               0.68
                                                                                                         9
               2
                     7.8
                           0.760
                                  0.04
                                             2.3
                                                     0.092
                                                              15.0
                                                                       54.0 0.99700
                                                                                    3.26
                                                                                               0.65
                                                                                                         9
                           0.280
                                  0.56
                                                     0.075
                                                                       60.0 0.99800
               3
                    11.2
                                             1.9
                                                              17.0
                                                                                    3.16
                                                                                               0.58
                                                                                                         9
                     7.4
                           0.700
                                  0.00
                                             1.9
                                                     0.076
                                                                            0.99780
               4
                                                              11.0
                                                                       34.0
                                                                                    3.51
                                                                                               0.56
                                                                                                         9
                                             ...
              ---
                      ...
                              ...
                                    ...
                                                        ...
                                                                ...
                                                                                 ...
                                                                                                 ...
            1594
                     6.2
                           0.600
                                  0.08
                                             2.0
                                                     0.090
                                                              32.0
                                                                       44.0
                                                                           0.99490
                                                                                    3.45
                                                                                               0.58
                                                                                                        10
            1595
                     5.9
                           0.550
                                  0.10
                                             2.2
                                                     0.062
                                                              39.0
                                                                       51.0 0.99512 3.52
                                                                                               0.76
                                                                                                        11
            1596
                     6.3
                                  0.13
                                             2.3
                                                     0.076
                                                              29.0
                                                                       40.0 0.99574 3.42
                                                                                               0.75
                           0.510
                                                                                                        11
                                  0.12
                                             2.0
                                                     0.075
                                                                       44.0 0.99547 3.57
            1597
                     5.9
                           0.645
                                                              32.0
                                                                                               0.71
                                                                                                        10
                           0.310
                                  0.47
                                             3.6
                                                     0.067
                                                                       42.0 0.99549 3.39
                                                                                               0.66
            1598
                     6.0
                                                              18.0
                                                                                                        11
           1599 rows × 12 columns
          x = df.drop(columns=['quality'])
 In [5]:
          y = df.quality
In [10]: # train and test split
           X_train,x_test,Y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=1
```

## BaggingClassifier with DecisionTreeClassifier

In [11]: # Creating a bagging classifier object 100 decision trees as base estimators
bag\_dt = BaggingClassifier(DecisionTreeClassifier(),n\_estimators=100)

```
In [13]: # Training the data
         bag_dt.fit(X_train,Y_train)
Out[13]: BaggingClassifier(base_estimator=DecisionTreeClassifier(), n_estimators=100)
In [17]: bag dt.base estimator # showing base estimator
Out[17]: DecisionTreeClassifier()
In [20]: bag dt.classes # showing classes or label values
Out[20]: array([3, 4, 5, 6, 7, 8], dtype=int64)
In [41]: # Numbers of features used
         bag_dt.n_features_in_
Out[41]: 11
In [42]: # we can check parameters which are used to train decision trees
         bag_dt.get_params()
Out[42]: {'base_estimator__ccp_alpha': 0.0,
           'base_estimator__class_weight': None,
          'base_estimator__criterion': 'gini',
          'base estimator max depth': None,
           'base_estimator__max_features': None,
           'base estimator max leaf nodes': None,
           'base_estimator__min_impurity_decrease': 0.0,
           'base_estimator__min_impurity_split': None,
          'base estimator min samples leaf': 1,
           'base estimator min samples split': 2,
           'base_estimator__min_weight_fraction_leaf': 0.0,
           'base estimator presort': 'deprecated',
           'base_estimator__random_state': None,
           'base estimator splitter': 'best',
           'base estimator': DecisionTreeClassifier(),
           'bootstrap': True,
           'bootstrap features': False,
           'max_features': 1.0,
           'max samples': 1.0,
           'n estimators': 100,
           'n jobs': None,
           'oob score': False,
           'random state': None,
           'verbose': 0,
           'warm_start': False}
In [43]: # accuracy score
         bag dt.score(x test,y test)
Out[43]: 0.69375
```

```
In [44]: # making prediction
         bag dt.predict(x test)
Out[44]: array([5, 5, 5, 5, 6, 6, 5, 6, 5, 5, 6, 6, 4, 5, 5, 5, 6, 6, 5, 5, 5, 7,
                5, 5, 6, 5, 5, 5, 6, 5, 6, 5, 5, 6, 5, 7, 6, 5, 6, 6, 5, 5, 6, 5,
                7, 5, 6, 6, 6, 6, 6, 5, 5, 5, 6, 5, 5, 5, 5, 5, 5, 5, 6, 5, 5,
                6, 5, 6, 6, 5, 6, 6, 6, 5, 7, 5, 6, 6, 6, 5, 7, 5, 6, 6, 6, 5, 5,
                5, 5, 5, 6, 5, 6, 6, 6, 6, 5, 7, 6, 5, 5, 7, 5, 6, 6, 5, 5, 5, 6,
                5, 5, 5, 5, 5, 6, 6, 5, 5, 6, 5, 6, 6, 5, 6, 6, 6, 6, 5, 7, 6, 5,
                5, 5, 5, 6, 6, 6, 5, 7, 5, 6, 6, 5, 6, 7, 4, 5, 5, 6, 5, 5, 6,
                6, 6, 5, 5, 5, 7, 6, 6, 6, 5, 6, 6, 5, 5, 5, 6, 7, 7, 5, 6, 6, 7,
                5, 6, 5, 5, 7, 7, 6, 6, 5, 5, 6, 6, 5, 5, 6, 5, 6, 6, 6, 7, 5, 6,
                7, 6, 7, 6, 6, 5, 6, 5, 5, 5, 6, 6, 6, 5, 6, 6, 5, 5, 6, 7, 6,
                5, 5, 7, 5, 6, 7, 5, 8, 6, 6, 5, 6, 6, 7, 5, 6, 7, 5, 6, 5, 5,
                7, 5, 6, 6, 5, 6, 5, 6, 6, 6, 5, 5, 6, 6, 5, 5, 7, 6, 5, 5, 6,
                6, 5, 5, 7, 6, 5, 6, 7, 7, 6, 5, 7, 6, 6, 6, 5, 5, 6, 6, 5, 5, 5,
                5, 6, 5, 7, 5, 6, 6, 7, 7, 7, 6, 5, 7, 5, 5, 5, 5, 5, 5, 5, 5,
                6, 6, 7, 6, 6, 5, 5, 6, 5, 5, 6], dtype=int64)
In [46]: # prediction probabilites of each class for each data row in the x test
         bag_dt.predict_proba(x_test)
Out[46]: array([[0. , 0.01, 0.46, 0.32, 0.2 , 0.01],
                [0., 0.06, 0.91, 0.03, 0., 0.],
                [0. , 0.04, 0.85, 0.11, 0. , 0.
                . . . ,
                [0. , 0.04, 0.71, 0.25, 0. , 0.
                [0.2, 0.05, 0.69, 0.04, 0.02, 0.]
                [0., 0.01, 0.22, 0.56, 0.2, 0.01]])
         BaggingClassifier using KNeighborsClassifier
```

```
In [55]: # classes or label in which classification has to be done
         bag knn.classes
Out[55]: array([3, 4, 5, 6, 7, 8], dtype=int64)
In [56]: # Numbers of features on which model is trained
         bag_knn.n_features_in_
Out[56]: 11
In [59]: # shows all the parameters availabe which we can play to influence the quality of
         bag_knn.get_params()
Out[59]: {'base_estimator__algorithm': 'auto',
           'base_estimator__leaf_size': 30,
           'base estimator metric': 'minkowski',
           'base_estimator__metric_params': None,
           'base estimator n jobs': None,
           'base_estimator__n_neighbors': 6,
           'base_estimator__p': 2,
           'base estimator weights': 'uniform',
           'base_estimator': KNeighborsClassifier(n_neighbors=6),
           'bootstrap': True,
           'bootstrap features': False,
           'max_features': 1.0,
           'max_samples': 1.0,
           'n estimators': 100,
           'n jobs': None,
           'oob score': False,
           'random state': None,
           'verbose': 0,
           'warm_start': False}
In [61]: # shows accuracy of the model
         bag_knn.score(x_test,y_test)
Out[61]: 0.5
```

```
In [63]: # prediction on x test dataset
         bag knn.predict(x test)
Out[63]: array([6, 5, 5, 5, 6, 5, 6, 5, 6, 6, 5, 6, 5, 5, 5, 5, 6, 5, 6, 5, 7,
                5, 6, 7, 6, 6, 5, 5, 5, 5, 5, 5, 6, 5, 7, 6, 5, 6, 6, 5, 5, 6, 5,
                7, 6, 5, 5, 6, 6, 6, 5, 5, 6, 6, 6, 5, 6, 5, 6, 5, 6, 5, 5, 5,
                5, 6, 7, 5, 5, 5, 6, 6, 5, 5, 6, 5, 6, 6, 6, 6, 7, 5, 6, 5, 5,
                5, 5, 5, 6, 5, 6, 6, 6, 5, 5, 7, 6, 5, 5, 7, 4, 6, 6, 6, 5, 7, 6,
                5, 6, 7, 5, 5, 6, 5, 5, 5, 6, 5, 5, 6, 5, 6, 5, 5, 5, 6, 5, 5,
                5, 5, 7, 5, 6, 5, 6, 6, 5, 5, 6, 6, 5, 6, 4, 6, 5, 6, 6, 5, 6, 6,
                6, 5, 5, 6, 5, 6, 6, 6, 6, 5, 5, 7, 5, 7, 5, 5, 6, 6, 5, 6, 6, 6,
                6, 7, 5, 5, 7, 7, 7, 5, 5, 6, 6, 5, 7, 5, 6, 5, 6, 6, 5, 7, 5, 5,
                6, 6, 6, 5, 7, 5, 6, 5, 6, 6, 5, 6, 6, 5, 6, 7, 5, 5, 7, 6, 6,
                5, 5, 6, 6, 6, 7, 5, 6, 7, 6, 5, 5, 7, 6, 6, 5, 8, 5, 6, 5, 5, 7,
                6, 5, 5, 6, 5, 6, 5, 5, 5, 5, 6, 5, 5, 6, 7, 6, 5, 6, 6, 5, 5, 5,
                6, 5, 5, 7, 6, 4, 7, 6, 7, 6, 5, 7, 7, 5, 6, 5, 6, 6, 6, 5, 5, 5,
                5, 6, 6, 6, 5, 6, 5, 7, 6, 6, 5, 5, 6, 6, 5, 5, 5, 5, 5, 5, 5, 6,
                5, 7, 5, 5, 6, 5, 5, 5, 5, 6], dtype=int64)
In [65]: # prediction probabilites of each class for each data row in the x_test
         bag knn.predict proba(x test)
Out[65]: array([[0.
                           , 0.15833333, 0.055
                                                   , 0.50666667, 0.10166667,
                 0.17833333],
                           , 0.00166667, 0.97833333, 0.02
                [0.
                 0.
                           ],
                [0.
                           , 0.195
                                       , 0.775
                                                   , 0.03
                 0.
                           ],
                . . . ,
                           , 0.
                                                   , 0.00166667, 0.12333333,
                [0.
                                       , 0.805
                 0.07
                           ],
                           , 0.24666667, 0.40166667, 0.19666667, 0.00333333,
                 0.15166667],
                [0.
                           , 0.
                                       , 0.07166667, 0.76333333, 0.165
                 0.
                           ]])
```

#### RandomForestClassifier

```
In [68]: # importing random forest
from sklearn.ensemble import RandomForestClassifier

In [91]: # creating random forest classifier object
    rf = RandomForestClassifier(100)

In [120]: # training model
    rf.fit(X_train,Y_train)

Out[120]: RandomForestClassifier()

In [93]: # base estimator is shown as decision tree classifer
    rf.base_estimator_
Out[93]: DecisionTreeClassifier()
```

```
In [94]: # classes
          rf.classes
 Out[94]: array([3, 4, 5, 6, 7, 8], dtype=int64)
 In [95]: # features
          rf.n_features_
 Out[95]: 11
 In [96]: # show all parameter used to create base estimator
          rf.get_params()
 Out[96]: {'bootstrap': True,
            'ccp_alpha': 0.0,
            'class_weight': None,
            'criterion': 'gini',
            'max depth': None,
            'max_features': 'auto',
            'max_leaf_nodes': None,
            'max_samples': None,
            'min_impurity_decrease': 0.0,
            'min_impurity_split': None,
            'min samples leaf': 1,
            'min_samples_split': 2,
            'min_weight_fraction_leaf': 0.0,
            'n_estimators': 100,
            'n jobs': None,
            'oob_score': False,
            'random state': None,
            'verbose': 0,
            'warm_start': False}
In [122]: # accuracy
          rf.score(x_test,y_test)
Out[122]: 0.6875
```

Out[101]: 0.5

Out[102]: 0.71875

```
In [123]: # prediction on x test
          rf.predict(x test)
Out[123]: array([6, 5, 5, 5, 6, 6, 5, 6, 5, 5, 6, 5, 5, 5, 5, 5, 5, 6, 5, 6, 5, 7,
                 5, 5, 6, 5, 6, 5, 6, 5, 5, 5, 5, 6, 5, 7, 6, 5, 6, 6, 5, 5, 7, 5,
                 7, 5, 6, 6, 6, 6, 6, 5, 5, 5, 6, 5, 5, 5, 5, 5, 5, 5, 6, 5, 5,
                 6, 5, 6, 6, 5, 6, 6, 6, 5, 7, 5, 6, 6, 6, 5, 7, 5, 6, 6, 6, 5, 5,
                 5, 5, 5, 6, 5, 6, 6, 6, 6, 5, 7, 6, 5, 5, 7, 4, 6, 6, 5, 5, 6,
                 5, 5, 5, 5, 5, 6, 6, 5, 5, 6, 5, 5, 6, 5, 5, 6, 6, 6, 5, 7, 6, 5,
                 5, 5, 5, 6, 6, 6, 5, 7, 5, 6, 6, 5, 6, 6, 6, 5, 5, 6, 5, 5, 6,
                 6, 6, 5, 5, 5, 7, 6, 6, 6, 5, 6, 6, 5, 5, 5, 6, 7, 6, 5, 6, 7,
                 6, 6, 5, 5, 7, 7, 6, 6, 5, 5, 6, 6, 6, 5, 6, 5, 6, 6, 6, 7, 5, 6,
                 7, 6, 7, 6, 6, 5, 6, 5, 5, 5, 6, 6, 6, 5, 6, 6, 5, 5, 6, 6, 6,
                 6, 5, 7, 5, 6, 6, 5, 7, 6, 6, 5, 6, 6, 7, 5, 5, 7, 5, 6, 5, 5, 7,
                 7, 5, 5, 6, 5, 5, 5, 5, 6, 6, 5, 5, 6, 6, 5, 5, 7, 6, 5, 5,
                 6, 5, 5, 7, 6, 5, 6, 7, 7, 6, 5, 7, 6, 6, 6, 5, 5, 6, 6, 5, 5, 5,
                 5, 6, 5, 7, 5, 6, 6, 7, 7, 7, 6, 6, 6, 5, 5, 5, 5, 5, 5, 5, 5,
                 6, 6, 7, 6, 6, 5, 5, 6, 6, 5, 5, 6], dtype=int64)
In [124]: | ## prediction probabilites of each class for each data row in the x_test
          rf.predict proba(x test)
Out[124]: array([[0. , 0.01, 0.35, 0.42, 0.2 , 0.02],
                 [0. , 0.03, 0.91, 0.06, 0. , 0.
                 [0. , 0.03, 0.9 , 0.07, 0. , 0.
                 . . . ,
                 [0. , 0.03, 0.72, 0.24, 0.01, 0.
                 [0.12, 0.13, 0.69, 0.06, 0. , 0.
                 [0.01, 0.01, 0.24, 0.58, 0.15, 0.01]])
          comparision between above created models
In [100]: # accuracy of bagging with 100 decision trees
          bag_dt.score(x_test,y_test)
Out[100]: 0.69375
In [101]: # accuracy of bagging with 100 knn
          bag_knn.score(x_test,y_test)
```

```
Observation - Random forest model gives better accuracy
```

In [102]: # accuracy of random forest where 100 decision trees is used

rf.score(x\_test,y\_test)

## **EXTRA EXPERIMENTS WITH RANDOM FOREST**

```
In [105]: # shows all decision trees classifiers
rf.estimators_
```

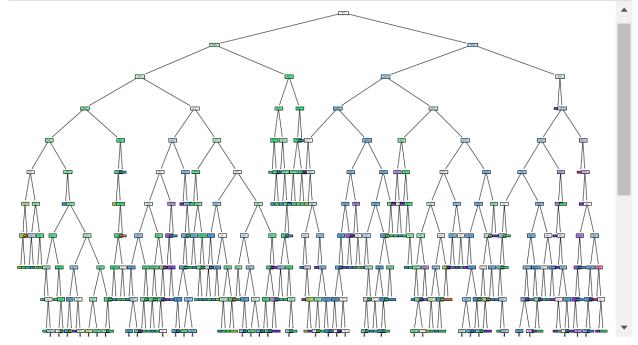
```
Out[105]: [DecisionTreeClassifier(max features='auto', random state=312740373),
           DecisionTreeClassifier(max_features='auto', random_state=501395558),
           DecisionTreeClassifier(max_features='auto', random_state=557653086),
           DecisionTreeClassifier(max_features='auto', random_state=540797481),
           DecisionTreeClassifier(max_features='auto', random_state=1511964624),
           DecisionTreeClassifier(max features='auto', random state=1504556210),
           DecisionTreeClassifier(max_features='auto', random_state=265342615),
           DecisionTreeClassifier(max_features='auto', random_state=2107936750),
           DecisionTreeClassifier(max_features='auto', random_state=912578988),
           DecisionTreeClassifier(max_features='auto', random_state=1667734836),
           DecisionTreeClassifier(max_features='auto', random_state=529365236),
           DecisionTreeClassifier(max_features='auto', random_state=1627003457),
           DecisionTreeClassifier(max features='auto', random state=542956394),
           DecisionTreeClassifier(max_features='auto', random_state=301313637),
           DecisionTreeClassifier(max_features='auto', random_state=1422842175),
           DecisionTreeClassifier(max_features='auto', random_state=553057207),
           DecisionTreeClassifier(max_features='auto', random_state=2076940166),
           DecisionTreeClassifier(max features='auto', random state=2071920328),
           DecisionTreeClassifier(max_features='auto', random_state=727855028),
           DecisionTreeClassifier(max_features='auto', random_state=1787214414),
           DecisionTreeClassifier(max_features='auto', random_state=661811813),
           DecisionTreeClassifier(max_features='auto', random_state=2081664800),
           DecisionTreeClassifier(max_features='auto', random_state=443620103),
           DecisionTreeClassifier(max_features='auto', random_state=663835942),
           DecisionTreeClassifier(max features='auto', random state=1627590358),
           DecisionTreeClassifier(max_features='auto', random_state=905918762),
           DecisionTreeClassifier(max_features='auto', random_state=2029246914),
           DecisionTreeClassifier(max_features='auto', random_state=476728404),
           DecisionTreeClassifier(max_features='auto', random_state=666656415),
           DecisionTreeClassifier(max features='auto', random state=873514685),
           DecisionTreeClassifier(max_features='auto', random_state=1540258354),
           DecisionTreeClassifier(max_features='auto', random_state=1144086889),
           DecisionTreeClassifier(max_features='auto', random_state=411263395),
           DecisionTreeClassifier(max_features='auto', random_state=1147673243),
           DecisionTreeClassifier(max features='auto', random state=1735133656),
           DecisionTreeClassifier(max_features='auto', random_state=1042472732),
           DecisionTreeClassifier(max features='auto', random state=159797054),
           DecisionTreeClassifier(max_features='auto', random_state=2101646931),
           DecisionTreeClassifier(max_features='auto', random_state=1978037760),
           DecisionTreeClassifier(max features='auto', random state=527641289),
           DecisionTreeClassifier(max_features='auto', random_state=47556576),
           DecisionTreeClassifier(max features='auto', random state=1043034224),
           DecisionTreeClassifier(max_features='auto', random_state=740185955),
           DecisionTreeClassifier(max_features='auto', random_state=1689296212),
           DecisionTreeClassifier(max_features='auto', random_state=1916029355),
           DecisionTreeClassifier(max_features='auto', random_state=2056178505),
           DecisionTreeClassifier(max features='auto', random state=528428343),
           DecisionTreeClassifier(max_features='auto', random_state=768326405),
           DecisionTreeClassifier(max_features='auto', random_state=2072593606),
           DecisionTreeClassifier(max_features='auto', random_state=623639442),
           DecisionTreeClassifier(max_features='auto', random_state=869824446),
           DecisionTreeClassifier(max_features='auto', random_state=1478189093),
           DecisionTreeClassifier(max features='auto', random state=1040143818),
```

```
DecisionTreeClassifier(max features='auto', random state=1491541750),
DecisionTreeClassifier(max_features='auto', random_state=304217153),
DecisionTreeClassifier(max_features='auto', random_state=79878926),
DecisionTreeClassifier(max_features='auto', random_state=333806279),
DecisionTreeClassifier(max features='auto', random state=78271061),
DecisionTreeClassifier(max_features='auto', random_state=2001447860),
DecisionTreeClassifier(max_features='auto', random_state=615288930),
DecisionTreeClassifier(max_features='auto', random_state=811092897),
DecisionTreeClassifier(max_features='auto', random_state=402753445),
DecisionTreeClassifier(max features='auto', random state=803241188),
DecisionTreeClassifier(max_features='auto', random_state=1991803271),
DecisionTreeClassifier(max_features='auto', random_state=1589823085),
DecisionTreeClassifier(max_features='auto', random_state=266782430),
DecisionTreeClassifier(max_features='auto', random_state=1923976267),
DecisionTreeClassifier(max_features='auto', random_state=1013439913),
DecisionTreeClassifier(max_features='auto', random_state=1091767324),
DecisionTreeClassifier(max features='auto', random state=1046333134),
DecisionTreeClassifier(max_features='auto', random_state=991988346),
DecisionTreeClassifier(max_features='auto', random_state=1216253569),
DecisionTreeClassifier(max_features='auto', random_state=1073359155),
DecisionTreeClassifier(max_features='auto', random_state=1238070029),
DecisionTreeClassifier(max features='auto', random state=232697891),
DecisionTreeClassifier(max_features='auto', random_state=416565981),
DecisionTreeClassifier(max_features='auto', random_state=1340735253),
DecisionTreeClassifier(max_features='auto', random_state=584043586),
DecisionTreeClassifier(max_features='auto', random_state=1396916814),
DecisionTreeClassifier(max_features='auto', random_state=1461070228),
DecisionTreeClassifier(max_features='auto', random_state=1848863425),
DecisionTreeClassifier(max features='auto', random state=510803409),
DecisionTreeClassifier(max_features='auto', random_state=1541029432),
DecisionTreeClassifier(max_features='auto', random_state=1035054934),
DecisionTreeClassifier(max_features='auto', random_state=2119037477),
DecisionTreeClassifier(max_features='auto', random_state=1186688076),
DecisionTreeClassifier(max_features='auto', random_state=578380015),
DecisionTreeClassifier(max_features='auto', random_state=102894615),
DecisionTreeClassifier(max_features='auto', random_state=2025988657),
DecisionTreeClassifier(max_features='auto', random_state=251211642),
DecisionTreeClassifier(max_features='auto', random_state=877618950),
DecisionTreeClassifier(max_features='auto', random_state=2043115325),
DecisionTreeClassifier(max_features='auto', random_state=1519876068),
DecisionTreeClassifier(max features='auto', random state=1561189135),
DecisionTreeClassifier(max_features='auto', random_state=336195706),
DecisionTreeClassifier(max_features='auto', random_state=1820559425),
DecisionTreeClassifier(max_features='auto', random_state=1445046562),
DecisionTreeClassifier(max_features='auto', random_state=948796864),
DecisionTreeClassifier(max features='auto', random state=1271323882),
DecisionTreeClassifier(max_features='auto', random_state=160613891)]
```

```
In [106]: rf.estimators_[0]
```

Out[106]: DecisionTreeClassifier(max\_features='auto', random\_state=312740373)

```
In [108]: #Plotting tree structure of one and first decision tree
          plt.figure(figsize=(20,20))
          tp = tree.plot_tree(rf.estimators_[0],filled=True)
```



### **GRID SEARCH**

```
In [109]:
          # setting some parameter which we want to tune for randome forest
          grid_param = {
              "n_estimators":[100,150,200],
              "criterion" : ["gini", "random"],
              "max_depth" : range(5,10),
              "min_samples_leaf" : range(5)
```

```
In [110]: rf = RandomForestClassifier()
```

```
In [111]: # gridsearchcv for hyperparameter tuning
          grid_search_rf = GridSearchCV(estimator=rf,param_grid=grid_param,cv=10,n_jobs=6,v
```

```
In [112]: # training
          grid search rf.fit(X train, Y train)
          Fitting 10 folds for each of 150 candidates, totalling 1500 fits
          C:\Users\aniyant\anaconda3\lib\site-packages\sklearn\model selection\ split.py:
          670: UserWarning: The least populated class in y has only 9 members, which is 1
          ess than n splits=10.
            warnings.warn(("The least populated class in y has only %d"
          [Parallel(n_jobs=6)]: Using backend LokyBackend with 6 concurrent workers.
          [Parallel(n jobs=6)]: Done 38 tasks
                                                     | elapsed:
                                                                  15.7s
          [Parallel(n_jobs=6)]: Done 188 tasks
                                                       elapsed:
                                                                  37.1s
          [Parallel(n jobs=6)]: Done 438 tasks
                                                     | elapsed: 1.3min
          [Parallel(n jobs=6)]: Done 788 tasks
                                                     | elapsed: 2.2min
                                                     | elapsed: 2.5min
          [Parallel(n jobs=6)]: Done 1238 tasks
          [Parallel(n_jobs=6)]: Done 1500 out of 1500 | elapsed: 2.7min finished
Out[112]: GridSearchCV(cv=10, estimator=RandomForestClassifier(), n_jobs=6,
                       param grid={'criterion': ['gini', 'random'],
                                    'max depth': range(5, 10),
                                    'min_samples_leaf': range(0, 5),
                                    'n estimators': [100, 150, 200]},
                       verbose=1)
In [113]: # best parameter we got from grid search
          grid search rf.best params
Out[113]: {'criterion': 'gini',
            'max depth': 9,
           'min samples leaf': 1,
            'n estimators': 200}
In [114]: # new random forest model
          rf new = RandomForestClassifier(n estimators=200,criterion='gini',max depth=9,min
In [116]: rf new.fit(X train, Y train)
Out[116]: RandomForestClassifier(max_depth=9, n_estimators=200)
In [118]: # accuracy of this model
          rf_new.score(x_test,y_test)
Out[118]: 0.696875
In [125]: # accuracy of the previous model
          rf.score(x_test,y_test)
Out[125]: 0.6875
```

# GridSearchCV may or may not be able to increase the accuracy of the model, it's just a matter of experiment

```
In [ ]:
```