Experiment:-7

Objective: Decision Tree

```
from sklearn.datasets import load_breast_cancer
dataset = load_breast_cancer()
dataset
    {'data': array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
            1.189e-01].
           [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
            8.902e-02],
           [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
            8.758e-02],
           [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
            7.820e-02],
           [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
            1.240e-01],
           [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
            7.039e-02]]),
     0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
           1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0,
           1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
           1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
           0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
           1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
           1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0,
           0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
           1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
           1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
           0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 1,\ 0,\ 1,\ 0,\ 1,\ 1,\ 0,\ 1,\ 1,\ 0,\ 1,\ 1,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0,
           0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0,
           0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,
           1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1,
           1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
           1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
           1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
           1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1]),
     'frame': None,
     'target_names': array(['malignant', 'benign'], dtype='<U9'),
     'DESCR': '.. breast cancer dataset:\n\nBreast cancer wisconsin (diagnostic) dataset\n------
    \n\n**Data Set Characteristics:**\n\n :Number of Instances: 569\n\n :Number of Attributes: 30 numeric, predictive attributes
    and the class\n\n :Attribute Information:\n - radius (mean of distances from center to points on the perimeter)\n
    texture (standard deviation of gray-scale values)\n
                                                      - perimeter\n - area\n
                                                                                      - smoothness (local variation in
                         - compactness (perimeter^2 / area - 1.0)\n
    radius lengths)\n
                                                                     - concavity (severity of concave portions of the

    concave points (number of concave portions of the contour)\n

    contour)\n
                                                                                - symmetry\n- fractal dimension
    ("coastline approximation" - 1)\n\n The mean, standard error, and "worst" or largest (mean of the three\n
    values) of these features were computed for each image,\n resulting in 30 features. For instance, field 0 is Mean Radius,
            10 is Radius SE, field 20 is Worst Radius.\n\n
                                                             - class:∖n
                                                                                       - WDBC-Malignant∖n
    field\n
    WDBC-Benign\n\n :Summary Statistics:\n\n =======\n
                 =======\n radius (mean):
                                                                                                      6.981 28.11\n
                                     9.71 39.28\n perimeter (mean):
    texture (mean):
                                                                                       43.79 188.5\n
                                                                                                     area (mean):
    143.5 2501.0\n
                     smoothness (mean):
                                                      0.053 0.163\n
                                                                      compactness (mean):
                                                                                                        0.019 0.345\n
    concavity (mean):
                                     0.0
                                            0.427\n
                                                     concave points (mean):
                                                                                       0.0
                                                                                             0.201\n
                                                                                                       symmetry (mean):
    0.106 0.304\n fractal dimension (mean):
                                                     0.05 0.097\n radius (standard error):
                                                                                                       0.112 2.873\n
                                     0.36 4.885\n
                                                     perimeter (standard error):
                                                                                      0.757 21.98\n
    texture (standard error):
                                                                                                       area (standard
                        6.802 542.2\n smoothness (standard error):
                                                                         0.002 0.031\n compactness (standard error):
    error):
import pandas as pd
import numpy as np
data1 = pd.DataFrame(data= np.c_[dataset['data']],columns= dataset['feature_names'])
```

mea

data1

```
mean
                   mean
                             mean
                                    mean
                                               mean
                                                           mean
                                                                    mean
                                                                          concav
                texture perimeter
                                                    compactness
          radius
                                    area
                                         smoothness
                                                               concavity
                                                                           point
      0
           17.99
                   10.38
                            122.80 1001.0
                                             0.11840
                                                        0.27760
                                                                  0.30010
                                                                          0.1471
      1
           20.57
                   17 77
                            132.90 1326.0
                                             0.08474
                                                        0.07864
                                                                  0.08690
                                                                          0.0701
      2
           19.69
                            130.00 1203.0
                                             0.10960
                                                        0.15990
                                                                  0.19740 0.1279
                   21.25
      3
           11 42
                   20.38
                             77 58
                                   386.1
                                             0.14250
                                                        0.28390
                                                                  0.24140 0.1052
      4
           20.29
                   14.34
                            135.10 1297.0
                                             0.10030
                                                        0.13280
                                                                  0.19800 0.1043
     564
           21.56
                   22.39
                            142.00 1479.0
                                             0.11100
                                                        0.11590
                                                                  0.24390
                                                                         0.1389
           20.13
                   28.25
                            131.20 1261.0
                                             0.09780
                                                                  0.14400
                                                                         0.0979
     565
                                                        0.10340
     566
           16.60
                   28.08
                            108.30
                                   858.1
                                             0.08455
                                                        0.10230
                                                                  0.09251
                                                                         0.0530
     567
           20.60
                   29.33
                            140.10 1265.0
                                             0.11780
                                                        0.27700
                                                                  0.35140
                                                                         0.1520
     568
           7.76
                   24 54
                             47.92
                                   181.0
                                             0.05263
                                                        0.04362
                                                                  0.00000 0.0000
    569 rows × 30 columns
    4
y = dataset.target
    0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
          1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1,
                                                      0, 0,
                                                            1,
           1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
           1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
           1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
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          0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
          1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
             1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
           0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1,
          1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1,
             1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0,
           0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0,
          0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0,
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          1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
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          1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1])
x = data1
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3)
print(x_train.shape,y_train.shape,y_test.shape,x_test.shape)
    (398, 30) (398,) (171,) (171, 30)
from sklearn.tree import DecisionTreeClassifier
clf = DecisionTreeClassifier()
clf.fit(x_train,y_train)
     ▼ DecisionTreeClassifier
     DecisionTreeClassifier()
from sklearn import tree
from matplotlib import pyplot as plt
feature names = x.columns
```

```
fig = plt.figure(figsize = (15, 10))
_ = tree.plot_tree(clf, feature_names = feature_names, class_names={0:'maligant', 1:'benign'}, filled = True, fontsize = 12)
```

