Experiment 2.4

Decision Trees and Random Forests

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Subject Name: Machine Learning Lab Subject Code: 20CSP-317

1. Aim: Decision Trees and Random Forests — Explained with Python Implementation.

- **2. Objective:** To prepare a model with Decision Trees and Random Forests algorithm.
- 3. Data Set Chosen: Breast Cancer Wisconsin (Diagnostic) Data Set
- 4. Result and output:

```
In [1]: import numpy as np
         import pandas as pd
         import seaborn as sns
In [2]: df = pd.read_csv('Breast_Cancer.csv')
In [3]: df.head()
Out[3]:
                                                                                                                              concave
                                                                                                                                      symmetry_mean
                  id radius_mean texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mean
         0 842302
                                                                                                                                               0.2419
                            17 99
                                         10.38
                                                       122 80
                                                                 1001.0
                                                                                  0.11840
                                                                                                   0.27760
                                                                                                                   0.3001
                                                                                                                               0.14710
             842517
                            20.57
                                         17.77
                                                       132.90
                                                                  1326.0
                                                                                  0.08474
                                                                                                   0.07864
                                                                                                                    0.0869
                                                                                                                               0.07017
                                                                                                                                               0.1812
         2 84300903
                           19.69
                                        21.25
                                                       130.00
                                                                 1203.0
                                                                                 0.10960
                                                                                                   0.15990
                                                                                                                   0.1974
                                                                                                                               0.12790
                                                                                                                                               0.2069
          3 84348301
                            11.42
                                         20.38
                                                       77.58
                                                                  386.1
                                                                                  0.14250
                                                                                                   0.28390
                                                                                                                    0.2414
                                                                                                                               0.10520
                                                                                                                                               0.2597
          4 84358402
                         20.29
                                         14.34
                                                       135.10
                                                                 1297.0
                                                                                  0.10030
                                                                                                    0.13280
                                                                                                                    0.1980
                                                                                                                               0.10430
                                                                                                                                               0.1809
         5 rows × 32 columns
In [4]: df.set_index(['id'], inplace = True)
In [6]: df['diagnosis'] = df['diagnosis'].map({'M':1, 'B':0})
```

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```
In [7]: df.apply(lambda x: x.isnull().sum())
 Out[7]: radius mean
                                     0
         texture mean
                                     0
         perimeter_mean
                                     0
         area mean
         smoothness mean
         compactness_mean
         concavity_mean
                                     0
         concave points mean
                                     0
         symmetry mean
                                     0
         fractal_dimension_mean
         radius se
                                     0
                                     0
         texture se
         perimeter se
                                     0
         area se
                                     0
         smoothness_se
                                     0
         compactness_se
         concavity se
                                     0
         concave points_se
         symmetry_se
                                     0
         fractal dimension se
         radius worst
                                     0
         texture_worst
                                     0
         perimeter_worst
         area worst
                                     0
         smoothness worst
         compactness worst
                                     0
         concavity_worst
                                     0
         concave points_worst
         symmetry_worst
         fractal dimension worst
         diagnosis
         dtype: int64
In [8]: df.diagnosis.unique()
Out[8]: array([1, 0], dtype=int64)
In [13]: feature_space = df.iloc[:, df.columns != 'diagnosis']
         feature_class = df.iloc[:, df.columns == 'diagnosis']
In [14]: from sklearn.model selection import train test split
```

```
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```

```
In [17]: from sklearn.ensemble import RandomForestClassifier
        Classifier = RandomForestClassifier(random state = 50)
        Classifier.fit(training set, class set)
Out[17]: RandomForestClassifier(random state=50)
Out[17]: RandomForestClassifier(random_state=50)
In [18]: predict=Classifier.predict(test set)
In [19]: predict
0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,
              0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0,
               0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0,
               0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1,
              0, 1, 1, 0], dtype=int64)
In [20]: from sklearn.metrics import accuracy score
        accuracy score(test class set,predict)
Out[20]: 0.956140350877193
```

Result: Accuracy of the model is approximately 95%.