

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]:
                   id radius_mean texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mean
                                                                                                                             points_mean
                                                                                    0.11840
                                                                                                      0.27760
                                                                                                                                                  0.2419
              842517
                                          17.77
                                                                                    0.08474
                                                                                                      0.07864
                                                                                                                                  0.07017
                                                                                                                                                  0 1812
                             20.57
                                                        132 90
                                                                   1326 0
                                                                                                                      0.0869
          2 84300903
                            19.69
                                          21.25
                                                                   1203.0
                                                                                                      0.15990
                                                                                                                      0.1974
                                                                                                                                  0.12790
                                                        130.00
                                                                                    0.10960
                                                                                                                                                  0.2069
          3 84348301
                                          20.38
                                                                    386.1
                                                                                                                      0.2414
                                                                                                                                  0.10520
                             11.42
                                                         77.58
                                                                                    0.14250
                                                                                                      0.28390
                                                                                                                                                  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})
```

```
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
```

```
In [7]: df.apply(lambda x: x.isnull().sum())
Out[7]: radius mean
                                    0
        texture mean
                                    0
        perimeter mean
                                    0
        area mean
                                    0
        smoothness mean
        compactness mean
                                   0
        concavity mean
                                    0
        concave points mean
        symmetry mean
        fractal dimension mean
        radius se
                                    0
        texture se
                                    0
        perimeter se
                                    0
        area se
                                    0
        smoothness se
        compactness se
                                    0
        concavity se
        concave points se
        symmetry se
                                    0
        fractal dimension se
        radius worst
                                    0
        texture worst
        perimeter worst
                                    0
        area worst
        smoothness worst
                                   0
        compactness worst
                                    0
        concavity worst
        concave points worst
        symmetry worst
        fractal dimension worst
                                    0
        diagnosis
        dtype: int64
```

```
In [17]: from sklearn.ensemble import RandomForestClassifier
    Classifier = RandomForestClassifier(random_state = 50)
    Classifier.fit(training_set,class_set)
```

Out[17]: RandomForestClassifier(random_state=50)

Result: Accuracy of the model is approximately 95%.