ASSIGNMENT 6

Data Analytics III

- 1. Implement Simple Naïve Bayes classification algorithm using Python/R on iris.csv dataset.
- 2. Compute Confusion matrix to find TP, FP, TN, FN, Accuracy, Error rate, Precision, Recall on the given dataset.

```
In [1]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.datasets import load_iris
        from sklearn.preprocessing import StandardScaler
         from sklearn.model_selection import train_test_split
        from sklearn.naive_bayes import GaussianNB
        from mlxtend.plotting import plot confusion matrix
        from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
         import warnings
        warnings.filterwarnings("ignore")
        %matplotlib inline
In [2]: iris = load_iris()
        iris.keys()
Out[2]: dict_keys(['data', 'target', 'frame', 'target_names', 'DESCR', 'feature_names', 'f
         ilename', 'data_module'])
In [3]: x = pd.DataFrame(iris['data'], columns=iris['feature_names'])
        y = pd.DataFrame(iris['target'], columns=['target'])
In [4]: x.head()
Out[4]:
           sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
         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
In [5]: x.tail()
```

```
Out[5]:
              sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
         145
                           6.7
                                           3.0
                                                            5.2
                                                                           2.3
         146
                           6.3
                                           2.5
                                                            5.0
                                                                           1.9
                                                            5.2
         147
                           6.5
                                           3.0
                                                                           2.0
         148
                           6.2
                                           3.4
                                                            5.4
                                                                           2.3
         149
                           5.9
                                           3.0
                                                            5.1
                                                                           1.8
In [6]: x.sample(5)
Out[6]:
              sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
           2
                                           3.2
                                                           1.3
                                                                           0.2
                           4.7
          32
                           5.2
                                           4.1
                                                           1.5
                                                                           0.1
         128
                           6.4
                                           2.8
                                                            5.6
                                                                           2.1
          10
                           5.4
                                           3.7
                                                            1.5
                                                                           0.2
           7
                           5.0
                                           3.4
                                                            1.5
                                                                           0.2
In [7]: x.shape, y.shape
Out[7]: ((150, 4), (150, 1))
In [8]: x.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 4 columns):
         # Column
                                Non-Null Count Dtype
        --- -----
                                -----
         0 sepal length (cm) 150 non-null
                                               float64
         1 sepal width (cm) 150 non-null float64
             petal length (cm) 150 non-null
                                               float64
            petal width (cm) 150 non-null
                                               float64
        dtypes: float64(4)
        memory usage: 4.8 KB
In [9]: y.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 1 columns):
         # Column Non-Null Count Dtype
         0 target 150 non-null
                                     int64
        dtypes: int64(1)
        memory usage: 1.3 KB
In [10]: x.describe()
```

| Out[10]: | | sepal length (cm) | sepal width (cm) | petal length (cm) | petal width (cm) |
|----------|-------|-------------------|------------------|-------------------|------------------|
| | count | 150.000000 | 150.000000 | 150.000000 | 150.000000 |
| | mean | 5.843333 | 3.057333 | 3.758000 | 1.199333 |
| | std | 0.828066 | 0.435866 | 1.765298 | 0.762238 |
| | min | 4.300000 | 2.000000 | 1.000000 | 0.100000 |
| | 25% | 5.100000 | 2.800000 | 1.600000 | 0.300000 |
| | 50% | 5.800000 | 3.000000 | 4.350000 | 1.300000 |
| | 75% | 6.400000 | 3.300000 | 5.100000 | 1.800000 |
| | max | 7.900000 | 4.400000 | 6.900000 | 2.500000 |

Data preparation

```
In [11]: scaler = StandardScaler()
    x = scaler.fit_transform(x.values)

In [12]: x_train, x_test, y_train, y_test = train_test_split(x, y.values, test_size=0.3, ran

In [13]: x_train.shape, x_test.shape, y_train.shape, y_test.shape

Out[13]: ((105, 4), (45, 4), (105, 1), (45, 1))
```

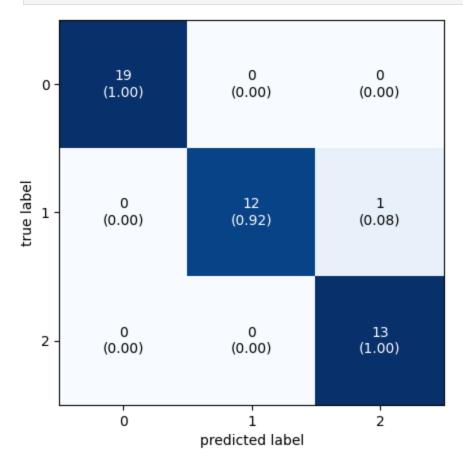
Model building

Confusion Matrix

```
In [17]: cm = confusion_matrix(y_test, y_pred)
    print(cm)
```

```
[[19 0 0]
[ 0 12 1]
[ 0 0 13]]
```

```
In [18]: plot_confusion_matrix(conf_mat=cm, figsize=(5,5), show_normed=True)
   plt.show()
```



```
In [19]: print(f"TP value is {cm[0,0]}")
         print(f"TN value is {cm[1,1] + cm[2,2]}")
         print(f"FP value is {cm[0,1] + cm[0,2]}")
         print(f"FN value is {cm[1,0] + cm[2,0]}")
        TP value is 19
        TN value is 25
        FP value is 0
        FN value is 0
In [20]: tp = 19
         tn = 25
         fp = 0
         fn = 0
In [21]: print('Accuracy score is :',(tn+tp)/(tn+fp+fn+tp))
        Accuracy score is : 1.0
In [22]: print('Error Rate: ',(fp+fn)/(tp+tn+fn+fp))
        Error Rate: 0.0
```