

Assignment N0. 6

Aim : 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.

Code :

```
In [6]: 1 import pandas as pd
        2 import matplotlib.pyplot as plt
```

```
In [11]: 1 data = pd.read_csv("https://raw.githubusercontent.com/plotly/datasets/master/iris-data.csv")
        2 data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column          Non-Null Count  Dtype  
---  --
 0   sepal length    150 non-null   float64
 1   sepal width     150 non-null   float64
 2   petal length    150 non-null   float64
 3   petal width     150 non-null   float64
 4   class           150 non-null   object  
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
In [12]: 1 data.shape
```

Out[12]: (150, 5)

```
In [13]: 1 data.head()
```

Out[13]:

| | sepal length | sepal width | petal length | petal width | class |
|---|--------------|-------------|--------------|-------------|-------------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |

```
In [14]: 1 data.tail()
```

Out[14]:

| | sepal length | sepal width | petal length | petal width | class |
|-----|--------------|-------------|--------------|-------------|----------------|
| 145 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

```
In [15]: 1 data.describe()
```

```
Out[15]:
```

| | sepal length | sepal width | petal length | petal width |
|-------|--------------|-------------|--------------|-------------|
| count | 150.000000 | 150.000000 | 150.000000 | 150.000000 |
| mean | 5.843333 | 3.054000 | 3.758667 | 1.198667 |
| std | 0.828066 | 0.433594 | 1.764420 | 0.763161 |
| 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 |

```
In [16]: 1 data.isnull().sum()
```

```
Out[16]: sepal length    0
sepal width    0
petal length    0
petal width    0
class          0
dtype: int64
```

```
In [17]: 1 X = data.drop(['class'], axis=1)
2 y = data.drop(['sepal length', 'sepal width', 'petal length', 'petal width'], axis=1)
3 print(X)
4 print(y)
5 print(X.shape)
6 print(y.shape)
```

| | sepal length | sepal width | petal length | petal width |
|-----|--------------|-------------|--------------|-------------|
| 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 |
| .. | ... | ... | ... | ... |
| 145 | 6.7 | 3.0 | 5.2 | 2.3 |
| 146 | 6.3 | 2.5 | 5.0 | 1.9 |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 |

```
[150 rows x 4 columns]
```

| | class |
|-----|----------------|
| 0 | Iris-setosa |
| 1 | Iris-setosa |
| 2 | Iris-setosa |
| 3 | Iris-setosa |
| 4 | Iris-setosa |
| .. | ... |
| 145 | Iris-virginica |
| 146 | Iris-virginica |
| 147 | Iris-virginica |
| 148 | Iris-virginica |
| 149 | Iris-virginica |

```
[150 rows x 1 columns]
```

```
(150, 4)
```

```
(150, 1)
```

```
In [18]: 1 from sklearn.model_selection import train_test_split
2 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, shuffle=True)
3 print(X_train.shape)
4 print(X_test.shape)
5 print(y_train.shape)
6 print(y_test.shape)
```

```
(120, 4)
(30, 4)
(120, 1)
(30, 1)
```

```
In [19]: 1 from sklearn.naive_bayes import GaussianNB
2 model = GaussianNB()
3 model.fit(X_train, y_train)
```

C:\Users\Welcome\anaconda3\Lib\site-packages\sklearn\utils\validation.py:1143: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
y = column_or_1d(y, warn=True)

```
Out[19]: ▾ GaussianNB
GaussianNB()
```

```
In [20]: 1 y_pred = model.predict(X_test)
2 model.score(X_test, y_test)
```

```
Out[20]: 0.9333333333333333
```

```
In [21]: 1 y_pred
```

```
Out[21]: array(['Iris-setosa', 'Iris-virginica', 'Iris-versicolor',
'Iris-virginica', 'Iris-virginica', 'Iris-setosa',
'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa',
'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
'Iris-virginica', 'Iris-virginica', 'Iris-virginica',
'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
'Iris-versicolor', 'Iris-virginica', 'Iris-versicolor',
'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica',
'Iris-virginica', 'Iris-virginica', 'Iris-virginica',
'Iris-virginica', 'Iris-virginica'], dtype='<U15')
```

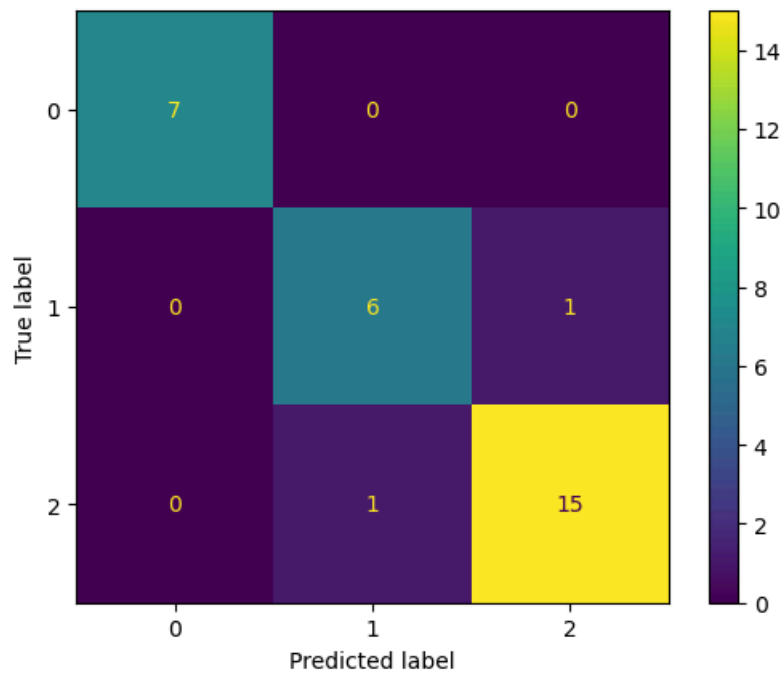
```
In [23]: 1 from sklearn.metrics import accuracy_score, confusion_matrix, ConfusionMatrixDisplay
2 print(accuracy_score(y_test, y_pred))
```

```
0.9333333333333333
```

```
In [24]: 1 cm = confusion_matrix(y_test, y_pred)
2 disp = ConfusionMatrixDisplay(confusion_matrix = cm)
3 print("Confusion matrix:")
4 print(cm)
```

Confusion matrix:
[[7 0 0]
[0 6 1]
[0 1 15]]

```
In [25]: 1 disp.plot()  
2 plt.show()
```



```
In [26]: 1 def get_confusion_matrix_values(y_true, y_pred):  
2     cm = confusion_matrix(y_true, y_pred)  
3     return(cm[0][0], cm[0][1], cm[1][0], cm[1][1])  
4  
5 TP, FP, FN, TN = get_confusion_matrix_values(y_test, y_pred)  
6 print("TP: ", TP)  
7 print("FP: ", FP)  
8 print("FN: ", FN)  
9 print("TN: ", TN)
```

```
TP: 7  
FP: 0  
FN: 0  
TN: 6
```

```
In [27]: 1 print("The Accuracy is ", (TP+TN)/(TP+TN+FP+FN))  
2 print("The precision is ", TP/(TP+FP))  
3 print("The recall is ", TP/(TP+FN))
```

```
The Accuracy is 1.0  
The precision is 1.0  
The recall is 1.0
```

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