Name :sandip gadadare. Div:A
Roll no: CO3A09. BATCH:B1

Assignment No: 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.

Source Code:-

```
import numpy as np

import pandas as pd

from sklearn.model_selection import train_test_split

x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,shuffle=True)

from sklearn.naive_bayes import GaussianNB

from sklearn.metrics import accuracy_score, confusion_matrix,ConfusionMatrixDisplay

data=pd.read_csv('C:\\Users\\sspm\\Downloads\\iris.csv')

data
```

OUTPUT:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
...	...	...	...	...	...
145	6.7	3.0	5.2	2.3	virginica
146	6.3	2.5	5.0	1.9	virginica
147	6.5	3.0	5.2	2.0	virginica

148 6.2 3.4 5.4 2.3 virginica

149 5.9 3.0 5.1 1.8 virginica

150 rows × 5 columns

CODE:

data.isnull()

OUTPUT:

	sepal_length	sepal_width	petal_length	petal_width	species
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
...	...	...	...	...	...
145	False	False	False	False	False
146	False	False	False	False	False
147	False	False	False	False	False
148	False	False	False	False	False
149	False	False	False	False	False

150 rows × 5 columns

CODE:

data.isnull().sum()

OUTPUT:

sepal_length 0

sepal_width 0

petal_length 0

petal_width 0

species 0

dtype: int64

CODE:

```
data.columns
```

OUTPUT:

```
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width', 'species'],dtype='object')
```

CODE:

```
x=data.drop(["species"],axis=1)
y=data.drop(["sepal_length","sepal_width","petal_length","petal_width"],axis=1)
print(x)
```

OUTPUT:

```
    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]
```

CODE:

```
print(y)
```

OUTPUT:

```
      species
0      setosa
1      setosa
2      setosa
```

3 setosa

4 setosa

.. ...

145 virginica

146 virginica

147 virginica

148 virginica

149 virginica

[150 rows x 1 columns]

CODE:

```
print(x.shape)

print(y.shape)
```

OUTPUT:

(150, 4)

(150, 1)

CODE:

```
print(x_train.shape)

print(x_test.shape)
```

OUTPUT:

(120, 4)

(30, 4)

CODE:

```
print(y_train.shape)

print(y_test.shape)
```

OUTPUT:

(120, 1)

(30, 1)

CODE:

```
model=GaussianNB()

model.fit(x_train,y_train)
```

```
y = column_or_1d(y, warn=True)
```

OUTPUT:

```
GaussianNB()
```

CODE:

```
y_pred = model.predict(x_test)
model.score(x_test,y_test)*100
```

OUTPUT:

```
93.33333333333333
```

CODE:

```
print(accuracy_score(y_test, y_pred)*100)
```

OUTPUT:

```
33.33333333333333
```

CODE:

```
cm = confusion_matrix(y_test, y_pred)
disp = ConfusionMatrixDisplay(confusion_matrix = cm)
print("Confusion matrix:")
print(cm)
```

OUTPUT:

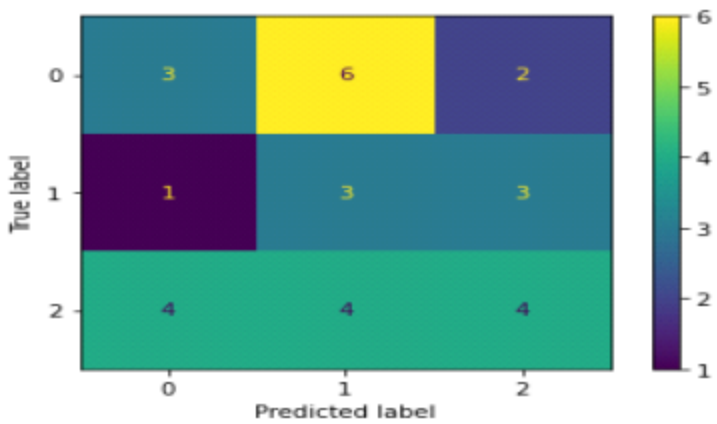
```
Confusion matrix:
[[3 6 2]
 [1 3 3]
 [4 4 4]]
```

CODE:

```
disp.plot()
```

OUTPUT:

```
<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x10aafbcb7f0>
```



CODE:

```
def get_confusion_matrix_values(y_true, y_pred):
    cm = confusion_matrix(y_true, y_pred)
    return(cm[0][0], cm[0][1], cm[1][0], cm[1][1])
TP, FP, FN, TN = get_confusion_matrix_values(y_test, y_pred)
print("TP: ", TP)
print("FP: ", FP)
print("FN: ", FN)
print("TN: ", TN)
```

OUTPUT:

TP: 3

FP: 6

FN: 1

TN: 3

CODE:

```
print("The Accuracy is ", (TP+TN)/(TP+TN+FP+FN))
print("The precision is ", TP/(TP+FP))
print("The recall is ", TP/(TP+FN))
```

OUTPUT:

The Accuracy is 0.46153846153846156
The precision is 0.3333333333333333
The recall is 0.75