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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.

```
import pandas as pd
```


```
df=pd.read_csv('/content/Iris.csv')
```

```
df.head()
```




	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

```
df.tail()
```




	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
145	146	6.7	3.0	5.2	2.3	Iris-virginica
146	147	6.3	2.5	5.0	1.9	Iris-virginica
147	148	6.5	3.0	5.2	2.0	Iris-virginica
148	149	6.2	3.4	5.4	2.3	Iris-virginica
149	150	5.9	3.0	5.1	1.8	Iris-virginica

```
df.shape
```


 (150, 6)

```
df.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Id               150 non-null   int64
1   SepalLengthCm    150 non-null   float64
2   SepalWidthCm     150 non-null   float64
3   PetalLengthCm    150 non-null   float64
4   PetalWidthCm     150 non-null   float64
5   Species          150 non-null   object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB
```

```
df.describe()
```



	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	150.000000	150.000000	150.000000	150.000000	150.000000
mean	75.500000	5.843333	3.054000	3.758667	1.198667
std	43.445368	0.828066	0.433594	1.764420	0.763161
min	1.000000	4.300000	2.000000	1.000000	0.100000
25%	38.250000	5.100000	2.800000	1.600000	0.300000
50%	75.500000	5.800000	3.000000	4.350000	1.300000
75%	112.750000	6.400000	3.300000	5.100000	1.800000
max	150.000000	7.900000	4.400000	6.900000	2.500000

```
x=df.drop(['Id', 'Species'], axis=1)
```

```
x.head()
```



	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
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



```
x.shape
```



(150, 4)

```
from sklearn.preprocessing import LabelEncoder
label=LabelEncoder()
df['Species']=label.fit_transform(df['Species'])
```

```
y=df['Species']
y.head()
```



	Species
0	0
1	0
2	0
3	0
4	0

df.head()



```
print(y.shape)
```



(150,)

```
from sklearn.model_selection import train_test_split
xtrain, xtest, ytrain, ytest=train_test_split(x,y,test_size=0.2, random_state=0)
```

```
xtrain.shape
```



(120, 4)

```
xtest.shape
```



(30, 4)

```
ytrain.shape
```



(120,)

```
ytest.shape
```



(30,)

```
from sklearn.naive_bayes import GaussianNB
model=GaussianNB()
```

```
model.fit(xtrain, ytrain)
```



▼ GaussianNB ⓘ ?

GaussianNB()



```
ypred=model.predict(xtest)
model.score(xtest, ytest)
```



0.9666666666666667

```
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
accuracy_score(ytest, ypred)
```

```
0.9666666666666667
```

```
cm=confusion_matrix(ytest, ypred)
print(cm)
```

```
[[11  0  0]
 [ 0 13  0]
 [ 0  1  5]]
```

Explanation of above 3*3 Confusion Matrix:

Class 0: TP=11, TN=19, FP=0, FN=0

Class 1: TP=13, TN=16, FP=1, FN=0

Class 2: TP=5, TN=24, FP=0, FN=1

For Class 0:

TP: Same actual value and predicted value

TN: The sum of values of all columns and rows except the values of Class 0

FP: The sum of values of the corresponding column except for the TP value

FN: The sum of values of corresponding rows except for the TP value

Similarly for remaining classes

```
print(classification_report(ytest, ypred))
```

```

              precision    recall  f1-score   support

     0       1.00        1.00        1.00         11
     1       0.93        1.00        0.96         13
     2       1.00        0.83        0.91          6

 accuracy          0.97
 macro avg          0.98
 weighted avg       0.97
```

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