

```
In [ ]: from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
from sklearn.naive_bayes import MultinomialNB, GaussianNB
from sklearn.metrics import accuracy_score
import seaborn as sb
import pandas as pd
```

```
In [ ]: data = pd.read_csv("Iris.csv")
```

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

```
Out[ ]:
```

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

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

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Out[ ]:
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	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
<b>0</b>	1	5.1	3.5	1.4	0.2	Iris-setosa
<b>1</b>	2	4.9	3.0	1.4	0.2	Iris-setosa
<b>2</b>	3	4.7	3.2	1.3	0.2	Iris-setosa
<b>3</b>	4	4.6	3.1	1.5	0.2	Iris-setosa
<b>4</b>	5	5.0	3.6	1.4	0.2	Iris-setosa

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

```
Out[ ]: Id          0
SepalLengthCm    0
SepalWidthCm     0
PetalLengthCm    0
PetalWidthCm     0
Species          0
dtype: int64
```

```
In [ ]: data.dtypes
```

```
Out[ ]: Id                int64
SepalLengthCm          float64
SepalWidthCm           float64
PetalLengthCm          float64
PetalWidthCm           float64
Species                object
dtype: object
```

```
In [ ]: # X, y = load_iris(return_X_y=True)
X = data.drop(columns=["Species"])
y = data["Species"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, ra
# X_train, X_test, y_train, y_test = train_test_split(data, test_size=0.3,
nb_classifier = GaussianNB()
# nb_classifier = MultinomialNB()
y_pred = nb_classifier.fit(X_train, y_train).predict(X_test)
print("Number of mislabeled points out of a total %d points : %d" % (X_test
```

Number of mislabeled points out of a total 45 points : 0

```
In [ ]: nb_classifier.score(X_test,y_test)
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```
Out[ ]: 1.0
```

```
In [ ]: accuracy_score(y_test, y_pred)
```

```
Out[ ]: 1.0
```

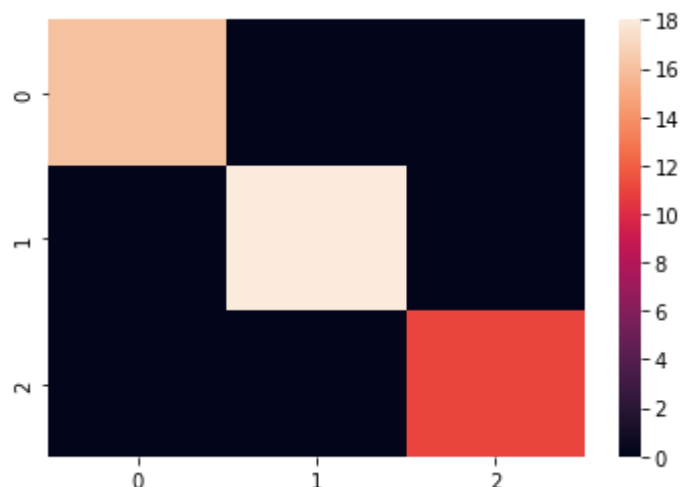
```
In [ ]: nb_classifier.score(X_train,y_train)
```

```
Out[ ]: 0.9904761904761905
```

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

```
In [ ]: sb.heatmap(cm)
```

```
Out[ ]: <AxesSubplot:>
```



```
In [ ]: tp_setosa = cm[0][0]
tp_verginica = cm[1][1]
```

```
tp_versicolor = cm[2][2]

tp = tp_setosa+tp_virginica+tp_versicolor
```

```
In [ ]: tn_setosa = cm[1][1]+cm[1][2]+cm[2][1]+cm[2][2]
tn_virginica = cm[0][0]+cm[0][2]+cm[2][0]+cm[2][2]
tn_versicolor = cm[0][0]+cm[0][1]+cm[1][0]+cm[1][1]

tn = tn_setosa+tn_virginica+tn_versicolor
```

```
In [ ]: fp_setosa = cm[1][0]+cm[2][0]
fp_virginica = cm[0][1]+cm[2][1]
fp_versicolor = cm[0][2]+cm[1][2]

fp = fp_setosa+fp_virginica+fp_versicolor
```

```
In [ ]: fn_setosa = cm[0][1]+cm[0][2]
fn_virginica = cm[1][0]+cm[1][2]
fn_versicolor = cm[2][0]+cm[2][1]

fn = fn_setosa+fn_virginica+fn_versicolor
```

```
In [ ]: print("TP : ",tp)
print("TN : ",tn)
print("FP : ",fp)
print("FN : ",fn)

print("Precision : ",tp/(tp+fp)*100,"%")
print("Recall : ",tp/(tp+fn)*100,"%")
print("Error Rate : ",(fn+fp)/(fn+fp+tn+tp)*100,"%")
print("Accuracy : ",(tp+tn)/(tp+tn+fp+fn)*100,"%")
```

```
TP : 45
TN : 90
FP : 0
FN : 0
Precision : 100.0 %
Recall : 100.0 %
Error Rate : 0.0 %
Accuracy : 100.0 %
```