import numpy as np

import pandas as pd
df=pd.read\_csv('\_/content/Iris.csv')

$\Rightarrow$		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
	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

150 rows × 6 columns

## 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.dtypes

int64 float64 Id SepalLengthCm SepalWidthCm float64 PetalLengthCm float64 PetalWidthCm float64 Species object dtype: object

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

```
SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                                Species
        0
                                                                              Iris-setosa
        1
                        4.9
                                        3.0
                                                        1.4
                                                                        0.2
                                                                              Iris-setosa
        2
                                        3.2
                                                                        0.2
                        4.7
                                                        1.3
                                                                              Iris-setosa
        3
                        4.6
                                        3.1
                                                         1.5
                                                                        0.2
                                                                              Iris-setosa
                        5.0
                                        36
                                                                        0.2
                                                        1.4
                                                                              Iris-setosa
                         ...
                                         ...
                                                         ...
                                                                         ...
       145
                        67
                                        3.0
                                                         52
                                                                        2.3 Iris-virginica
df.isna().sum()
     SepalLengthCm
     SepalWidthCm
                         0
     PetalLengthCm
                         a
     PetalWidthCm
                         0
     Species
                         0
     dtype: int64
df.describe()
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
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

## Separating input and output samples

```
'Iris-virginica', 'Iris-virgin
```

## Training and testing data

```
from sklearn.model_selection import train_test_split
x\_train, x\_test, y\_train, y\_test=train\_test\_split(x, y, test\_size=0.30, random\_state=42)
x_train
             [6. , 2.7, 5.1, 1.6],
             [6.1, 2.6, 5.6, 1.4],
             [7.7, 3., 6.1, 2.3],
             [5.5, 2.5, 4. , 1.3],
             [4.4, 2.9, 1.4, 0.2],
             [4.3, 3., 1.1, 0.1],
             [6., 2.2, 5., 1.5],
             [7.2, 3.2, 6., 1.8], [4.6, 3.1, 1.5, 0.2],
             [5.1, 3.5, 1.4, 0.3],
             [4.4, 3., 1.3, 0.2],
             [6.3, 2.5, 4.9, 1.5],
             [6.3, 3.4, 5.6, 2.4],
             [4.6, 3.4, 1.4, 0.3],
             [6.8, 3., 5.5, 2.1],
             [6.3, 3.3, 6., 2.5],
             [4.7, 3.2, 1.3, 0.2],
             [6.1, 2.9, 4.7, 1.4],
             [6.5, 2.8, 4.6, 1.5],
[6.2, 2.8, 4.8, 1.8],
             [7., 3.2, 4.7, 1.4],
             [6.4, 3.2, 5.3, 2.3],
             [5.1, 3.8, 1.6, 0.2],
             [6.9, 3.1, 5.4, 2.1],
             [5.9, 3., 4.2, 1.5],
[6.5, 3., 5.2, 2.],
             [5.7, 2.6, 3.5, 1.],
             [5.2, 2.7, 3.9, 1.4],
             [6.1, 3. , 4.6, 1.4],
             [4.5, 2.3, 1.3, 0.3],
             [6.6, 2.9, 4.6, 1.3],
             [5.5, 2.6, 4.4, 1.2],
             [5.3, 3.7, 1.5, 0.2],
             [5.6, 3., 4.1, 1.3],
             [7.3, 2.9, 6.3, 1.8],
             [6.7, 3.3, 5.7, 2.1],
             [5.1, 3.7, 1.5, 0.4],
             [4.9, 2.4, 3.3, 1.],
             [6.7, 3.3, 5.7, 2.5],
             [7.2, 3., 5.8, 1.6],
             [4.9, 3.1, 1.5, 0.1],
             [6.7, 3.1, 5.6, 2.4],
             [4.9, 3., 1.4, 0.2],
             [6.9, 3.1, 4.9, 1.5],
             [7.4, 2.8, 6.1, 1.9],
             [6.3, 2.9, 5.6, 1.8],
             [5.7, 2.8, 4.1, 1.3],
             [6.5, 3., 5.5, 1.8],
             [6.3, 2.3, 4.4, 1.3],
             [6.4, 2.9, 4.3, 1.3],
             [5.6, 2.8, 4.9, 2.],
             [5.9, 3., 5.1, 1.8],
             [5.4, 3.4, 1.7, 0.2],
             [6.1, 2.8, 4. , 1.3],
             [4.9, 2.5, 4.5, 1.7],
             [5.8, 4., 1.2, 0.2],
             [5.8, 2.6, 4., 1.2],
[7.1, 3., 5.9, 2.1]])
```

```
array([[6.1, 2.8, 4.7, 1.2], [5.7, 3.8, 1.7, 0.3], [7.7, 2.6, 6.9, 2.3], [6. , 2.9, 4.5, 1.5], [6.8, 2.8, 4.8, 1.4],
```

[5.4, 3.4, 1.5, 0.4],

x test

```
[5.6, 2.9, 3.6, 1.3],
[6.9, 3.1, 5.1, 2.3],
[6.2, 2.2, 4.5, 1.5],
[5.8, 2.7, 3.9, 1.2],
[6.5, 3.2, 5.1, 2.],
[4.8, 3., 1.4, 0.1],
[5.5, 3.5, 1.3, 0.2],
[4.9, 3.1, 1.5, 0.1],
[5.1, 3.8, 1.5, 0.3],
[6.3, 3.3, 4.7, 1.6],
[6.5, 3., 5.8, 2.2],
[5.6, 2.5, 3.9, 1.1],
[5.7, 2.8, 4.5, 1.3],
[6.4, 2.8, 5.6, 2.2],
[4.7, 3.2, 1.6, 0.2],
[6.1, 3., 4.9, 1.8],
[5., 3.4, 1.6, 0.4],
[6.4, 2.8, 5.6, 2.1],
[7.9, 3.8, 6.4, 2.],
[6.7, 3., 5.2, 2.3],
[6.7, 2.5, 5.8, 1.8],
[6.8, 3.2, 5.9, 2.3],
[4.8, 3., 1.4, 0.3],
[4.8, 3.1, 1.6, 0.2],
[4.6, 3.6, 1., 0.2],
[5.7, 4.4, 1.5, 0.4],
[6.7, 3.1, 4.4, 1.4],
[4.8, 3.4, 1.6, 0.2],
[4.4, 3.2, 1.3, 0.2],
[6.3, 2.5, 5., 1.9],
[6.4, 3.2, 4.5, 1.5],
[5.2, 3.5, 1.5, 0.2],
[5., 3.6, 1.4, 0.2],
[5.2, 4.1, 1.5, 0.1],
[5.8, 2.7, 5.1, 1.9],
[6., 3.4, 4.5, 1.6],
[6.7, 3.1, 4.7, 1.5],
[5.4, 3.9, 1.3, 0.4],
[5.4, 3.7, 1.5, 0.2]])
```

#### y\_train

### y\_test

```
'Iris-setosa', 'Iris-virginica', 'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa'], dtype=object)
```

 $\hbox{\tt [-0.29283662, -0.25367584, -0.15856401, 0.08358997],}$ [1.27470056, 0.2306144, 0.71771076, 1.43388941],[ 0.43064208, -1.9486917 , 0.36720086, 0.35364985], [-0.05167705, -0.73796609, 0.01669095, -0.05143998], [0.79238143, 0.47275953, 0.71771076, 1.02879957],

```
Normalisation
from sklearn.preprocessing import StandardScaler
 scaler=StandardScaler()
scaler.fit(x_train)
x train=scaler.transform(x train)
x_test=scaler.transform(x_test)
 x_train
                             [ 0.18948252, -0.73796609, 0.71771076,
                                                                                                                           0.4886798 ],
                                                                                                                           0.21861991],
                            [ 0.3100623 , -0.98011121, 1.00980236,
                            [ 2.23933883, -0.01153072, 1.30189395, 1.43388941],
[-0.4134164, -1.22225633, 0.07510927, 0.08358997],
                            [-1.73979401, -0.25367584, -1.443767 , -1.40173942],
[-1.8603738 , -0.01153072, -1.61902196, -1.53676936],
                            [ 0.18948252, -1.9486917 , 0.65929245, 0.35364985], [ 1.63643991, 0.47275953, 1.24347563, 0.75873969],
                             \hbox{\tt [-1.49863445, 0.2306144, -1.38534869, -1.40173942],}
                             [-0.89573553, 1.19919489, -1.443767 , -1.26670948],
                             [-1.73979401, -0.01153072, -1.50218532, -1.40173942],
                            [ 0.55122187, -1.22225633, 0.60087413, 0.35364985],
[ 0.55122187, 0.95704977, 1.00980236, 1.56891935],
                            [-1.49863445, 0.95704977, -1.443767 , -1.26670948],
[1.15412078, -0.01153072, 0.95138404, 1.16382952],
                            [ 0.55122187, 0.71490465, 1.24347563, 1.7039493 ], [-1.37805466, 0.47275953, -1.50218532, -1.40173942],
                                0.3100623, -0.25367584, 0.48403749, 0.21861991], 0.79238143, -0.49582097, 0.42561917, 0.35364985],
                                0.43064208, -0.49582097, 0.54245581, 0.75873969],
                                1.39528035, 0.47275953, 0.48403749, 0.21861991],
                            [ 0.67180165, 0.47275953, 0.8345474, 1.43388941],
[-0.89573553, 1.92563026, -1.32693037, -1.40173942],
[ 1.27470056, 0.2306144, 0.89296572, 1.16382952],
                            [ 0.06890273, -0.01153072, 0.1919459 , 0.35364985], [ 0.79238143, -0.01153072, 0.77612908, 1.02879957],
                            [-0.17225683, -0.98011121, -0.21698232, -0.32149987],
[-0.77515575, -0.73796609, 0.01669095, 0.21861991],
                             [ \ 0.3100623 \ , \ -0.01153072, \ \ 0.42561917, \ \ 0.21861991],
                             \hbox{\tt [-1.61921423, -1.70654658, -1.50218532, -1.26670948],}\\
                             [ \ 0.91296121, \ -0.25367584, \ \ 0.42561917, \ \ 0.08358997],
                             [-0.4134164 , -0.98011121, 0.30878254, -0.05143998],
                            [-0.65457597, 1.68348514, -1.38534869, -1.40173942],
[-0.29283662, -0.01153072, 0.13352758, 0.08358997],
                            [ 1.7570197 , -0.25367584, 1.41873058, 0.75873969],
                            [ 1.033541 , 0.71490465, 1.06822067, 1.16382952], [-0.89573553, 1.68348514, -1.38534869, -1.13167953],
                            [-1.1368951 , -1.46440146 , -0.33381896 , -0.32149987],
[ 1.033541 , 0.71490465 , 1.06822067 , 1.7039493 ],
                             [ 1.63643991, -0.01153072, 1.12663899, 0.4886798 ],
                              [-1.1368951 , 0.2306144 , -1.38534869, -1.53676936],
                            [ 1.033541 , 0.2306144 , 1.00980236, 1.56891935],
[-1.1368951 , -0.01153072, -1.443767 , -1.40173942],
                            [ 1.27470056, 0.2306144, 0.60087413, 0.35364985], [ 1.87759948, -0.49582097, 1.30189395, 0.89376963], [ 0.55122187, -0.25367584, 1.00980236, 0.75873969], 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.40582007, 0.405820
                            [-0.17225683, -0.49582097, 0.13352758, 0.08358997], [0.79238143, -0.01153072, 0.95138404, 0.75873969],
                                  0.55122187, \ -1.70654658, \ \ 0.30878254, \ \ 0.08358997], 
                             [ \ 0.67180165, \ -0.25367584, \ \ 0.25036422, \ \ 0.08358997],
                            [-0.29283662, -0.49582097, 0.60087413, 1.02879957], [ 0.06890273, -0.01153072, 0.71771076, 0.75873969],
                             [-0.53399618, 0.95704977, -1.26851205, -1.40173942],
                             [ \ 0.3100623 \ , \ -0.49582097, \ \ 0.07510927, \ \ 0.08358997],
                             [-1.1368951 , -1.22225633, 0.36720086, 0.62370974],
                            [-0.05167705, 2.40992051, -1.56060364, -1.40173942],
[-0.05167705, -0.98011121, 0.07510927, -0.05143998],
                             [ 1.51586013, -0.01153072, 1.18505731, 1.16382952]])
 x test
             array([[ 0.3100623 , -0.49582097, 0.48403749, -0.05143998],
                            [-0.17225683, 1.92563026, -1.26851205, -1.26670948],
[2.23933883, -0.98011121, 1.76924049, 1.43388941],
[0.18948252, -0.25367584, 0.36720086, 0.35364985],
                            [ 1.15412078, -0.49582097, 0.54245581, 0.21861991],
[-0.53399618, 0.95704977, -1.38534869, -1.13167953],
```

```
[-1.25747488, -0.01153072, -1.443767
                                                                                   , -1.53676936],
[-0.4134164 , 1.19919489, -1.50218532, -1.40173942],
[-1.1368951 , 0.2306144 , -1.38534869, -1.53676936],
[-0.89573553, 1.92563026, -1.38534869, -1.26670948],
[0.55122187, 0.71490465, 0.48403749, 0.4886798],
[0.79238143, -0.01153072, 1.12663899, 1.29885946],
[-0.29283662, -1.22225633, 0.01669095, -0.18646992],
[-0.17225683, -0.49582097, 0.36720086, 0.08358997],
[ 0.67180165, -0.49582097, 1.00980236, 1.29885946], [-1.37805466, 0.47275953, -1.32693037, -1.40173942],
[ 0.3100623 , -0.01153072, 0.60087413, 0.75873969], [-1.01631531, 0.95704977, -1.32693037, -1.13167953], [ 0.67180165, -0.49582097, 1.00980236, 1.16382952],
[ 2.4804984 , 1.92563026, 1.4771489 , 1.02879957],
[ 1.033541 , -0.01153072, 0.77612908, 1.43388941],
[ 1.033541 , -1.22225633, 1.12663899, 0.75873969],
[ 1.15412078, 0.47275953, 1.18505731, 1.43388941],
[-1.25747488, -0.01153072, -1.443767 , -1.26670948],
[-1.25747488, 0.2306144 , -1.32693037, -1.40173942],
[-1.49863445, 1.44134002, -1.67744028, -1.40173942],

[-0.17225683, 3.378501, -1.38534869, -1.13167953],

[1.033541, 0.2306144, 0.30878254, 0.21861991],

[-1.25747488, 0.95704977, -1.32693037, -1.40173942],
[-1.73979401, 0.47275953, -1.50218532, -1.40173942],
[ 0.55122187, -1.22225633, 0.65929245, 0.89376963], [ 0.67180165, 0.47275953, 0.36720086, 0.35364985], [-0.77515575, 1.19919489, -1.38534869, -1.40173942], [-1.01631531, 1.44134002, -1.443767, -1.40173942],
[-0.77515575, 2.65206563, -1.38534869, -1.53676936],
[-0.05167705, -0.73796609, 0.71771076, 0.89376963],
[ 0.18948252, 0.95704977, 0.36720086, 0.4886798 ],
   1.033541 , 0.2306144 , 0.48403749, 0.35364985],
[-0.53399618, 2.16777538, -1.50218532, -1.13167953],
[-0.53399618, 1.68348514, -1.38534869, -1.40173942]])
```

#### Model creation

```
from sklearn.naive bayes import GaussianNB
model=GaussianNB()
model.fit(x_train,y_train)
y_pred=model.predict(x_test)
y_pred
                'Iris-setosa', 'Iris-setosa', 'Iris-virginica', 'Iris-virginica',
                                       'Iris-versicolor', 'Iris-versicolor', 'Iris-virginica',
                                        'Iris-setosa', 'Iris-virginica', 'Iris-setosa', 'Iris-virginica',
                                       'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',
                                        'Iris-virginica', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
                                       'Iris-setosa', 'Iris-virginica', 'Iris-versicolor', 'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa'], dtype='<U15')
y_test
                'Iris-setosa', 'Iris-setosa', 'Iris-versicolor', 'Iris-virginica',
                                       'Iris-versicolor', 'Iris-versicolor', 'Iris-virginica',
                                       'Iris-verginica', 'Iris-virginica', 'Iris-setosa', 'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-virginica', 'Iris-setosa', 'Iris-setosa
                                       'Iris-virginica', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
                                       'Iris-setosa', 'Iris-virginica', 'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa'], dtype=object)
```

# ▼ Performance evaluation

```
from sklearn.metrics import confusion_matrix,accuracy_score
result=confusion_matrix(y_test,y_pred)
score=accuracy_score(y_test,y_pred)
result,score
```

```
(array([[19, 0, 0],
[ 0, 12, 1],
```

## [ 0, 0, 13]]), 0.9777777777777)

from sklearn.metrics import classification\_report
print(classification\_report(y\_test,y\_pred))

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	19
Iris-versicolor	1.00	0.92	0.96	13
Iris-virginica	0.93	1.00	0.96	13
accuracy			0.98	45
macro avg	0.98	0.97	0.97	45
weighted avg	0.98	0.98	0.98	45