In [4]: import io import numpy as np import matplotlib.pyplot as plt %matplotlib inline import seaborn as sns df=pd.read_csv("Iris.csv") print(df.shape) df

(150, 6)

Out[4]:

	ld	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

```
In [5]: print('The DataFrame contains %d rows and %d columns'%(df.shape))
        print(df.dtypes)
        print(df.info())
        print(df.head(7))
        The DataFrame contains 150 rows and 6 columns
                           int64
        Ιd
        SepalLengthCm
                         float64
        SepalWidthCm
                         float64
        PetalLengthCm
                         float64
        PetalWidthCm
                         float64
        Species
                          object
        dtype: object
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 6 columns):
                            Non-Null Count Dtype
             Column
             ----
                             -----
                                             ----
         0
             Ιd
                            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
        None
           Id
               SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                              Species
        0
            1
                         5.1
                                        3.5
                                                       1.4
                                                                     0.2 Iris-setosa
            2
                         4.9
        1
                                        3.0
                                                       1.4
                                                                     0.2 Iris-setosa
        2
            3
                         4.7
                                        3.2
                                                       1.3
                                                                     0.2 Iris-setosa
            4
        3
                         4.6
                                        3.1
                                                       1.5
                                                                     0.2 Iris-setosa
        4
            5
                         5.0
                                        3.6
                                                       1.4
                                                                     0.2 Iris-setosa
        5
            6
                         5.4
                                        3.9
                                                       1.7
                                                                     0.4 Iris-setosa
            7
                         4.6
                                        3.4
                                                       1.4
                                                                     0.3 Iris-setosa
In [6]: features=df.iloc[:,0:4]
        print(features.head())
           Ιd
               SepalLengthCm SepalWidthCm PetalLengthCm
        0
            1
                         5.1
                                        3.5
                                                       1.4
        1
            2
                         4.9
                                        3.0
                                                       1.4
        2
            3
                         4.7
                                        3.2
                                                       1.3
        3
            4
                         4.6
                                                       1.5
                                        3.1
            5
                         5.0
                                        3.6
                                                       1.4
In [7]: | target=df.iloc[:,5]
        print(target.head())
        0
             Iris-setosa
        1
             Iris-setosa
        2
             Iris-setosa
        3
             Iris-setosa
             Iris-setosa
        Name: Species, dtype: object
```

```
In [8]: print('The initial dataframe contained %d rows and %d columns '%(df.shape))
         print('The features matrix contains %d rows and %d columns '%(features.shape))
         print('The target vector contains %d rows and %d columns'%(np.array(target).res
         The initial dataframe contained 150 rows and 6 columns
         The features matrix contains 150 rows and 4 columns
         The target vector contains 150 rows and 1 columns
 In [9]: from sklearn.naive bayes import GaussianNB
         algorithm=GaussianNB(priors=None, var smoothing=1e-9)
         algorithm.fit(features, target)
 Out[9]:
          ▼ GaussianNB
          GaussianNB()
In [10]: print(algorithm.classes_)
         ['Iris-setosa' 'Iris-versicolor' 'Iris-virginica']
In [11]: print('The Gaussain model has acheieved %.2f percent accuracy'%(algorithm.score
         The Gaussain model has acheieved 0.99 percent accuracy
In [12]: | observation=[[5.0,3.7,1.6,0.1]]
         predictions=algorithm.predict(observation)
         print(predictions)
         ['Iris-setosa']
         C:\Users\vinay\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
         arn\base.py:450: UserWarning: X does not have valid feature names, but Gaussi
         anNB was fitted with feature names
           warnings.warn(
In [13]: print(algorithm.predict proba(observation).round())
         [[1. 0. 0.]]
         C:\Users\vinay\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
         arn\base.py:450: UserWarning: X does not have valid feature names, but Gaussi
         anNB was fitted with feature names
           warnings.warn(
In [14]: x = df.iloc[:,1:5]
         y = df.iloc[:,5:]
In [15]: from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.3,random_sta
```

```
In [16]: from sklearn.metrics import confusion matrix, classification report
         naive bayes = GaussianNB()
         naive_bayes.fit(x_train,y_train)
         pred = naive bayes.predict(x test)
         C:\Users\vinay\AppData\Local\Programs\Python\Python311\Lib\site-packages\skle
         arn\utils\validation.py:1111: DataConversionWarning: A column-vector y was pa
         ssed when a 1d array was expected. Please change the shape of y to (n_sample
         s, ), for example using ravel().
           y = column_or_1d(y, warn=True)
In [17]: from sklearn.preprocessing import LabelEncoder
         cm = confusion_matrix(y_test,pred,labels = naive_bayes.classes_)
         print(cm)
         [[16 0 0]
          [ 0 18 0]
          [ 0 0 11]]
In [18]: | print(classification_report(y_test,pred))
                          precision
                                        recall f1-score
                                                           support
             Iris-setosa
                                1.00
                                          1.00
                                                    1.00
                                                                16
                               1.00
         Iris-versicolor
                                          1.00
                                                    1.00
                                                                18
          Iris-virginica
                                1.00
                                          1.00
                                                    1.00
                                                                11
                accuracy
                                                    1.00
                                                                45
               macro avg
                                1.00
                                          1.00
                                                    1.00
                                                                45
                                                                45
            weighted avg
                                          1.00
                                                    1.00
                                1.00
In [19]: from sklearn.metrics import confusion matrix, ConfusionMatrixDisplay, classificat
         print('\nAccuracy: {:.2f}'.format(accuracy_score(y_test,pred)))
         err_rate=1-accuracy_score(y_test,pred)
         print('Error Rate: ',err_rate)
```

Accuracy: 1.00 Error Rate: 0.0