## **Logistic Regression**

```
In [74]:
          import pandas as pd
           import numpy as np
           from sklearn.linear_model import LogisticRegression
           from sklearn.preprocessing import StandardScaler
In [75]:
           df=pd.read csv(r"C:\Users\sneha\Downloads\ionosphere.csv")
Out[75]:
                                                                     column_f column_g
                                                                                                                colı
                 column_a column_b column_c column_d column_e
                                                                                           column_h column_i
              0
                                        0.99539
                                                   -0.05889
                                                                        0.02306
                                                                                                        1.00000
                                                                                                                  0.
                                False
                                                              0.85243
                                                                                   0.83398
                                                                                             -0.37708
                      True
                                        1.00000
                                                   -0.18829
                                                              0.93035
                                                                                  -0.10868
              1
                      True
                                False
                                                                        -0.36156
                                                                                             -0.93597
                                                                                                        1.00000
                                                                                                                 -0.
              2
                      True
                                False
                                        1.00000
                                                   -0.03365
                                                              1.00000
                                                                        0.00485
                                                                                   1.00000
                                                                                             -0.12062
                                                                                                        0.88965
                                                                                                                  0.
              3
                                        1.00000
                                                              1.00000
                                                                        1.00000
                                                                                             -1.00000
                                                                                                        0.00000
                                                                                                                  0.
                      True
                                False
                                                   -0.45161
                                                                                   0.71216
              4
                                False
                                        1.00000
                                                   -0.02401
                                                              0.94140
                                                                        0.06531
                                                                                   0.92106
                                                                                             -0.23255
                                                                                                        0.77152
                                                                                                                 -0.
                      True
              5
                      True
                                False
                                        0.02337
                                                   -0.00592
                                                             -0.09924
                                                                        -0.11949
                                                                                  -0.00763
                                                                                             -0.11824
                                                                                                        0.14706
                                                                                                                  0.
              6
                      True
                                False
                                        0.97588
                                                   -0.10602
                                                              0.94601
                                                                        -0.20800
                                                                                   0.92806
                                                                                             -0.28350
                                                                                                        0.85996
                                                                                                                  -0.
              7
                     False
                                False
                                        0.00000
                                                   0.00000
                                                              0.00000
                                                                        0.00000
                                                                                   1.00000
                                                                                             -1.00000
                                                                                                        0.00000
                                                                                                                  0.
              8
                      True
                                False
                                        0.96355
                                                   -0.07198
                                                              1.00000
                                                                        -0.14333
                                                                                   1.00000
                                                                                             -0.21313
                                                                                                        1.00000
                                                                                                                 -0.
              9
                      True
                                False
                                        -0.01864
                                                   -0.08459
                                                              0.00000
                                                                        0.00000
                                                                                   0.00000
                                                                                             0.00000
                                                                                                        0.11470
                                                                                                                 -0.
           pd.set_option('display.max_rows',10000000000)
           pd.set_option('display.max_columns',10000000000)
           pd.set option('display.width',95)
In [77]: print('this DataFrame has %d Rowws and %d columns'%(df.shape))
           this DataFrame has 351 Rowws and 35 columns
In [78]:
           df.head()
Out[78]:
               column_a
                         column_b
                                    column_c column_d column_e column_f column_g
                                                                                         column_h column_i
                                                                                                              column_j
            0
                    True
                              False
                                      0.99539
                                                -0.05889
                                                            0.85243
                                                                      0.02306
                                                                                 0.83398
                                                                                           -0.37708
                                                                                                      1.00000
                                                                                                                0.03760
                                                            0.93035
            1
                    True
                              False
                                      1.00000
                                                -0.18829
                                                                     -0.36156
                                                                                -0.10868
                                                                                           -0.93597
                                                                                                      1.00000
                                                                                                               -0.04549
            2
                              False
                                      1.00000
                                                -0.03365
                                                            1.00000
                                                                      0.00485
                                                                                 1.00000
                                                                                                      0.88965
                    True
                                                                                           -0.12062
                                                                                                                0.01198
```

1.00000

0.06531

1.00000

0.94140

0.71216

0.92106

-1.00000

-0.23255

0.00000

0.77152

3

4

True

True

1.00000

1.00000

-0.45161

-0.02401

False

False

0.00000

-0.16399

```
features matrix=df.iloc[:,0:34]
In [79]:
In [80]: | target_vector=df.iloc[:,-1]
In [81]: print('The Features Matrix Has %d Rows and %d Column(s)'%(features_matrix.shape))
         print('The Target Matrix Has %d Rows and %d Column(s)'
               %(np.array(target vector).reshape(-1,1).shape))
         The Features Matrix Has 351 Rows and 34 Column(s)
         The Target Matrix Has 351 Rows and 1 Column(s)
In [82]: | features matrix standardized=StandardScaler().fit transform(features matrix)
In [83]: algorithm=LogisticRegression(penalty=None,dual=False,tol=1e-4,C=1.0,fit_intercept=True,
                                      intercept scaling=1, class weight=None, random state=None,
                                      solver='lbfgs',max iter=1000,multi class='auto',verbose=0,
                                      warm start=False,n jobs=None,l1 ratio=None)
In [84]: logistic Regression Mode=algorithm.fit(features matrix standardized, target vector)
In [85]: observation=[[1,0,0.99539,-0.5889,0.85242999999999,0.02306,0.8339799999999,-0.37708,
                       1.0,0.0376,0.8524-29999999999,-0.17755,0.59755,-0.44945,0.60536,-0.38223
                       ,0.8435600000000001,-0.38542,0.58212,-0.32192,0.56971,-0.29674,0.36946,
                       -0.47357, 0.56811, -0.51171, 0.4107800000000003, -0.4616800000000003, 0.21256,
                       -0.3409,0.112267,-0.54487,0.18641,-0.453]]
In [86]: predictions=logistic Regression Mode.predict(observation)
         print('The Model predicted the observation to belog to class %s'%(predictions))
         The Model predicted the observation to belog to class ['g']
In [87]: print('The algorithm was trained to predict one of the two classes:%s'%(algorithm.classes
         The algorithm was trained to predict one of the two classes:['b' 'g']
In [88]: print(""The model says the probability of the observation we passed belonging to
         class['b']is %s"""%(algorithm.predict_proba(observation)[0][0]))
         print("""The model says the probability of the observation we passed belonging to
         class['g']is %s"""%(algorithm.predict_proba(observation)[observation[0][1]]))
         The model says the probability of the observation we passed belonging to
         class['b']is 0.0
         The model says the probability of the observation we passed belonging to
         class['g']is [0. 1.]
 In [ ]:
 In [ ]:
```