```
In [1]: import pandas as pd
import numpy as np
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

In [2]: df=pd.read\_csv(r"C:\Users\sweet\Downloads\ionosphere.csv")
 df

## Out[2]:

|     | column_a | column_b | column_c | column_d | column_e | column_f | column_g | column_h | column_i | column_j | <br>column_z | column_aa | colur       |
|-----|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|--------------|-----------|-------------|
| 0   | True     | False    | 0.99539  | -0.05889 | 0.85243  | 0.02306  | 0.83398  | -0.37708 | 1.00000  | 0.03760  | <br>-0.51171 | 0.41078   | -0.         |
| 1   | True     | False    | 1.00000  | -0.18829 | 0.93035  | -0.36156 | -0.10868 | -0.93597 | 1.00000  | -0.04549 | <br>-0.26569 | -0.20468  | -0.         |
| 2   | True     | False    | 1.00000  | -0.03365 | 1.00000  | 0.00485  | 1.00000  | -0.12062 | 0.88965  | 0.01198  | <br>-0.40220 | 0.58984   | -0.         |
| 3   | True     | False    | 1.00000  | -0.45161 | 1.00000  | 1.00000  | 0.71216  | -1.00000 | 0.00000  | 0.00000  | <br>0.90695  | 0.51613   | 1.          |
| 4   | True     | False    | 1.00000  | -0.02401 | 0.94140  | 0.06531  | 0.92106  | -0.23255 | 0.77152  | -0.16399 | <br>-0.65158 | 0.13290   | -0.         |
|     |          |          |          |          |          |          |          |          |          |          | <br>         |           |             |
| 346 | True     | False    | 0.83508  | 0.08298  | 0.73739  | -0.14706 | 0.84349  | -0.05567 | 0.90441  | -0.04622 | <br>-0.04202 | 0.83479   | 0.          |
| 347 | True     | False    | 0.95113  | 0.00419  | 0.95183  | -0.02723 | 0.93438  | -0.01920 | 0.94590  | 0.01606  | <br>0.01361  | 0.93522   | 0.          |
| 348 | True     | False    | 0.94701  | -0.00034 | 0.93207  | -0.03227 | 0.95177  | -0.03431 | 0.95584  | 0.02446  | <br>0.03193  | 0.92489   | 0.          |
| 349 | True     | False    | 0.90608  | -0.01657 | 0.98122  | -0.01989 | 0.95691  | -0.03646 | 0.85746  | 0.00110  | <br>-0.02099 | 0.89147   | <b>-</b> 0. |
| 350 | True     | False    | 0.84710  | 0.13533  | 0.73638  | -0.06151 | 0.87873  | 0.08260  | 0.88928  | -0.09139 | <br>-0.15114 | 0.81147   | -0.         |

351 rows × 35 columns

```
In [3]: pd.set_option('display.max_rows',10000000000)
    pd.set_option('display.max_columns',10000000000)
    pd.set_option('display.width',95)
```

```
print('This DataFrame has %d Rows and %d columns'%(df.shape))
 In [4]:
          This DataFrame has 351 Rows and 35 columns
          df.head()
 In [5]:
 Out[5]:
              column_a column_b column_c column_d column_e column_f column_g column_h column_i column_j column_k column_l column_m
           0
                                   0.99539
                                             -0.05889
                                                       0.85243
                                                                 0.02306
                                                                           0.83398
                                                                                    -0.37708
                                                                                              1.00000
                                                                                                       0.03760
                                                                                                                  0.85243
                                                                                                                          -0.17755
                                                                                                                                     0.59755
                   True
                            False
                   True
                                   1.00000
                                             -0.18829
                                                       0.93035
                                                                -0.36156
                                                                           -0.10868
                                                                                    -0.93597
                                                                                              1.00000
                                                                                                       -0.04549
                                                                                                                  0.50874
                                                                                                                          -0.67743
                                                                                                                                     0.34432
                            False
                   True
                            False
                                   1.00000
                                             -0.03365
                                                       1.00000
                                                                 0.00485
                                                                           1.00000
                                                                                    -0.12062
                                                                                              0.88965
                                                                                                        0.01198
                                                                                                                  0.73082
                                                                                                                           0.05346
                                                                                                                                     0.85443
                                             -0.45161
                                                       1.00000
                                                                                    -1.00000
                                                                                              0.00000
                                                                                                       0.00000
           3
                   True
                            False
                                   1.00000
                                                                 1.00000
                                                                           0.71216
                                                                                                                  0.00000
                                                                                                                           0.00000
                                                                                                                                     0.00000
                   True
                                   1.00000
                                             -0.02401
                                                       0.94140
                                                                 0.06531
                                                                           0.92106
                                                                                    -0.23255
                                                                                              0.77152
                                                                                                       -0.16399
                                                                                                                  0.52798
                                                                                                                          -0.20275
                                                                                                                                     0.56409
                            False
          features matrix = df.iloc[:,0:34]
 In [6]:
          target vector = df.iloc[:,-1]
 In [8]: print('The Features Matrix Has %d Rows And %d columnsIn [11]:(s)'%(features_matrix.shape))
          The Features Matrix Has 351 Rows And 34 columnsIn [11]:(s)
In [10]: print('The Target Matrix Has %d Rows And %d Columns(s)'%(np.array(target_vector).reshape(-1,1).shape))
          The Target Matrix Has 351 Rows And 1 Columns(s)
In [11]: | features_matrix_standardized = StandardScaler().fit_transform(features_matrix)
```

```
In [13]: 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=10000,
        multi class='auto',verbose=0, warm start=False, n jobs=None,l1 ratio=None)
In [14]: Logistic Regression Model = algorithm.fit(features matrix standardized,target vector)
In [15]: observation = [[1, 0, 0.99539, -0.05889, 0.85242999999999, 0.02306, 0.83397999999999, -0.37708,1.0,0.0376,
         0.58212, -0.32192, 0.56971, -0.29674, 0.36946, -0.47357, 0.56811, -0.51171, 0.4107800000000003,
         -0.461680000000003, 0.21266, -0.3409, 0.112267, -0.54487, 0.18641, -0.453]]
In [16]: predictions = Logistic Regression Model.predict(observation)
        print('The Model predicted The observation To Belong To Class %s'%(predictions))
         The Model predicted The observation To Belong To Class ['g']
In [17]: print('The Algorithm Was Trained To predict The One Of The Classes: %s'%(algorithm.classes ))
        The Algorithm Was Trained To predict The One Of The Classes: ['b' 'g']
In [19]: print("""The Model Says The Probability Of The observation We Passed belonging To The Class ['b'] is %s"""
        %(algorithm.predict proba(observation)[0][0]))
        print()
        The Model Says The Probability Of The observation We Passed belonging To The Class ['b'] is 2.5317757538667607e-05
In [22]: print("""The Model Says The Probability Of The observation We Passed belonging To The Class ['g'] is %s"""
        %(algorithm.predict proba(observation)[0][1]))
        The Model Says The Probability Of The observation We Passed belonging To The Class ['g'] is 0.9999746822424613
```