```
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\Gowthami\Downloads\ionosphere_data.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 | column_ab | С |
|-----|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|--------------|-----------|-----------|---|
| 0   | True     | False    | 0.99539  | -0.05889 | 0.85243  | 0.02306  | 0.83398  | -0.37708 | 1.00000  | 0.03760  | <br>-0.51171 | 0.41078   | -0.46168  | _ |
| 1   | True     | False    | 1.00000  | -0.18829 | 0.93035  | -0.36156 | -0.10868 | -0.93597 | 1.00000  | -0.04549 | <br>-0.26569 | -0.20468  | -0.18401  |   |
| 2   | True     | False    | 1.00000  | -0.03365 | 1.00000  | 0.00485  | 1.00000  | -0.12062 | 0.88965  | 0.01198  | <br>-0.40220 | 0.58984   | -0.22145  |   |
| 3   | True     | False    | 1.00000  | -0.45161 | 1.00000  | 1.00000  | 0.71216  | -1.00000 | 0.00000  | 0.00000  | <br>0.90695  | 0.51613   | 1.00000   |   |
| 4   | True     | False    | 1.00000  | -0.02401 | 0.94140  | 0.06531  | 0.92106  | -0.23255 | 0.77152  | -0.16399 | <br>-0.65158 | 0.13290   | -0.53206  |   |
|     |          |          |          |          |          |          |          |          |          |          | <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   | 0.00123   |   |
| 347 | True     | False    | 0.95113  | 0.00419  | 0.95183  | -0.02723 | 0.93438  | -0.01920 | 0.94590  | 0.01606  | <br>0.01361  | 0.93522   | 0.04925   |   |
| 348 | True     | False    | 0.94701  | -0.00034 | 0.93207  | -0.03227 | 0.95177  | -0.03431 | 0.95584  | 0.02446  | <br>0.03193  | 0.92489   | 0.02542   |   |
| 349 | True     | False    | 0.90608  | -0.01657 | 0.98122  | -0.01989 | 0.95691  | -0.03646 | 0.85746  | 0.00110  | <br>-0.02099 | 0.89147   | -0.07760  |   |
| 350 | True     | False    | 0.84710  | 0.13533  | 0.73638  | -0.06151 | 0.87873  | 0.08260  | 0.88928  | -0.09139 | <br>-0.15114 | 0.81147   | -0.04822  |   |

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)
```

#### In [6]:

```
print('The DataFrame has %d Rows and %d columns'%(df.shape))
```

The DataFrame has 351 Rows and 35 columns

# In [7]:

df.head()

# Out[7]:

|   | 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 | column_r |
|---|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| 0 | True     | False    | 0.99539  | -0.05889 | 0.85243  | 0.02306  | 0.83398  | -0.37708 | 1.00000  | 0.03760  | 0.85243  | -0.17755 | 0.59755  | -0.4494  |
| 1 | True     | False    | 1.00000  | -0.18829 | 0.93035  | -0.36156 | -0.10868 | -0.93597 | 1.00000  | -0.04549 | 0.50874  | -0.67743 | 0.34432  | -0.69707 |
| 2 | True     | False    | 1.00000  | -0.03365 | 1.00000  | 0.00485  | 1.00000  | -0.12062 | 0.88965  | 0.01198  | 0.73082  | 0.05346  | 0.85443  | 0.00827  |
| 3 | True     | False    | 1.00000  | -0.45161 | 1.00000  | 1.00000  | 0.71216  | -1.00000 | 0.00000  | 0.00000  | 0.00000  | 0.00000  | 0.00000  | 0.00000  |
| 4 | True     | False    | 1.00000  | -0.02401 | 0.94140  | 0.06531  | 0.92106  | -0.23255 | 0.77152  | -0.16399 | 0.52798  | -0.20275 | 0.56409  | -0.00712 |
| 4 |          |          |          |          |          |          |          |          |          |          |          |          |          |          |

# In [8]:

df.tail()

# Out[8]:

|     | 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 | columr |
|-----|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|--------|
| 346 | True     | False    | 0.83508  | 0.08298  | 0.73739  | -0.14706 | 0.84349  | -0.05567 | 0.90441  | -0.04622 | 0.89391  | 0.13130  | 0.81197  | 0.067  |
| 347 | True     | False    | 0.95113  | 0.00419  | 0.95183  | -0.02723 | 0.93438  | -0.01920 | 0.94590  | 0.01606  | 0.96510  | 0.03281  | 0.94171  | 0.073  |
| 348 | True     | False    | 0.94701  | -0.00034 | 0.93207  | -0.03227 | 0.95177  | -0.03431 | 0.95584  | 0.02446  | 0.94124  | 0.01766  | 0.92595  | 0.04€  |
| 349 | True     | False    | 0.90608  | -0.01657 | 0.98122  | -0.01989 | 0.95691  | -0.03646 | 0.85746  | 0.00110  | 0.89724  | -0.03315 | 0.89061  | -0.014 |
| 350 | True     | False    | 0.84710  | 0.13533  | 0.73638  | -0.06151 | 0.87873  | 0.08260  | 0.88928  | -0.09139 | 0.78735  | 0.06678  | 0.80668  | -0.003 |
| 4.6 |          |          |          |          |          | 1        |          |          |          |          |          |          |          |        |

```
In [9]:
 features_matrix=df.iloc[:,0:34]
 In [10]:
 target_vector=df.iloc[:,-1]
 In [11]:
 print('The Feature 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 Feature Matrix Has 351 Rows and 34 Column(s)
 The Target Matrix Has 351 Rows and 1 Column(s)
 In [12]:
 features_matrix_standardized=StandardScaler().fit_transform(features_matrix)
 In [18]:
 algorithm=LogisticRegression(penalty='12',dual=False,tol=1e-4,C=1.0,fit intercept=True,intercept scaling=1,class weight=None,random states
 Logistic_Regression_Model=algorithm.fit(features_matrix_standardized,target_vector)
 In [20]:
 0.59755, -0.44945, 0.60536, -0.38223, 0.8435600000000001, -0.38542, 0.58212, -0.32192, 0.56971, -0.29674, 0.36946, -0.47357, -0.29674, 0.36946, -0.47357, -0.29674, 0.36946, -0.47357, -0.29674, 0.36946, -0.47357, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0.29674, -0
                            0.56811,-0.51171,0.4107800000000003,-0.4616800000000003,0.21260,-0.3409,0.42267,-0.54487,0.18641,-0.453]]
 In [24]:
 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 [25]:
 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 [26]:
The Model says The probabilty of the observation we passed Belonging to class['b'] Is %s"""%(algorithm.predict_proba(observation)[0][0]))
The Model says The probabilty of the observation we passed Belonging to class['g'] Is %s"""%(algorithm.predict_proba(observation)[0][1]))
   The Model says The probabilty of the observation we passed Belonging to class['b'] Is 0.00777308403249255
   The Model says The probabilty of the observation we passed Belonging to class['g'] Is 0.9922269159675075
 In [ ]:
```