```
In [1]:
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
           import seaborn as sns
 In [2]: | from sklearn.linear_model import LogisticRegression
 In [3]: | df=pd.read_csv(r"1_ionosphere.csv")
           df
 Out[3]:
                       0.99539 -0.05889
                                          0.85243
                                                   0.02306
                                                            0.83398 -0.37708
                                                                                       0.03760 ...
                                                                                                   -0.511
                 1
                    0
                                                                                  1.1
                    0
                       1.00000
                                -0.18829
                                          0.93035
                                                  -0.36156
                                                            -0.10868
                                                                     -0.93597
                                                                              1.00000
                                                                                      -0.04549
                                                                                                   -0.265
              0
                 1
              1
                    0
                       1.00000 -0.03365
                                          1.00000
                                                   0.00485
                                                            1.00000 -0.12062
                                                                             0.88965
                                                                                       0.01198
                                                                                                   -0.402
                 1
                       1.00000
                                -0.45161
                                          1.00000
                                                   1.00000
                                                            0.71216 -1.00000
                                                                              0.00000
                                                                                       0.00000
                                                                                                    0.906
                                          0.94140
                                                            0.92106
                       1.00000
                                -0.02401
                                                   0.06531
                                                                    -0.23255
                                                                              0.77152
              3
                                                                                      -0.16399
                                                                                                   -0.651
                       0.02337
                                -0.00592
                                         -0.09924
                                                   -0.11949
                                                            -0.00763
                                                                              0.14706
                    0
                                                                     -0.11824
                                                                                       0.06637
                                                                                                   -0.015
            345
                    0
                       0.83508
                                 0.08298
                                          0.73739
                                                  -0.14706
                                                            0.84349
                                                                     -0.05567
                                                                              0.90441
                                                                                      -0.04622 ...
                                                                                                   -0.042
                 1
            346
                        0.95113
                                 0.00419
                                          0.95183
                                                  -0.02723
                                                            0.93438
                                                                     -0.01920
                                                                              0.94590
                                                                                       0.01606
                                                                                                    0.013
                 1
                                -0.00034
            347
                    0
                       0.94701
                                          0.93207 -0.03227
                                                            0.95177
                                                                     -0.03431
                                                                              0.95584
                                                                                       0.02446 ...
                                                                                                    0.031
                 1
            348
                       0.90608
                                -0.01657
                                          0.98122
                                                 -0.01989
                                                            0.95691
                                                                     -0.03646
                                                                              0.85746
                                                                                       0.00110 ...
                                                                                                   -0.020
                                          0.73638 -0.06151
            349
                       0.84710
                                 0.13533
                                                            0.87873
                                                                     0.08260
                                                                              0.88928
                                                                                      -0.09139 ...
                                                                                                   -0.15
           350 rows × 35 columns
           feature matrix=df.iloc[:,0:34]
 In [4]:
           target_vector=df.iloc[:,-1]
 In [5]: | feature_matrix.shape
 Out[5]: (350, 34)
 In [6]: target_vector.shape
 Out[6]: (350,)
 In [9]:
           from sklearn.preprocessing import StandardScaler
           fs=StandardScaler().fit transform(feature matrix)
In [12]:
```

```
In [13]: logr=LogisticRegression()
         logr.fit(fs,target_vector)
Out[13]: LogisticRegression()
In [17]: observation=[[1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,2
In [18]:
         prediction=logr.predict(observation)
         print(prediction)
         ['g']
In [19]: logr.classes_
Out[19]: array(['b', 'g'], dtype=object)
In [20]: logr.predict_proba(observation)[0][0]
Out[20]: 7.175815497362237e-12
In [21]: logr.predict_proba(observation)[0][1]
Out[21]: 0.999999999928242
In [33]:
         import re
         from sklearn.datasets import load digits
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.linear model import LogisticRegression
         from sklearn.model_selection import train_test_split
```

```
In [34]: | digit=load_digits()
         digit
Out[34]: {'data': array([[ 0., 0., 5., ..., 0., 0., 0.],
                 [0., 0., 0., \ldots, 10., 0., 0.],
                        0., 0., ..., 16., 9.,
                 [ 0.,
                        0., 1., ..., 6., 0., 0.],
                 [ 0.,
                 [0., 0., 2., ..., 12., 0., 0.],
                        0., 10., ..., 12., 1., 0.]]),
           'target': array([0, 1, 2, ..., 8, 9, 8]),
           'frame': None,
           'feature_names': ['pixel_0_0',
           'pixel_0_1',
            'pixel_0_2',
            'pixel_0_3',
            'pixel_0_4'
            'pixel_0_5',
            'pixel 0 6',
            'pixel_0_7',
            'pixel_1_0',
            'pixel_1_1',
In [48]: plt.figure(figsize=(69,69))
         for index,(image,label) in enumerate(zip(digit.data[0:5],digit.target[0:5])):
             plt.subplot(1,5,index+1)
             plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
             plt.title('Number:%i\n'%label,fontsize=15)
In [39]: |x_train,x_test,y_train,y_test=train_test_split(digit.data,digit.target,test_size
         print(x_train.shape)
In [40]:
         print(x test.shape)
         print(y train.shape)
         print(y_test.shape)
         (1257, 64)
         (540, 64)
         (1257,)
         (540,)
In [41]: logre=LogisticRegression(max_iter=10000)
         logre.fit(x_train,y_train)
Out[41]: LogisticRegression(max_iter=10000)
```

In [42]: |print(logre.predict(x\_test)) [9 1 9 1 7 5 0 6 3 1 8 9 9 2 8 9 5 7 7 0 7 2 4 4 2 4 4 4 6 5 3 5 4 8 9 5 6  $6\ 6\ 0\ 8\ 2\ 0\ 9\ 9\ 4\ 1\ 9\ 9\ 8\ 4\ 5\ 7\ 7\ 0\ 4\ 4\ 0\ 4\ 3\ 4\ 9\ 2\ 1\ 6\ 3\ 6\ 1\ 2\ 0\ 6\ 7\ 5\ 6$ 8 7 4 2 6 7 0 5 3 1 8 5 3 6 6 3 8 0 9 5 9 2 2 3 3 5 4 8 6 1 1 5 6 8 1 5 3 0 1 6 1 8 5 6 4 5 5 3 4 2 4 1 5 5 0 5 6 8 7 0 8 7 0 6 9 5 5 1 1 4 6 6 0 3 8 2 3 3 4 6 2 0 1 7 5 8 4 7 9 0 8 2 7 4 6 6 8 8 3 8 1 3 0 6 3 4 2 7 0 3 6 7 2 7 7 7 4 2 2 7 8 4 0 3 9 6 0 3 0 6 2 7 5 7 9 5 9 8 0 1 6 9 7 7 8 6 1 0 9 7 6 0 8 2 7 3 8 7 2 9 3 6 5 1 1 9 8 2 0 1 8 2 0 1 1 2 3 3 8 1 1 3 4 4 4 5 7 1 8 0 8 3 1 7 6 8 7 1 1 9 7 3 1 7 7 3 6 4 2 0 7 5 0 2 1 4 6 2 1 3 2 3 4 5 7 0 2 3 1 3 8 1 9 9 8 3 9 9 4 4 7 6 3 5 9 5 9 7 0 6 3 9 7 8 3 9 8 5 1 4 6 2 0 9 8 0 4 6 7 6 3 4 5 2 2 7 2 3 3 3 3 1 8 4 5 2 1 7 7 5 3 9 9 8 5 7 6 1 4 8 3 7 1 6 1 8 4 0 0 8 0 6 7 4 0 4 1 5 7 1 5 1 5 4 7 5 5 9 7 6 2 4 3 2 3 8 7 8 0 5 6 1 0 0 1 1 2 2 1 2 6 5 4 6 0 7 8 4 0 0 8 1 0 2 7 5 2 6 0 3 9 8 8 0 6 8 0 6 2 4 5 2 1 0 3 1 5 5 8 4 9 1 0 9 1 8 2 9 8 2 3 1 0 3 6 8 4 5 0 6 4 2 9 6 9 3 9 2 1 2 8 3 9 8 8 8 4 1 6 7 5 2 2 5 7 5 6 8 9 8 4 6 3 9 9 6 8 1 9 0 2 1 8 0 7 8 1 8 5 2 7 2 8 3 2 0 In [43]: |print(logre.score(x\_test,y\_test)) 0.9629629629629 In [ ]: