Logistic Rgression Using Excel

In [29]: import pandas as pd import numpy as np from sklearn.linear model import LogisticRegression from sklearn.preprocessing import StandardScaler 1 df=pd.read csv(r"C:\Users\yoshitha lakshmi\OneDrive\Desktop\python\ionosphere data.csv") In [30]: 2 df Out[30]: 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_l -0.17755 0 True False 0.99539 -0.05889 0.85243 0.02306 0.83398 -0.37708 0.03760 0.85243 1.00000 1 True False 1.00000 -0.18829 0.93035 -0.36156 -0.10868 -0.93597 1.00000 -0.04549 0.50874 -0.67743 C 2 True 1.00000 -0.03365 1.00000 0.73082 0.05346 False 1.00000 0.00485 -0.12062 0.88965 0.01198 3 1.00000 0.71216 0.00000 0.00000 True False -0.45161 1.00000 1.00000 -1.00000 0.00000 0.00000 True 1.00000 -0.02401 0.06531 0.92106 -0.23255 0.77152 -0.16399 0.52798 -0.20275 False 0.94140 5 True False 0.02337 -0.00592 -0.09924 -0.11949 -0.00763 -0.11824 0.14706 0.06637 0.03786 -0.06302 C -0.10602 6 True False 0.97588 0.94601 -0.20800 0.92806 -0.28350 0.85996 -0.273420.79766 -0.47929 C 7 False False 0.00000 0.00000 0.00000 0.00000 1.00000 -1.00000 0.00000 0.00000 -1.00000 -1.00000 C 8 True False 0.96355 -0.07198 1.00000 -0.14333 1.00000 -0.21313 1.00000 -0.36174 0.92570 -0.43569 9 -0.01864 True False -0.08459 0.00000 0.00000 0.00000 0.00000 0.11470 -0.26810 -0.45663-0.38172 1 00000 0 06655 0 18388 1 00000 1 00000 0.43107 1 00000 U 11310 10 Truc Falco 1 00000 በ ኃፓ፯ኃበ

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In [31]:
            1 pd.set option('display.max rows',100000000000)
            pd.set option('display.max columns',100000000000)
               pd.set option('display.width',95)
            1 print('This DataFrame has %d Rows and %d columns'%(df.shape))
In [32]:
          This DataFrame has 351 Rows and 35 columns
In [33]:
            1 df.head()
Out[33]:
              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
                  True
                            False
                                   0.99539
                                             -0.05889
                                                       0.85243
                                                                 0.02306
                                                                           0.83398
                                                                                    -0.37708
                                                                                              1.00000
                                                                                                       0.03760
                                                                                                                 0.85243
                                                                                                                          -0.17755
                                                                                                                                     0.59755
                                                       0.93035
                  True
                            False
                                   1.00000
                                             -0.18829
                                                                -0.36156
                                                                          -0.10868
                                                                                    -0.93597
                                                                                              1.00000
                                                                                                       -0.04549
                                                                                                                 0.50874
                                                                                                                          -0.67743
                                                                                                                                     0.34432
           2
                                   1.00000
                                             -0.03365
                                                       1.00000
                                                                 0.00485
                                                                                    -0.12062
                                                                                              0.88965
                                                                                                       0.01198
                                                                                                                           0.05346
                                                                                                                                     0.85443
                  True
                            False
                                                                           1.00000
                                                                                                                 0.73082
                                                                                              0.00000
                                                                                                                                     0.00000
           3
                  True
                            False
                                   1.00000
                                             -0.45161
                                                       1.00000
                                                                 1.00000
                                                                           0.71216
                                                                                    -1.00000
                                                                                                       0.00000
                                                                                                                 0.00000
                                                                                                                           0.00000
                  True
                            False
                                   1.00000
                                             -0.02401
                                                       0.94140
                                                                 0.06531
                                                                           0.92106
                                                                                    -0.23255
                                                                                              0.77152
                                                                                                       -0.16399
                                                                                                                 0.52798
                                                                                                                          -0.20275
                                                                                                                                     0.56409
In [34]:
            1 features matrix = df.iloc[:,0:34]
In [35]:
            1 target vector = df.iloc[:,-1]
In [36]:
            1 print('The Features Matrix Has %d Rows And %d Column(S)'%(features matrix.shape))
            2 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 [37]:
           1 | features matrix standardized = StandardScaler().fit transform(features matrix)
In [38]:
           1 algorithm = LogisticRegression(penalty='12',dual=False,tol=1e-4,C=1.0,fit intercept=True,intercept scaling=1,cla
In [47]:
           1 Logistic Regression Model=algorithm.fit(features matrix standardized,target vector)
In [48]:
              observation=[[1,0,0.99539,-0.05889,0.85242999999999,0.02306,0.83397999999999,
                            -0.37708,1.0,0.0376,0.85242999999999,-0.17755,0.59755,-0.44945,
                            0.60536, -0.38223, 0.8435600000000001, -0.38542, 0.58212, -0.32192, 0.56971
           3
                            ,-0.29674,0.36946,-0.47357,0.56811,-0.51171,0.4107800000000003,
                            -0.4616800000000003,0.21266,-0.3409,0.42267,-0.54487,0.18641,
           6
                            -0.45311
In [49]:
           1 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 [51]:
           1 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 [52]:
           1 print("""The Model says The probability of the observation We passed Belonging To class['b'] Is %s"""%(algorithm
           2 print()
             print("""The Model says The probability of the observation We passed Belonging To class['g'] Is %s"""%(algorithm
         The Model says The probability of the observation We passed Belonging To class['b'] Is 0.00777393160013784
```

The Model says The probability of the observation We passed Belonging To class['g'] Is 0.00777393160013784

localhost:8888/notebooks/Python Note/Logistics Regression.ipynb

Logistic Regression

(with out using Excel for visualisation)

```
In [66]:
              plt.figure(figsize=(20,4))
              for index,(image,label) in enumerate(zip(digits.data[0:5],digits.target[0:5])):
            3
                   plt.subplot(1,5,index+1)
                   plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
            4
                   plt.title('Training:%i\n'%label,fontsize=10)
            5
            6
                    Training:0
                                             Training:1
                                                                      Training:2
                                                                                              Training:3
                                                                                                                       Training:4
          2 -
          5 -
In [68]:
           1 from sklearn.model selection import train test split
           2 x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.30,random_state=2)
In [69]:
           1 print(x_train.shape)
          (1257, 64)
In [71]:
           1 print(y_train.shape)
          (1257,)
In [72]:
           1 print(x_test.shape)
          (540, 64)
```

```
1 print(y test.shape)
In [73]:
          (540,)
In [74]:
           1 from sklearn.linear model import LogisticRegression
           2 logisticRegr=LogisticRegression(max iter=10000)
           3 logisticRegr.fit(x train,y train)
Out[74]: LogisticRegression(max iter=10000)
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
           1 print(logisticRegr.predict(x test))
In [75]:
          \lceil 4 \ 0 \ 9 \ 1 \ 8 \ 7 \ 1 \ 5 \ 1 \ 6 \ 6 \ 7 \ 6 \ 1 \ 5 \ 5 \ 8 \ 6 \ 2 \ 7 \ 4 \ 6 \ 4 \ 1 \ 5 \ 2 \ 9 \ 5 \ 4 \ 6 \ 5 \ 6 \ 3 \ 4 \ 0 \ 9 \ 9
           8 4 6 8 8 5 7 9 8 9 6 1 7 0 1 9 7 3 3 1 8 8 8 9 8 5 8 4 9 3 5 8 4
           7 3 3 0 8 7 2 8 5 3 8 7 6 4 6 2 2 0 1 1 5 3 5 7 1 8 2 2 6 4 6 7 3 7 3 9 4
           7 0 3 5 1 5 0 3 9 2 7 3 2 0 8 1 9 2 1 5 1 0 3 4 3 0 8 3 2 2 7 3 1 6 7 2 8
           3 1 1 6 4 8 2 1 8 4 1 3 1 1 9 5 4 8 7 4 8 9 5 7 6 9 4 0 4 0 0 9 0 6 5 8 8
           3 7 9 2 0 8 2 7 3 0 2 1 9 2 7 0 6 9 3 1 1 3 5 2 5 5 2 1 2 9 4 6 5 5 5 9 7
           1 5 9 6 3 7 1 7 5 1 7 2 7 5 5 4 8 6 6 2 8 7 3 7 8 0 9 5 7 4 3 4
           4 1 3 1 2 5 1 4 0 3 1 5 5 7 4 0 1 0 9 5 5 5 4 0 1 8 6 2 1 1 1 7 9 6 7 9 7
           0496927210828657845786426930080667145
           6 9 7 2 8 5 1 2 4 1 8 8 7 6 0 8 0 6 1 5 7 8 0 4 1 4 5 9 2 2 3 9 1 3 9 3 2
           8 0 6 5 6 2 5 2 3 2 6 1 0 7 6 0 6 2 7 0 3 2 4 2 3 6 9 7 7 0 3 5 4 1 2 2 1
           2 7 7 0 4 9 8 5 6 1 6 5 2 0 8 2 4 3 3 2 9 3 8 9 9 5 9 0 3 4 7 9 8 5 7 5 0
           5 3 5 0 2 7 3 0 4 3 6 6 1 9 6 3 4 6 4 6 7 2 7 6 3 0 3 0 1 3 6 1 0 4 3 8 4
           3 3 4 8 6 9 6 3 3 0 5 7 8 9 1 5 3 2 5 1 7 6 0 6 9 5 2 4 4 7 2 0 5 6 2 0 8
           4 4 4 7 1 0 4 1 9 2 1 3 0 5 3 9 8 2 6 0 0 4
           1 score=logisticRegr.score(x test,y test)
In [85]:
              print(score)
```

0.9537037037037037

In []: