logistic regression2

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In [1]: import re
         from sklearn.datasets import load digits
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn import metrics
         %matplotlib inline
         digits = load_digits()
In [2]: print("Image Data Shape",digits.data.shape)
         print("Label Data Shape",digits.data.shape)
         Image Data Shape (1797, 64)
         Label Data Shape (1797, 64)
In [4]: plt.figure(figsize=(20,4))
         for index,(image,label)in enumerate(zip(digits.data[0:5],digits.target[0:5])):
             plt.subplot(1,5,index+1)
             plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
             plt.title('Training:%i/n'%label,fontsize=10)
                   Training:0/r
                                                                      Training:2/n
                                                                                                Training:3/n
                                                                                                                         Training:4/n
In [6]: from sklearn.model selection import train test split
         x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,test_size=0.30,random_state=2)
In [7]: print(x_train.shape)
         (1257, 64)
In [8]: print(y_train.shape)
         (1257,)
In [9]: print(x_test.shape)
         (540, 64)
In [10]: print(y_test.shape)
         (540,)
In [11]: from sklearn.linear_model import LogisticRegression
In [14]: logisticRegr = LogisticRegression(max_iter=10000)
         logisticRegr.fit(x_train,y_train)
Out[14]: 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.
In [15]: print(logisticRegr.predict(x_test))
         [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 3 1 3 8
            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
            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
          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 1 0 3 3 5
          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
          8065625232610760627032423697
                                                                  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]
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