## **Logistic Regression**

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In [1]: import numpy as np

macro avg

weighted avg

0.92

0.93

0.92

0.93

```
import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.linear_model import LogisticRegression
         from sklearn.metrics import confusion matrix, accuracy score, classification report
 In [2]: | train = pd.read csv("D:/BITS/DL/Assignments/Assignment1/mnist train.csv")
         test = pd.read csv("D:/BITS/DL/Assignments/Assignment1/mnist test.csv")
 In [3]: X train = train.iloc[:,1:].values
         y train = train.iloc[:,0].values
 In [4]: X test = test.iloc[:,1:].values
         y_test = test.iloc[:,0].values
 In [5]: | X_train, X_test = X_train/255.0, X test/255.0
 In [6]: | lr = LogisticRegression(max iter=1000, verbose=2)
         lr.fit(X train, y train)
          [Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
          [Parallel(n_jobs=1)]: Done 1 out of 1 | elapsed: 3.4min remaining:
                                      1 out of 1 | elapsed: 3.4min finished
          [Parallel(n jobs=1)]: Done
 Out[6]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=True,
                             intercept scaling=1, l1 ratio=None, max iter=1000,
                             multi_class='auto', n_jobs=None, penalty='12',
                             random_state=None, solver='lbfgs', tol=0.0001, verbose=2,
                             warm start=False)
         pred = lr.predict(X_test)
 In [7]:
 In [8]:
         acc = accuracy_score(y_test,pred)
         acc
Out[8]: 0.9256
 In [9]: sns.heatmap(confusion_matrix(y_test,pred),cmap='Blues_r',fmt='',annot=True,cbar=False)
         plt.ylabel("Correct Label")
         plt.xlabel("Predicted Label")
         plt.title("Confusion Matrix for Logistic Regression\nModel Accuracy={}\".format(acc*100))
Out[9]: Text(0.5, 1.0, 'Confusion Matrix for Logistic Regression\nModel Accuracy=92.56%')
                    Confusion Matrix for Logistic Regression
                          Model Accuracy=92.56%
            0 - 955
                                    10
                   1110
                       930
                            14
                                10
                                         12
                                                      4
                                             10
                                                 34
                        16
                           925
                                     23
                                             10
                                                 19
          Correct Label
                                921
                                         6
            4
                1
                                     0
                                                 6
                                                     30
                                        15
           S
                            35
                                    777
                                                 31
                8
                        8
                                 6
                                    16
                                        912
                                             2
                                                      0
            9
                                                 4
                                                      32
                        23
                                            947
                                                     12
                                     29
                                                     924
                              Predicted Label
In [11]: print("Classification Report for Logistic Regression:\n{}".format(classification report(y test,pred)))
         Classification Report for Logistic Regression:
                        precision
                                     recall f1-score
                                                          support
                     0
                             0.95
                                        0.97
                                                   0.96
                                                              980
                     1
                             0.96
                                        0.98
                                                   0.97
                                                             1135
                     2
                             0.93
                                        0.90
                                                   0.91
                                                             1032
                                                             1010
                     3
                             0.90
                                        0.92
                                                   0.91
                             0.94
                                        0.94
                                                   0.94
                                                              982
                     5
                             0.90
                                                              892
                                        0.87
                                                   0.88
                     6
                             0.94
                                        0.95
                                                   0.95
                                                              958
                     7
                             0.93
                                        0.92
                                                   0.93
                                                             1028
                     8
                             0.88
                                        0.88
                                                  0.88
                                                              974
                     9
                             0.91
                                        0.92
                                                  0.91
                                                             1009
             accuracy
                                                   0.93
                                                            10000
```

10000

0.92

0.93