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```
In [3]: #Author : Sayali Kudale

#loading the required libraries

import scipy.io
import numpy as np
import pandas as pd
from sklearn.multiclass import OneVsRestClassifier
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report
import time
```

```
In [4]: #Data Loading

mat = scipy.io.loadmat('Data/USPS_all.mat')
    df_features = pd.DataFrame(mat['fea'])
    df_gnd = pd.DataFrame(mat['gnd'])

#Split data into training and testing set

train_fea= df_features.iloc[:7291]
    test_fea= df_features.iloc[-2007:]
    train_dec= df_gnd.iloc[:7291]
    test_dec= df_gnd.iloc[-2007:]
```

## **SVM Model - Linear Kernel**

```
In [5]: # Creating the SVM model
    model = OneVsRestClassifier(SVC(kernel='linear'))

trainStart = time.time()
    # Fitting the model with training data
    model.fit(train_fea, train_dec)

trainEnd = time.time()

print("Training time : ", trainEnd-trainStart)
```

Training time : 20.14596390724182

```
In [6]: # Making a prediction on the test set
prediction = model.predict(test_fea)
```

```
In [7]: # Evaluating the model
    score=accuracy_score(test_dec, prediction) * 100
    print("Accuracy Score : %s" %score)
```

Accuracy Score : 91.6292974589

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```
In [10]: classification_report=classification_report(test_dec, prediction)
In [11]: print('Classification report : \n',classsification report)
          Classification report :
                         precision
                                       recall
                                                f1-score
                                                            support
                     1
                              0.95
                                        0.97
                                                   0.96
                                                                359
                                                    0.98
                     2
                              0.98
                                        0.97
                                                                264
                     3
                              0.87
                                        0.85
                                                   0.86
                                                                198
                     4
                              0.86
                                        0.86
                                                   0.86
                                                                166
                     5
                              0.86
                                        0.89
                                                   0.87
                                                                200
                              0.89
                                        0.89
                                                                160
                     6
                                                   0.89
                     7
                              0.96
                                        0.95
                                                   0.96
                                                                170
                     8
                              0.93
                                        0.90
                                                   0.92
                                                                147
                     9
                              0.88
                                        0.87
                                                   0.87
                                                                166
                    10
                              0.91
                                        0.93
                                                   0.92
                                                               177
          avg / total
                              0.92
                                        0.92
                                                   0.92
                                                               2007
In [16]: from sklearn.metrics import confusion matrix
          cm= confusion matrix(test dec, prediction)
          print(cm)
          [[350
                       2
                           1
                                             0
                                                 3
                                                      0]
              0 255
                       1
                           3
                                        2
                                             0
                                                 1
                                                      0]
           [
              5
                   0 168
                           6
                                9
                                    1
                                                      01
           [
                       4 142
              2
                   0
                                2
                                    8
                                        0
                                             3
                                                      2]
                  2
                       7
                           0 177
                                    1
                                        2
                                             2
              1
                                                 3
                                                      5]
           [
              3
                  0
                       0
                           7
                                3 143
                                        0
                                             0
           [
                                                      4]
              0
                  0
                       3
                           0
                                2
                                    2 162
                                                      0]
           [
             2
                       2
                         2
           [
                  0
                                    0
                                        0 133
                                                      41
              5
                  0
                       4
                                1
                                             0 144
                           4
                                    6
                                        1
                                                      11
                       2
                                    0
                                                 3 165]]
                                        0
                                             3
```

## **SVM Model - Polynomial Kernel function**

```
In [26]: # Creating the SVM model
    modelPoly = OneVsRestClassifier(SVC(kernel='poly'))
    trainStart = time.time()
    # Fitting the model with training data
    modelPoly.fit(train_fea, train_dec)
    trainEnd = time.time()

    print("Training time : ", trainEnd-trainStart)

Training time : 18.499656915664673

In [18]: # Making a prediction on the test set
    predictionPoly = modelPoly.predict(test_fea)
```

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In [19]: # Evaluating the model
 scorePoly=accuracy\_score(test\_dec, predictionPoly) \* 100
 print("Accuracy Score : %s" %scorePoly)

Accuracy Score : 93.2735426009

In [20]: classsification\_reportPoly=classification\_report(test\_dec, predictionPol
 y)
 print('Classification report of Polynomial kernel function: \n',classsif
 ication\_reportPoly)

Classification report of Polynomial kernel function: precision recall f1-score support 1 0.94 359 0.99 0.96 2 0.98 0.97 0.98 264 3 0.90 0.89 0.90 198 4 0.92 0.90 0.91 166 5 0.88 0.93 0.90 200 6 0.90 0.91 0.90 160 7 0.96 0.92 0.94 170 8 0.96 0.91 147 0.94 9 0.94 0.88 0.91 166 10 0.93 0.95 0.94 177 avg / total 0.93 0.93 0.93 2007

In [21]: cm\_poly= confusion\_matrix(test\_dec, predictionPoly)
 print("Confusion Matrix\n",cm\_poly)

Confusion Matrix 2 0 0 0 0 0 [[354 0 2 1] 1 0 255 1 4 0 3 0 0 0] 6 0 177 3 5 2 1 1 3 0] 4 149 3 0 0 6 2 11 [ [ 1 2 6 0 185 0 2 1 3 ] 5 0 0 4 1 146 0 0 4] 4 0 3 0 3 2 157 0 1 01 [ 0 0 2 0 6 0 0 134 2 3] [ 4 1 2 5 1 6 0 0 146 1] [ 1 0 3 0 1 1 2 169]]