

# Project - Devanagari Handwritten Character Classification

December 7, 2019

```
[0]: from google.colab import drive
drive.mount('/content/drive')

[0]: !unzip -q "/content/drive/My Drive/Colab Notebooks/
↳DevanagariHandwrittenCharacterDataset.zip"

[0]: import numpy as np
import cv2
from google.colab.patches import cv2_imshow
import os
import random
import matplotlib.pyplot as plt
from PIL import Image
from sklearn.svm import SVC
from sklearn.metrics import classification_report
import shutil
from sklearn.metrics import accuracy_score

[0]: #Extract names of classes from dataset folders
def fetchName(folderName):
    if folderName[0] == 'c':
        return str(folderName[(folderName.index('_'),folderName.index('_')+1))+1:
↳])
    else:
        return str("d"+folderName[folderName.index('_')+1:])

[0]: #Function to perform fetch the entire dataset
wholeDataset = []
def fetchDataset(rootPath):
    folders = os.listdir(rootPath)
    for fold in folders:
        files = os.listdir(rootPath+fold)
        for eachFile in files:
            openImage = cv2.imread(rootPath+fold+"/"+eachFile)
            grayscale = cv2.cvtColor(openImage, cv2.COLOR_BGR2GRAY)
            ret,thresh1 = cv2.threshold(grayscale,100,255,cv2.THRESH_BINARY)
            singleLineImage = np.array(thresh1).flatten()
```

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        replacedValues = [1 if singleLineImage[i]==255 else_
↪singleLineImage[i] for i in range(len(singleLineImage))]
        wholeDataset.append([replacedValues,fetchName(fold)])

```

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[0]: #Function defined in such a way that each fold can be performed seperately
startPoint = int(len(wholeDataset)/5)
def crossValidation(k,currentFold):
    for i in range(k):
        if i == (currentFold-1):
            print("(" +str(startPoint*i)+", "+str(startPoint*(i+1)-1)+")")
            performingFold([startPoint*i,startPoint*(i+1)-1],i)

```

```

[0]: #Creating train and test Arrays and performing SVM
def performingFold(startEndPoints,k):
    trainDataset = []
    testDataset = []
    for j in range(len(wholeDataset)):
        if j >= startEndPoints[0] and j <= startEndPoints[1]:
            testDataset.append(wholeDataset[j])
        else:
            trainDataset.append(wholeDataset[j])
    performSVM(trainDataset, testDataset, k)

```

```

[0]: #Collecting the Dataset into a Single Array
fetchDataset('DevanagariHandwrittenCharacterDataset/')
#Shuffling the Dataset
random.shuffle(wholeDataset)

```

```

[0]: accuracyTrace = []
def performSVM(trainDataset, testDataset, k):
    print("Performing Fold-{}".format(k+1))
    print("-----")
    x_train,y_train,x_test,y_test = [],[],[],[]
    for p in range(len(trainDataset)):
        x_train.append(trainDataset[p][0])
        y_train.append(trainDataset[p][1])
    for q in range(len(testDataset)):
        x_test.append(testDataset[q][0])
        y_test.append(testDataset[q][1])
    print(np.shape(x_train),np.shape(y_train),np.shape(x_test),np.shape(y_test))

    #Building a Support Vector Machine having Gaussian Kernel
    model = SVC(kernel='rbf',decision_function_shape='ovo',gamma='scale')
    model.fit(x_train, y_train)

    #Predicting
    y_pred = model.predict(x_test)

    #Showing Entire Fold Accuracy as well as Class Wise

```

```

print(classification_report(y_test, y_pred))

#Storing Accuracy
accuracyTrace.append(accuracy_score(y_test,y_pred))
print("Fold Accuracy: {}".format(accuracy_score(y_test,y_pred)))

```

```

[0]: #5-Fold Cross Validation
#Fold No. 1
crossValidation(5,1)

```

(0,18399)

Performing Fold-1

-----  
(73600, 1024) (73600,) (18400, 1024) (18400,)

	precision	recall	f1-score	support
adna	0.89	0.96	0.92	368
ba	0.89	0.84	0.87	404
bha	0.91	0.91	0.91	405
cha	0.91	0.95	0.93	454
chha	0.91	0.91	0.91	383
chhya	0.91	0.92	0.91	400
d0	0.98	0.99	0.99	385
d1	0.97	0.99	0.98	415
d2	0.94	0.96	0.95	408
d3	0.96	0.95	0.95	403
d4	0.97	0.98	0.97	383
d5	0.97	0.99	0.98	409
d6	0.98	0.97	0.97	461
d7	0.99	0.98	0.98	374
d8	0.97	0.99	0.98	403
d9	0.98	0.98	0.98	407
da	0.90	0.88	0.89	394
daa	0.92	0.92	0.92	368
dha	0.90	0.90	0.90	397
dhaa	0.95	0.93	0.94	374
ga	0.94	0.90	0.92	386
gha	0.86	0.92	0.89	409
gya	0.94	0.94	0.94	415
ha	0.93	0.92	0.93	377
ja	0.94	0.95	0.95	371
jha	0.98	0.95	0.97	427
ka	0.96	0.96	0.96	385
kha	0.94	0.90	0.92	444
kna	0.94	0.90	0.92	397
la	0.98	0.94	0.96	387
ma	0.90	0.89	0.89	413
motosaw	0.89	0.94	0.92	401

na	0.93	0.91	0.92	439
pa	0.86	0.92	0.89	413
patalosaw	0.87	0.90	0.88	389
petchiryakha	0.91	0.91	0.91	384
pha	0.96	0.97	0.96	423
ra	0.96	0.95	0.96	413
taamatar	0.94	0.95	0.94	385
tabala	0.97	0.93	0.95	374
tha	0.89	0.87	0.88	429
thaa	0.95	0.94	0.94	407
tra	0.92	0.90	0.91	378
waw	0.90	0.91	0.90	379
yaw	0.86	0.86	0.86	390
yna	0.96	0.90	0.93	390
accuracy			0.93	18400
macro avg	0.93	0.93	0.93	18400
weighted avg	0.93	0.93	0.93	18400

Fold Accuracy: 0.931141304347826

```
[0]: #Fold No. 2
crossValidation(5,2)
```

(18400,36799)

Performing Fold-2

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(73600, 1024) (73600,) (18400, 1024) (18400,)

	precision	recall	f1-score	support
adna	0.90	0.95	0.93	432
ba	0.88	0.86	0.87	416
bha	0.93	0.90	0.91	413
cha	0.90	0.93	0.92	372
chha	0.93	0.89	0.91	404
chhya	0.92	0.93	0.93	424
d0	0.97	0.99	0.98	378
d1	0.97	0.99	0.98	388
d2	0.94	0.97	0.95	396
d3	0.96	0.96	0.96	422
d4	0.97	0.98	0.98	387
d5	0.97	0.98	0.98	393
d6	0.98	0.98	0.98	383
d7	0.99	0.97	0.98	392
d8	0.97	0.98	0.97	406
d9	0.98	1.00	0.99	412
da	0.86	0.88	0.87	405

daa	0.88	0.89	0.89	390
dha	0.88	0.89	0.89	421
dhaa	0.92	0.92	0.92	398
ga	0.94	0.93	0.94	396
gha	0.85	0.87	0.86	393
gya	0.92	0.94	0.93	385
ha	0.92	0.94	0.93	407
ja	0.94	0.93	0.94	381
jha	0.98	0.95	0.97	374
ka	0.97	0.97	0.97	402
kha	0.91	0.94	0.93	397
kna	0.92	0.86	0.89	370
la	0.95	0.93	0.94	421
ma	0.89	0.86	0.88	383
motosaw	0.90	0.93	0.91	381
na	0.89	0.92	0.91	384
pa	0.84	0.90	0.87	378
patalosaw	0.88	0.87	0.88	416
petchiryakha	0.90	0.94	0.92	405
pha	0.97	0.95	0.96	377
ra	0.95	0.94	0.95	394
taamatar	0.97	0.92	0.95	426
tabala	0.94	0.95	0.95	442
tha	0.88	0.84	0.86	409
thaa	0.96	0.94	0.95	417
tra	0.89	0.90	0.89	393
waw	0.91	0.88	0.89	423
yaw	0.89	0.84	0.86	420
yna	0.95	0.93	0.94	394
accuracy			0.93	18400
macro avg	0.93	0.93	0.93	18400
weighted avg	0.93	0.93	0.93	18400

Fold Accuracy: 0.9265217391304348

```
[0]: #Fold No. 3
crossValidation(5,3)
```

(36800,55199)

Performing Fold-3

-----

(73600, 1024) (73600,) (18400, 1024) (18400,)

	precision	recall	f1-score	support
adna	0.91	0.96	0.93	390
ba	0.88	0.85	0.87	406

bha	0.93	0.93	0.93	368
cha	0.92	0.96	0.94	374
chha	0.92	0.88	0.90	392
chhya	0.94	0.91	0.93	438
d0	0.99	1.00	0.99	422
d1	0.97	0.99	0.98	406
d2	0.95	0.95	0.95	400
d3	0.94	0.95	0.95	387
d4	0.96	0.99	0.98	433
d5	0.98	0.99	0.99	387
d6	0.97	0.97	0.97	409
d7	0.98	0.99	0.99	410
d8	0.97	0.97	0.97	390
d9	0.99	0.99	0.99	403
da	0.92	0.89	0.91	404
daa	0.86	0.92	0.89	371
dha	0.90	0.89	0.90	379
dhaa	0.92	0.94	0.93	422
ga	0.95	0.92	0.94	423
gha	0.89	0.90	0.90	409
gya	0.95	0.92	0.93	433
ha	0.94	0.91	0.93	435
ja	0.93	0.95	0.94	408
jha	0.97	0.98	0.98	427
ka	0.96	0.97	0.96	377
kha	0.95	0.89	0.92	403
kna	0.92	0.88	0.90	399
la	0.97	0.94	0.96	391
ma	0.92	0.92	0.92	409
motosaw	0.86	0.93	0.89	352
na	0.93	0.90	0.91	375
pa	0.87	0.91	0.89	433
patalosaw	0.89	0.87	0.88	397
petchiryakha	0.88	0.93	0.91	393
pha	0.96	0.96	0.96	410
ra	0.97	0.95	0.96	382
taamatar	0.94	0.96	0.95	400
tabala	0.93	0.95	0.94	413
tha	0.87	0.85	0.86	356
thaa	0.94	0.94	0.94	377
tra	0.90	0.93	0.91	401
waw	0.90	0.87	0.88	391
yaw	0.89	0.88	0.89	403
yna	0.96	0.93	0.95	412
accuracy			0.93	18400
macro avg	0.93	0.93	0.93	18400
weighted avg	0.93	0.93	0.93	18400

Fold Accuracy: 0.9323369565217391

```
[30]: #Fold No. 4
      crossValidation(5,4)
```

(55200,73599)

Performing Fold-4

-----  
(73600, 1024) (73600,) (18400, 1024) (18400,)

	precision	recall	f1-score	support
adna	0.92	0.94	0.93	422
ba	0.89	0.88	0.88	392
bha	0.91	0.91	0.91	411
cha	0.91	0.94	0.92	409
chha	0.93	0.93	0.93	404
chhya	0.93	0.97	0.95	360
d0	0.99	0.99	0.99	395
d1	0.98	0.99	0.99	406
d2	0.94	0.97	0.95	381
d3	0.96	0.96	0.96	400
d4	0.97	0.98	0.98	425
d5	0.98	0.97	0.98	446
d6	0.99	0.96	0.97	353
d7	0.99	1.00	0.99	421
d8	0.97	0.98	0.98	402
d9	0.99	0.99	0.99	371
da	0.91	0.89	0.90	377
daa	0.90	0.93	0.91	427
dha	0.87	0.87	0.87	389
dhaa	0.95	0.94	0.95	414
ga	0.95	0.92	0.93	428
gha	0.88	0.86	0.87	413
gya	0.94	0.93	0.94	375
ha	0.91	0.94	0.93	378
ja	0.94	0.93	0.94	414
jha	0.97	0.96	0.97	364
ka	0.97	0.97	0.97	408
kha	0.91	0.93	0.92	373
kna	0.94	0.91	0.92	398
la	0.97	0.92	0.94	428
ma	0.89	0.84	0.86	389
motosaw	0.89	0.92	0.90	437
na	0.93	0.91	0.92	413
pa	0.83	0.93	0.87	405
patalosaw	0.88	0.85	0.87	372

petchiryakha	0.89	0.91	0.90	406
pha	0.96	0.94	0.95	358
ra	0.97	0.95	0.96	402
taamatar	0.97	0.95	0.96	408
tabala	0.93	0.96	0.95	400
tha	0.89	0.88	0.88	384
thaa	0.96	0.96	0.96	403
tra	0.91	0.90	0.90	415
waw	0.88	0.87	0.88	391
yaw	0.87	0.87	0.87	415
yna	0.95	0.94	0.94	418
accuracy			0.93	18400
macro avg	0.93	0.93	0.93	18400
weighted avg	0.93	0.93	0.93	18400

Fold Accuracy: 0.9309782608695653

```
[31]: #Fold No. 5
      crossValidation(5,5)
```

(73600,91999)

Performing Fold-5

```
-----
(73600, 1024) (73600,) (18400, 1024) (18400,)
      precision    recall  f1-score   support

 adna      0.90      0.96      0.93      388
  ba      0.86      0.86      0.86      382
 bha      0.93      0.92      0.92      403
  cha      0.92      0.93      0.92      391
 chha      0.92      0.88      0.90      417
 chhya      0.95      0.94      0.95      378
   d0      0.98      1.00      0.99      420
   d1      0.96      0.99      0.97      385
   d2      0.94      0.98      0.96      415
   d3      0.96      0.96      0.96      388
   d4      0.97      0.97      0.97      372
   d5      0.96      0.98      0.97      365
   d6      0.98      0.97      0.97      394
   d7      0.98      0.99      0.99      403
   d8      0.97      0.98      0.98      399
   d9      0.98      0.99      0.98      407
   da      0.92      0.92      0.92      420
  daa      0.90      0.89      0.90      444
  dha      0.89      0.87      0.88      414
 dhaa      0.94      0.93      0.93      392
```



ga	0.93	0.93	0.93	367
gha	0.86	0.86	0.86	376
gya	0.93	0.95	0.94	392
ha	0.92	0.90	0.91	403
ja	0.96	0.94	0.95	426
jha	0.98	0.96	0.97	408
ka	0.98	0.96	0.97	428
kha	0.94	0.91	0.92	383
kna	0.93	0.87	0.90	436
la	0.96	0.95	0.96	373
ma	0.89	0.89	0.89	406
motosaw	0.90	0.94	0.92	429
na	0.91	0.91	0.91	389
pa	0.83	0.91	0.86	371
patalosaw	0.90	0.87	0.88	426
petchiryakha	0.88	0.94	0.91	412
pha	0.97	0.95	0.96	432
ra	0.96	0.91	0.94	409
taamatar	0.96	0.97	0.96	381
tabala	0.96	0.95	0.96	371
tha	0.90	0.84	0.87	422
thaa	0.95	0.96	0.95	396
tra	0.91	0.91	0.91	413
waw	0.90	0.89	0.89	416
yaw	0.86	0.90	0.88	372
yna	0.96	0.94	0.95	386
accuracy			0.93	18400
macro avg	0.93	0.93	0.93	18400
weighted avg	0.93	0.93	0.93	18400

Fold Accuracy: 0.9302717391304348

```
[35]: print("Average Accuracy: {}".format(np.mean(accuracyTrace)))
```

Average Accuracy: 0.93025