assignment10

December 6, 2018

- 1 Assignment09
- 2 ID: 20155093
- 3 Name: Sonjeongseol
- 4 Github Link: https://github.com/wjdtjf1234/assignment01.gitű
- 5 Importing APIs

6 Reading data from mnist train&test data sets

```
In [3]: h_data_train = open(file_data_train, "r")
    h_data_test = open(file_data_test, "r")

data_train = h_data_train.readlines()
    data_test = h_data_test.readlines()

h_data_train.close()
    h_data_test.close()

size_row = 28  # height of the image
    size_col = 28  # width of the image
```

```
num_train = len(data_train) # number of training images
num_test = len(data_test) # number of testing images
```

7 Arrays for storing labels&image data

8 Data normalizing & assigning labels for each data

```
In [5]: def label_assign(n):
           count = 0
            for line in data_train:
                line_data = line.split(',')
                          = line_data[0]
                label
                im_vector = np.asfarray(line_data[1:])
                im_vector = normalize(im_vector)
                list_label_train[count]
                                           = label
                list_image_train[:, count] = im_vector
                                                          #image sets
                count += 1
            for x in range(count):
                if(list_label_train[x]!=n):
                    list_label_train[x]=-1
                    list_label_train[x]=1
            count = 0
            for line in data_test:
                line_data = line.split(',')
                label=line_data[0]
                im_vector = np.asfarray(line_data[1:])
                           = normalize(im_vector)
                im_vector
                list_label_test[count]
                                          = label
                list_label_test2[count]
                                          = label
                list_image_test[:, count] = im_vector
                count += 1
            for x in range(count):
                if(list_label_test[x]!=n):
                    list_label_test[x]=-1
                else:
                    list_label_test[x]=1
```

9 Generating random normally distributed vectors

10 Some Executions to get model parameters from training data via pseudo-inverse

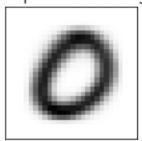
11 Defining a Bi-partitioning function

```
In [8]: def bpf(n,a):
            tp_average=np.zeros((size_row*size_col),dtype=float)
            d_t=np.zeros((1,a),dtype=float)
            table=np.zeros(11,dtype=int)
            label_test=np.empty(num_test,dtype=int)
            mp=ps_inv(a)
            for i in range(num_test):
                for 1 in range(a):
                    d_t[0,1]=np.inner(r[:,1],list_image_test[:,i])
                d=np.matmul(d_t,mp)
                if(d>=0):
                    if(list_label_test[i] == 1):
                         table[n] += 1
                         for j in range(size_row*size_col):
                             tp_average[j]+=list_image_test[j, i]
                         table[list_label_test2[i]-1]+=1
                                                              #FP
                elif(d<0):</pre>
                    if(list_label_test[i]==1):
                        table[10] += 1
                                                 #FN
```

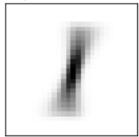
12 A result using 100 random vectors

In [11]: pf(100)

true positive average

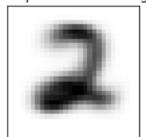


886 12 7 1 10 24 6 4 5 0 94



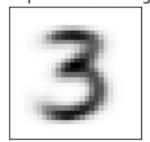
0 1054 1 3 8 5 23 11 1 0 108

true positive average

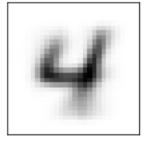


2 0 627 4 3 7 8 0 3 0 421

true positive average

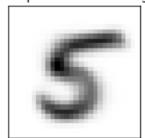


1 20 0 610 50 1 7 5 4 0 417



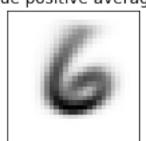
1 5 0 0 574 15 14 1 11 0 414

true positive average



1 0 11 1 0 333 1 3 1 0 590

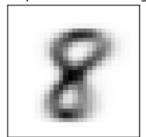
true positive average



4 36 10 5 18 0 783 5 1 0 212

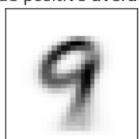
0 10 11 0 14 1 0 769 8 0 264

true positive average



18 21 6 2 39 7 3 0 326 0 668

true positive average



0 0 3 24 11 0 41 0 0 360 653 13 The confusion matrix M, which of partition function using 100 parameters:

14 886/12/7/1/10/24/6/4/5/0

15 0/1054/1/3/8/5/23/11/1/0

16 2/0/627/4/3/7/8/0/3/0

17 1/20/0/610/50/1/7/5/4/0

18 1/5/0/0/574/15/14/1/11/0

19 1/0/11/1/0/333/1/3/1/0

20 4/36/10/5/18/0/783/5/1/0

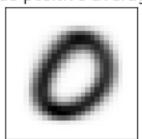
21 0/10/11/0/14/1/0/769/8/0

22 18/21/6/2/39/7/3/0/326/0

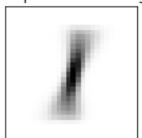
23 0/0/3/24/11/0/41/0/0/360

In [12]: pf(300)

true positive average



918 15 6 1 6 20 10 5 6 0 62 true positive average

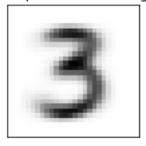


0 1073 2 7 8 5 17 13 2 0 81

true positive average

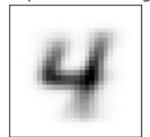


3 0 686 6 3 8 3 1 0 356



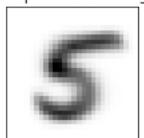
1 17 0 672 41 1 12 4 3 0 345

true positive average



1 5 0 0 718 15 11 3 16 0 270

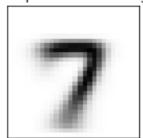
true positive average



1 1 6 0 0 408 0 15 1 0 508

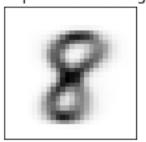
4 46 4 6 18 0 814 8 0 0 175

true positive average



0 10 13 0 15 0 0 798 14 0 232

true positive average



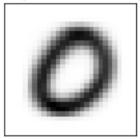
12 22 6 6 46 5 1 0 395 0 600

0 3 3 25 14 0 51 2 0 533 478

- 24 The confusion matrix M, which of partition function using 300 parameters:
- 25 918/15/6/1/6/20/10/5/6/0
- 26 0/1073/2/7/8/5/17/13/2/0
- 27 3/0/686/6/3/8/3/3/1/0
- 28 1/17/0/672/41/1/12/4/3/0
- 29 1/5/0/0/718/15/11/3/16/0
- 30 1/1/6/0/0/408/0/15/1/0
- 31 4/46/4/6/18/0/814/8/0/0
- 32 0/10/13/0/15/0/0/798/14/0
- 33 12/22/6/6/46/5/1/0/395/0
- 34 0/3/3/25/14/0/51/2/0/533

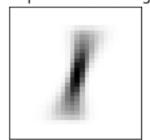
In [10]: pf(600)

true positive average



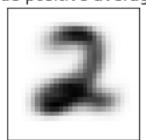
922 15 7 0 9 16 8 6 7 0 58

true positive average



0 1074 2 8 7 6 15 13 2 0 84

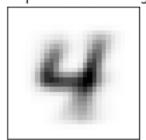
true positive average



2 0 678 8 5 11 1 4 3 0 368

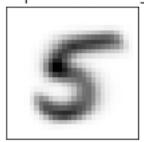
2 25 0 672 41 2 9 3 4 0 346

true positive average



1 8 0 0 751 18 15 4 16 0 240

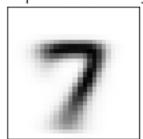
true positive average



0 1 6 4 0 461 1 25 2 0 458

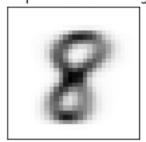
4 51 5 7 18 0 812 7 0 0 169

true positive average

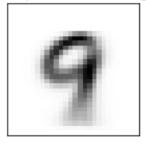


0 8 14 1 15 2 0 804 26 0 228

true positive average



14 25 9 7 49 8 3 0 404 0 594



0 4 4 42 17 0 60 4 0 569 442

- 35 The confusion matrix M, which of partition function using 600 parameters:
- 36 922/15/7/0/9/16/8/6/7/0
- 37 0/1074/2/8/7/6/15/13/2/0
- 38 2/0/678/8/5/11/1/4/3/0
- 39 2/25/0/672/41/2/9/3/4/0
- 40 1/8/0/0/751/18/15/4/16/0
- 41 0/1/6/4/0/461/1/25/2/0
- 42 4/51/5/7/18/0/812/7/0/0
- 43 0/8/14/1/15/2/0/804/26/0
- 44 14/25/9/7/49/8/3/0/404/0
- 45 0/4/4/42/17/0/60/4/0/569
- 46 The best F1 Score can be found when p = 600, and the value is 0.786.