

Assignment08

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- Description: Binary-classification using MNIST data
- github: https://github.com/mydream757/Computer_Vision

1. Import libraries

- import needed libraries.

```
In [1]: import matplotlib.pyplot as plt
import numpy as np
import numpy.linalg as lin
from matplotlib.image import imread
```

2. Read data from CSV files

- get ready for using MNIST data

```
In [2]: file_data_train = "mnist_train.csv"
file_data_test  = "mnist_test.csv"

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
```

3. Define functions

- the function of data normalizing

```
In [3]: # normalize the values of the input data to be [0, 1]
def normalize(data):
    data_normalized = (data - min(data)) / (max(data) - min(data))
    return(data_normalized)
```

- the function of computing distance

```
In [4]: def distance(x, y):

    d = (x - y) ** 2
    s = np.sum(d)
    # r = np.sqrt(s)
    return(s)
```

4. Define classifier

- make containers which contain MNIST image data

```
In [5]: #make a matrix each column of which represents an images
list_image_train = np.empty((num_train, size_row * size_col), dtype=float)
list_label_train = np.empty(num_train, dtype=int)

list_image_test = np.empty((num_test, size_row * size_col), dtype=float)
list_label_test = np.empty(num_test, dtype=int)
```

- parse the data sets

```
In [6]: count = 0
for line in data_train:

    line_data = line.split(',')
    label = line_data[0]
    im_vector = np.asfarray(line_data[1:])
    im_vector = normalize(im_vector)

    list_label_train[count] = label
    list_image_train[count,:] = im_vector

    count += 1

count = 0
for line in data_test:

    line_data = line.split(',')
    label = line_data[0]
    im_vector = np.asfarray(line_data[1:])
    im_vector = normalize(im_vector)

    list_label_test[count] = label
```

```
list_image_test[count,:] = im_vector

count += 1
```

- compute average images of each digits and label 1 or -1

```
In [7]: im_average = np.zeros((10, size_col*size_row), dtype=float)
        im_count   = np.zeros(10, dtype=int)

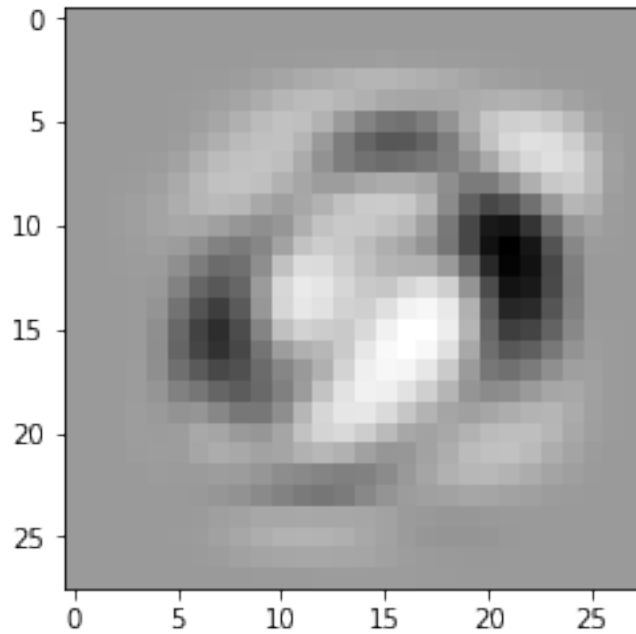
        for i in range(num_train):
            im_average[list_label_train[i],:] += list_image_train[i,:]
            im_count[list_label_train[i]] += 1

        im_label = np.zeros((10,1), dtype=float)
        for i in range(10):
            im_average[i,:] /= im_count[i]
            #label = 0 : 1, others : -1
            if i==0:
                im_label[0,0] = 1
            else:
                im_label[i,0] = -1
```

- compute coefficient.

```
In [8]: #compute coefficient using average image vectors
        inverse = lin.pinv(im_average)
        coefficient = np.dot(inverse, im_label)

        #plot the coefficient image
        image_coeff = coefficient[:,0].reshape((size_row, size_col))
        plt.figure()
        plt.imshow(image_coeff, cmap='Greys', interpolation='None')
        plt.show()
```



5. Evaluate the classifier using the test data

- ready for test

```
In [9]: #Ready for test
        experiment_label_test = np.empty(num_test, dtype=int)
        experiment_average_test = np.zeros((4,size_row * size_col), dtype=float)
        experiment_count_test = np.zeros(4, dtype=int)
```

- label 1 or -1 to test data and check TP, FP, FN, TN

```
In [10]: #Evaluate my classifier using MNIST test set
        result = np.dot(list_image_test, coefficient)
        num = 0
        for i in range(result.size):
            #sign(f(x))
            if result[i] >= 0:
                experiment_label_test[i] = 1
            else:
                experiment_label_test[i] = -1
            #check TP, FP, TN, FN
            if experiment_label_test[i] == 1 and list_label_test[i] == 0:
                num = 0 #then TP
            elif experiment_label_test[i] == 1 and list_label_test[i] !=0:
                num = 1 #then FP
            elif experiment_label_test[i] == -1 and list_label_test[i] == 0:
```

```

        num = 2 #then FN
    elif experiment_label_test[i] == -1 and list_label_test[i] !=0:
        num = 3 #then TN
    experiment_average_test[num,:] += list_image_test[i,:]
    experiment_count_test[num] += 1

```

- plot the result

```

In [11]: plt.figure()
        for i in range(4):
            experiment_average_test[i, :] /= experiment_count_test[i]
            title = ['TP', 'FP', 'FN', 'TN']
            plt.subplot(1, 4, i+1)
            plt.title(title[i])
            plt.imshow(experiment_average_test[i,:].reshape((size_row, size_col)), cmap='Greys')

            frame = plt.gca()
            frame.axes.get_xaxis().set_visible(False)
            frame.axes.get_yaxis().set_visible(False)

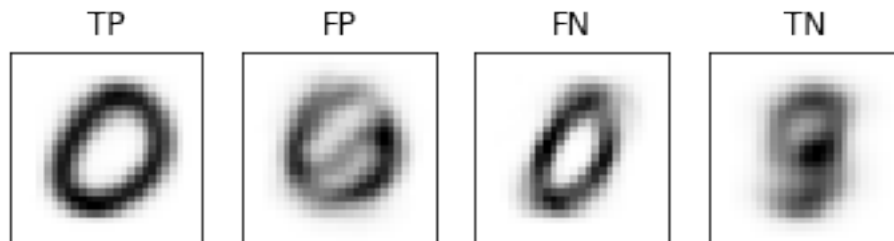
            print(title[i], ': ', experiment_count_test[i])
        table = imread('Eval_table.png')
        plt.figure()
        plt.imshow(table)
        plt.show()

```

```

TP : 871
FP : 417
FN : 109
TN : 8603

```



	Original data label		
	Label = '0'	Label = '2,3,4...9'	
Sign = 1	TP = 871	FP = 417	Total Positive = 1288
Sign = -1	FN = 109	TN = 8603	Total negative = 8712
	Total L = '0' = 980	Total L = '2,3,4...9' = 9020	Grand total = 10000

- compute evaluations

```
In [12]: #print the evaluation
print('Error rate: ', (experiment_count_test[1]+experiment_count_test[2])*100/num_test, '%')
print('Precision: ', experiment_count_test[0]*100/(experiment_count_test[0]+experiment_count_test[1]), '%')
print('Recall: ', experiment_count_test[0]*100/(experiment_count_test[0]+experiment_count_test[2]), '%')
print('False positive rate: ', experiment_count_test[1]*100/(experiment_count_test[1]+experiment_count_test[2]), '%')
print('Accuracy: ', (experiment_count_test[0]+experiment_count_test[3])*100/num_test, '%')
plt.show()
```

```
Error rate: 5.26 %
Precision: 67.62422360248448 %
Recall: 88.87755102040816 %
False positive rate: 4.623059866962306 %
Accuracy: 94.74 %
```