Assignment11

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1 Information

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Project: Build a binary classifier to classify digit 0 against all the other digits at MNIST dataset.

2 import library

3 Load files

Training data, Testing data

```
In [2]: file_data = "mnist_train.csv"
       handle_file = open(file_data, "r")
       train_data = handle_file.readlines()
       handle_file.close()
        file_data = "mnist_test.csv"
       handle_file = open(file_data, "r")
        test_data = handle_file.readlines()
       handle_file.close()
       num_train = len(train_data)
       num_test = len(test_data)
        train_image = np.zeros((28 * 28, num_train), dtype=float)
        train_label = np.zeros(num_train, dtype=int)
        test_image = np.zeros((28 * 28, num_test), dtype=float)
        test_label = np.zeros(num_test, dtype=int)
        count = 0
        for line in train_data:
           line_data = line.split(',')
```

```
label = line_data[0]
            train_label[count] = label
            im_vector = np.asfarray(line_data[1:])
            train_image[:,count] = im_vector
            count += 1
        count = 0
        for line in test data:
            line_data = line.split(',')
            label = line_data[0]
            test_label[count] = label
            im_vector = np.asfarray(line_data[1:])
            test_image[:,count] = im_vector
            count += 1
In [3]: class classifier:
            def __init__(self, k=64):
                self.k = k
                self.filter = self.generate_filter(k)
            def normalize(self, data):
                data_normalized = (data - np.min(data)) / (np.max(data) - np.min(data))
                return(data_normalized)
            def calculate_average(self, data):
                size x, size y = data.shape
                size = size_x*size_y
                average = sum(sum(data))/size
                return(average)
            def calculate_standard(self, average, data):
                variance = self.calculate_average(data*data) - (average*average)
                std = np.sqrt(variance)
                return(std)
            def whitening(self, data):
                avgerage = self.calculate_average(data)
                standard = self.calculate_standard(avgerage, data)
                whiten_data = (data - avgerage) / standard
                return(whiten_data)
            def generate_filter(self, k=64):
                data_filter = np.random.normal(0, 1, (k, 28*28))
                return(np.asmatrix(data_filter))
            def extract_feature(self, data, f):
                feature = f * data
                size_x, size_y = feature.shape
                for i in range(size_x):
```

```
for j in range(size_y):
            feature[i,j] = max(0, feature[i,j])
    return(feature)
def reshape data(self, data):
    num = max(data.shape)
    reshape_data = np.zeros((self.k + 1, num), dtype=float)
    for i in range(num):
        reshape_data[:,i] = np.insert(data[:,i], 0, 1)
    return(reshape_data)
def reshape_label(self, label, select):
   num = len(label)
    reshape_label = np.zeros(num, dtype=int)
    for i in range(num):
        if(int(label[i]) == select):
            reshape_label[i] = 1
        else:
            reshape_label[i] = -1
    return(reshape_label)
def train(self, train_data, label, num_digit=10):
    count = 0
    normalized_data = self.normalize(train_data)
    whiten_data = self.whitening(normalized_data)
    self.x = np.zeros((num_digit, self.k+1), dtype=float)
    feature = self.extract_feature(whiten_data, self.filter)
    train_image = self.reshape_data(feature)
    A = np.asmatrix(train_image.transpose())
    pinv_A = np.linalg.pinv(A)
    for i in range(num_digit):
        train_label = self.reshape_label(label, i)
        y = np.asmatrix(train_label)
        buf = np.array(pinv_A * y.transpose())
        self.x[i,:] = buf.T
    self.x = np.asmatrix(self.x)
    #return(self.x)
def predict(self, test_data):
   normalized_data = self.normalize(test_data)
    whiten_data = self.whitening(normalized_data)
    feature = self.extract_feature(whiten_data, self.filter)
    test_image = self.reshape_data(feature)
    A = np.asmatrix(test_image.transpose())
    v = A * self.x.T
    label = []
    for i in range(max(test_data.shape)):
```

```
label = np.array(label)
                return(label)
            def evaulation(self, prediction, label):
                tp = 0
                error = 0
                result =np.zeros((11,11), dtype=int)
                for i in range(len(prediction)):
                    result[prediction[i]][label[i]] +=1
                     if(prediction[i] == label[i]):
                         tp += 1
                    else:
                         error += 1
                for i in range(10):
                    result[10][i] = sum(result.T[:][i])
                    result[i][10] = sum(result[:][i])
                result[10][10] = sum(result[:,10])
                # Plot
                print("True Possitive: ", tp/result[10][10])
                print("Error Rate: ", error/result[10][10])
                chart = pd.DataFrame(result.T)
                return(chart)
In [4]: binary_classifier = classifier(64)
        binary_classifier.train(train_image, train_label)
In [5]: y = binary_classifier.predict(train_image)
        binary_classifier.evaulation(y, train_label)
True Possitive: 0.77035
Error Rate: 0.22965
Out[5]:
              0
                     1
                           2
                                       4
                                              5
                                                    6
                                                          7
                                                                              10
                                 3
                                                                 8
                                                                       9
        0
            5355
                    22
                                                   144
                                                                            5923
                           91
                                 70
                                       14
                                              99
                                                                       25
                                                          65
                                                                 38
        1
                           89
                                       35
                                              24
                                                    27
               2
                  6415
                                 50
                                                          20
                                                                 69
                                                                       11
                                                                            6742
        2
                   335
                        4279
                                              32
                                                   417
                                                                            5958
             131
                                169
                                      163
                                                         204
                                                                160
                                                                       68
        3
             133
                    219
                          209
                               4674
                                       48
                                             285
                                                    87
                                                         131
                                                                171
                                                                      174
                                                                            6131
        4
              43
                    204
                           68
                                     4470
                                              79
                                                   173
                                                         141
                                                                      507
                                                                            5842
                                 68
                                                                 89
        5
             373
                   337
                           70
                                515
                                      163
                                            3218
                                                   216
                                                         115
                                                                224
                                                                      190
                                                                            5421
        6
             141
                   312
                          201
                                 60
                                      113
                                              88
                                                 4917
                                                          15
                                                                 51
                                                                       20
                                                                            5918
        7
                   222
                                              26
                                                    15 5332
                                                                      240
              71
                           98
                                 52
                                      190
                                                                 19
                                                                            6265
        8
                          206
                                425
                                             221
                                                   145
                                                         139
                                                              3879
                                                                      235
             133
                    338
                                      130
                                                                            5851
                                                    73
        9
              84
                    153
                           62
                                160
                                      806
                                             112
                                                         706
                                                                111
                                                                     3682
                                                                            5949
                               6243
                                     6132 4184 6214
                                                        6868
                                                              4811
                                                                     5152
        10
           6466
                  8557 5373
                                                                           60000
```

label.append(np.argmax(y[i,:]))

True Possitive: 0.7766 Error Rate: 0.2234

Out[6]:		0	1	2	3	4	5	6	7	8	9	10
	0	881	2	12	17	2	18	28	15	4	1	980
	1	0	1086	11	13	2	2	3	0	14	4	1135
	2	30	83	718	27	30	7	64	34	32	7	1032
	3	29	30	24	804	5	37	11	22	25	23	1010
	4	3	30	7	6	758	17	35	23	17	86	982
	5	56	40	19	79	26	559	27	24	34	28	892
	6	39	35	33	6	26	15	787	3	11	3	958
	7	5	49	25	6	26	3	4	853	3	54	1028
	8	37	32	24	76	16	30	29	31	668	31	974
	9	14	18	6	22	150	18	17	91	21	652	1009
	10	1094	1405	879	1056	1041	706	1005	1096	829	889	10000

True Possitive: 0.88595 Error Rate: 0.11405

Out[8]:	0	1	2	3	4	5	6	7	8	9	10
0	5645	2	15	23	16	49	101	11	50	11	5923
1	1	6600	28	21	15	20	12	15	24	6	6742
2	67	169	5087	124	95	31	99	118	138	30	5958
3	49	92	182	5222	26	150	56	101	158	95	6131
4	12	70	31	4	5205	21	85	19	21	374	5842
5	115	74	45	287	106	4355	155	60	136	88	5421
6	56	58	42	7	47	115	5562	6	22	3	5918
7	36	132	66	24	128	18	11	5583	17	250	6265
8	49	247	86	198	70	204	66	44	4750	137	5851
9	41	47	23	95	259	42	17	220	57	5148	5949
10	6071	7491	5605	6005	5967	5005	6164	6177	5373	6142	60000

True Possitive: 0.8955 Error Rate: 0.1045 Out [9]: 1016 1247

Out[11]:		0	1	2	3	4	5	6	7	8	9	10
0) 5	5801	2	14	5	7	9	31	5	44	5	5923
1	-	0	6636	36	13	16	5	2	9	16	9	6742
2	2	31	34	5588	50	48	6	35	62	84	20	5958
3	3	8	21	83	5698	8	95	16	49	99	54	6131
4	<u> </u>	6	31	15	3	5551	5	33	7	20	171	5842
5	5	33	19	17	96	30	5024	86	13	63	40	5421
6	3	32	18	9	2	15	53	5754	0	35	0	5918
7	7	15	58	47	12	57	7	4	5938	14	113	6265
8	}	15	66	53	78	30	94	44	15	5389	67	5851
9)	25	13	12	73	118	39	4	118	48	5499	5949
1	.0 5	5966	6898	5874	6030	5880	5337	6009	6216	5812	5978	60000

True Possitive: 0.9443 Error Rate: 0.0557

Out[12]:

5	3	2	0	24	6	826	10	5	10	6	892
6	7	2	1	0	7	7	927	1	6	0	958
7	1	16	15	2	11	0	2	952	4	25	1028
8	5	2	6	22	3	11	9	8	901	7	974
9	7	7	1	7	23	14	2	16	8	924	1009
10	994	1157	999	1022	978	875	978	1009	980	1008	10000

In []: