## CS289A\_HW07\_prob1

## April 28, 2017

## 1 Homework 7: Problem 1

In [1]: import HW07\_utils as ut

## 1.1 *k*-means Clustering

Programmatic overhead; import this homework's utility module as well as math, numpy, scipy and matplotlib.

```
import math
        import numpy as np
        from matplotlib import pyplot as plt
        from scipy import ndimage as ndi
        from scipy import io as spio
  Load the MNIST data into memory.
In [2]: BASE_DIR = '/Users/mitch/Documents/Cal/2_2017_Spring/COMPSCI 289A - Intro t
        DATA_PATH = 'Data/hw7_data/mnist_data/images.mat'
        mnistdataraw = ut.load_data(DATA_PATH, BASE_DIR, 'images')
        mnistdata = np.empty((60000,784))
        for i in range(60000):
            mnistdata[i]=np.reshape(mnistdataraw[:,:,i],(784,))
In [6]: class Cluster:
            """ A class to perform k-means clustering"""
            def __init__(self,k):
                self.k = k
                self.means = None
                self.clusters = None
            def update_means(self):
                """Method to calculate the means of each cluster"""
                for i in range(self.k):
                    cluster_i = np.array(self.clusters[i])
                    mu_i = np.mean(cluster_i,axis=0)
                    self.means[i] = mu_i
```

```
def update_clusters(self):
    """ Method to take in a set of means and reclassify points according
    new_clusters = [[] for k_i in range(self.k)]
    clusters changed = 0
    for i in range(self.k):
        while True:
            try:
                x_j = self.clusters[i].pop(0)
                # Create object to store the i-value of the closest med
                cluster_index = self.assign_cluster(x_j)
                # Reclassify point
                if cluster_index != i:
                    clusters_changed += 1
                new_clusters[cluster_index].append(x_j)
            except:
                break
    self.clusters = new clusters
    return clusters_changed
def forgy_init(self, data):
    Execute the Forgy initializaiton method:
            -choose k random sample points from data to be initial mean
    select = np.random.choice(len(data), self.k, replace=False)
    self.means = data[select]
    # Assign points according to these random means
    self.clusters = [[] for k_i in range(self.k)]
    for datapoint in data:
        cluster_index = self.assign_cluster(datapoint)
        self.clusters[cluster_index].append(datapoint)
def assign_cluster(self, x_j):
    """Assign point x_j to the cluster with nearest mean"""
    nearest_mean_index = -1
    nearest_mean_dist = math.inf
    for i in range(self.k):
        # Check the distance of the point to each mean
        mu_i = self.means[i]
        d = np.linalg.norm(x_j-mu_i)
```

```
if d < nearest_mean_dist:</pre>
                         # Reassign closest mean
                         nearest_mean_index = i
                         nearest_mean_dist = d
                return nearest mean index
            def lloyd_alg(self):
                 11 11 11
                Execute Lloyd's algorithm to construct k clusters:
                         -Minimize the sum of squared distances of points from clust
                         -Use k-means heuristic to alternate between updating means
                clusters_changed = 1
                counter = 0
                while clusters_changed != 0:
                     print('Iteration', counter)
                     self.update_means()
                     clusters_changed = self.update_clusters()
                     print (clusters_changed)
                     counter += 1
                print('Finished')
                 for cluster in self.clusters:
                         print(np.shape(cluster))
  Test the clustering algorithm on a simple dataset.
In [7]: simpledata = np.array([[3,10,10],[9,10,10],[9,9,10],[9,4,10],[10,3,10],[4,8]
        clustering = Cluster(3)
        clustering.forgy_init(simpledata)
        clustering.lloyd_alg()
        print (np.array (clustering.clusters))
Iteration 0
Iteration 1
Finished
(3, 3)
(3, 3)
```

(3, 3)

[[[ 3 10 10] [ 4 8 10] [ 2 8 10]]

```
[[ 9 10 10]
  [ 9 9 10]
  [ 9 8 10]]
 [[ 9 4 10]
  [10 3 10]
  [ 8 2 10]]]
In [9]: clustering_k5 = Cluster(5)
        clustering_k5.forgy_init(mnistdata)
        clustering_k5.lloyd_alg()
Iteration 0
11738
Iteration 1
6207
Iteration 2
4116
Iteration 3
3457
Iteration 4
3170
Iteration 5
3138
Iteration 6
3042
Iteration 7
2661
Iteration 8
2285
Iteration 9
1963
Iteration 10
1740
Iteration 11
1604
Iteration 12
1488
Iteration 13
1457
Iteration 14
1653
Iteration 15
1828
Iteration 16
```

```
Iteration 17
2088
Iteration 18
2104
Iteration 19
1910
Iteration 20
1686
Iteration 21
1479
Iteration 22
1239
Iteration 23
970
Iteration 24
666
Iteration 25
490
Iteration 26
329
Iteration 27
239
Iteration 28
179
Iteration 29
126
Iteration 30
75
Iteration 31
54
Iteration 32
2.7
Iteration 33
25
Iteration 34
24
Iteration 35
Iteration 36
Iteration 37
3
Iteration 38
```

Iteration 40

```
Iteration 41
Iteration 42
Iteration 43
Iteration 44
Iteration 45
Iteration 46
Iteration 47
Iteration 48
Iteration 49
Iteration 50
Finished
(10901, 784)
(11630, 784)
(16619, 784)
(5563, 784)
(15287, 784)
In [21]: k5 = plt.figure()
         for k in range(5):
             cluster = clustering_k5.means[k]
             center = np.reshape(cluster, (28,28))
             k5.add\_subplot(1,5,k+1)
             plt.imshow(center)
         plt.savefig(BASE_DIR+'Figures/MNIST_k05.jpg')
         plt.show()
         0
```

```
clustering_k10.forgy_init(mnistdata)
         clustering_k10.lloyd_alg()
Iteration 0
16393
Iteration 1
7565
Iteration 2
5146
Iteration 3
4241
Iteration 4
4000
Iteration 5
3634
Iteration 6
2980
Iteration 7
2302
Iteration 8
1701
Iteration 9
1351
Iteration 10
1042
Iteration 11
883
Iteration 12
754
Iteration 13
734
Iteration 14
675
Iteration 15
651
Iteration 16
674
Iteration 17
713
Iteration 18
718
Iteration 19
765
Iteration 20
816
Iteration 21
825
```

In [10]: clustering\_k10 = Cluster(10)

821

Iteration 23

824

Iteration 24

814

Iteration 25

831

Iteration 26

827

Iteration 27

746

Iteration 28

778

Iteration 29

725

Iteration 30

661

Iteration 31

547

Iteration 32

488

Iteration 33

414

Iteration 34

360

Iteration 35

291

Iteration 36

242

Iteration 37

235

Iteration 38

195

Iteration 39

172

Iteration 40

160

Iteration 41

160

Iteration 42

150

Iteration 43

129

Iteration 44

118

Iteration 45

104

Iteration 47

103

Iteration 48

104

Iteration 49

97

Iteration 50

89

Iteration 51

79

Iteration 52

87

Iteration 53

7 2

Iteration 54

71

Iteration 55

55

Iteration 56

54

Iteration 57

54

Iteration 58

60

Iteration 59

54

Iteration 60

69

Iteration 61

71

Iteration 62

86

Iteration 63

84

Iteration 64

63

Iteration 65

57

Iteration 66

48

Iteration 67

43

Iteration 68

45

Iteration 69

63

Iteration 71

51

Iteration 72

49

Iteration 73

64

Iteration 74

67

Iteration 75

73

Iteration 76

84

Iteration 77

104

Iteration 78

103

Iteration 79

110

Iteration 80

107

Iteration 81

112

Iteration 82

99

Iteration 83

91

Iteration 84

84

Iteration 85

76

Iteration 86

79

Iteration 87

72

Iteration 88

69

Iteration 89

72

Iteration 90

50

Iteration 91

29

Iteration 92

25

Iteration 93

29

Iteration 95

29

Iteration 96

35

Iteration 97

31

Iteration 98

35

Iteration 99

26

Iteration 100

32

Iteration 101

2 E

Iteration 102

44

Iteration 103

44

Iteration 104

45

Iteration 105

38

Iteration 106

20

Iteration 107

20

Iteration 108

19

Iteration 109

21

Iteration 110

29

Iteration 111

33

Iteration 112

31

Iteration 113

18

Iteration 114

14

Iteration 115

15

Iteration 116

15

Iteration 117

20

Iteration 119

33

Iteration 120

48

Iteration 121

43

Iteration 122

41

Iteration 123

51

Iteration 124

57

Iteration 125

63

Iteration 126

62

Iteration 127

57

Iteration 128

57

Iteration 129

60

Iteration 130

65

Iteration 131

66

Iteration 132

65

Iteration 133

79

Iteration 134

73

Iteration 135

66

Iteration 136

69

Iteration 137

67

Iteration 138

66

Iteration 139

60

Iteration 140

66

Iteration 141

```
Iteration 142
Iteration 143
75
Iteration 144
Iteration 145
79
Iteration 146
63
Iteration 147
66
Iteration 148
44
Iteration 149
Iteration 150
35
Iteration 151
25
Iteration 152
13
Iteration 153
Iteration 154
Iteration 155
13
Iteration 156
15
Iteration 157
12
Iteration 158
13
Iteration 159
12
Iteration 160
Iteration 161
16
Iteration 162
17
Iteration 163
20
Iteration 164
21
Iteration 165
```

```
Iteration 166
11
Iteration 167
10
Iteration 168
Iteration 169
Iteration 170
Iteration 171
Iteration 172
Iteration 173
Iteration 174
10
Iteration 175
16
Iteration 176
21
Iteration 177
Iteration 178
18
Iteration 179
16
Iteration 180
Iteration 181
Iteration 182
12
Iteration 183
16
Iteration 184
Iteration 185
10
Iteration 186
12
Iteration 187
Iteration 188
Iteration 189
1
```

```
Iteration 190
Iteration 191
2
Iteration 192
Finished
(6311, 784)
(6208, 784)
(7178, 784)
(5043, 784)
(7409, 784)
(5041, 784)
(4793, 784)
(5927, 784)
(5435, 784)
(6655, 784)
In [19]: k10 = plt.figure()
         for k in range(10):
             cluster = clustering_k10.means[k]
             center = np.reshape(cluster, (28,28))
             k10.add_subplot(2,5,k+1)
             plt.imshow(center)
         plt.savefig(BASE_DIR+'Figures/MNIST_k10.jpg')
         plt.show()
          0
         20
          0
        20
                      Ò
                               25 0
                                          25 0
                                                     25 0
           Ó
```

```
clustering_k20.forgy_init(mnistdata)
         clustering_k20.lloyd_alg()
Iteration 0
18870
Iteration 1
9819
Iteration 2
6728
Iteration 3
4908
Iteration 4
3726
Iteration 5
2961
Iteration 6
2337
Iteration 7
1944
Iteration 8
1743
Iteration 9
1593
Iteration 10
1424
Iteration 11
1201
Iteration 12
1081
Iteration 13
986
Iteration 14
968
Iteration 15
897
Iteration 16
829
Iteration 17
721
Iteration 18
650
Iteration 19
535
Iteration 20
483
Iteration 21
430
```

In [11]: clustering\_k20 = Cluster(20)

377

Iteration 23

303

Iteration 24

275

Iteration 25

276

Iteration 26

243

Iteration 27

195

Iteration 28

204

Iteration 29

190

Iteration 30

158

Iteration 31

155

Iteration 32

156

Iteration 33

161

Iteration 34

139

Iteration 35

110

Iteration 36

121

Iteration 37

101

Iteration 38

95

Iteration 39

99

Iteration 40

88

Iteration 41

73

Iteration 42

56

Iteration 43

50

Iteration 44

4 ∩

Iteration 45

56

Iteration 47

67

Iteration 48

68

Iteration 49

79

Iteration 50

85

Iteration 51

63

Iteration 52

64

Iteration 53

55

Iteration 54

44

Iteration 55

40

Iteration 56

39

Iteration 57

48

Iteration 58

46

Iteration 59

48

Iteration 60

38

Iteration 61

29

Iteration 62

30

Iteration 63

12

Iteration 64

9

Iteration 65

10

Iteration 66

9

Iteration 67

5

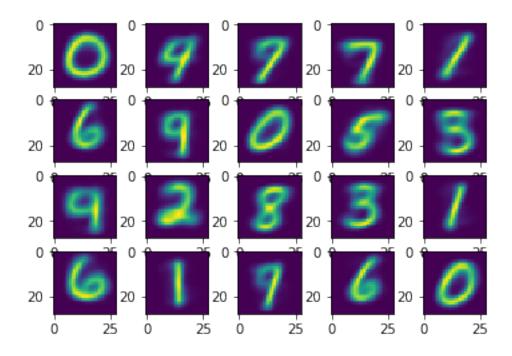
Iteration 68

8

Iteration 69

```
Iteration 70
Iteration 71
Iteration 72
Iteration 73
Iteration 74
Iteration 75
Iteration 76
Iteration 77
Iteration 78
Iteration 79
Iteration 80
Iteration 81
Iteration 82
Iteration 83
Iteration 84
Iteration 85
Iteration 86
Iteration 87
Iteration 88
Iteration 89
Iteration 90
Iteration 91
Iteration 92
Iteration 93
2
```

```
Iteration 94
Iteration 95
Finished
(2222, 784)
(2929, 784)
(2962, 784)
(2377, 784)
(2611, 784)
(2583, 784)
(3917, 784)
(1763, 784)
(3109, 784)
(4386, 784)
(2936, 784)
(3919, 784)
(4390, 784)
(3958, 784)
(3011, 784)
(1739, 784)
(3453, 784)
(3645, 784)
(2310, 784)
(1780, 784)
In [23]: k20 = plt.figure()
         for k in range(20):
             cluster = clustering_k20.means[k]
             center = np.reshape(cluster, (28,28))
             k20.add\_subplot(4,5,k+1)
             plt.imshow(center)
         plt.savefig(BASE_DIR+'Figures/MNIST_k20.jpg')
         plt.show()
```



In [ ]: