DATASCI W261: Machine Learning at Scale

David Rose david.rose@berkeley.edu W261-1 Week 04 2015.09.23

HW4.0

- MRJob: a Python API framework for accessing the Hadoop streaming capabilities.
 It differs from MapReduce in that it acts as a higher-level interface to MapReduce,
 yet utilizes the Hadoop MapReduce functionality. It provides a mechanism for
 creating data processing pipelines that MapReduce, on its own, cannot.
- The *_final() methods are defined in the MRJob class, and as such can be
 overridden by classes that extend MRJob. The *_final() methods are executed when
 the input stream to the respective task is closed.

HW4.1

- **Serialization** converts an object into a bytestream that can be used for transporting the object over a network or to and from disk storage.
- Within Hadoop and MRJob processes data must be **serialized**, at a minimum, when it is first submitted to a map task, when it is spilled to disk, when it is submitted to a reduce task, and again when the results are written to disk.
- Default modes for MRJob serialization are:
 - INPUT_PROTOCOL = mrjob.protocol.RawValueProtocol
 - INTERNAL_PROTOCOL = mrjob.protocol.JSONProtocol
 - OUTPUT_PROTOCOL = mrjob.protocol.JSONProtocol

```
In [45]:
           1 %%writefile hw 4 2 flatten.py
           2 #!/usr/bin/python
           3 '''
           4 '''
           5 from future import print function
           6 import csv
           7 import sys
           8 with open(sys.argv[1], 'rb') as fin, open(sys.argv[2], 'w') as fou
           9
                 csvreader = csv.reader(fin, delimiter = ',', quotechar = '"')
          10
                 visitorid = ''
                 for line in csvreader:
          11
          12
                     if line[0] == 'C':
          13
                         visitorid = line[2]
          14
                         continue
          15
                     if line[0] == 'V':
          16
                         line.append(visitorid)
          17
                     print(','.join(line), file=fout)
```

Overwriting hw 4 2 flatten.py

```
In [46]: 1 # combine the page id and visitor id onto a single line for
2 # subsequent processing
3 !python hw_4_2_flatten.py anonymous-msweb.data flattened.data
```

HW4.3

In [47]:		

```
1 %%writefile hw 4 3 mrjob.py
 2 ''' count and list the top five most frequently visited pages
 3 '''
 4 from future import print function
 5 from mrjob.job import MRJob
 6 from mrjob.step import MRStep
 7 import sys
8
9 class FrequentPages(MRJob):
10
11
       def steps(self):
12
           return [
13
               MRStep(mapper=self.mapper,
14
                       combiner=self.combiner,
15
                       reducer=self.reducer,
                       reducer final=self.reducer final)
16
17
           ]
18
19
       def mapper(self, _, line):
           ''' enumerate page visits
20
21
22
           row = line.split(',')
23
           if row[0] == 'V':
24
               yield row[1], 1
25
       def combiner(self, page, i):
26
27
           ''' combine local results
           . . .
28
29
           yield page, sum(i)
30
31
       # track top five frequent pages in sorted order
       topfive = [['',0]]
32
33
       def inserttopfive(self, page, total):
34
35
           ''' data structure and logic to maintain list of
36
               top five most frequent pages.
               This operation is performed here in the reducer
37
38
               and again in the driver to capture output from
               multiple reducers
39
40
           for j in range(0, len(self.topfive)):
41
42
               if total > self.topfive[j][1]:
43
                    self.topfive.insert(j, (page, total))
44
                    if len(self.topfive) > 5:
                        self.topfive.pop()
45
46
                   break
47
       def reducer(self, page, i):
48
49
           ''' sum and sort page counts
50
51
           total = sum(i)
52
           self.inserttopfive(page, total)
53
54
       def reducer final(self):
```

Overwriting hw 4 3 mrjob.py

```
1 %%writefile hw_4_3_driver.py
In [48]:
           2 from future import print function
           3 from mrjob import util
           4 import sys
           5 from hw 4 3 mrjob import FrequentPages
           6 util.log to null() # to suppress a 'no handler found' message
           7
           8 # list for storing most frequent pages
           9 # we do this step here since multiple reducer tasks may run and th
          10 # combined output needs to be processed
          11 topfive = [['',0]]
          12 def inserttopfive(page, total):
                 for j in range(0, len(topfive)):
          13
          14
                      if total > topfive[j][1]:
          15
                          topfive.insert(j, (page, total))
                          if len(topfive) > 5:
          16
          17
                              topfive.pop()
          18
                              break
          19
          20 mr job = FrequentPages(args=sys.argv[1:])
          21 with mr_job.make_runner() as runner:
          22
                 runner.run()
          23
                 for line in runner.stream output():
          24
                     page, total = line.split()
          25
                     inserttopfive(page, int(total))
          26 for i in range(0, len(topfive)):
          27
                 print('page: {}, visits: {}'.format(topfive[i][0],
          28
                                                      topfive[i][1]), file=sys.s
          29
```

Overwriting hw 4 3 driver.py

```
In [49]: 1 # HW 4.3: Find the 5 most frequently visited pages using mrjob fro
2 !python hw_4_3_driver.py flattened.data --strict-protocols -r loca

page: "1008", visits: 10836
page: "1034", visits: 9383
page: "1004", visits: 8463
page: "1018", visits: 5330
page: "1017", visits: 5108
```

In [50]:	

```
1 %%writefile hw 4 4 mrjob.py
 2 ''' count the number of page-visitor combinations and
       for each page list the most frequent visitors
 3
 4
 5
       the data does not effectively support this operation since it
       only lists unique page visits, therefore every visitor
       will show up as having visited once, and therefore every visit
 7
       is the most frequent visitor
8
   1 1 1
9
10 from future import print function
11 from mrjob.job import MRJob
12 import sys
13
14 class FrequentVisitors(MRJob):
15
16
       def mapper(self, , line):
17
           ''' enumerate page visitors
18
19
           row = line.split(',')
20
           if row[0] == 'V':
21
               # page ID, visitor ID
22
               yield row[1], row[3]
23
24
       # data structures to manage reducer logic
       visitors = {}
25
       currentpage = ''
26
27
28
       def reducer(self, page, visitor):
           ''' sum page visitor counts
29
30
           if not page == self.currentpage:
31
               ''' page id has changed in the stream, so process and
32
                   the information for the current page
33
               . . .
34
35
               if len(self.visitors) > 0:
                   frequentv = []
36
                   maxv = 0
37
38
                   for v in self.visitors:
                       if self.visitors[v] > maxv:
39
40
                            frequentv = [v]
                           maxv = self.visitors[v]
41
42
                       elif self.visitors[v] == maxv:
43
                           frequentv.append(v)
44
                   # emit results
                   for v in frequentv:
45
46
                       yield self.currentpage, v
47
               # reset counters
               self.visitors = {}
48
               self.currentpage = page
49
          for v in visitor:
50
51
               if not v in self.visitors:
                   self.visitors[v] = 0
52
               self.visitors[v] += 1
53
54
```

```
- -
55
       # process any remaining values after stream closes
       def reducer final(self):
56
           if len(self.visitors) > 0:
57
               frequentv = []
58
59
               maxv = 0
               for v in self.visitors:
60
61
                    if self.visitors[v] > maxv:
62
                        frequentv = [v]
63
                    elif self.visitors[v] == maxv:
                        frequentv.append(v)
64
               for v in frequentv:
65
                   yield self.currentpage, v
66
67
68 if __name__ == ' main ':
69
       FrequentVisitors.run()
70
```

Overwriting hw 4 4 mrjob.py

```
In [51]:
           1 %%writefile hw 4 4 driver.py
           2 import csv
           3 from mrjob import util
           4 import sys
           5 from hw 4 4 mrjob import FrequentVisitors
           7 util.log to null() # to suppress a 'no handler found' message
           9 # construct list of page ids and urls to satisfy the requirement
          10 # this is essentially a join; it makes better sense to do this in
          11 # the driver as doing so in hadoop offers no advantages and incurs
          12 # additional network overhead; mrjob output is parsed and augmente
          13 # the url information
          14 pageinfo = {}
          15 # read in the page attributes including the url; this could also h
          16 # been preprocessed so that the page attributes were in their own
          17 with open('flattened.data', 'rb') as fin:
                 csvreader = csv.reader(fin, delimiter = ',', quotechar = '"')
          18
                 for line in csvreader:
          19
                     if line[0] == 'A':
          20
          21
                         pageid = line[1]
          22
                         url = line[4]
          23
                         pageinfo[pageid] = url
          24
          25 mr job = FrequentVisitors(args=sys.argv[1:])
          26 with mr job.make runner() as runner:
          27
                 runner.run()
          28
                 for line in runner.stream output():
                     page, visitor = line.replace('"', '').split()
          29
          30
                     print pageinfo[page], page, visitor
          31
```

```
In [52]: 1 # HW 4.4: Find the most frequent visitor of each page using
2 # mrjob and the output of 4.2
3 # In this output please include the webpage URL, webpageID
4 # and Visitor ID.
5 !python hw_4_4_driver.py flattened.data --strict-protocols -r loca
6 !head -n50 mrjob_4_4_output
7 !tail -n50 mrjob_4_4_output
```

```
/regwiz 1000 10001
/regwiz 1000 10010
/regwiz 1000 10039
/regwiz 1000 10073
/regwiz 1000 10087
/regwiz 1000 10101
/regwiz 1000 10132
/regwiz 1000 10141
/regwiz 1000 10154
/regwiz 1000 10162
/regwiz 1000 10166
/regwiz 1000 10201
/regwiz 1000 10218
/regwiz 1000 10220
/regwiz 1000 10324
/regwiz 1000 10348
/regwiz 1000 10376
/regwiz 1000 10384
/regwiz 1000 10409
```

HW4.5

```
1 ''' utility script to get some statistics on the data set
In [53]:
           2
           3 with open('topUsers Apr-Jul 2014 1000-words.txt', 'rb') as fin:
           4
                 max = 0
                 total = 0
           5
                 linecount = 0
           6
           7
                 classes = {}
                 for line in fin:
           8
           9
                      row = line.split(',')
          10
                      count = int(row[2])
          11
                      clazz = row[1]
          12
                      if count > max: max = count
          13
                      total += count
          14
                     linecount += 1
                     if not clazz in classes:
          15
          16
                          classes[clazz] = 0
          17
                      classes[clazz] += 1
          18 print(linecount, max, total, str(classes))
          19
         (1000, 1724608, 61819567, "{'1': 91, '0': 752, '3': 103, '2': 54}")
In [54]:
              ''' utility script to get some statistics on the data set
           1
           2
                  look for the maximum word ratio after the data is normalized
           3
           4 with open('topUsers Apr-Jul 2014 1000-words.txt', 'rb') as fin:
           5
                 max = 0.0
                 for line in fin:
           6
           7
                      row = line.split(',')
                      count = int(row[2])
           8
           9
                      for i in range(3, len(row)):
```

p = float(row[i]) / count

if p > max:

max = p

0.409909665665

13 print(max)

10

11

12

```
In [55]:
           1 %%writefile hw 4 5 preprocess.py
           2 ''' preprocess the data set, normalizing the word counts
           3 '''
           4 from __future__ import print_function
           5 import sys
           6 with open(sys.argv[1], 'rb') as fin, open(sys.argv[2], 'w') as fou
           7
                  for line in fin:
                      row = map(int, line.split(','))
           8
           9
                      userid = row[0]
          10
                      code = row[1]
          11
                     total = row[2]
                      for j in range(2, len(row)):
          12
          13
                          row[j] = float(row[j]) / total
                     print('{}'.format(row[0]), file = fout, end = '')
          14
          15
                     for j in range(1, 3):
          16
                         print(',{}'.format(row[j]), file = fout, end = '')
                      for j in range(3, len(row)):
          17
                         print(',{:0.8f}'.format(row[j]), file = fout, end = ''
          18
          19
                     print('', file = fout)
          20
```

Overwriting hw_4_5_preprocess.py

```
In [56]: 1 !python hw_4_5_preprocess.py topUsers_Apr-Jul_2014_1000-words.txt
```

In [57]:	

```
1 %%writefile hw 4 5 mrjob.py
 2 ''' map/reduce approach to determining stable centroids using
       a K-means algorithm
 3
 4
 5 from future import print function
 6 from numpy import argmin, array, random
 7 from mrjob.job import MRJob
 8 from mrjob.step import MRStep
 9 from itertools import chain
10 import sys
11
12 # Calculate find the nearest centroid for data point
13 def MinDist(datapoint, centroid points):
14
       datapoint = array(datapoint)
15
       centroid points = array(centroid points)
       diff = datapoint - centroid points
16
17
       diffsq = diff*diff
       # Get the nearest centroid for each instance
18
19
       sumofsquares = list(diffsq.sum(axis = 1))
20
       minindex = argmin(sumofsquares)
21
       return minindex
22
23 # Check whether centroids converge
24 def stop criterion(centroid points old, centroid points new, T):
25
       oldvalue = list(chain(*centroid points old))
26
       newvalue = list(chain(*centroid points new))
27
       Diff = [abs(x-y) \text{ for } x, y \text{ in } zip(oldvalue, newvalue)]
28
       Flag = True
29
       for i in Diff:
30
           if i > T:
               Flaq = False
31
32
               break
33
       return Flag
34
35 class MRKmeans(MRJob):
36
37
       centroid points=[]
38
       k = -1
39
       CENTROIDFILE = '/tmp/centroids.txt'
40
41
       def steps(self):
42
           return [
43
               MRStep(mapper init = self.mapper init,
44
                      mapper=self.mapper,
45
                      combiner = self.combiner,
46
                      reducer=self.reducer)
47
                   1
       # load initial centroids
48
       def mapper init(self):
49
50
           self.centroid points=[]
51
           with open(self.CENTROIDFILE, 'rb') as fin:
52
               header = True
               for line in fin:
53
                   if header:
54
```

```
55
                         self.k = int(line.strip())
 56
                         header = False
 57
                     else:
 58
                         self.centroid_points.append(map(float,line.st)
            # print the value of k back to the cache file
 59
 60
            with open(self.CENTROIDFILE, 'w') as fout:
 61
                print('{}'.format(self.k), file = fout)
 62
 63
        #load data and output the nearest centroid index and data poil
        def mapper(self, _, line):
 64
 65
            D = (map(float,line.split(',')[3:]))
            centroid = MinDist(D, self.centroid points)
 66
 67
            yield centroid, (D, 1)
 68
        # aggregate data points locally
 69
        def combiner(self, centroid, inputdata):
 70
 71
            count = 0
 72
            bucket = [0] * 1000
 73
            for data, n in inputdata:
 74
                count += n
 75
                data = map(float, data)
 76
                for j in range(0, len(data)):
 77
                    bucket[j] += data[j]
 78
            yield centroid, (bucket, count)
 79
 80
 81
        # aggregate values for each centroid, then recalculate centrol
 82
        def reducer(self, idx, inputdata):
 83
            centroids = []
 84
            with open(self.CENTROIDFILE, 'rb') as fin:
                self.k = int(fin.readline().strip())
 85
 86
            num = [0] * self.k
            for i in range(self.k):
 87
                centroids.append([0.0]*1000)
 88
 89
            for data, n in inputdata:
 90
                num[idx] += n
 91
                data = map(float, data)
                for j in range(0, len(data)):
 92
 93
                    centroids[idx][j] += data[j]
            for j in range(0, len(centroids[idx])):
 94
 95
                centroids[idx][j] = centroids[idx][j] / num[idx]
            with open(self.CENTROIDFILE, 'a') as fout:
 96
 97
                print(','.join(str(i) for i in centroids[idx]), file =
 98
            yield idx, (centroids[idx])
 99
100 if name == ' main ':
101
        MRKmeans.run()
```

T [F0].	
In [58]:	

```
1 %%writefile hw 4 5 mrjob labeler.py
 2 ''' assigns centroid labels to data points
 3
       this step is done separately from the centroid calculations
 4
       to simplify the data processing
 5
 6 '''
 7 from future import print function
 8 from numpy import argmin, array, random
 9 from mrjob.job import MRJob
10 from mrjob.step import MRStep
11 import sys
12
13 from hw 4 5 mrjob import MinDist
14
15 class MRLabeler(MRJob):
16
17
       centroid points=[]
18
       k = -1
19
       CENTROIDFILE = '/tmp/centroids.txt'
20
21
       def steps(self):
22
           return [
23
               MRStep(mapper init = self.mapper init,
24
                      mapper=self.mapper,
25
                      combiner = self.combiner,
                      reducer=self.reducer,
26
27
                      reducer final=self.reducer final)
28
                   ]
29
       #load centroids info from file
       def mapper init(self):
30
           self.centroid points=[]
31
           with open(self.CENTROIDFILE, 'rb') as fin:
32
33
               header = True
               for line in fin:
34
35
                   if header:
36
                        self.k = int(line.strip())
                       header = False
37
38
                   else:
                        self.centroid points.append(map(float,line.str
39
40
       # determine the closest centroid for each data point
41
42
       def mapper(self, , line):
43
           row = line.strip().split(',')
44
           label = int(row[1])
           D = (map(float, row[3:]))
45
46
           centroid = MinDist(D, self.centroid_points)
           yield centroid, (label, 1)
47
48
49
       # combine sum of data points locally
       def combiner(self, centroid, inputdata):
50
51
           counts = [0, 0, 0, 0]
           for label, n in inputdata:
52
53
               counts[label] += n
54
           for i in range(len(counts)):
```

```
- ------, - - - - - - - , , -
55
               yield centroid, (i, counts[i])
56
57
       # sum the counts for centroids and classes
58
59
       currentcentroid = ''
60
       counts = []
61
       def reducer(self, centroid, inputdata):
           if not centroid == self.currentcentroid:
62
63
                for i in range(len(self.counts)):
                    yield self.currentcentroid, (i, self.counts[i])
64
                self.counts = [0, 0, 0, 0]
65
               self.currentcentroid = centroid
66
           for label, n in inputdata:
67
               self.counts[label] += n
68
69
       def reducer final(self):
70
           for i in range(len(self.counts)):
71
               yield self.currentcentroid, (i, self.counts[i])
72
73
74
75 if __name__ == '__main__':
76
       MRKmeans.run()
```

Overwriting hw 4 5 mrjob labeler.py

In [59]:		

```
1 %%writefile hw 4 5 driver.py
 2 ''' driver script for Kmeans job
 3 '''
 4 from future import print function
 5 from numpy import random
 6 from hw 4 5 mrjob import MRKmeans, stop criterion
 7 from hw 4 5 mrjob labeler import MRLabeler
 8 from mrjob import util
9 import sys
10
11 util.log to null() # to suppress a 'no handler found' message
12 CENTROIDFILE = '/tmp/centroids.txt'
13
14 def countlabels():
15
       ''' count the number of each classification as labeled in the
           original data set
16
       1 1 1
17
18
       with open('topUsers Apr-Jul 2014 1000-words.txt', 'rb') as f:
19
           labels = {}
20
           for line in f:
21
               row = line.split(',')
22
               label = row[1]
23
               if not label in labels:
24
                    labels[label] = 0
25
               labels[label] += 1
26
       return labels
27
28 def init centroids random internal(k):
       ''' select initial centroids by choosing data points
29
30
           randomly from the data set
       . . .
31
32
       randoms = sorted(random.randint(0, 1000, size = k))
33
       centroids = []
34
       with open('topUsers Apr-Jul 2014 1000-words.txt', 'rb') as f:
35
           lineno = 0
           count = 0
36
           for line in f:
37
38
               if lineno in randoms:
39
                   row = line.strip().split(',')
40
                   data = map(float, row[3:])
41
                   # normalize the values
42
                   data = [i / float(row[2]) for i in data]
                   centroids.append(data)
43
44
                   count += 1
45
                    if count == k:
                        break
46
47
               lineno += 1
       with open(CENTROIDFILE, 'w') as f:
48
49
           print('{}'.format(k), file = f)
           for tuple in centroids:
50
51
               print(','.join(str(i) for i in tuple), file = f)
52
       return centroids
53
54 def init centroids random external(k):
```

```
''' create initial centroids by generating random values
 55
 56
 57
        centroids = []
 58
        for i in range(k):
 59
            centroid = []
 60
            for j in range(1000):
 61
                centroid.append(random.uniform(0.0, .5))
 62
            centroids.append(centroid)
        with open(CENTROIDFILE, 'w') as fout:
 63
 64
            print('{}'.format(k), file = fout)
 65
            for tuple in centroids:
 66
                print(','.join(str(i) for i in tuple), file = fout)
 67
        return centroids
 68
 69 def init centroids perturbed(k):
        ''' create initial centroids by using aggregated values and
 70
 71
            perturbing them with random noise
        . . .
 72
 73
        centroids = []
 74
        with open('topUsers Apr-Jul 2014 1000-words summaries.txt', '1
 75
            for line in f:
 76
                row = line.strip().split(',')
 77
                if row[0] == 'ALL CODES':
                    data = map(float, row[3:])
 78
                    # normalize the values
 79
 80
                    data = [i / float(row[2]) for i in data]
                    for i in range(k):
 81
 82
                         centroid = []
 83
                         for j in range(len(data)):
 84
                             # modify each value with random number in
                             # range of +- the value; avoids a small no
 85
 86
                             # by a large number
                             centroid.append(data[j] + random.uniform(-
 87
 88
                         centroids.append(centroid)
 89
                    break
        with open(CENTROIDFILE, 'w') as f:
 90
 91
            print('{}'.format(k), file = f)
 92
            for tuple in centroids:
 93
                print(','.join(str(i) for i in tuple), file = f)
 94
        return centroids
 95
 96 def init centroids trained(k):
        ''' create initial centroids by choosing the class-specific a
 97
 98
 99
        centroids = []
        with open('topUsers Apr-Jul 2014 1000-words summaries.txt', '1
100
101
            for line in f:
102
                row = line.strip().split(',')
103
                if row[0] == 'CODE':
104
                    code = row[1]
105
                    total = int(row[2])
106
                    data = map(float, row[3:])
107
                    # normalize the values
                    data = [i / total for i in data]
108
```

```
109
                    centroids.append(data)
110
        with open(CENTROIDFILE, 'w') as f:
            print('{}'.format(k), file = f)
111
112
            for tuple in centroids:
                print(','.join(str(i) for i in tuple), file = f)
113
114
        return centroids
115
116 def getpurity(clusters):
117
        majority = 0
118
        for c in clusters:
119
            counts = map(int, clusters[c]);
120
            majority += max(counts)
121
        return majority / float(1000)
122
123 def go(centroid points):
124
        ''' submit the centroid calculation job;
125
            follow that with the data labeling job
126
127
        mr job = MRKmeans(args = ['normalized.txt'])
128
        iteration = 0
129
        while(1):
130
            # save previous centroids to check convergency
131
            centroid points old = centroid points[:]
            with mr job.make runner() as runner:
132
133
                runner.run()
134
                # capture reducer output
135
                for line in runner.stream output():
136
                    key,value = mr job.parse output line(line)
137
                    centroid points[key] = value
138
            iteration += 1
139
            if stop criterion(centroid points old, centroid points, 0
140
141
        print('centroids converged: {} iterations'.format(iteration),
142
        mr job = MRLabeler(args = ['normalized.txt'])
143
        labels = countlabels()
144
        clusters = {}
145
        with mr job.make runner() as runner:
146
            runner.run()
147
            for line in runner.stream output():
                centroid, value = mr job.parse output line(line)
148
149
                label = int(value[0])
150
                count = int(value[1])
151
                if not centroid in clusters:
152
                    clusters[centroid] = [0,0,0,0]
153
                clusters[centroid][label] += count
            for c in sorted(clusters):
154
155
                results = map(float, clusters[c])
156
                print('centroid {}: label 0: {:.4f} label 1: {:.4f} label
157
                       .format(c, results[0] / labels['0'], results[1]
158
                               results[2] / labels['2'], results[3] / 1
159
160
                # print the actual counts
161 #
                  print('centroid {}: counts: {}'.format(c, clusters[
162 #
                         file = sys.stderr)
```

```
163
            print('purity: {:.4f}'.format(getpurity(clusters)))
164
            print('', file = sys.stderr)
165
166 print('k = 4, random initialization, v1', file = sys.stderr)
167 go(init_centroids_random_internal(4))
168 print('k = 4, random initialization, v2', file = sys.stderr)
169 go(init centroids random external(4))
170 print('k = 2, perturbed initialization', file = sys.stderr)
171 go(init centroids perturbed(2))
172 print('k = 4, perturbed initialization', file = sys.stderr)
173 go(init centroids perturbed(4))
174 print('k = 4, trained initialization', file = sys.stderr)
175 go(init_centroids_trained(4))
176
177
```

Overwriting hw_4_5_driver.py

```
In [60]:
           1 !python hw 4 5 driver.py
         k = 4, random initialization, v1
         centroids converged: 9 iterations
         centroid 0: label 0: 0.0000 label 1: 0.1868 label 2: 0.2778 label 3:
         0.0388
         centroid 1: label 0: 0.0000 label 1: 0.5604 label 2: 0.0000 label 3:
         centroid 2: label 0: 0.9987 label 1: 0.0330 label 2: 0.0370 label 3:
         0.9612
         centroid 3: label 0: 0.0013 label 1: 0.2198 label 2: 0.6852 label 3:
         0.0000
         purity: 0.8560
         k = 4, random initialization, v2
         centroids converged: 3 iterations
         centroid 0: label 0: 1.0000 label 1: 1.0000 label 2: 1.0000 label 3:
         1.0000
         purity: 0.7520
         k = 2, perturbed initialization
         centroids converged: 4 iterations
         centroid 0: label 0: 0.9987 label 1: 0.0330 label 2: 0.2593 label 3:
         0.9612
         centroid 1: label 0: 0.0013 label 1: 0.9670 label 2: 0.7407 label 3:
         0.0388
         purity: 0.8390
         k = 4, perturbed initialization
         centroids converged: 4 iterations
         centroid 0: label 0: 0.0000 label 1: 0.0000 label 2: 0.0370 label 3:
         0.0000
         centroid 1: label 0: 0.8816 label 1: 0.0110 label 2: 0.0000 label 3:
         0.6214
         centroid 2: label 0: 0.0013 label 1: 0.9670 label 2: 0.7037 label 3:
         0.0388
         centroid 3: label 0: 0.1170 label 1: 0.0220 label 2: 0.2593 label 3:
         0.3398
         purity: 0.8410
         k = 4, trained initialization
         centroids converged: 5 iterations
         centroid 0: label 0: 0.9960 label 1: 0.0330 label 2: 0.2593 label 3:
         0.3689
         centroid 1: label 0: 0.0000 label 1: 0.5604 label 2: 0.0000 label 3:
         0.0000
         centroid 2: label 0: 0.0013 label 1: 0.4066 label 2: 0.7407 label 3:
         0.0388
         centroid 3: label 0: 0.0027 label 1: 0.0000 label 2: 0.0000 label 3:
         0.5922
```

purity: 0.9010

Discussion: centroid initialization has a significant impact on the results. The purity scores of the five runs range from 0.75 to 0.90. Not surprisingly the best results came from the centroids initialized with the class-specific aggregate data, and the worst results came from using randomized synthetic data. Ideally the results would suggest a diagonal of populated cells with all others being zero. The final scenario above comes closest, but there is still room for improvement.

In []:

1