EECS E6893 HW2

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Question 1

(1) Screenshots of the code

```
In [0]: from pyspark import SparkConf, SparkContext
        import pyspark
        from collections import defaultdict
In [0]: # Finished. Return RDD
        def getData(sc, filename):
            Load data from raw text file into RDD and transform.
            Hint: transfromation you will use: map(<lambda function>).
                 sc (SparkContext): spark context.
                 filename (string): hw2.txt cloud storage URI.
            Returns:
                 RDD: RDD list of tuple of (<User>, [friend1, friend2, ...]),
                 each user and a list of user's friends
            # read text file into RDD
            data = sc.textFile(filename)
            # TODO: implement your logic here
data = data.map(lambda line: np.array([str(x) for x in line.replace('\n','').split('\t')]))
            data = data.map(lambda p:(int(p[0]), p[1].split(',')))
```

```
In [0]: def mapFriends(line):
    """
    List out every pair of mutual friends, also record direct friends.
    Args:
        line (tuple): tuple in data RDD
    Yields:
        RDD: rdd like a list of (A, (B, 0)) or (A, (C, 1))
    """
    friends = line[1]
    user = line[0]

    if friends != ['']:
        for i in range(len(friends)):
            # Direct friend
            # TODO: implement your logic here
            yield((user,(int(friends[i]),0)))

        for j in range(i+1, len(friends)):
            # Mutual friend in both direction
            # TODO: implement your logic here
            yield((int(friends[i]), (int(friends[j]),1)))
            yield((int(friends[j]), (int(friends[i]),1)))
```

```
In [0]: def findMutual(line):
             Find top 10 mutual friend for each person.
                line (tuple): a tuple of (<User1>, [(<User2>, 0), (<User3>, 1)....])
                RDD of tuple (line[0], returnList),
                returnList is a list of recommended friends
             # friendDict, Key: user, value: count of mutual friends
             friendDict = defaultdict(int)
             # set of direct friends
             directFriend = set()
             # initialize return list
             returnList = []
             # TODO: Iterate through input to aggregate counts
            # save to friendDict and directFriend
user = line[0]
             friends = list(line[1])
             for i in range(len(friends)):
                 len(friends[i])
                 if friends[i][1] == 0:
                    directFriend.add(friends[i][0])
                    friendDict[friends[i][0]] = friendDict.get(friends[i][0],0) + 1
             # TODO: Formulate output
             sorted_friendDict = sorted(friendDict.items(), key = lambda x:(-x[1],x[0]))
             for i in sorted friendDict:
                 if len(returnList) == 10:
                    break
                 elif i[0] in directFriend:
                 else:
                     returnList.append(i[0])
            return (line[0], returnList)
In [0]: #def main():
         # Configure Spark
        conf = SparkConf()
         sc = SparkContext.getOrCreate(conf=conf)
         # The directory for the file
         filename = "/content/gdrive/My Drive/BigData/q1.txt"
In [0]: # Get data in proper format
         data = getData(sc, filename)
In [0]: # Get set of all mutual friends
      mapData = data.flatMap(mapFriends).groupByKey()
In [0]: # For each person, get top 10 mutual friends
getFriends = mapData.map(findMutual)
        #getFriends.take(5)
In [0]: # Only save the ones we want
        wanted = [924, 8941, 8942, 9019, 49824, 13420, 44410, 8974, 5850, 9993]
        result = getFriends.filter(lambda x: x[0] in wanted).collect()
In [0]: sc.stop()
```

(2) Screenshots of the recommendation results.

Question 2 Graph Analysis

(1) There are 917 clusters/connected components in total for this dataset.

(1). Number of clusters in this dataset

```
In [81]: result.select("component").distinct().count()
Out[81]: 917
```

(2) There are 49045 users in the top 10 clusters.

```
In [99]: count = result.groupBy("component").count().orderBy("count",ascending=False)
          (2) Top 10 clusters
In [100]: count.show(10)
          |component|count|
                   0 | 48860 |
               38403 İ
                        66
               18466
                        31
               18233
                        25
                864
               49297
               19199
                        6
                7658
                         5
               22897
In [104]: # number of users in the top 10 clusters
           from pyspark.sql.functions import sum as
           count.limit(10).agg(_sum("count")).show()
           sum(count)
                 49045
```

(3) The user ids for the cluster which has 25 users are: 18233 - 18257.

```
(3) List all 25 user IDS in cluster 18233
In [113]: count.filter("count=25").select("component").show()
          component
              18233
In [114]: result.filter("component=18233").select("id").show(25)
          | id|
          |18233|
          18234
          18235
          18236
          18237
          18238
           18239
          18240
          18241
          18242
          18243
          18244
        |18245|
        18246
        18247
        18248
        18249
         18250
        18251
         18252
        18253
         18254
        18255
         18256
        18257
```

(4) The most important user is the one with User ID 10164.

(4). Top 10 important users

(5) Using different parameters setting for PageRank would lead to differences in the result. ???未完成

(5). Try different parameters

```
In [116]: pr1 = g.pageRank(resetProbability=0.15, tol=0.01)
          prl.vertices.select("id", "pagerank").orderBy("pagerank",ascending=False).show(10)
             id
                          pagerank
           |10164|17.315312963089895|
           15496 14.866327204150846
           14689 12.685692559698428
           24966 12.26882183906656
            7884 11.827780808752543
             934 11.49589135687648
           45870 11.27397140801791
            5148 11.222433130678017
           20283 11.14062997830236
          46039 11.02696924843223
          only showing top 10 rows
In [117]: pr2 = g.pageRank(resetProbability=0.5, tol=0.01)
          pr2.vertices.select("id", "pagerank").orderBy("pagerank",ascending=False).show(10)
             idl
                           pagerank|
          |10164|18.539756319902864|
          15496 15.895700017529919
           14689 13.814565627780183
           24966 12.594967254720714
            5148 12.13232924938358
          38123 12.107079705652753
            7884 11.988217312291413
             934 | 11.939041942106776
             910 11.207783548336854
          44815 11.092504432507283
          only showing top 10 rows
In [118]: pr3 = g.pageRank(resetProbability=0.15, tol=0.1)
          pr3.vertices.select("id", "pagerank").orderBy("pagerank",ascending=False).show(10)
             id
          |10164|19.200290615258158|
          15496 16.546851217080825
           14689 14.940716809515001
           24966 13.124783956624656
            5148 12.759229785981626
          38123 12.556966112921204
             934 | 12.430209408516708
            7884 12.380173406826115
             910 11.995515035966134
          44815 11.990097101490727
          only showing top 10 rows
In [120]: pr4 = g.pageRank(resetProbability=0.15, tol=0.01, sourceId=10164)
          pr4.vertices.select("id", "pagerank").orderBy("pagerank",ascending=False).show(10)
          | id|
                            pagerank
          10164 0.5405405405405407
           10239 0.004594594594594596
           10182 0.004594594594594596
           10246 0.004594594594594596
           10178 0.004594594594594596
          10176 0.004594594594594596
           10168 0.004594594594594596
           10166 0.004594594594594596
          110237 | 0.004594594594594596
             222 0.004594594594594596
          only showing top 10 rows
```

(6) ???未完成. 10164 是最 important 的 也不是特别清楚,就是它在最大的

(7) PageRank Calculation

For this question, I used self-defined function to do the calculation and the result is as following along with the code.

```
Iteration 0 . Page rank: {'ID1': 0.2, 'ID2': 0.2, 'ID3': 0.2, 'ID4': 0.2, 'ID5': 0.2}
Iteration 1 . Page rank: {'ID1': 0.07, 'ID2': 0.29, 'ID3': 0.41, 'ID4': 0.07, 'ID5': 0.16}
Iteration 2 . Page rank: {'ID1': 0.09, 'ID2': 0.45, 'ID3': 0.25, 'ID4': 0.09, 'ID5': 0.12}
Iteration 3 . Page rank: {'ID1': 0.13, 'ID2': 0.29, 'ID3': 0.29, 'ID4': 0.13, 'ID5': 0.16}
Iteration 4 converges.
Page rank: {'ID1': 0.09, 'ID2': 0.34, 'ID3': 0.33, 'ID4': 0.09, 'ID5': 0.15}
```

```
Т
1 iteration = 0
    while True:
 3
       flag = True
 4
        newPR = dict()
       print("Iteration", iteration, ". Page rank:", PR)
for i in L.keys():
 6
            pr = (1-d)/N
 7
 8
             if type(M[i])==str:
 9
                pr += d*PR[M[i]]/L[M[i]]
10
           else:
           for j in M[i]:
11
12
                    pr += d*PR[j]/L[j]
         pr += d*PR[j]/L[j]
pr = float("{0:.2f}".format(pr))
if abs(pr - PR[i]) > tol:
13
                 flag = False
15
             newPR[i] = pr
16
      newPl
if flag:
17
        print("Iteration", iteration+1, "converges.")
print("Page rank:", newPR)
19
20
             break
      PR = newPR
21
22 iteration += 1
```