Anubha Bhargava – Homework 3

1. I installed and tested Spark. I ran a few tests on Spark, as shown below to test the installation. I imported a textfile into Spark as shown below.

```
>>> lines = sc.textFile("README.md")
15/11/05 07:46:08 INFO MemoryStore: ensureFreeSpace(143840) called with curMem=6,
, maxMem=560497950
15/11/05 07:46:08 INFO MemoryStore: Block broadcast_0 stored as values in memory
(estimated size 140.5 KB, free 534.4 MB)
15/11/05 07:46:08 INFO MemoryStore: ensureFreeSpace(12673) called with curMem=14
3840, maxMem=560497950
15/11/05 07:46:08 INFO MemoryStore: Block broadcast_0_piece0 stored as bytes in
memory (estimated size 12.4 KB, free 534.4 MB)
15/11/05 07:46:08 INFO BlockManagerInfo: Added broadcast_0_piece0 in memory on l
ocalhost:41056 (size: 12.4 KB, free: 534.5 MB)
15/11/05 07:46:08 INFO SparkContext: Created broadcast 0 from textFile at Native
MethodAccessorImpl.java:-2
```

I ran spark-submit, Spark in Python and Spark and Scala successfully as shown:

```
bigdata@ubuntu:/usr/local/spark-1.5.1-bin-hadoop2.4$ ./bin/spark-submit examples
/src/main/python/pi.py 10
Using Spark's default log4j profile: org/apache/spark/log4j-defaults.properties
15/11/05 07:22:23 INFO SparkContext: Running Spark version 1.5.1
15/11/05 07:22:31 INFO TaskSetManager: Finished task 7.0 in stage 0.0 (TID 7) in
245 ms on localhost (8/10)
15/11/05 07:22:31 INFO Executor: Running task 8.0 in stage 0.0 (TID 8)
15/11/05 07:22:31 INFO PythonRunner: Times: total = 78, boot = 4, init = 6, fini
15/11/05 07:22:31 INFO Executor: Finished task 8.0 in stage 0.0 (TID 8). 998 byt
es result sent to driver
15/11/05 07:22:31 INFO TaskSetManager: Starting task 9.0 in stage 0.0 (TID 9, lo
calhost, PROCESS_LOCAL, 503631 bytes)
Welcome to
                              version 1.5.1
Using Python version 2.7.6 (default, Mar 22 2014 22:59:56)
SparkContext available as sc, HiveContext available as sqlContext.
>>>
    adjust toggting tevet use setsettoglevet( INFO
 Welcome to
                               version 1.5.1
 Using Scala version 2.10.4 (Java HotSpot(TM) 64-Bit Server VM, Java 1.7.0 79)
 Type in expressions to have them evaluated.
 Type thelp for more information
```

Then, I used the Wikipedia data to calculate TF-IDF and performed clustering. First, I parsed the Wikipedia dataset using the Wikipedia-extractor.py python script.

```
wikipedia-extractor.py ×
# Codes starts here
import lxml.etree
import os
tree = lxml.etree.parse('/home/bigdata/Documents/Homework2/Wikipedia/
Wikipedia20150602/enwiki-latest-pages-articles1.xml-
p000000010p000010000')
namespaces = {'ns':'http://www.mediawiki.org/xml/export-0.10/'}
i = 0
el_list = tree.xpath('//ns:page', namespaces = namespaces)
for el in el_list:
    title = el.xpath('.//ns:title', namespaces = namespaces)[0].text
    print title
    f = file('parse/' + str(i), 'w')
    text = el.xpath('.//ns:revision/ns:text', namespaces = namespaces)
[0].text
    f.write(title.encode('utf-8') + '\n')
    f.write(text.encode('utf-8'))
    i = i + 1
    f.close()
# Codes ends here
bigdata@ubuntu:~/Documents/Homework3/parse$ ls
      1432 1868 2301 2737 3171
                                    3606
                                                4476
                                                      4910 5345
                                                                  5780
                                                                        6214
      1433 1869 2302
                      2738 3172
                                    3607
                                          4041
                                                4477
                                                      4911
                                                            5346
                                                                  5781
                                                                        6215
10
      1434 187
                                                4478
                  2303
                       2739
                             3173
                                    3608
                                          4042
                                                      4912
                                                           5347
                                                                  5782
                                                                        6216
```

1870 2304

1876 231

Then, I created a script called cluster_tfidf.py which completed both clustering and TFIDF.

```
cluster_tfidf.py x
from pyspark.mllib.feature import HashingTF, IDF
from pyspark.mllib.clustering import KMeans, KMeansModel
from numpy import array
from math import sqrt
from pyspark import SparkContext
sc=SparkContext()
rdd=sc.wholeTextFiles("/home/bigdata/Documents/Homework3/parse2/").map(lambda
(name,text): text.split())
tf=HashingTF()
tfVectors=tf.transform(rdd).cache()
idf=IDF()
idfModel=idf.fit(tfVectors)
tfIdfVectors=idfModel.transform(tfVectors)
model=KMeans.train(tfIdfVectors, 2, maxIterations=10, runs=10,
initializationMode="random")
clusters = model.predict(tfIdfVectors)
dataWithClusters = tfIdfVectors.zip(clusters)
dataWithClusters.saveAsTextFile("/home/bigdata/Documents/Homework3/
dataWithClusters/")
tfIdfVectors.saveAsTextFile("/home/bigdata/Documents/Homework3/tfIdfVectors/")
```

This is the output that was received from the script:

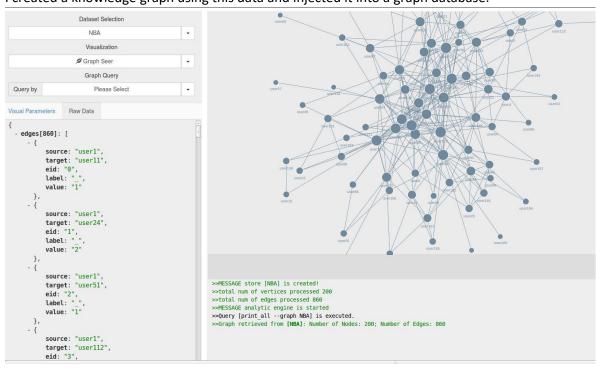
```
(SparseVector(1048576, {2550: 12.0817, 3932: 0.3463, 19267: 3.0204,
22582: 3.0204, 30674: 1.2287, 36754: 0.3463, 38034: 1.2287, 43344:
3.0204, 46446: 2.1041, 49181: 1.411, 50570: 1.0388, 50583: 0.6247,
72971: 3.0204, 89423: 3.0204, 93626: 3.0204, 95620: 2.615, 95965:
6.3124, 102217: 3.0204, 121051: 3.0204, 137800: 3.0204, 141964: 3.0204 148055: 3.0204, 152402: 2.615, 159190: 3.0204, 160072: 3.0204, 167368:
3.0765, 168186: 15.1021, 176055: 3.0204, 182263: 6.0408, 189356:
0.6225, 195374: 3.0204, 196347: 11.6364, 200383: 1.3157, 211696:
3.0204, 213150: 3.0204, 218450: 3.0204, 225722: 3.0204, 234013: 0.8355,
235870: 3.0204, 236823: 12.0817, 251696: 3.0204, 261722: 1.9218,
263998: 3.0204, 267062: 5.2299, 280048: 3.0204, 284784: 2.615, 292923:
0.7178, 293156: 0.1582, 323305: 6.2475, 326588: 2.3123, 338406: 2.1041,
340523: 2.3273, 348047: 4.0221, 354917: 1.3381, 356253: 2.1041, 358193:
3.0204, 362969: 0.669, 382094: 3.0204, 385604: 3.0204, 393486: 3.0204,
413733: 36.6094, 417558: 1.6341, 420024: 3.0204, 420843: 0.669, 429729:
2.615, 442539: 6.0408, 443614: 3.0204, 448789: 1.2287, 448852: 3.0204,
452315: 8.4165, 457375: 3.0204, 470253: 2.615, 497087: 3.0204, 498760:
```

This algorithm is similar to that in Mahout, as it dumps out the data in clusters.

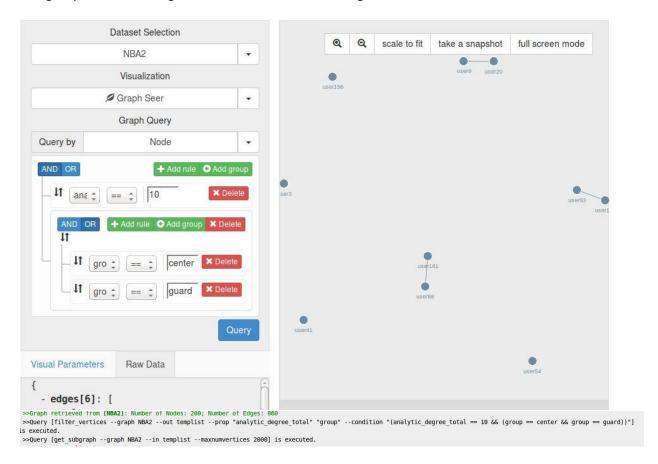
2. I downloaded IBM System G Graph Tools and the Basketball data. I used the following for the edges and nodes in the software:



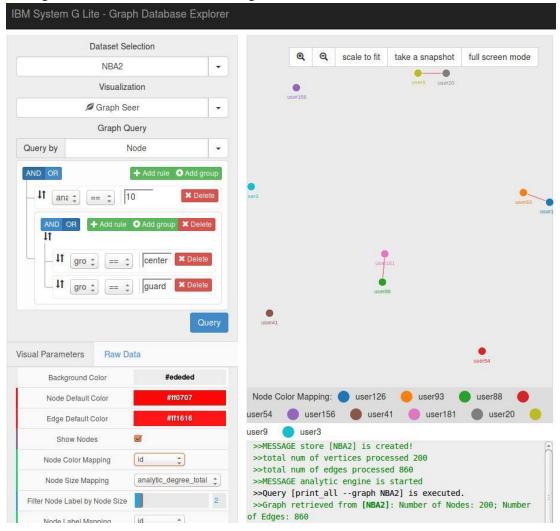
I created a knowledge graph using this data and injected it into a graph database.



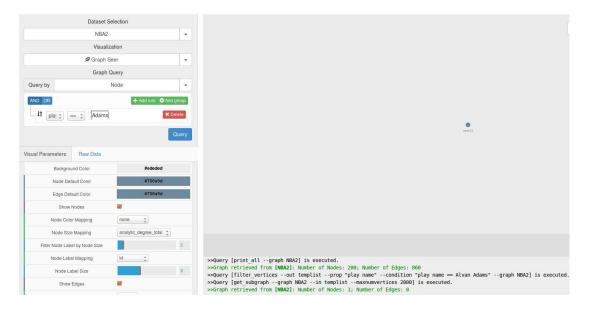
I then tried some visualizations with my queries, as shown below. I set the analytic degree total to 10, and group to center and guard and received the following results.



I changed the colors of the nodes and edges.



I looked up the player Alvan Adams and received one player, as displayed.



I was unable to use the Wikipedia dataset, as I couldn't convert it from a .sql file. I successfully analyzed the Basketball data, however.