

EECS E6893 Big Data Analytics HW2: Friend Recommendations, GraphFrames

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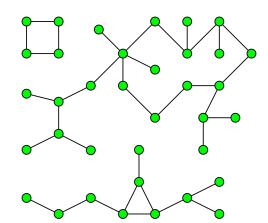
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<u>GraphFrames</u>

- DataFrame-based Graph
- GraphX is to RDDs as GraphFrames are to DataFrames
- Represent graphs: vertices (e.g. users) and edges (e.g. relationships between users)
- GraphFrames package separate from core Apache Spark

Connected components

- A subgraph where any two vertices are connected to each other by edges, but not connected to other vertices in the graph
- In a social network, connected components can approximate clusters
- In the GraphFrame, the connected components algorithm labels each connected component of the graph with the ID of its lowest-numbered vertex



Reference: https://en.wikipedia.org/wiki/Component (graph theory)

PageRank

- PageRank measures the importance of each vertex in a graph
- An edge from u to v represents an endorsement of v's importance by u

$$PR(p_i) = rac{1-d}{N} + d\sum_{p_j \in M(p_i)} rac{PR(p_j)}{L(p_j)}$$

d: damping factor;

default = 0.85 - 15% chance that a typical users won't follow any links on the page and instead navigate to a new random URL.

Convergence occurs when all PageRank values are within the margin of error.

Reference: https://en.wikipedia.org/wiki/PageRank

PageRank (Spark)

pageRank(resetProbability=0.15, sourceId=None, maxIter=None, tol=None)

Parameters:

resetProbability: 1-d, Probability of resetting to a random vertex, default=0.15

maxIter: If set, the algorithm is run for a fixed number of iterations.

tol: If set, the algorithm is run until the given tolerance/margin of error.

NOTE: Exactly one of *maxIter* or *tol* must be set.

HW2

- Question 1: Friend Recommendations
- Question 2: Graph Analysis using GraphFrames

Environment Setup

Cloud Shell:

- Create multiple workers on Dataproc instead of single node, otherwise it will take long time to run.
- 2. Install graphframe package in spark when create the cluster. (You can reference to config Spark properties)

- Write a Spark program that implements a simple "People You Might Know" social network friendship recommendation algorithm. The key idea is that if two people have a lot of mutual friends, then the system should recommend that they connect with each other.
- Question: Give recommendation for 10 Users

- Dataset Format
 - <User> <Tab> <Friends>
 - <User> is a unique ID; <Friends> are comma separated list of unique IDs

Q1 - Code Skeleton

```
#Configure Spark
conf = SparkConf()
sc = SparkContext(conf=conf)

# The directory that save the hw2.txt
filename = "<your hw2.txt cloud storage URI>"
```

```
# Get data in proper format
data = getData(sc, filename)
```

```
# Get set of all mutual friends
mapData = data.flatMap(mapFriends).groupByKey()
```

```
# For each person, get top 10 mutual friends
getFriends = mapData.map(findMutual)
```

```
# The final results
wanted = [924, 8941, 8942, 9019, 49824, 13420, 44410, 8974, 5850, 9993]
getFriends.filter(lambda x: x[0] in wanted).collect()
```

Q1 - Function example

```
def findMutual(line):
   Find top 10 mutual friend for each person.
   Hint: For each <User>, input is a list of tuples of friend relations,
   whether direct friend (count = 0) or has friend in common (count = 1)
   Use friendDict to store the number of mutual friend that the current <User>
   has in common with each other <User> in tuple.
   Input:(User1, [(User2, 1), (User3, 1), (User2, 1), (User3, 0), (User2, 1)])
   friendDict stores: {User2:3, User3:1}
    directFriend stores: User3
   If a user has many mutual frineds and is not a direct friend, we recommend
    them to be friends.
    Args:
       line (tuple): a tuple of (<User1>, [(<User2>, 0), (<User3>, 1)....])
   Returns:
       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
    # TODO: Formulate output
   return (line[0], returnList)
```

- Use the Q1 dataset again do the graph analysis
- Connected Component
- PageRank

- Steps 1
 - Format the provided dataset into two Spark DataFrames: edges and vertices
 - Notice: For the vertices, if there is no other properties for vertices (like in our case), then we should create tuple like this, otherwise a string inside a tuple will not be identified as a tuple but as a single string. If there are other properties, then no need for that extra comma.

```
vertices = data.map(lambda x: (x[0], ))

vertices.take(5)

[('0',), ('1',), ('2',), ('3',), ('4',)]

edges.take(5)

[('0', '1'), ('0', '2'), ('0', '3'), ('0', '4'), ('0', '5')]
```

- Step 2
 - Convert the RDD to DataFrame
 - Directly convert to DataFrame
 - Save RDD to csv, then read csv to DataFrame

```
v = spark.createDataFrame(vertices, ["id"])
```

- Step 3
 - Create graph

```
from graphframes import *
```

```
g = GraphFrame(v, e)
```

If you set the environment correctly following the instructions above, there should be no problem with Jupyter.

If you are using Spark shell and it doesn't work, you could try running:

```
pyspark --packages
graphframes:0.6.0-spark2.3-s_2.11
running Spark
```

Q2 - Connected components

Notice

If you are using Connected components, and get the error like

```
Py4JJavaError: An error occurred while calling o151.run.
: java.io.IOException: Checkpoint directory is not set. Please set it first using sc.setCheckpointDir().
```

You could reference the following answer on stackoverflow

https://stackoverflow.com/questions/49159896/how-to-set-checkpiont-dir-pyspark-data-science-experience

Q2 - PageRank

- results = g.pageRank(resetProbability=0.15, tol=0.01)
- There are multiple parameters. You can play with them, see whether there are different result.

References

- https://graphframes.github.io/graphframes/docs/_site/index.html
- https://www.cs.princeton.edu/~chazelle/courses/BIB/pagerank.htm