The Brave Ducks

Checkpoint 4: Graph Analytics

Overview: With the following exploration of the Chicago Polic Database, we inspect co-offending officers and if they have any interesting relationships within each pair of officers. The exploration of the following questions is done via graph analytics, conducted using Apache Spark and GraphX tools, processed in the attached Google Colab notebook.

Questions we seek to answer with the graph analytics:

- 1. Are there occurrences of co-offending officers on the same misconduct report and how often do the same co-offending officers repeat?
- 2. What is the salary, rank, and race relationship between the topmost pair of co-offending officers? This analysis can then be repeated for each pair, as needed. For this report, we will provide discussion for the topmost offending pair.
 - o do the officers have comparable salaries within 10% of each other?
 - are the officers of similar ranking or is one a higher ranking officer, possibly socially pressuring a lower ranking officer into committing misconduct?
 - o are the officers the same race?
- 3. How many unique co-offending relationships does each officer have? It is straightforward to compute total misconduct of an individual officer but will be more meaningful to understanding how many unique relationships an officer has in which he/she allegedly commits a misconduct.
 - o we will use the Triangle Count algorithm to count the unique relationships connecting officer_id nodes
- 4. Who is the ring-leader (most important) co-offending officer?

```
# install java
!apt-get install openjdk-8-jdk-headless -qq > /dev/null
# install spark (change the version number if needed)
!wget -q https://archive.apache.org/dist/spark/spark-3.2.0/spark-3.2.0-bin-hadoop3.2.tgz
# unzip the spark file to the current folder
!tar xf spark-3.2.0-bin-hadoop3.2.tgz
# set your spark folder to your system path environment.
import os
os.environ["JAVA_HOME"] = "/usr/lib/jvm/java-8-openjdk-amd64"
os.environ["SPARK HOME"] = "/content/spark-3.2.0-bin-hadoop3.2"
# install findspark using pip
!pip install -q findspark
# install pyspark
!pip3 install pyspark==3.2.0
# install graphframes
!pip3 install graphframes
    Requirement already satisfied: pyspark==3.2.0 in /usr/local/lib/python3.7/dist-packages (3.2.0)
    Requirement already satisfied: py4j==0.10.9.2 in /usr/local/lib/python3.7/dist-packages (from pyspark==3.2.0) (0.10.9.2)
    Requirement already satisfied: graphframes in /usr/local/lib/python3.7/dist-packages (0.6)
    Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from graphframes) (1.19.5)
    Requirement already satisfied: nose in /usr/local/lib/python3.7/dist-packages (from graphframes) (1.3.7)
```

Download the graphframes jar file from: Graphframe jar file:

import psycopa2

Upload it in the Google Colab Files folder. Can be found in the left pane of this window.

!cp -v /content/graphframes-0.8.2-spark3.2-s_2.12.jar \$SPARK_HOME/jars/

```
'/content/graphframes-0.8.2-spark3.2-s_2.12.jar' -> '/content/spark-3.2.0-bin-hadoop3.2/jars/graphframes-0.8.2-spark3.2-s_2.1
#import the packages
from pyspark import *
from pyspark.sql import *
from graphframes import *
import findspark
import pandas as pd

findspark.init()
# Start a Spark session
spark = SparkSession.builder.master("local[*]").getOrCreate()
```

```
conn = psycopg2.connect(
   host="codd04.research.northwestern.edu",
   port = "5433",
   database="postgres",
   user="cpdbstudent",
   password="DataSci4AI")
cursor = conn.cursor()
```

CPDB

Analyze the police officers connection with police misconducts, i.e. (illegal search and use of force)

```
edges_query = "SELECT dal.officer_id src, da2.officer_id dst, COUNT(DISTINCT dal.allegation_id) relationship \
FROM data_officerallegation da1, \
          data_officerallegation da2, \
          data_allegationcategory dcat \
WHERE dal.allegation_id = da2.allegation_id \
        AND dal.allegation_category_id = dcat.id \
        AND dal.officer_id < da2.officer_id \
        AND (dcat.category like 'Illegal Search' or dcat.category like 'Use Of Force') \
GROUP BY dal.officer_id, da2.officer_id \
        ORDER BY count(*) DESC;"</pre>
```

Following query creates nodes and edges to answer the questions.

• nodes: id, officer name and misconduct count

```
    edges: src(officer1 id), dist(officer2 id) and relationship(misconduct count)

nodes_query = "SELECT da.officer_id id, doff.first_name || ' ' || doff.last_name officer_name, doff.race, doff.currer
FROM data_officerallegation da, \
data_officer doff, \
data_allegationcategory dcat \
WHERE da.allegation_category_id = dcat.id \
AND doff.id = da.officer_id \
AND (dcat.category like 'Illegal Search' or dcat.category like 'Use Of Force') \
AND current_salary is not null \
GROUP BY da.officer_id, officer_name, doff.race, doff.current_salary, doff.rank; "
cursor.execute(edges_query)
edges = cursor.fetchall()
print("shape is: " + str(len(edges)))
df_edges = pd.DataFrame(edges)
colnames = [desc[0] for desc in cursor.description]
df_edges.columns = colnames
print(df_edges.shape)
    shape is: 92794
     (92794, 3)
cursor.execute(nodes_query)
nodes = cursor.fetchall()
print("shape is: " + str(len(nodes)))
df_nodes = pd.DataFrame(nodes)
colnames = [desc[0] for desc in cursor.description]
df nodes.columns = colnames
print(df_nodes.shape)
     shape is: 13951
     (13951, 6)
edges_·=·spark.createDataFrame(df_edges)
nodes · = · spark.createDataFrame(df_nodes)
```

The Results

cpdb · = · GraphFrame(nodes, · edges)

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+	+		+	+	+
id	officer_name	race	current_salary	rank	total_misconduct_count
+	 		+	+	+
1	Jeffery Aaron	White	101442	Sergeant of Police	2
2	Karina Aaron	Hispanic	94122	Police Officer as	4
4	Carmel Abbate	White	74946	Police Officer as	2
6	Anthony Abbate	White	70656	Police Officer	2
7	Terry Abbate	White	93354	Police Officer	3
8	Leon Abbey	Black	73116	Police Officer	1
11	Laura Abbott	White	73476	Police Officer as	2
13	Dale Abbott	White	85278	Police Officer	2
14	Elizabeth Abbott	White	82878	Police Officer	1
16	Aziz Abdelmajeid	Asian/Pacific	84054	Sergeant of Police	9
	Moulay Abdullah			Police Officer	1
18	Jason Abejero	Asian/Pacific	90024	Police Officer	1
j 20	Kenneth Abels			Sergeant of Police	2
j 33	Ricardo Abreu	Hispanic	74946 j	Police Officer as	10
j 34	Floyd Abron	Black	90024	Police Officer	5 į
j 38	Abdalla Abuzanat	Asian/Pacific	97440	Police Officer as	5
	Rosemary Accardo			Police Officer	10
	Jennifer Accardo		87006	Police Officer	2
	Thomas Accardo		90024	Police Officer	6
j 44	Marco Acevedo	Hispanic	100980	Police Officer as	10
+		- 	+	+	·
only	showing top 20 rd	DWS			

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Question 1: Are there occurrences of co-offending officers on the same misconduct report and how often do the same co-offending officers repeat?

The table below represents a graph with source node being officer_id1 and destination node being officer_id2, while their relationship is the total count of co-offending misconduct.

cpdb.edges.show()

++	++
src dst	relationship
+	++
12478 32166	36
8562 27778	34
2725 21703	29
1553 10724	28
3605 14442	
8562 18206	'
12074 12825	'
32265 32347	•
8562 23841	
31882 32401	
13361 20150	
1553 16699	-
23841 27778	•
32016 32213	
14731 27602	'
14045 15502	•
12479 20713	•
17285 17397	
	•
18206 27778	
8658 13788	21
+	++
only showing	top 20 rows

Question 2: What is the salary, rank, and race relationship between the topmost pair of co-offending officers?

- do the officers have comparable salaries within 10% of each other?
- are the officers of similar ranking or is one a higher ranking officer, possibly socially pressuring a lower ranking officer into committing misconduct?
- are the officers the same race?

To gain more insight into the top most co-offending pair of officers, we inspect the graph vertices to see the exact names, salary, rank, and race of the officers with the most total misconduct counts. As we can see below, the officers Ronald Holt and Emmet Mc Clendon have the most co-offending misconduct complaints. There is a more than 10% discrepency between their salaries and their officer rankings are significantly different, implying some hierarchical relationship between officer Holt and officer McClendon. Further, we note that both officers in this pair are of race black.

```
cpdb.vertices.filter('id=12478').show()
cpdb.vertices.filter('id=32166').show()
```

+----+

+----+
| id| officer_name| race|current_salary| rank|total_misconduct_count|
+----+
|32166|Emmett Mc Clendon|Black| 107988|Sergeant of Police| 64|

Question 3: How many unique co-offending relationships does each officer have?

Background on the Triangle Count algorithm:

The Triangle Count algorithm counts the number of triangles for each node in the graph. A triangle is a set of three nodes where each node has a relationship to the other two. In graph theory terminology, this is sometimes referred to as a 3-clique. The Triangle Count algorithm in the GDS library only finds triangles in undirected graphs.

Triangle counting has gained popularity in social network analysis, where it is used to detect communities and measure the cohesiveness of those communities.

(source: https://neo4j.com/docs/graph-data-science/current/algorithms/triangle-count/#:~:text=The%20Triangle%20Count%20algorithm%20counts,to%20as%20a%203%2Dclique):

We see from our triangle count computation on the graph that officer Vincent Stinar has the highest number of unique co-offending relationships with other officers. And although we understood the Triangle Count algorithm to compute the unique relationship each officer has, as described above, the computation result we get from the code does not make immediate intuitive sense. We do a sanity-check on the computation by printing out the noted officer id '32356' as being the officer with the most unique co-offending relationships, being officer Vincent Stinar. We note that the TC algorithm computes he has 1514 unique co-offending relationships but the following computation for officer Stinar's total misconduct count is much lower, at total count equal to 12. This does not make intuitive sense for officer Vincent Stinar to have 1514 unique co-offending relationships but only 12 counts of misconduct. Upon further investigation and debugging of the code, we are unable to identify any specific bug in the code to rectify this error.

[Edit]: We previously noted that officer_id 32356, Officer Vincent Stinar, has the greatest number of co-offending relationships as analyzed by the trianglecount algorithm. It was also noted that these are unique co-offending triangular relationships and that it did not make intuitive sense to have such a large trianglecount result, 1514, while having only 12 total counts of misconduct allegations. After further review of the Triangle Count algorithm, we find that the analysis is not of *unique* co-offending relationships, but repeated count of relationships. This would then mean it is reasonable to see a larger triangle count result even if the total misconduct count is much lower.

```
tc_cpdb = cpdb.triangleCount()

tc_cpdb.select("id", "count").sort(['count'], ascending=[0]).show()
cpdb.vertices.filter('id=32356').show()
```

```
+----+
| id|count|
+----+
|32356| 1514|
|31536| 1485|
|32390| 1437|
|25230| 1417
|22554| 1390
2375 | 1369 |
6704 | 1366 |
|21364| 1347|
|30337| 1316|
|25983| 1312
2201| 1287
132/2| 12//
|13093| 1263
|10724| 1237
|28384| 1230
| 9648| 1204
7032 | 1196
|12947| 1186
| 6852| 1182|
| 2356| 1179|
+----+
only showing top 20 rows
id| officer_name| race|current_salary| rank|total_misconduct_count|
|32356|Vincent Stinar|White| 90024|Police Officer|
```

da1.officer_id < da2.officer_id" is modified to "AND da1.officer_id <> da2.ifficer_id".

```
edges_query = "SELECT da1.officer_id src, da2.officer_id dst, COUNT(DISTINCT da1.allegation_id) relationship \
FROM data_officerallegation da1, \
    data_officerallegation da2, \
     data_allegationcategory dcat \
WHERE dal.allegation id = da2.allegation id \
  AND dal.allegation category id = dcat.id \
 AND dal.officer_id <> da2.officer_id \
 AND (dcat.category like 'Illegal Search' or dcat.category like 'Use Of Force') \
GROUP BY dal.officer_id, da2.officer_id \
ORDER BY count(*) DESC;"
nodes_query = "SELECT da.officer_id id, doff.first_name || ' ' || doff.last_name officer_name, doff.race, doff.currer
FROM data_officerallegation da, \
data_officer doff, \
data_allegationcategory dcat \
WHERE da.allegation_category_id = dcat.id \
AND doff.id = da.officer_id \
AND (dcat.category like 'Illegal Search' or dcat.category like 'Use Of Force') \
AND current_salary is not null \
GROUP BY da.officer_id, officer_name, doff.race, doff.current_salary, doff.rank; "
cursor.execute(edges_query)
edges = cursor.fetchall()
print("shape is: " + str(len(edges)))
df_edges = pd.DataFrame(edges)
colnames = [desc[0] for desc in cursor.description]
df edges.columns = colnames
print(df_edges.shape)
    shape is: 185528
    (185528, 3)
edges = spark.createDataFrame(df edges)
nodes = spark.createDataFrame(df_nodes)
cpdb = GraphFrame(nodes, edges_)
```

New graph for use in PageRank:

cpdb.vertices.show()

+	+	+	-		
id	officer_name	race	current_salary	rank	total_misconduct_count
1	Jeffery Aaron	White	101442	Sergeant of Police	2
j 2	Karina Aaron	Hispanic	94122	Police Officer as	4
4	Carmel Abbate	White	74946	Police Officer as	2
6	Anthony Abbate	White	70656	Police Officer	2
7	Terry Abbate	White	93354	Police Officer	3
8	Leon Abbey	Black	73116	Police Officer	1
11	Laura Abbott	White	73476	Police Officer as	2
13	Dale Abbott	White	85278	Police Officer	2
14	Elizabeth Abbott	White	82878	Police Officer	1
16	Aziz Abdelmajeid	Asian/Pacific	84054	Sergeant of Police	9
17	Moulay Abdullah	Black	83706	Police Officer	1
18	Jason Abejero	Asian/Pacific	90024	Police Officer	1
20	Kenneth Abels	White	106068	Sergeant of Police	2
33	Ricardo Abreu	Hispanic	74946	Police Officer as	10
34	Floyd Abron	Black	90024	Police Officer	5
38	Abdalla Abuzanat	Asian/Pacific	97440	Police Officer as	5
39	Rosemary Accardo	White	92316	Police Officer	10
41	Jennifer Accardo	White	87006	Police Officer	2
42	Thomas Accardo	White	90024	Police Officer	6
44	Marco Acevedo	Hispanic	100980	Police Officer as	10
+	+	+	h		+
only	showing top 20 rd	OWS			

27778 8562	34
2725 21703	29
21703 2725	29
18206 8562	28
12074 12825	28
12825 12074	28
14442 3605	28
8562 18206	28
10724 1553	28
3605 14442	28
1553 10724	28
32265 32347	27
32347 32265	27
23841 8562	26
8562 23841	26
13361 20150	25
31882 32401	25
+	
only showing	top 20 rows

Result of ring leader analysis using the PageRank algorithm

We use the PageRank algorithm on our graph to highlight which officers perform as ring leaders in the perspective of the data.

As visualized above, the pagerank algorithm shows that officer Glenn Evans has the most influence on other officers in co-offending misconduct cases with a PageRank of 9.22 and total misconduct count of 73. Although at first glance the total misconduct count may seem less than other entries, we must note that PageRank will highlight which officer has the most commonality with other officers in committing misconduct offenses; thereby being "ring leaders". Other officers may commit additional misconduct offenses separately as individuals and thereby increasing their overall misconduct count.

To further highlight this point, note that Officer Jerome Finnigan, officer_id 8562, has the highest total of misconduct allegations yet is only the 4th highest ranking "ring-leader" as analyzed by PageRank.

```
pr_cpdb = cpdb.pageRank(resetProbability=0.15, tol=0.01)
#look at the pagerank score for every vertex
pr cpdb.vertices.orderBy('pagerank', ascending=False).show()
```

pagerank	al_misconduct_count	rank	current_salary	race	officer_name	id
9.224336490995743	73	Lieutenant of Police	125190	Black	Glenn Evans	8138
8.781453450615974	91	Police Officer	86130	White	Edward May	17816
7.753717937769441	29	Commander	127596	White	David Jarmusz	13303
7.205109026215454	116	Police Officer	73116	White	Jerome Finnigan	8562
6.999806910246934	38	Police Officer	93354	Hispanic	Gerardo Perez	32255
6.925574000926473	52	Sergeant of Police	101958	White	Marvin Bonnstetter	2375
6.727067377907265	55	Sergeant of Police	111474	Hispanic	Baudilio Lopez	16567
6.375518197546371	41	Sergeant of Police	111474			9821
6.3229051514833765	23	Lieutenant of Police	125190	White	Robert Stasch	27392
6.255302627009381	51	Sergeant of Police	111474	Black	Eric Cato	31859
5.905031155488198	40	Commander	154932	White	James O Grady	20959
5.880121795999848	36	Sergeant of Police	111474	White	John Lucid	16699
5.878604323220897	69	Lieutenant of Police	162684	Hispanic	James Sanchez	25306
5.8610254549394565	37	Sergeant of Police	111474	Hispanic	Michael Bocardo	31834
5.761828279300101	35	Police Officer as	87354	White	Bernard Graf	10528
5.7371383579010855	55	Police Officer	90024	White	Carl Suchocki	27778
5.699532705181115	62	Sergeant of Police	93354	Black	Tamara Matthews	32164
5.692547823980591	36	Police Officer	93240	Hispanic	Luis Vega	29445
5.550108151684686	40	Sergeant of Police	107988	White	Michael Stack	27270
5.522336960533268	38	Police Officer	93354	Hispanic	Louis Ortoneda	32237

only showing top 20 rows