```
In [20]: !pip install neo4j
!pip install graphdatascience
%matplotlib inline

import json
import time
import networkx as nx
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from tqdm import tqdm
from neo4j import GraphDatabase
from graphdatascience import GraphDataScience as gds
#from neo4j import GraphDataScience as gds
```

Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheel s/public/simple/ Requirement already satisfied: neo4j in /usr/local/lib/python3.9/dist-packages (5. 7.0) Requirement already satisfied: pytz in /usr/local/lib/python3.9/dist-packages (fro m neo4j) (2022.7.1) Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheel s/public/simple/ Requirement already satisfied: graphdatascience in /usr/local/lib/python3.9/dist-p ackages (1.6) Requirement already satisfied: tqdm<5.0,>=4.0 in /usr/local/lib/python3.9/dist-pac kages (from graphdatascience) (4.65.0) Requirement already satisfied: neo4j<6.0,>=4.4.2 in /usr/local/lib/python3.9/distpackages (from graphdatascience) (5.7.0) Requirement already satisfied: pyarrow<11.0,>=4.0 in /usr/local/lib/python3.9/dist -packages (from graphdatascience) (9.0.0) Requirement already satisfied: multimethod<2.0,>=1.0 in /usr/local/lib/python3.9/d ist-packages (from graphdatascience) (1.9.1) Requirement already satisfied: pandas<2.0,>=1.0 in /usr/local/lib/python3.9/dist-p ackages (from graphdatascience) (1.5.3) Requirement already satisfied: pytz in /usr/local/lib/python3.9/dist-packages (fro m neo4j < 6.0, >=4.4.2-> graphdatascience) (2022.7.1) Requirement already satisfied: numpy>=1.20.3 in /usr/local/lib/python3.9/dist-pack ages (from pandas<2.0,>=1.0->graphdatascience) (1.22.4) Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.9/ dist-packages (from pandas<2.0,>=1.0->graphdatascience) (2.8.2) Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.9/dist-packages

```
In []: # A wrapper class to manage connection
    class Neo4jConnection:

    def __init__(self, uri, user, pwd):
        self._uri = uri
        self._user = user
        self._pwd = pwd
        self._driver = None
        try:
            self._driver = GraphDatabase.driver(self._uri, auth=(self._user, selexcept Exception as e:
            print("Failed to create the driver:", e)

    def close(self):
        if self._driver is not None:
            self._driver.close()
```

(from python-dateutil>=2.8.1->pandas<2.0,>=1.0->graphdatascience) (1.16.0)

```
def query(self, query, parameters=None, db=None):
    assert self.__driver is not None, "Driver not initialized!"
    session = None
    response = None
    try:
        session = self.__driver.session(database=db) if db is not None else sel
        response = list(session.run(query, parameters))
    except Exception as e:
        print("Query failed:", e)
    finally:
        if session is not None:
            session.close()
        return response
In []: # Actually create a connection. Copy uri (with bolt port), username and passwords in the content of th
```

Q1 read node (character) files for all eight seasons and merge them to the same database.

conn = Neo4jConnection(uri="bolt://18.204.247.82:7687", user="neo4j", pwd="airplane

Q2. read edge (character interaction) files for all eight seasons and merge them to the same database. However, character interactions in different seasons should have different types.

```
In [ ]: # clear data
        # DETACH will remove add edges as well
        query string = '''
        MATCH (c:Character)
        DETACH DELETE c;
        conn.query(query string)
        # # read characters from season1 to season 8 of GOT and add as Character nodes
        # # MERGE: This operation creates a node with label "Character" and a property "id
        # # If a node with the same "id" already exists in the graph database, it updates \mathfrak{t}
        # # SET: This operation sets the value of the "name" property of the created/update
        # Q 1 & Q 2
        seasons = ['s1', 's2', 's3', 's4', 's5', 's6', 's7', 's8']
        # loop through the seasons and add nodes and edges
        for season in seasons:
            # Q1
            # read characters from the season and add as Character nodes
            query string = f'''
            WITH "https://raw.githubusercontent.com/mathbeveridge/gameofthrones/master/data
            LOAD CSV WITH HEADERS FROM uri AS row
            MERGE (c:Character {{id:row.Id}})
            SET c.name = row.Label
```

```
# Q2

# read character interactions from the season and add as edges
query_string = f'''
WITH "https://raw.githubusercontent.com/mathbeveridge/gameofthrones/master/data
LOAD CSV WITH HEADERS FROM uri AS row
MATCH (source:Character {{id: row.Source}})
MATCH (target:Character {{id: row.Target}})
MERGE (source)-[:SEASON{season[1].upper()} {{weight: toInteger(row.Weight)}}]-(
'''
conn.query(query_string)
```

Q3. Compute degrees of nodes separately for each season. That means, compute degrees separately for each of eight subgraphs (each subgraph denotes character interactions in each season). The notebook already has an example for season 1. Also compute the degrees of nodes for the entire graph created from all seasons.

```
In [ ]: # Degrees of nodes - separately for each of the 8 seasons
        query_string = '''
        MATCH (c:Character)
        RETURN c.name,
               SIZE([(c)-[:SEASON1]->() | 1]) AS season1InDegree,
               SIZE([(c)-[:SEASON2]->() | 1]) AS season2InDegree,
               SIZE([(c)-[:SEASON3]->() | 1]) AS season3InDegree,
               SIZE([(c)-[:SEASON4]->() | 1]) AS season4InDegree,
               SIZE([(c)-[:SEASON5]->() | 1]) AS season5InDegree,
               SIZE([(c)-[:SEASON6]->() | 1]) AS season6InDegree,
               SIZE([(c)-[:SEASON7]->() | 1]) AS season7InDegree,
               SIZE([(c)-[:SEASON8]->() | 1]) AS season8InDegree
        ORDER BY season1InDegree DESC,
                 season2InDegree DESC,
                 season3InDegree DESC,
                 season4InDegree DESC,
                 season5InDegree DESC,
                 season6InDegree DESC,
                 season7InDegree DESC,
                 season8InDegree DESC
        top_cat_df = pd.DataFrame([dict(_) for _ in conn.query(query_string)])
        print(top cat df.head(30))
```

				W4		
	c.name	season1InDegree	season2InDe		on3InDegree \	
0	Catelyn	29		19	16	
1	Arya	28		25	15	
2	Cersei	23		28	18	
3	Ned	23		10	7	
4	Joffrey	18		23	14	
5	Bran	18		13	17	
6	Daenerys	17		13	10	
7	Jaime	16		12	13	
8	Petyr	15		13	9	
9	Jon	14		6	12	
10	Sandor	13		10	7	
11	Bronn	12		15	14	
12	Aerys	12		4	5	
13	Jeor	10		4	4	
14	Drogo	10		1	0	
15	Barristan	10		0	8	
16	Greatjon	10		0	0	
17	Robb	9		9	10	
18	Jorah	9		6	5	
19	Benjen	9		1	0	
20	Luwin	8		6	0	
21	Robert	8		4	3	
22	Dareon	8		0	2	
23	Alliser	8		0	0	
24	Baelor	8		0	0	
25	Pycelle	7		5	4	
26	Doreah	7		4	0	
27	Jon Arryn	7		3	0	
28	Loras	6		9	10	
		6		4	3	
29	Ilyn	Ö		4	.5	
	-	-		-	•	
0	season4InD	egree season5InD	•	n6InDegree	season7InDegree	
0		egree season5InI 11	4	n6InDegree 5	season7InDegree	
1		egree season5InD 11 24	4 11	n6InDegree 5 17	season7InDegree 3 18	
1 2		egree season5InD 11 24 28	4 11 29	n6InDegree 5 17 24	season7InDegree 3 18 26	
1 2 3		egree season5InD 11 24 28 5	4 11 29 4	n6InDegree 5 17 24 10	season7InDegree 3 18 26 4	
1 2 3 4		egree season5InI 11 24 28 5 26	4 11 29 4 7	n6InDegree 5 17 24 10 4	season7InDegree 3 18 26 4 8	
1 2 3 4 5		egree season5InE 11 24 28 5 26 15	4 11 29 4 7 3	n6InDegree 5 17 24 10 4 14	season7InDegree 3 18 26 4 8 16	
1 2 3 4 5 6		egree season5InE 11 24 28 5 26 15	4 11 29 4 7 3	n6InDegree 5 17 24 10 4 14 16	season7InDegree 3 18 26 4 8 16 28	
1 2 3 4 5 6 7		egree season5InI 11 24 28 5 26 15 14	4 11 29 4 7 3	n6InDegree 5 17 24 10 4 14 16 16	season7InDegree 3 18 26 4 8 16 28	
1 2 3 4 5 6		egree season5InE 11 24 28 5 26 15	4 11 29 4 7 3	n6InDegree 5 17 24 10 4 14 16	season7InDegree 3 18 26 4 8 16 28	
1 2 3 4 5 6 7		egree season5InI 11 24 28 5 26 15 14	4 11 29 4 7 3 11	n6InDegree 5 17 24 10 4 14 16 16	season7InDegree 3 18 26 4 8 16 28	
1 2 3 4 5 6 7 8		egree season5InI 11 24 28 5 26 15 14 19 8	4 11 29 4 7 3 11 11	n6InDegree 5 17 24 10 4 14 16 16 7	season7InDegree 3 18 26 4 8 16 28 12	
1 2 3 4 5 6 7 8		egree season5InI 11 24 28 5 26 15 14 19 8 20	4 11 29 4 7 3 11 11 20 18	n6InDegree 5 17 24 10 4 14 16 16	season7InDegree 3 18 26 4 8 16 28 12 9 24	
1 2 3 4 5 6 7 8 9		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14	4 11 29 4 7 3 11 11 20 18 0	n6InDegree 5 17 24 10 4 14 16 16 7 21	season7InDegree 3 18 26 4 8 16 28 12 9 24 15	
1 2 3 4 5 6 7 8 9 10 11		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4	4 11 29 4 7 3 11 11 20 18 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6	season7InDegree 3 18 26 4 8 16 28 12 9 24 15 15	
1 2 3 4 5 6 7 8 9 10 11 12		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4	4 11 29 4 7 3 11 11 20 18 0 11 2	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1	season7InDegree 3 18 26 4 8 16 28 12 9 24 15 15 8 4	
1 2 3 4 5 6 7 8 9 10 11 12 13		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1	4 11 29 4 7 3 11 11 20 18 0 11 2	n6InDegree 5 17 24 10 4 14 16 16 7 21 3 6	season7InDegree 3 18 26 4 8 16 28 12 9 24 15 15 8 4	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6	n6InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 8 4 2 0	
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1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2	
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1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 8 4 2 0 0 0 2 11	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0	4 11 29 4 7 3 11 11 20 18 0 11 2 0 6 0 0 2 6 0 0 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 8 4 2 0 0 2 11 1	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0	n6InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0 0 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11	n6InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0 0 0 9	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24		egree season5Inf  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14  4	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 2 7 0 0 0 9 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23		egree season5InI  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11	n6InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0 0 0 9	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24		egree season5Inf  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14  4	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 2 7 0 0 0 9 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0 0 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25		egree season5Inf  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14  4  4  6	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11 0 2	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 2 7 0 0 0 9 0 2	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 2 11 1 0 3 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26		egree season5Inf  11  24  28  5  26  15  14  19  8  20  14  14  4  2  1  9  0  2  3  0  0  1  0  14  4  6  0	4 11 29 4 7 3 11 11 20 18 0 11 2 2 0 6 0 2 6 0 0 3 0 11 0 2 0	16InDegree 5 17 24 10 4 14 16 16 7 21 3 6 1 1 1 0 0 2 2 7 0 0 0 9 0 2 0 0	season7InDegree  3 18 26 4 8 16 28 12 9 24 15 15 15 8 4 2 0 0 0 2 11 1 0 3 0 0 0 0 0	

season8InDegree 

 27
 0

 28
 0

 29
 0

```
In []: # See the degrees of nodes for all seasons together
    query_string = '''
MATCH (c:Character)
RETURN c.name, SIZE([(c)-[]->(s) | s]) AS inDegree
ORDER BY inDegree DESC
'''
top_cat_df = pd.DataFrame([dict(_) for _ in conn.query(query_string)])
print(top_cat_df.head(30))
```

	c.name	inDegree
0	Cersei	192
1	Arya	176
2	Daenerys	141
3	Jon	135
4	Bran	125
5	Jaime	115
6	Davos	111
7	Brienne	105
8	Joffrey	102
9	Bronn	98
10	Catelyn	87
11	Petyr	82
12	Sandor	77
13	Ned	67
14	Eddison	55
15	Jorah	54
16	Grey Worm	52
17	Gendry	49
18	Beric	47
19	Gilly	47
20	Sansa	46
21	Ellaria	44
22	Loras	43
23	Alliser	42
24	Aerys	38
25	Robb	36
26	Barristan	33
27	Podrick	33
28	Hodor	31
29	Edmure	30

Q4. From season 1 find 10 nodes with the highest degrees. Then, for these 10 nodes, plot their degrees found in every season subgraph as well as for on the overall graph. That means, for each of the selected 10 nodes, you will have 9 numbers (one for each season and the last one for the overall graph). Please line plots where one line denotes one of the ten characters. x axis of the plot denotes seasons, and y axis denotes degree.

```
In []: import matplotlib.pyplot as plt

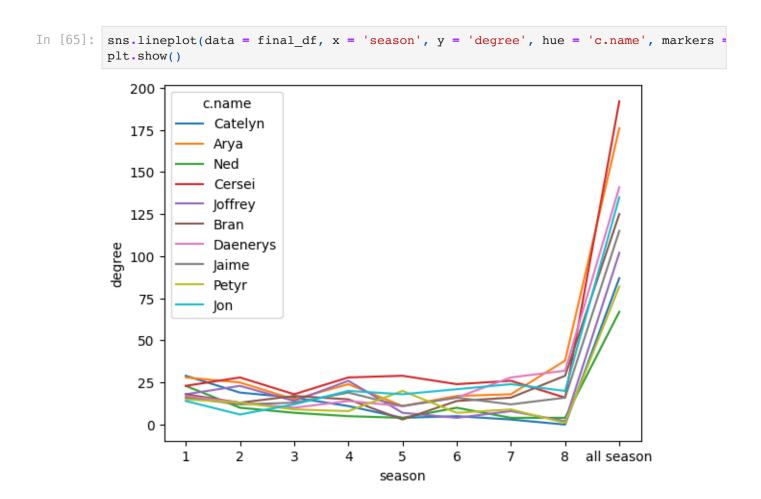
# Query to get the top 10 nodes with highest degrees in Season 1
query_string = '''
MATCH (c:Character)
RETURN c.name, SIZE([(c)-[:SEASON1]->(s) | s]) AS inDegree
ORDER BY inDegree DESC LIMIT 10

# Get the top 10 nodes with highest degrees in Season 1
```

```
top 10 df = pd.DataFrame([dict() for in conn.query(query string)])
          # List of the top 10 nodes
         top_10_nodes = top_10_df['c.name'].tolist()
         print(top_10_nodes)
         ['Catelyn', 'Arya', 'Ned', 'Cersei', 'Joffrey', 'Bran', 'Daenerys', 'Jaime', 'Pety
         r', 'Jon']
In [66]: final df = pd.DataFrame()
         for character_name in top_10_nodes:
              # Iterate over all seasons (1 to 8)
             for season in range(1, 9):
               query_string = f'''
               MATCH (c:Character)
               WHERE c.name = '{character_name}'
               RETURN c.name,
               SIZE([(c)-[:SEASON{season}]->() | 1]) AS degree
               1.1.1
               query_params = {'character_name': character_name, 'season': season}
               season_df = pd.DataFrame([dict(_) for _ in conn.query(query_string, query_par
               season df['season'] = str(season)
               final df = pd.concat([final df, season df])
             query_string = f'''
               MATCH (c:Character)
               WHERE c.name = '{character name}'
               RETURN c.name,
               SIZE([(c)-[]->() | 1]) AS degree
               1.1.1
             query_params = {'character_name': character_name}
              season_df = pd.DataFrame([dict(_) for _ in conn.query(query_string, query_param
             season df['season'] = 'all season'
              final df = pd.concat([final df, season df])
          final df = final df.reset index(drop=True)
         final df
```

Out[66]: c.name degree season Catelyn 29 2 Catelyn 19 Catelyn 16 3 Catelyn 11 4 Catelyn 4 5 85 Jon 18 5 86 Jon 21 6 87 Jon 24 7 8 88 Jon 20 89 Jon 135 all season

90 rows × 3 columns



Q5. Write a Cypher query to find the pair of characters that had the most interactions in each season and their corresponding weights.

Interactions are represented as edges in the graph. The greater the weight of the edge, the greater the interaction between the nodes(characters) joining that edge.

```
In [45]: final_df_for_interactions = pd.DataFrame()

for season in range(1,9):
    query_string = f'''
    MATCH (cl:Character)-[r:SEASON{season}]-(c2:Character)
    RETURN cl.name, c2.name, r.weight
    ORDER BY r.weight DESC
    LIMIT 1
    '''
    query_params = {'season': season}
    season_df = pd.DataFrame([dict(_) for _ in conn.query(query_string, query_params)
    season_df['season'] = str(season)
    final_df_for_interactions = pd.concat([final_df_for_interactions, season_df])

# The following dataframe outputs the pairs of character that. had the most interactional_df_for_interactions
```

$\cap$	1+	[/5]	
Uι	1 L	[47]	1

	c1.name	c2.name	r.weight	season
0	Ned	Robert	192	1
0	Cersei	Tyrion	177	2
0	Jon	Ygritte	159	3
0	Arya	Sandor	162	4
0	Arya	Jaqen	148	5
0	Jon	Sansa	151	6
0	Daenerys	Tyrion	179	7
0	Daenerys	Jon	194	8

Q6. For the combined graph of all seasons, write a Cypher query to find the top 10 characters (nodes) that have interactions with the most number of distinct characters.

c1.name num\_distinct\_characters Out[56]: 0 Tyrion 128 1 105 Jon 2 Sansa 101 3 Arya 97 93 4 Daenerys 5 Jaime 90 6 Cersei 86 7 Sam 78 8 Ned 78 9 Davos 78

```
In [ ]: # #Create a projecttion of the season 1 subgraph
        # query_string = '''
        # CALL gds.graph.project.cypher(
              'season1_graph',
        #
              'MATCH (n) RETURN id(n) AS id',
        #
              'MATCH (n)-[e:SEASON1]-(m) RETURN id(n) AS source, e.weight AS weight, id(m)
        # )
        # conn.query(query_string)
In [ ]: # # Compute pagerank of the season 1 subgraph
        # query_string = '''
        # CALL gds.pageRank.stream('season1_graph', {
        # maxIterations: 100,
          dampingFactor: 0.85,
           relationshipWeightProperty: 'weight'
        # })
        # YIELD nodeId, score
        # RETURN gds.util.asNode(nodeId).id AS id, gds.util.asNode(nodeId).name as nam
        # ORDER BY full pagerank DESC
        # '''
        # conn.query(query string)
In []: # # Detect communities from the season 1 subgraph
        # query_string = '''
        # CALL gds.louvain.write('season1 graph',
        #
             {relationshipWeightProperty: 'weight',
        #
              writeProperty: 'full community id'
        # })
        # '''
        # conn.query(query string)
In []:
```