

Knowledge_Graph

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1 Assignment-2

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2 Question

Download the dataset provided in this [webpage](#). Follow the instructions in the page and create a knowledge graph. Identify the important personalities and locations based on PageRank and Betweenness centrality.

3 Setup

```
[1]: import re
import pandas as pd
import bs4
import requests
import spacy
from spacy import displacy
nlp = spacy.load('en_core_web_sm')

from spacy.matcher import Matcher
from spacy.tokens import Span

import networkx as nx
from networkx.algorithms.link_analysis.pagerank_alg import pagerank
from networkx.algorithms centrality import betweenness centrality

import matplotlib.pyplot as plt
from tqdm import tqdm

pd.set_option('display.max_colwidth', 200)
%matplotlib inline
```

```
[2]: # import wikipedia sentences
candidate_sentences = pd.read_csv("wiki_sentences_v2.csv")
candidate_sentences.shape
```

[2]: (4318, 1)

4 Extract Entities (Subject & Object)

```
[3]: def get_entities(sent):

    # Finding named entities in sent
    doc = nlp(sent)
    entity_names = [ent.text for ent in doc.ents]
    entity_labels = [ent.label_ for ent in doc.ents]
    ## chunk 1
    ent1 = ""
    ent2 = ""

    prv_tok_dep = ""    # dependency tag of previous token in the sentence
    prv_tok_text = ""    # previous token in the sentence

    prefix = ""
    modifier = ""

    #####

    for tok in nlp(sent):
        ## chunk 2
        # if token is a punctuation mark then move on to the next token
        if tok.dep_ != "punct":
            # check: token is a compound word or not
            if tok.dep_ == "compound":
                prefix = tok.text
                # if the previous word was also a 'compound' then add the
                → current word to it
                if prv_tok_dep == "compound":
                    prefix = prv_tok_text + " " + tok.text

            # check: token is a modifier or not
            if tok.dep_.endswith("mod") == True:
                modifier = tok.text
                # if the previous word was also a 'compound' then add the current
                → word to it
                if prv_tok_dep == "compound":
                    modifier = prv_tok_text + " " + tok.text

        ## chunk 3
        if tok.dep_.find("subj") == True:
            ent1 = modifier + " " + prefix + " " + tok.text
            prefix = ""
```

```

        modifier = ""
        prv_tok_dep = ""
        prv_tok_text = ""

    ## chunk 4
    if tok.dep_.find("obj") == True:
        ent2 = modifier + " " + prefix + " " + tok.text

    ## chunk 5
    # update variables
    prv_tok_dep = tok.dep_
    prv_tok_text = tok.text

#####
for element in ent1.split(" "):
    entity_type_1 = 'None'
    if element in entity_names:
        index = entity_names.index(element)
        ent_label = entity_labels[index]
        if ent_label == 'NORP':
            entity_type_1 = 'Personality'
        elif ent_label == 'GPE':
            entity_type_1 = 'Location'

for element in ent2.split(" "):
    entity_type_2 = 'None'
    if element in entity_names:
        index = entity_names.index(element)
        ent_label = entity_labels[index]
        if ent_label == 'NORP' or ent_label == 'ORG':
            entity_type_2 = 'Personality'
        elif ent_label == 'GPE':
            entity_type_2 = 'Location'

return ((ent1.strip(), entity_type_1), (ent2.strip(), entity_type_2))

```

```

[4]: # Extract entity pairs from each sentence
entity_pairs = []
entity_types = {}

for i in tqdm(candidate_sentences["sentence"]):
    entity_1, entity_2 = get_entities(i)
    entity_pairs.append([entity_1[0], entity_2[0]])
    entity_types[entity_1[0]] = entity_1[1]
    entity_types[entity_2[0]] = entity_2[1]

```

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5 Extract Relations (Verbs)

```
[5]: def get_relation(sent):

    doc = nlp(sent)

    # Matcher class object
    matcher = Matcher(nlp.vocab)

    #define the pattern
    pattern = [{'DEP': 'ROOT'},
               {'DEP': 'prep', 'OP': "?"},
               {'DEP': 'agent', 'OP': "?"},
               {'POS': 'ADJ', 'OP': "?"}]

    matcher.add("matching_1", [pattern], on_match=None)

    matches = matcher(doc)
    k = len(matches) - 1

    span = doc[matches[k][1]:matches[k][2]]

    return(span.text)

[6]: # Extract relations from each sentence
relations = [get_relation(i) for i in tqdm(candidate_sentences['sentence'])]
```

100%| | 4318/4318 [00:20<00:00, 209.31it/s]

```
[7]: # extract subject
source = [i[0] for i in entity_pairs]

# extract object
target = [i[1] for i in entity_pairs]

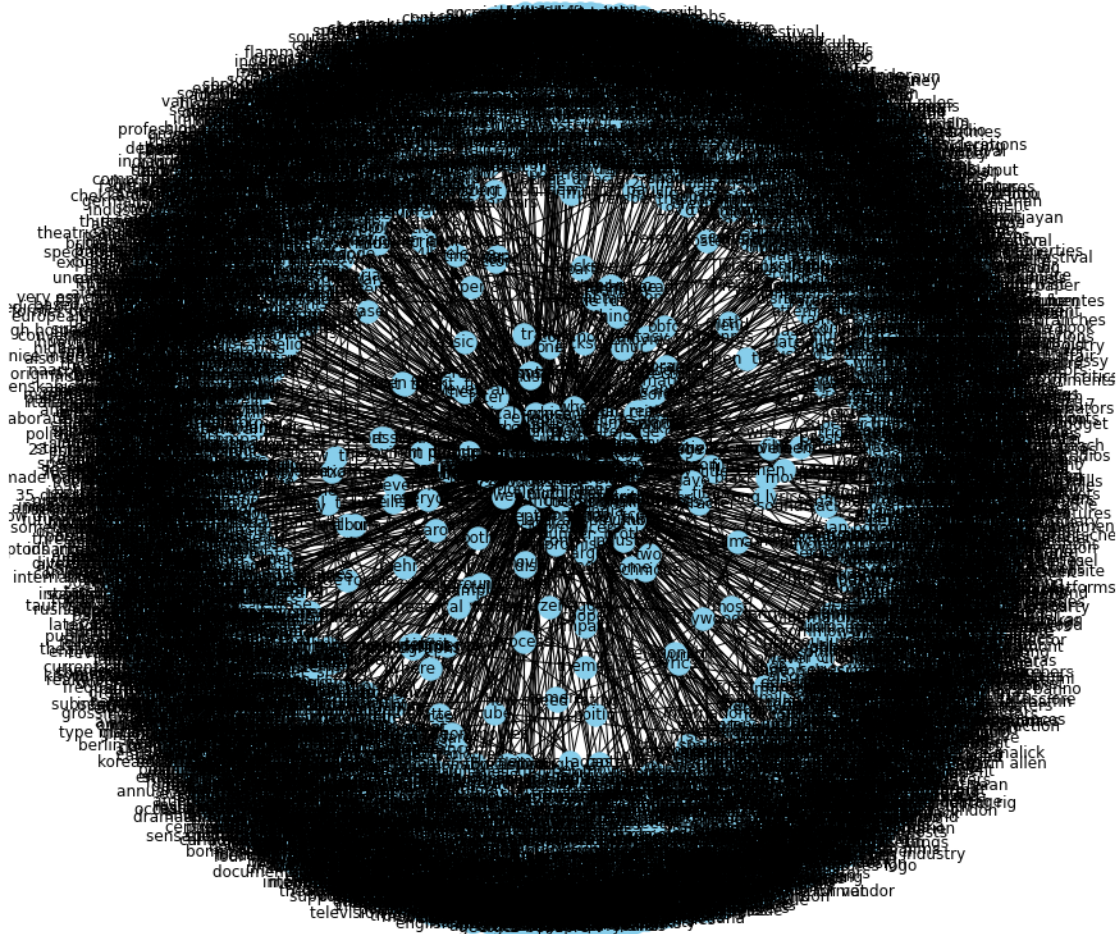
kg_df = pd.DataFrame({'source':source, 'target':target, 'edge':relations,
                      'source_type':[entity_types[x] for x in source],
                      'target_type':[entity_types[x] for x in target]})
```

6 Plot Knowledge Graph

```
[8]: # create a directed-graph from a dataframe
G=nx.from_pandas_edgelist(kg_df, "source", "target",
                          edge_attr=True, create_using=nx.DiGraph())
nx.set_node_attributes(G, values = entity_types, name='type')
```

```
plt.figure(figsize=(12,12))

pos = nx.spring_layout(G)
nx.draw(G, with_labels=True, node_color='skyblue', edge_cmap=plt.cm.Blues, pos_
    => pos)
plt.show()
```



```
[9]: # Betweenness Centrality
bn = betweenness_centrality(G)
bn = sorted(bn.items(), key=lambda item: item[1], reverse = True)
```

```
[10]: # Top entities considering all kinds of nodes
print("Top 10 entites based on Betweenness centrality:")
bn[:10]
```

Top 10 entites based on Betweenness centrality:

```
[10]: [('film', 0.001440642260991783),
      ('', 0.0013697462228097347),
      ('it', 0.0005731280424664372),
      ('this', 0.0003816507974873087),
      ('that', 0.00028048012303545964),
      ('music', 0.0001271824432769614),
      ('eilis', 0.00010396430117530981),
      ('films', 0.00010019807848145902),
      ('sequel', 7.718687169265637e-05),
      ('soundtrack', 7.577971156528355e-05)]
```

```
[11]: # Pagerank Centrality
pr = pagerank(G)
pr = sorted(pr.items(), key=lambda item: item[1], reverse = True)
```

```
[12]: # Top entities considering all kinds of nodes
print("Top 10 entites based on Page Rank centrality:")
pr[:10]
```

Top 10 entites based on Page Rank centrality:

```
[12]: [('', 0.006427768843810376),
      ('film', 0.0036383935101247703),
      ('him', 0.0017784284109153136),
      ('it', 0.001740290585411589),
      ('that', 0.0012468549227105698),
      ('december', 0.0011080809791448347),
      ('time', 0.0010417619936354497),
      ('april', 0.0010308549358775672),
      ('them', 0.0010217986208158698),
      ('november', 0.0009868703651856783)]
```

7 Plotting Sub Graphs

7.1 Personality Entites

```
[13]: # Plotting relations with target "Personality" Entites only
G=nx.from_pandas_edgelist(kg_df[kg_df['target_type']=="Personality"], "source",
    ↪ "target",
                        edge_attr=True, create_using=nx.DiGraph())
```


Top 10 entites based on Betweenness centrality:

```
[15]: [('examples', 0.0),
      ('all april house', 0.0),
      ('first alfred thriller', 0.0),
      ('silent london fog', 0.0),
      ('same method', 0.0),
      ('mclaren', 0.0),
      ('story', 0.0),
      ('miller', 0.0),
      ('western theatre', 0.0),
      ('considerably romans', 0.0)]
```

```
[16]: # Pagerank Centrality
pr = pagerank(G)
pr = sorted(pr.items(), key=lambda item: item[1], reverse = True)
```

```
[17]: # Top entities
print("Top 10 entites based on Page Rank centrality:")
pr[:10]
```

Top 10 entites based on Page Rank centrality:

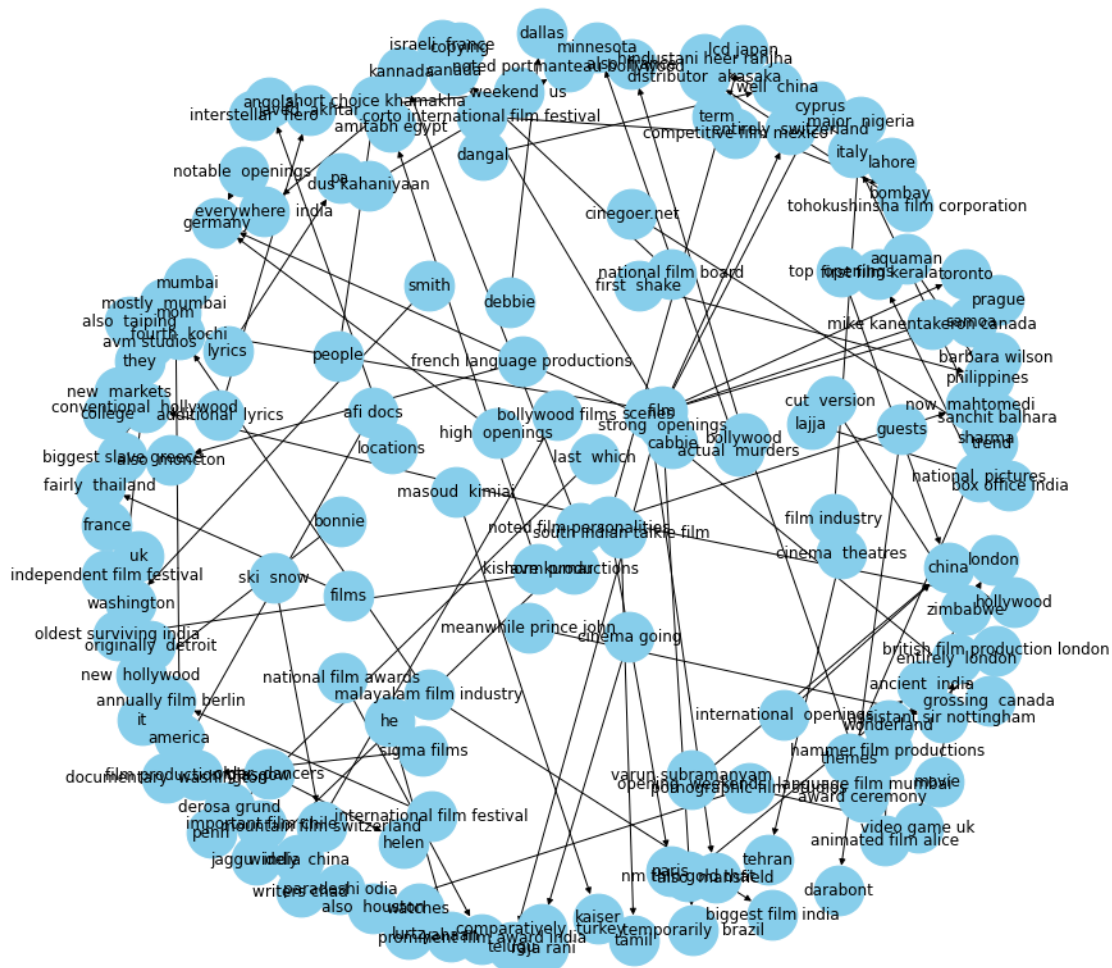
```
[17]: [('all april house', 0.020252015364126713),
      ('silent london fog', 0.020252015364126713),
      ('mclaren', 0.020252015364126713),
      ('miller', 0.020252015364126713),
      ('considerably romans', 0.020252015364126713),
      ('film mahesh manjrekar', 0.020252015364126713),
      ('ilaiyaraaja', 0.020252015364126713),
      ('indian survival drama blessy', 0.020252015364126713),
      ('3-d rko', 0.020252015364126713),
      ('nvidia', 0.020252015364126713)]
```

7.2 Location Entities

```
[18]: # Plotting relations with target "Location" Entites only

G=nx.from_pandas_edgelist(kg_df[kg_df['target_type']=="Location"], "source",
    ↪ "target",
                        edge_attr=True, create_using=nx.DiGraph())

plt.figure(figsize=(12,12))
pos = nx.spring_layout(G, k = 0.5)
nx.draw(G, with_labels=True, node_color='skyblue', node_size=1500,
    ↪ edge_cmap=plt.cm.Blues, pos = pos)
plt.show()
```

```
[19]: # Betweenness Centrality
bn = betweenness_centrality(G)
bn = sorted(bn.items(), key=lambda item: item[1], reverse = True)
```

```
[20]: # Top entities
print("Top 10 entites based on Betweenness centrality:")
bn[:10]
```

Top 10 entites based on Betweenness centrality:

```
[20]: [('last which', 0.0),
      ('penn', 0.0),
      ('', 0.0),
```

```
('paris', 0.0),  
('national film awards', 0.0),  
('prominent film award india', 0.0),  
('new markets', 0.0),  
('france', 0.0),  
('first shake', 0.0),  
('philippines', 0.0)]
```

```
[21]: # Pagerank Centrality  
pr = pagerank(G)  
pr = sorted(pr.items(), key=lambda item: item[1], reverse = True)
```

```
[22]: # Top entities  
print("Top 10 entites based on Page Rank centrality:")  
pr[:10]
```

Top 10 entites based on Page Rank centrality:

```
[22]: [('china', 0.015596634939584221),  
('germany', 0.015596634939584221),  
('philippines', 0.011862369818150255),  
('oldest surviving india', 0.011862369818150255),  
('washington', 0.009372859737194277),  
('penn', 0.00812810469671629),  
('prominent film award india', 0.00812810469671629),  
('france', 0.00812810469671629),  
('video game uk', 0.00812810469671629),  
('uk', 0.00812810469671629)]
```