Notebook9-checkpoint

May 22, 2020

[]: # Text Analysis | SNA | Sentiment analysis

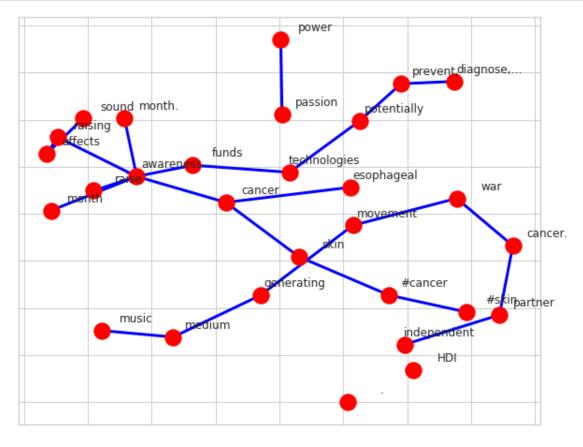
```
[89]: #modules import
      import os
      import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      import itertools
      import collections
      import tweepy as tw
      import nltk
      from nltk import bigrams
      from nltk.corpus import stopwords
      import re
      import networkx as nx
      import warnings
      warnings.filterwarnings("ignore")
      sns.set(font_scale=1.5)
      sns.set_style("whitegrid")
[90]: #defining my keys
      consumer_key= 'wDfiB8LIdPZkqngnZR2g9doVM'
      consumer_secret= '9vGVZwosoZuNiSHpP7wY6W0FgpdNH21wF0eDcICOZRc10IIjCp'
      access_token= '2735159228-UV7M2Wv9wZ5WEPx6lt9p4kdEnQ58euVXmK0hdZW'
      access_token_secret= 'GHLlOmT8eOIFtQ3u664zf5aFX1rFdgontow9I6WZCywZf'
[91]: auth = tw.OAuthHandler(consumer_key, consumer_secret)
      auth.set_access_token(access_token, access_token_secret)
      api = tw.API(auth, wait_on_rate_limit=True)
```

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[92]: # Create a custom search term cancer and qdp and hdi and define the number of
      \rightarrow tweets
      search_term = "#cancer+awareness -filter:retweets"
      tweets = tw.Cursor(api.search,
                         q=search_term,
                         lang="en",
                         since='2018-11-01').items(1000)
[93]: # grab and clean up 1000 recent tweets
[94]: def remove_url(txt):
          """Replace URLs found in a text string with nothing
          (i.e. it will remove the URL from the string).
          Parameters
          _____
          txt : string
              A text string that you want to parse and remove urls.
          Returns
          ____
          The same txt string with url's removed.
          url_pattern = re.compile(r'https?://\S+|www\.\S+')
          no_url = url_pattern.sub(r'', txt)
          return no_url
[95]: # Remove URLs
      tweets_no_urls = [remove_url(tweet.text) for tweet in tweets]
      # Create a sublist of lower case words for each tweet
      words_in_tweet = [tweet.lower().split() for tweet in tweets_no_urls]
      # Download stopwords
      nltk.download('stopwords')
      stop_words = set(stopwords.words('english'))
      # Remove stop words from each tweet list of words
      tweets_nsw = [[word for word in tweet_words if not word in stop_words]
                    for tweet_words in words_in_tweet]
      # Remove collection words
      collection_words = ['climatechange', 'climate', 'change']
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tweets_nsw_nc = [[w for w in word if not w in collection_words]
                        for word in tweets_nsw]
      [nltk data] Downloading package stopwords to /home/haziz/nltk data...
                    Package stopwords is already up-to-date!
      [nltk_data]
[96]: #Explore Co-occurring Words
[97]: # Create list of lists containing bigrams in tweets
       terms_bigram = [list(bigrams(tweet)) for tweet in tweets_nsw_nc]
[98]: # View bigrams for the first tweet
       terms_bigram[0]
[98]: [('support', 'emerging'),
        ('emerging', 'musicians,'),
        ('musicians,', 'bring'),
        ('bring', 'power'),
        ('power', 'passion'),
        ('passion', 'people'),
        ('people', '#music'),
        ('#music', 'together'),
        ('together', 'raise'),
        ('raise', 'funds...')]
[99]: # Original tweet without URLs
       tweets_no_urls[0]
[99]: 'With the support of emerging musicians, we bring the power and passion of
       people and #music together to raise funds... '
[100]: # Clean tweet
       tweets_nsw_nc[0]
[100]: ['support',
        'emerging',
        'musicians,',
        'bring',
        'power',
        'passion',
        'people',
        '#music',
        'together',
        'raise',
        'funds...']
```

```
[101]: # Flatten list of bigrams in clean tweets
       bigrams = list(itertools.chain(*terms_bigram))
       # Create counter of words in clean bigrams
       bigram_counts = collections.Counter(bigrams)
[102]: bigram_counts.most_common(25)
[102]: [(('raise', 'awareness'), 14),
        (('cancer', 'awareness'), 11),
        (('awareness', 'funds'), 10),
        (('sound', 'affects'), 7),
        (('awareness', 'month.'), 6),
        (('raising', 'awareness'), 5),
        (('skin', 'cancer'), 5),
        (('awareness', 'month'), 5),
        (('power', 'passion'), 4),
        (('skin', '#cancer'), 4),
        (('esophageal', 'cancer'), 4),
        (('#skin', '#cancer'), 4),
        (('.', '.'), 4),
        (('affects', 'raising'), 4),
        (('funds', 'technologies'), 4),
        (('technologies', 'potentially'), 4),
        (('potentially', 'prevent,'), 4),
        (('prevent,', 'diagnose,...'), 4),
        (('music', 'medium'), 4),
        (('medium', 'generating'), 4),
        (('generating', 'movement'), 4),
        (('movement', 'war'), 4),
        (('war', 'cancer.'), 4),
        (('cancer.', 'partner'), 4),
        (('partner', 'independent'), 4)]
[103]: #data frames
       bigram_df = pd.DataFrame(bigram_counts.most_common(25),
                                     columns=['bigram', 'count'])
       bigram_df
[103]:
                                bigram count
       0
                    (raise, awareness)
                                            14
                   (cancer, awareness)
       1
                                            11
                    (awareness, funds)
                                            10
       3
                      (sound, affects)
                                             7
       4
                   (awareness, month.)
                                             6
```

```
5
                  (raising, awareness)
                                              5
       6
                         (skin, cancer)
                                              5
       7
                                              5
                     (awareness, month)
                       (power, passion)
                                              4
       8
       9
                        (skin, #cancer)
                  (esophageal, cancer)
       10
                                              4
                       (#skin, #cancer)
                                              4
       11
       12
                                 (., .)
                                              4
                     (affects, raising)
       13
                                              4
       14
                  (funds, technologies)
                                              4
           (technologies, potentially)
                                              4
       15
       16
               (potentially, prevent,)
       17
                 (prevent,, diagnose,...)
                                              4
                                              4
       18
                        (music, medium)
       19
                   (medium, generating)
                                              4
                 (generating, movement)
       20
                                              4
       21
                        (movement, war)
                                              4
       22
                         (war, cancer.)
                                              4
                                              4
       23
                     (cancer., partner)
       24
                 (partner, independent)
[104]: #Visualize Networks of Bigrams
       # Create dictionary of bigrams and their counts
       d = bigram_df.set_index('bigram').T.to_dict('records')
[105]: # Create network plot
       G = nx.Graph()
[106]: # Create connections between nodes
       for k, v in d[0].items():
           G.add_edge(k[0], k[1], weight=(v * 10))
       G.add_node("HDI", weight=100)
[107]: fig, ax = plt.subplots(figsize=(10, 8))
       pos = nx.spring_layout(G, k=2)
       # Plot networks
       nx.draw_networkx(G, pos,
                         font_size=16,
                         width=3,
                         edge_color='blue',
                         node_color='red',
                         with labels = False,
                          ax=ax)
       # Create offset labels
```



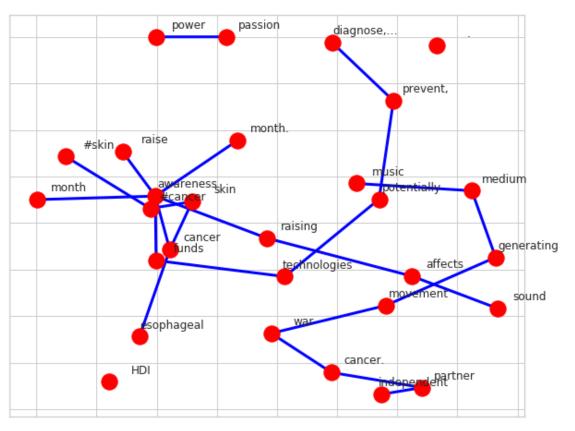
```
[108]: # Create network plot
H = nx.Graph()

[109]: # Create connections between nodes
for k, v in d[0].items():
        H.add_edge(k[0], k[1], weight=(v * 10))

H.add_node("GDP", weight=100)

[110]: fig, ax = plt.subplots(figsize=(10, 8))
        pos = nx.spring_layout(G, k=2)
```

```
# Plot networks
nx.draw_networkx(G, pos,
                 font_size=16,
                 width=3,
                 edge_color='blue',
                 node_color='red',
                 with_labels = False,
                  ax=ax)
# Create offset labels
for key, value in pos.items():
    x, y = value[0] + .135, value[1] + .045
    ax.text(x, y,
            s=key,
            #bbox=dict(facecolor='red', alpha=0.25),
            horizontalalignment='center', fontsize=12)
plt.show()
```

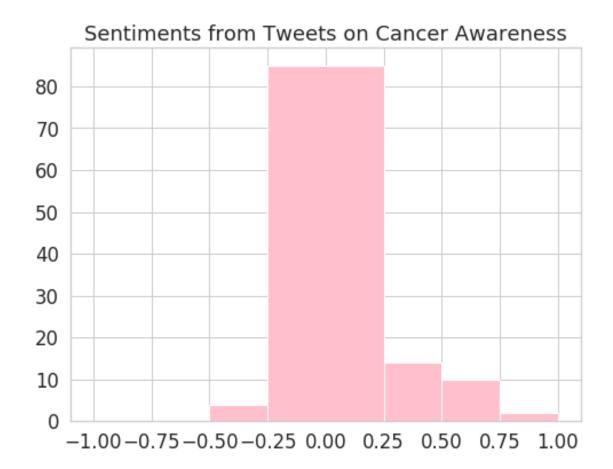


[]:

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[112]: def remove_url(txt):
           """Replace URLs found in a text string with nothing
           (i.e. it will remove the URL from the string).
           Parameters
           _____
           txt : string
              A text string that you want to parse and remove urls.
          Returns
           The same txt string with url's removed.
          return " ".join(re.sub("([^0-9A-Za-z \t])|(w+:\/\S+)", "", txt).split())
[113]: | # Create a custom search term and define the number of tweets
       search term = "#cancer+awareness -filter:retweets"
       tweets = tw.Cursor(api.search,
                          q=search_term,
                          lang="en",
                          since='2018-11-01').items(1000)
       # Remove URLs
       tweets_no_urls = [remove_url(tweet.text) for tweet in tweets]
[115]: from textblob import TextBlob
[116]: # Create textblob objects of the tweets
       sentiment_objects = [TextBlob(tweet) for tweet in tweets_no_urls]
       sentiment_objects[0].polarity, sentiment_objects[0]
[116]: (0.0,
       TextBlob("With the support of emerging musicians we bring the power and passion
       of people and music together to raise funds"))
[117]: # Create list of polarity valuesx and tweet text
       sentiment_values = [[tweet.sentiment.polarity, str(tweet)] for tweet in_
       →sentiment_objects]
```

[]: #Sentiment analysis

```
sentiment_values[0]
[117]: [0.0,
        'With the support of emerging musicians we bring the power and passion of
      people and music together to raise funds']
[118]: # Create dataframe containing the polarity value and tweet text
       sentiment_df = pd.DataFrame(sentiment_values, columns=["polarity", "tweet"])
       sentiment_df.head()
[118]:
         polarity
                                                                 tweet
               0.0 With the support of emerging musicians we brin...
       1
               0.0 Did you know You might wanna research a bit wa...
               0.1 Skin Cancer is Americas most common cancer wit...
       3
               0.2 What you may not know about breast cancer and ...
               0.0 QampA Determining foods to avoid when managing...
[119]: fig, ax = plt.subplots(figsize=(8, 6))
       # Plot histogram of the polarity values
       sentiment_df.hist(bins=[-1, -0.75, -0.5, -0.25, 0.25, 0.5, 0.75, 1],
                    ax=ax,
                    color="pink")
       plt.title("Sentiments from Tweets on Cancer Awareness")
       plt.show()
```



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