## topic clustering

March 23, 2021

#### 0.1 Topic Clustering

```
[1]: # import needed libraries
     from bson import json_util
     import requests, json, re
     import pandas as pd
     import matplotlib
     import nltk
     from nltk.corpus import stopwords
     from nltk.stem.wordnet import WordNetLemmatizer
     import string
     from gensim.models import ldamodel
     from gensim.corpora.dictionary import Dictionary
     import pyLDAvis.gensim
     import os
     import random
     import numpy as np
     import datetime
     from nltk import word_tokenize, sent_tokenize, ngrams, pos_tag
     from nltk.corpus import stopwords
     import gensim, operator
     import matplotlib.pyplot as plt
     import seaborn as sns
     from scipy import spatial
     from gensim.models import KeyedVectors
```

#### 0.1.1 Load data

```
[2]: data = []
with open('./tweets.json', 'r') as f:
    for line in f.readlines():
        data.append(json.loads(line))
print('There are', len(data),'tweets for topic clustering.')
print('Here is an example of the tweet:')
print(data[0])
```

There are 5071 tweets for topic clustering. Here is an example of the tweet:

{'url': 'https://twitter.com/AshleyDHeck1/status/1244776214222553088', 'text': 'Ya know how everyone kept saying "hold my beer", as a way to show things could most definitely get worse in situations? ... Well they named this stupid virus after a beer. #randomthoughts #coronavirus #wth #COVID19 #corona #lol #imlosingit https://t.co/gZJIKJc8md', 'date': 1585612750000, 'tweet\_id': 1244776214222553088, 'user\_name': 'AshleyDHeck1', 'cleaned\_text': 'Ya know how everyone kept saying hold my beer as a way to show things could most definitely get worse in situations Well they named this stupid virus after a beer randomthoughts coronavirus wth COVID corona lol imlosingit '}

#### 0.1.2 Save data to a data frame

```
[3]: df = pd.DataFrame(data[0:])
     df = df.iloc[: , 0:6]
     df.columns =["url","text","date","tweet_id","user_name","cleaned_text"]
     df['date']=df['date'].apply(lambda d: datetime.datetime.fromtimestamp(int(d)/
     \hookrightarrow1000).strftime('%Y-\%m-\%d'))
     df.head()
[3]:
                                                       url \
     0 https://twitter.com/AshleyDHeck1/status/124477...
     1 https://twitter.com/CffeeAndSarcasm/status/124...
     2 https://twitter.com/KarinBrauns/status/1244757...
     3 https://twitter.com/AspenBrewingCo/status/1244...
     4 https://twitter.com/Didanmeg/status/1244748406...
                                                                   date \
                                                       text
     O Ya know how everyone kept saying "hold my beer... 2020-03-30
     1 AA is going to need a bouncer when this is ove... 2020-03-30
     2 Don't drown in your own thoughts.. Stay close ...
                                                           2020-03-30
     3 Mix it up this week with a delicious pizza and...
                                                          2020-03-30
     4 The only beer we will be drinking until this w...
                                                           2020-03-30
                                    user_name
                   tweet_id
```

0 1244776214222553088 AshleyDHeck1 1 1244758836516397056 CffeeAndSarcasm

2 1244757544431751171 KarinBrauns

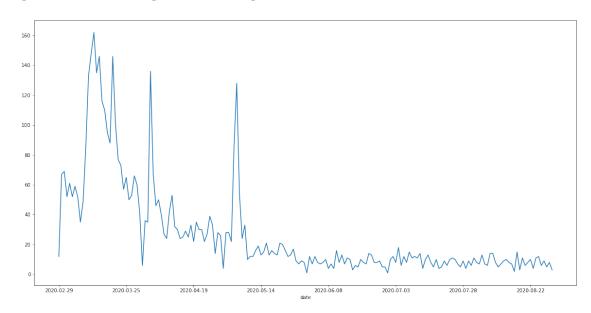
3 1244753120669958144 AspenBrewingCo

4 1244748406435852289 Didanmeg

cleaned\_text

- O Ya know how everyone kept saying hold my beer ...
- 1 AA is going to need a bouncer when this is ove...
- 2 Don't drown in your own thoughts Stay close at...
- 3 Mix it up this week with a delicious pizza and...
- 4 The only beer we will be drinking until this w...
  - The trend of tweets

[4]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7fdf505ce220>



#### 0.1.3 LDA model

```
[5]: # Convert tweet to lowercase
text = []
for item in df['cleaned_text']:
    text.append(item.lower())
```

```
[6]: # Remove stopwords
word=[]
for t in text:
    word.append(word_tokenize(t))
stop_words = set(stopwords.words('english'))
non_stop_word = []
for s in word:
    sent = []
    for w in s:
        if w not in stop_words:
            sent.append(w)
        non_stop_word.append(sent)
```

```
[7]: tweets = pd.Series(non_stop_word)
 [8]: dictionary = Dictionary(tweets)
      dictionary.filter_extremes(no_below=10, no_above=90)
      corpora = [dictionary.doc2bow(doc) for doc in tweets]
      lda_model = ldamodel.LdaModel(corpora, num_topics=6, id2word = dictionary,_
       →passes=40, random_state=4)
[21]: pyLDAvis.enable_notebook()
      vis = pyLDAvis.gensim.prepare(lda_model, corpora, dictionary)
      vis
[21]: PreparedData(topic_coordinates=
                                                             y topics cluster
                                                   х
      Freq
      topic
      3
           -0.047169 -0.063469
                                      1
                                               1 24.049944
      5
            -0.139312 -0.139897
                                      2
                                               1 19.156610
      2
           -0.163437 -0.037910
                                      3
                                                 19.048662
      0
                                      4
                                               1 16.160795
            0.254900 -0.059986
      1
            0.151720 0.017722
                                      5
                                               1 14.678576
      4
           -0.056702 0.283540
                                      6
                                                   6.905409, topic_info=
                                               1
                                                                                 Term
      Freq
                 Total Category
                                 logprob loglift
                    3211.000000
                                 3211.000000
                                                        30.0000
                                                                30.0000
      0
              beer
                                              Default
      21
            alcohol
                    1364.000000 1364.000000
                                              Default
                                                        29.0000
                                                                29.0000
      178
              hand
                      287.000000
                                   287.000000
                                              Default
                                                        28.0000 28.0000
      179
             hands
                                              Default
                                                        27.0000
                      283.000000
                                   283.000000
                                                                27.0000
      35
                      853.000000
                                   853.000000
                                              Default
                                                        26.0000
                                                                26.0000
      . .
      44
                      43.235523
                                   245.244720
                                               Topic6 -4.4701
                                                                 0.9373
           drinking
      145
             beers
                      38.207447
                                   210.015656
                                               Topic6 -4.5938
                                                                 0.9687
      209
                                                Topic6 -4.7911
              long
                      31.364672
                                    57.613319
                                                                 2.0648
      21
            alcohol
                       36.658817 1364.692017
                                                Topic6 -4.6352 -0.9442
      456
              sales
                      31.516981
                                   130.545853
                                                Topic6 -4.7863
                                                                 1.2517
      [352 rows x 6 columns], token_table=
                                                Topic
                                                           Freq
                                                                     Term
      term
      537
                1 0.962883
                            actually
      21
                1 0.266727
                             alcohol
                4 0.203709
      21
                              alcohol
      21
                5 0.501945
                              alcohol
      21
                6 0.027112
                              alcohol
                                    ,
      35
                2 0.485972
                3 0.154574
      35
      109
                1 0.996281
                1 0.990161
      251
```

```
869 1 0.987095
```

[451 rows x 3 columns], R=30, lambda\_step=0.01, plot\_opts={'xlab': 'PC1', 'ylab': 'PC2'}, topic\_order=[4, 6, 3, 1, 2, 5])

```
[9]: # LDA result
output = []
for topic_id in range(lda_model.num_topics):
    topk = lda_model.show_topic(topic_id, 15)
    topk_words = [ w for w, _ in topk ]
    output.append(topk_words)
    print('{}: {}'.format(topic_id, ' '.join(topk_words)))
```

- 0: corona covid coronavirus hand hands alcohol sanitizer virus wash use soap water alcoholbased based china
- 1: corona alcohol covid lockdown coronavirus liquor amp india pandemic people due virus government alcoholic socialdistancing
- 2: corona beer coronavirus virus ' mexico one beers summer bar would production get free amp
- 3: corona beer covid coronavirus alcohol 'people virus drinking " " know get one like
- 4: corona beer mexican covid virus hold modelo us de name amp pandemic named new mayo
- 5: corona beer ' day coronabeer coronavirus covid quarantine drink home happy love time cincodemayo cheers

```
[10]: # store result in a dataframe
      dictionary.filter_extremes(no_below=10, no_above=90)
      corpora = [dictionary.doc2bow(doc) for doc in tweets]
      # show which topic cluster each article belongs to
      topics = []
      for i in range(len(tweets)):
          topic = lda_model.get_document_topics(corpora[i])
          topics.append(topic)
      # transfer topics into a list of dictionaries
      topics_dic = []
      for topic in topics:
          dic = \{\}
          for item in topic:
              dic[item[0]] = item[1]
          topics_dic.append(dic)
      topic_cluster = []
      for i in range(len(tweets)):
          cluster = max(topics_dic[i], key=topics_dic[i].get)
          1 = [data[i]['text'], cluster]
```

```
topic_cluster.append(1)

topic_cluster = pd.DataFrame(topic_cluster)
topic_cluster.columns= ['tweet', 'cluster']
topic_cluster.head()
```

[10]: tweet cluster

O Ya know how everyone kept saying "hold my beer... 3

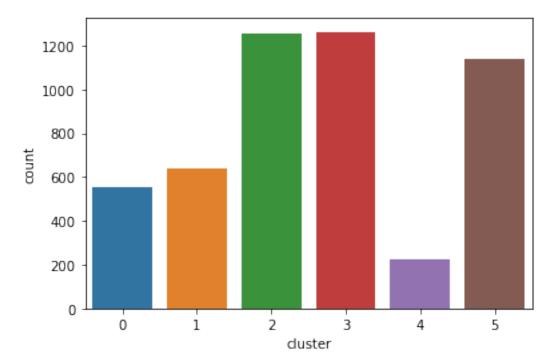
1 AA is going to need a bouncer when this is ove... 5

2 Don't drown in your own thoughts.. Stay close ... 3

3 Mix it up this week with a delicious pizza and... 5

4 The only beer we will be drinking until this w... 3

```
[11]: # result visualization
    cluster_count = topic_cluster['cluster'].value_counts()
    p = sns.countplot(data=topic_cluster, x = 'cluster')
    plt.show()
```



## 0.1.4 Implementation of topic classifier

```
[12]: # Load the pre-trained Word2Vec model
model_path = '/Users/feiyid/desktop/'
def load_wordvec_model(modelName, modelFile, flagBin):
    print('Loading ' + modelName + ' model...')
```

Loading Word2Vec model...
Finished loading Word2Vec model...

```
[13]: # Define taxonomy
      topic_taxonomy = {
          "Celebration Beer":
               "celebration day": "corona beer drink Cinco de Mayo"
          },
          "Corona Virus":
               "quarantine": "corona covid beer coronavirus quarantine virus sales_{\sqcup}
       →pandemic stayhome staysafe socialdistancing"
          },
          "Disinfect":
               "sanitize": "corona alcohol covid hand hands sanitizer coronavirus wash_{\sqcup}
       ⇒soap water use virus alcoholbased mask amp clean handsanitizer sanitizers⊔
       ⇔health"
          },
          "Corona drinking":
               "alcohol drinking": "alcohol corona lockdown covid liquor people ampu
       →drinking india us money government shops wine"
          },
          "Mexican":
              "Mexican beer": "corona beer covid coronavirus virus people amp,
       →mexican"
          },
          "Sentiment":
              "joy": "corona beer covid coronavirus love happy like china new memes⊔
       \hookrightarrowfollow wuhan viruscorona"
          },
      }
```

```
[14]: # Build functions
      # function checks whether the input words are present in the vocabulary for the
      \rightarrowmodel
      def vocab check(vectors, words):
          output = list()
          for word in words:
              if word in vectors.vocab:
                  output.append(word.strip())
          return output
      # Calculate semantic similarity of a pair of strings
      def calc_similarity(input1, input2, vectors):
          s1words = set(vocab_check(vectors, input1.split()))
          s2words = set(vocab_check(vectors, input2.split()))
          output = vectors.n_similarity(s1words, s2words)
          return output
      # Classify topics (run similarity of input string against taxonomy, sort and
      →return top 3 result)
      def classify_topics(input, vectors):
          feed_score = dict()
          for key, value in topic_taxonomy.items():
              max_value_score = dict()
              for label, keywords in value.items():
                  max value score[label] = 0
                  topic = (key + ' ' + keywords).strip()
                  max_value_score[label] += float(calc_similarity(input, topic,__
       →vectors))
              sorted_max_score = sorted(max_value_score.items(), key=operator.
       →itemgetter(1), reverse=True)[0]
              feed_score[sorted_max_score[0]] = sorted_max_score[1]
          return sorted(feed_score.items(), key=operator.itemgetter(1),__
       →reverse=True)[:3]
```

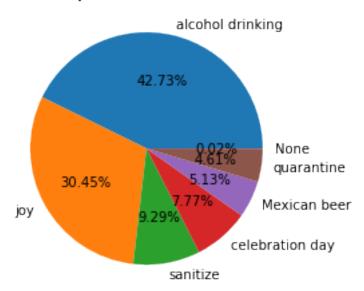
• Topic Similarity

```
[15]: # calculate the topic similarity for each article and choose the best one
topic_similarity = []

tweet = []
for item in df['cleaned_text']:
    text = item.split("#")
    tweet.append(item)
```

```
for i in range(len(tweet)):
         title = tweet[i]
         try:
             output = classify_topics(title, model_word2vec)
         except:
             output = [["None", 0]]
         topic = list(output[0])
         topic.append(title)
         topic_similarity.append(topic)
      # build the data frame
     Classifier = pd.DataFrame(topic_similarity, columns=['topic', 'similarity', __
      Classifier.head()
[15]:
                   topic similarity \
                     joy
                            0.663286
     1 alcohol drinking
                            0.625928
                     joy 0.696251
     2
                     joy 0.638301
     3
     4 alcohol drinking 0.564316
                                                    tweet
     O Ya know how everyone kept saying hold my beer ...
     1 AA is going to need a bouncer when this is ove...
     2 Don't drown in your own thoughts Stay close at...
     3 Mix it up this week with a delicious pizza and...
     4 The only beer we will be drinking until this w...
[16]: # Visualization
     topic_count = Classifier['topic'].value_counts()
     plt.pie(topic_count.values,
             labels=topic_count.index,
             autopct='%1.2f%%')
     plt.title('Topic Classifier', fontsize=20)
     plt.show()
```

# Topic Classifier



## • Topic Trend

```
[17]: Classifier['date'] = df['date']
  classifier_plot = Classifier[Classifier['topic'] != 'None']
  classifier_plot.head()
```

```
tweet date

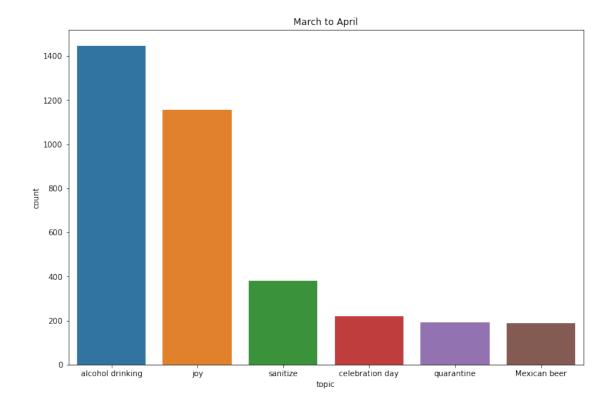
Va know how everyone kept saying hold my beer ... 2020-03-30

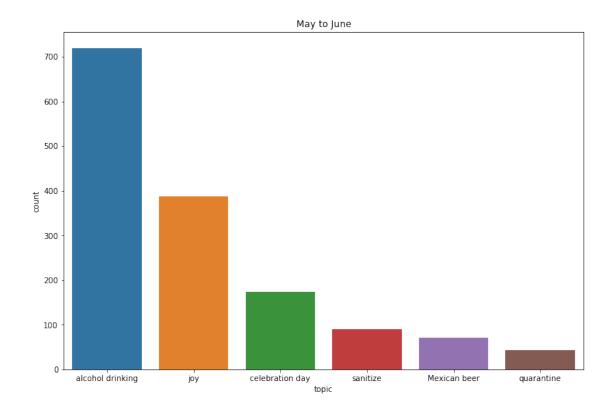
AA is going to need a bouncer when this is ove... 2020-03-30

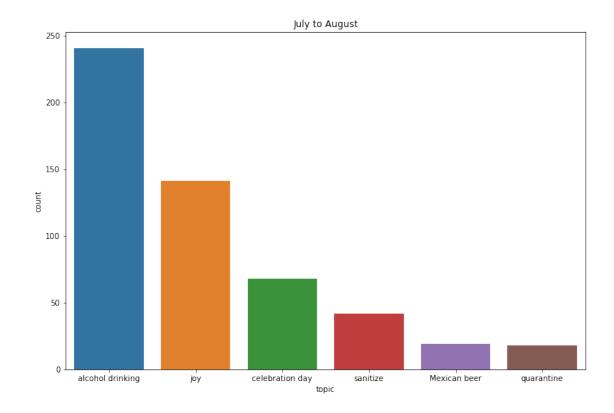
Don't drown in your own thoughts Stay close at... 2020-03-30

Mix it up this week with a delicious pizza and... 2020-03-30

The only beer we will be drinking until this w... 2020-03-30
```







[]: