## FIN3210 Week 5 Assignment

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October 20, 2023

```
[1]: import numpy as np
import pandas as pd
import nltk
import matplotlib.pyplot as plt
from nltk.tokenize import word_tokenize
from nltk.stem import WordNetLemmatizer
from wordcloud import WordCloud
from readability import Readability
```

### 0.0.1 Q1. Present word cloud

```
[2]: # Design a function in order to transfer the tag generated by nltk to
    # input in the WordNetLemmatizer

def transfer_to_pos(df):
    if df['Tag'].startswith('J'):
        df['Pos'] = 'a'
    elif df['Tag'].startswith('V'):
        df['Pos'] = 'v'
    elif df['Tag'].startswith('N'):
        df['Pos'] = 'n'
    elif df['Tag'].startswith('R'):
        df['Pos'] = 'r'
    else:
        df['Pos'] = 'x'
    return df
```

```
Tesla's third-quarter sales jumped 44%\ as global demand for its electric vehicles outpaced that of most other automakers. The company reported Friday that it had delivered 139,000 SUVs and sedans from July through September, compared with 97,000 deliveries during the same period a year ago. The sales topped even some of the most optimistic projections coming from Wall Street. Analysts polled by data provider FactSet expected the company to sell closer to 137,000. Tesla has been rewriting the script throughout the year amidst a pandemic that has closed factories and scrambled
```

```
supply lines. This puts Musk & Co. in prime position to
     hit the area code of 500k units for the year which six months
     ago was not even on the map for the bulls, Daniel Ives of
     Wedbush wrote Friday. China was likely a major source of
     strength in the quarter, Ives said. Tesla could post its fifth
     consecutive quarter of profits later this month.
[4]: words_list = word_tokenize(article_str)
     words_list[:10]
[4]: ['Tesla',
      "'s",
      'third-quarter',
      'sales',
      'jumped',
      '44',
      '%',
      '\\',
      'as',
      'global']
[5]: # Generate the postag of the words list
     words_tags = nltk.pos_tag(words_list)
     words tags[:10]
[5]: [('Tesla', 'NNP'),
      ("'s", 'POS'),
      ('third-quarter', 'JJ'),
      ('sales', 'NNS'),
      ('jumped', 'VBD'),
      ('44', 'CD'),
      ('%', 'NN'),
      ('\\', 'CC'),
      ('as', 'IN'),
      ('global', 'JJ')]
[6]: words_tags_df = pd.DataFrame(words_tags, columns=['Word_original', 'Tag'])
     words_tags_df = words_tags_df.apply(transfer_to_pos, axis=1)
     # Only preserve the relevant tags of words
     words_tags_df = words_tags_df[words_tags_df['Pos'] != 'x']
     words_tags_df = words_tags_df.reset_index(drop=True)
     words_tags_df
[6]:
         Word_original Tag Pos
                 Tesla NNP
                              n
     1
        third-quarter
                         JJ
```

```
2
              sales
                      NNS
3
                      VBD
             jumped
4
                  %
                       NN
                             n
. .
94
       consecutive
                       JJ
95
           quarter
                       NN
96
           profits
                      NNS
                             n
97
              later
                       RB
                             r
98
             month
                       NN
```

[99 rows x 3 columns]

```
[7]: # Lemmatize the words
words_tags_df['words_lemmatized'] = words_tags_df.apply(lambda x:

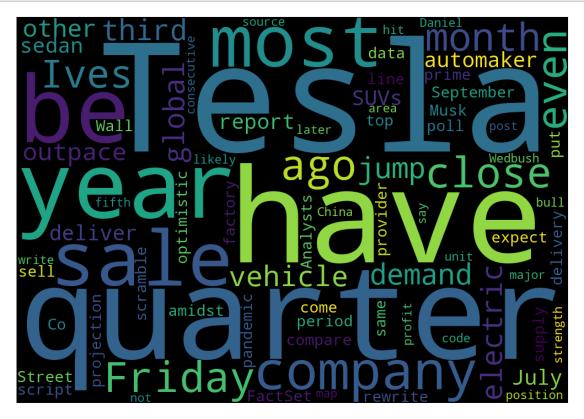
→WordNetLemmatizer().lemmatize(x['Word_original'], pos=x['Pos']), axis=1)
words_tags_df
```

```
[7]:
          Word_original
                          Tag Pos words_lemmatized
     0
                   Tesla
                          NNP
                                 n
                                                Tesla
     1
          third-quarter
                            JJ
                                       third-quarter
     2
                   sales
                          NNS
                                                 sale
     3
                  jumped
                          VBD
                                                 jump
                                 V
     4
                       %
                           NN
                                                    %
     . .
                           . .
     94
            consecutive
                            JJ
                                         consecutive
                                 а
     95
                quarter
                           NN
                                              quarter
                                 n
     96
                profits
                          NNS
                                              profit
                                 n
     97
                   later
                           RB
                                                later
     98
                  month
                                                month
                           NN
```

[99 rows x 4 columns]

```
[8]: # Convert the result dataframe of words into a list
lemmatized_list = words_tags_df['words_lemmatized'].tolist()
lemmatized_str = ' '.join(lemmatized_list)
lemmatized_str
```

[8]: 'Tesla third-quarter sale jump % global demand electric vehicle outpace most other automaker company report Friday have deliver SUVs sedan July September compare delivery same period year ago sale top even most optimistic projection come Wall Street Analysts poll data provider FactSet expect company sell close Tesla have be rewrite script year amidst pandemic have close factory scramble supply line put Musk Co. prime position hit area code unit year month ago be not even map bull Daniel Ives Wedbush write Friday China be likely major source strength quarter Ives say Tesla post fifth consecutive quarter profit later month'



# 0.0.2 Q2. Calculate the news sentiment variable using Loughran and McDonald Sentiment Word Lists

```
[10]: pos_words_list = pd.read_excel('LoughranMcDonald_SentimentWordLists_2018.xlsx', usheet_name='Positive', header=None)
pos_words_list
```

```
[10]:

0 ABLE
1 ABUNDANCE
2 ABUNDANT
3 ACCLAIMED
4 ACCOMPLISH
... ...
```

```
349
                  WIN
      350
               WINNER
      351
              WINNERS
      352
              WINNING
      353
               WORTHY
      [354 rows x 1 columns]
[11]: neg_words_list = pd.read_excel('LoughranMcDonald_SentimentWordLists_2018.xlsx',__
       ⇒sheet_name='Negative', header=None)
      neg words list
                       0
[11]:
                 ABANDON
      0
               ABANDONED
      1
      2
              ABANDONING
      3
             ABANDONMENT
            ABANDONMENTS
              WRONGDOING
      2350
      2351
             WRONGDOINGS
      2352
                WRONGFUL
      2353
              WRONGFULLY
      2354
                 WRONGLY
      [2355 rows x 1 columns]
[12]: # If positive, give a 1; If negative give -1; Else 0
      def judge_sentiment(df):
          if df['words_lemmatized'].upper() in pos_words_list[0].tolist():
              df['sentiment'] = 1
          elif df['words_lemmatized'].upper() in neg_words_list[0].tolist():
              df['sentiment'] = -1
              df['sentiment'] = 0
          return df
[13]: words_tags_df = words_tags_df.apply(judge_sentiment, axis=1)
      # Sum the positive and negative weights together
      num = words_tags_df['sentiment'].sum()
      denom = len(words_tags_df)
      new_sentiment = num / denom
      new_sentiment
```

[13]: 0.020202020202020204

### 0.0.3 Q3. Calculate the Fog index

```
[14]: # Calculate the Gunning Fog Index using Readability
r = Readability(article_str)
fog_index = r.gunning_fog().score
fog_index
```

#### [14]: 12.928205128205128

0.0.4 Q4. Using the data set of Tesla, a) report the summary statistics of sentiment, novelty, and impact; b) present the correlation coefficient among sentiment, novelty, and impact; and c) show the frequency and fraction of top 10 news categories.

```
[15]: # Provide the statistics required in the Homework requirement
  tesla_data = pd.read_excel('FIN3210 Week 5 Tesla.xlsx')
  summary_stat_data = tesla_data[['Sentiment','Novelty','Impact']]
  summary_stat = summary_stat_data.describe()
  summary_stat
```

```
[15]:
               Sentiment
                               Novelty
                                              Impact
                          1292.000000 1292.000000
      count 1292.000000
               53.845975
                             28.877709
                                          45.277864
      mean
      std
               14.824257
                             38.537946
                                           9.996097
                2,000000
                              0.000000
                                          13.000000
      min
      25%
               40.000000
                              0.000000
                                          39.000000
      50%
               50.000000
                              3.000000
                                          45.000000
      75%
               64.000000
                             56.000000
                                          52.000000
              100.000000
                                          77.000000
      max
                            100.000000
```

```
[16]: summary_stat_data.corr()
```

```
[16]: Sentiment Novelty Impact
Sentiment 1.000000 0.165557 -0.130972
Novelty 0.165557 1.000000 -0.064844
Impact -0.130972 -0.064844 1.000000
```

```
[17]:
                               Category Frequency Fractions
                             stock-loss
      0
                                               409
                                                     0.316563
      1
                             stock-gain
                                               232
                                                     0.179567
      2
                       product-release
                                               111
                                                     0.085913
      3
                     business-contract
                                                63
                                                     0.048762
```

4	capital-increase	55	0.042570
5	legal-verdict-favored	42	0.032508
6	price-target-upgrade	40	0.030960
7	fundraising	33	0.025542
8	acquisition-interest-acquirer	33	0.025542
9	product-price-cut	30	0.023220