## 3.Q\_Mean\_W2V

## February 27, 2019

3.6 Featurizing text data with tfidf weighted word-vectors

```
In [3]: df=pd.read_csv('final_features.csv')
In [4]: df.shape
Out[4]: (404290, 797)
In [2]: import pandas as pd
        import matplotlib.pyplot as plt
        import re
        import time
        import warnings
        import numpy as np
       from nltk.corpus import stopwords
       from sklearn.preprocessing import normalize
       from sklearn.feature_extraction.text import CountVectorizer
       from sklearn.feature_extraction.text import TfidfVectorizer
       warnings.filterwarnings("ignore")
        import sys
        import os
        import pandas as pd
        import numpy as np
       from tqdm import tqdm
        # exctract word2vec vectors
        # https://github.com/explosion/spaCy/issues/1721
        # http://landinghub.visualstudio.com/visual-cpp-build-tools
        import spacy
In [3]: # avoid decoding problems
       df = pd.read_csv("train.csv")
        # encode questions to unicode
        # https://stackoverflow.com/a/6812069
        # ----- python 2 -----
        \# df['question1'] = df['question1'].apply(lambda x: unicode(str(x), "utf-8"))
        \# df['question2'] = df['question2'].apply(lambda x: unicode(str(x), "utf-8"))
```

```
# ----- python 3 -----
       df['question1'] = df['question1'].apply(lambda x: str(x))
       df['question2'] = df['question2'].apply(lambda x: str(x))
In [4]: df.head()
Out [4]:
          id qid1 qid2
                                                                 question1 \
                       2 What is the step by step guide to invest in sh...
       1
                       4 What is the story of Kohinoor (Koh-i-Noor) Dia...
           1
                       6 How can I increase the speed of my internet co...
       3
          3
                 7
                      8 Why am I mentally very lonely? How can I solve...
                      10 Which one dissolve in water quikly sugar, salt...
                                                  question2 is_duplicate
       0 What is the step by step guide to invest in sh...
       1 What would happen if the Indian government sto...
                                                                       0
       2 How can Internet speed be increased by hacking...
                                                                       0
       3 Find the remainder when [math] 23^{24} [/math] i...
                                                                       0
                    Which fish would survive in salt water?
                                                                       0
In [5]: from sklearn.feature_extraction.text import TfidfVectorizer
       from sklearn.feature_extraction.text import CountVectorizer
        # merge texts
       questions = list(df['question1']) + list(df['question2'])
       tfidf = TfidfVectorizer(lowercase=False, )
       tfidf.fit_transform(questions)
        # dict key:word and value:tf-idf score
       word2tfidf = dict(zip(tfidf.get_feature_names(), tfidf.idf_))
```

- After we find TF-IDF scores, we convert each question to a weighted average of word2vec vectors by these scores.
- here we use a pre-trained GLOVE model which comes free with "Spacy". https://spacy.io/usage/vectors-similarity
- It is trained on Wikipedia and therefore, it is stronger in terms of word semantics.

```
vec1 = word1.vector
                                          # fetch df score
                                          try:
                                                     idf = word2tfidf[str(word1)]
                                          except:
                                                     idf = 0
                                          # compute final vec
                                          mean_vec1 += vec1 * idf
                               mean_vec1 = mean_vec1.mean(axis=0)
                                vecs1.append(mean_vec1)
                     df['q1_feats_m'] = list(vecs1)
100%|| 404290/404290 [1:24:04<00:00, 80.14it/s]
In [8]: vecs2 = []
                     for qu2 in tqdm(list(df['question2'])):
                                doc2 = nlp(qu2)
                               mean_vec2 = np.zeros([len(doc2), 384])
                                for word2 in doc2:
                                          # word2vec
                                          vec2 = word2.vector
                                          # fetch df score
                                          try:
                                                     idf = word2tfidf[str(word2)]
                                          except:
                                                     #print word
                                                     idf = 0
                                          # compute final vec
                                          mean_vec2 += vec2 * idf
                               mean_vec2 = mean_vec2.mean(axis=0)
                                vecs2.append(mean_vec2)
                     df['q2_feats_m'] = list(vecs2)
100%|| 404290/404290 [1:26:23<00:00, 77.99it/s]
In [9]: #prepro_features_train.csv (Simple Preprocessing Feartures)
                     #nlp_features_train.csv (NLP Features)
                     if os.path.isfile('nlp_features_train.csv'):
                                dfnlp = pd.read_csv("nlp_features_train.csv",encoding='latin-1')
                     else:
                               print("download nlp_features_train.csv from drive or run previous notebook")
                     if os.path.isfile('df_fe_without_preprocessing_train.csv'):
                                dfppro = pd.read_csv("df_fe_without_preprocessing_train.csv",encoding='latin-1')
                     else:
                               print("download df_fe_without_preprocessing_train.csv from drive or run previous named to be a superior of the contract of the
```

```
In [10]: df1 = dfnlp.drop(['qid1','qid2','question1','question2'],axis=1)
         df2 = dfppro.drop(['qid1','qid2','question1','question2','is_duplicate'],axis=1)
         df3 = df.drop(['qid1','qid2','question1','question2','is_duplicate'],axis=1)
         df3_q1 = pd.DataFrame(df3.q1_feats_m.values.tolist(), index= df3.index)
         df3_q2 = pd.DataFrame(df3.q2_feats_m.values.tolist(), index= df3.index)
In [11]: # dataframe of nlp features
         df1.head()
Out[11]:
                is_duplicate
                                                                           ctc_min
            id
                                 cwc_min
                                           cwc max
                                                      \mathtt{csc\_min}
                                                                 csc max
                               0.999980
             0
                                          0.833319
                                                     0.999983
                                                               0.999983
                                                                          0.916659
         1
             1
                               0.799984
                                          0.399996
                                                     0.749981
                                                               0.599988
                                                                          0.699993
                                                               0.249997
         2
             2
                            0 0.399992
                                          0.333328
                                                     0.399992
                                                                          0.399996
         3
             3
                               0.000000
                                          0.000000
                                                     0.000000
                                                               0.000000
                                                                          0.000000
             4
                            0
                               0.399992
                                          0.199998
                                                     0.999950
                                                               0.666644
                                                                          0.571420
             ctc max
                       last_word_eq first_word_eq
                                                     abs_len_diff
                                                                    mean len \
         0 0.785709
                                0.0
                                                 1.0
                                                                2.0
                                                                         13.0
                                0.0
                                                                         12.5
         1 0.466664
                                                 1.0
                                                               5.0
         2 0.285712
                                0.0
                                                 1.0
                                                               4.0
                                                                         12.0
         3 0.000000
                                                                         12.0
                                0.0
                                                 0.0
                                                               2.0
         4 0.307690
                                0.0
                                                 1.0
                                                               6.0
                                                                         10.0
                                                              fuzz_partial_ratio
            token_set_ratio
                             token_sort_ratio fuzz_ratio
         0
                         100
                                             93
                                                          93
                                                                               100
                          86
                                             63
                                                          66
                                                                                75
         1
         2
                          63
                                             63
                                                          43
                                                                                47
         3
                          28
                                             24
                                                           9
                                                                                14
         4
                          67
                                             47
                                                          35
                                                                                56
            longest_substr_ratio
                         0.982759
         0
                         0.596154
         1
         2
                         0.166667
         3
                         0.039216
                         0.175000
In [12]: # data before preprocessing
         df2.head()
Out[12]:
                                        q1len
                                               q21en
            id
                 freq_qid1
                            freq_qid2
                                                      q1_n_words
                                                                    q2_n_words
                                                   57
         0
             0
                         1
                                     1
                                           66
                                                                14
                                                                            12
                         4
         1
             1
                                     1
                                                   88
                                                                8
                                                                            13
                                           51
         2
             2
                         1
                                     1
                                           73
                                                   59
                                                                14
                                                                            10
         3
             3
                         1
                                     1
                                                                11
                                                                             9
                                           50
                                                   65
                                                                             7
                         3
                                           76
                                                   39
                                                                13
            word_Common word_Total word_share freq_q1+q2 freq_q1-q2
         0
                    10.0
                                 23.0
                                         0.434783
                                                             2
```

```
1
                   4.0
                               20.0
                                      0.200000
                                                          5
                                                                      3
         2
                   4.0
                              24.0
                                                          2
                                      0.166667
                                                                      0
                                                          2
         3
                   0.0
                              19.0
                                      0.000000
                                                                      0
                   2.0
                              20.0
                                      0.100000
                                                          4
                                                                      2
In [13]: # Questions 1 tfidf weighted word2vec
        df3_q1.head()
Out[13]:
                  0
                                          2
                              1
                                                       3
                                                                             5
          121.929927 100.083906
                                   72.497900 115.641795 -48.370865 34.619070
                                               98.191855 -51.234840
           -78.070935
                        54.843787
                                    82.738495
                                                                      55.013509
                                                                       35.229116
         2
            -5.355015
                        73.671810
                                    14.376365 104.130241
                                                            1.433537
         3
            5.778359 -34.712038
                                    48.999631
                                                59.699204 40.661263 -41.658731
                                                53.333041 -47.062739 37.356212
            51.138220
                        38.587312 123.639488
                   6
                              7
                                          8
                                                      9
                                                                           374 \
        0 -172.057790 -92.502626 113.223311 50.562456
                                                                     12.397640
         1 -39.140733
                       -82.692374
                                    45.161483 -9.556298
                                                                   -21.987079
                                                             . . .
                       -97.124595
        2 -148.519385
                                    41.972195 50.948731
                                                                     3.027700
                                                             . . .
        3 -36.808594
                        24.170655
                                     0.235601 -29.407290
                                                                   13.100007
                                                             . . .
         4 -298.722753 -106.421119 106.248914 65.880707
                                                             . . .
                                                                    13.906532
                 375
                            376
                                       377
                                                  378
                                                             379
                                                                        380
                                                                                  381
        0 40.909521
                      8.150260 -15.170692
                                           18.007711
                                                        6.166997 -30.124162 3.700903
         1 -12.389279 20.667980
                                  2.202720 -17.142453 -5.880972 -10.123963 -4.890665
                                             4.355141
                                                       2.936152 -20.199555 9.816351
          14.025767 -2.960312
                                -3.206544
           1.405671 -1.891076 -7.882638
                                            18.000561 12.106918 -10.507835 5.243834
         3
         4 43.461721 11.519207 -22.468284 45.431128
                                                       8.161224 -35.373910 7.728865
                  382
                            383
        0 -1.757692 -1.818057
         1 -13.018393 -5.219309
        2 11.894366 -8.798819
        3 10.158340 5.886351
            9.592849 5.447336
         [5 rows x 384 columns]
In [14]: # Questions 2 tfidf weighted word2vec
        df3_q2.head()
Out[14]:
                  0
                                         2
                                                   3
                             1
                                                                           5
        0 125.983301 95.636484 42.114717 95.449986 -37.386301
                                                                    39.400084
         1 -106.871899 80.290340
                                  79.066300 59.302100 -42.175332 117.616657
         2
             7.072875 15.513378
                                   1.846914 85.937583 -33.808811
                                                                     94.702337
            39.421539 44.136990 -24.010927 85.265864 -0.339028
                                                                    -9.323141
         3
                                                                    66.674880
            31.950109 62.854102
                                   1.778147 36.218763 -45.130861
                  6
                              7
                                          8
                                                     9
                                                                           374 \
```

```
0 -148.116068 -87.851481 110.371972 62.272816
                                                                                                                              . . .
                                                                                                                                                16.165591
                  1 -144.364242 -127.131506 22.962531 25.397579
                                                                                                                              . . .
                                                                                                                                                -4.901130
                  2 -122.256856 -114.009530 53.922293 60.131814
                                                                                                                                                  8.359966
                                                                                                                              . . .
                  3 -60.499653 -37.044767 49.407847 -23.350149
                                                                                                                                                  3.311411
                  4 -106.342344 -22.901031
                                                                           59.835921 62.663957
                                                                                                                                                -2.403871
                                     375
                                                           376
                                                                                  377
                                                                                                         378
                                                                                                                                379
                                                                                                                                                      380 \
                  0 33.030668
                                                 7.019997 -14.793958 15.437512
                                                                                                                     8.199657 -25.070835
                  1 -4.565391 41.520751 -0.727564 -16.413779 -7.373777
                                                                                                                                            2.638877
                  2 -2.165985 10.936580 -16.531660 14.681230 15.633759 -1.210901
                  3 3.788878 13.398598 -6.592596
                                                                                              6.437365 5.993291
                                                                                                                                            2.732392
                  4 11.991207
                                                 8.088480 -15.090196
                                                                                              8.375168 1.727225 -6.601130
                                     381
                                                            382
                                                                                  383
                       1.571618
                                                 1.603736
                                                                        0.305646
                  1 -7.403460 2.703069 0.408038
                  2 14.183826 11.703135 10.148075
                  3 -3.727647
                                              5.614115 6.023694
                  4 11.317404 11.544597
                                                                        2.478685
                  [5 rows x 384 columns]
In [15]: print("Number of features in nlp dataframe :", df1.shape[1])
                  print("Number of features in preprocessed dataframe :", df2.shape[1])
                  print("Number of features in question1 w2v dataframe: ", df3_q1.shape[1])
                  print("Number of features in question2 w2v dataframe :", df3_q2.shape[1])
                  print("Number of features in final dataframe :", df1.shape[1]+df2.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3_q1.shape[1]+df3
Number of features in nlp dataframe: 17
Number of features in preprocessed dataframe: 12
Number of features in question1 w2v dataframe: 384
Number of features in question2 w2v dataframe: 384
Number of features in final dataframe : 797
In [16]: # storing the final features to csv file
                  if not os.path.isfile('final_features.csv'):
                          df3_q1['id']=df1['id']
                          df3_q2['id']=df1['id']
                          df1 = df1.merge(df2, on='id',how='left')
                          df2 = df3_q1.merge(df3_q2, on='id',how='left')
                          result = df1.merge(df2, on='id',how='left')
                          result.to_csv('final_features.csv')
In [1]: df1.head()
```

------

```
NameError

<ipython-input-1-eda2b5cc1a42> in <module>
----> 1 df1.head()

NameError: name 'df1' is not defined
```

In []:

Traceback (most recent call last)