3.Q_Mean_W2V

April 18, 2019

3.6 Featurizing text data with tfidf weighted word-vectors

```
In [42]: 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 [43]: # 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 [44]: df.head()
```

```
Out [44]:
           id qid1 qid2
                                                                   question1 \
                        2 What is the step by step guide to invest in sh...
        0
                        4 What is the story of Kohinoor (Koh-i-Noor) Dia...
                  3
        1
            1
         2
                       6 How can I increase the speed of my internet co...
                       8 Why am I mentally very lonely? How can I solve...
         3 3
                       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 [45]: # 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_))
```

In []: word2tfidf

- 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.

```
In [23]: # en_vectors_web_lg, which includes over 1 million unique vectors.
         nlp = spacy.load(r"C:\Users\Byron\Applications\PythonMaster\Lib\site-packages\en_core
In [32]: vecs1 = []
         # https://github.com/noamraph/tqdm
         # tqdm is used to print the progress bar
         for qu1 in tqdm(list(df['question1'])):
             doc1 = nlp(qu1)
             # 384 is the number of dimensions of vectors
             mean_vec1 = np.zeros([len(doc1), 300])
             for word1 in doc1:
                 # word2vec
                 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 [56:22<00:00, 119.54it/s]
In [33]: vecs2 = []
         for qu2 in tqdm(list(df['question2'])):
             doc2 = nlp(qu2)
             mean_vec2 = np.zeros([len(doc2), 300])
             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 [58:32<00:00, 115.11it/s]
In [34]: #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 :
In [35]: 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 [36]: # dataframe of nlp features
         df1.head()
Out[36]:
                is_duplicate
            id
                                cwc_min
                                           cwc_max
                                                     csc_min
                                                                csc_max
                                                                          ctc_min \
                               0.999980
                                                                         0.916659
         0
                                          0.833319
                                                    0.999983
                                                               0.999983
         1
             1
                            0
                               0.799984
                                          0.399996
                                                    0.749981
                                                               0.599988
                                                                          0.699993
         2
             2
                            0
                               0.399992
                                          0.333328
                                                    0.399992
                                                               0.249997
                                                                          0.399996
         3
             3
                            0
                              0.000000
                                          0.000000
                                                    0.000000
                                                               0.000000
                                                                         0.000000
         4
             4
                               0.399992 0.199998
                                                    0.999950
                                                               0.666644
                                                                         0.571420
                       last_word_eq first_word_eq abs_len_diff
             ctc_max
                                                                    mean_len \
                                0.0
                                                1.0
                                                               2.0
                                                                         13.0
         0 0.785709
                                0.0
                                                               5.0
         1
            0.466664
                                                1.0
                                                                         12.5
         2 0.285712
                                0.0
                                                1.0
                                                               4.0
                                                                         12.0
         3 0.000000
                                0.0
                                                0.0
                                                               2.0
                                                                         12.0
         4 0.307690
                                0.0
                                                1.0
                                                               6.0
                                                                         10.0
            token_set_ratio token_sort_ratio fuzz_ratio fuzz_partial_ratio
         0
                         100
                                             93
                                                          93
                                                                              100
         1
                          86
                                             63
                                                          66
                                                                               75
         2
                          63
                                             63
                                                          43
                                                                               47
         3
                          28
                                             24
                                                           9
                                                                               14
         4
                          67
                                             47
                                                          35
                                                                               56
            longest_substr_ratio
         0
                         1.000000
                         0.607843
         1
         2
                         0.169492
         3
                         0.040000
                         0.179487
In [37]: # data before preprocessing
         df2.head()
Out [37]:
            id
                freq_qid1
                            freq_qid2
                                        q1len q2len q1_n_words
                                                                   q2_n_words
         0
             0
                         1
                                     1
                                           66
                                                  57
                                                               14
                                                                            12
         1
             1
                         4
                                     1
                                           51
                                                  88
                                                                8
                                                                            13
         2
             2
                                     1
                                           73
                                                  59
                                                               14
                         1
                                                                            10
         3
             3
                         1
                                     1
                                           50
                                                  65
                                                               11
                                                                             9
         4
             4
                         3
                                     1
                                           76
                                                  39
                                                               13
                                                                             7
            word_Common word_Total word_share
                                                   freq_q1+q2 freq_q1-q2
                                         0.434783
         0
                    10.0
                                23.0
                                                             2
                                                                          0
                                                             5
                                                                          3
         1
                     4.0
                                20.0
                                         0.200000
                                                             2
         2
                     4.0
                                24.0
                                                                          0
                                         0.166667
                                                             2
                                                                          0
         3
                     0.0
                                19.0
                                         0.000000
                     2.0
                                20.0
                                         0.100000
                                                             4
                                                                          2
In [38]: # Questions 1 tfidf weighted word2vec
```

df3_q1.head()

```
Out [38]:
                                                                       5
                 0
                            1
                                       2
                                                 3
                                                            4
                                 4.862720 7.971019 20.345586 -5.514759 -4.077800
        0 -5.856872 17.449559
           9.356103 13.098566 18.945098 -2.079594 -15.703841
                                                                 -2.173409
        1
                                                                             8.969065
            0.909520 16.050299 -8.126856 -4.848289 -2.806190
                                                                  9.752280
                                                                             4.349992
        3 -4.950745 17.098874 -15.474965 1.044680 -2.392017
                                                                 -0.051889
                                                                             2.650595
        4 -11.520302 19.769948 -4.510997 -6.548994 -20.835286 33.663909 -30.390504
                 7
                            8
                                        9
                                                              290
                                                                        291
        0 - 2.820742
                       8.029026
                                146.599092
                                                       -17.370964 5.393082
                                               . . .
        1 -20.458267 -20.674299
                                  13.760798
                                                        25.948247 0.603713
        2 -5.120332
                       6.785252 106.342974
                                                       -20.942061 2.398984
        3 -8.451192
                                116.184408
                       2.584123
                                               . . .
                                                        -2.551312 -4.971480
            0.826553 -19.571472
                                  84.458577
                                                        -8.331733 -4.866335
                                               . . .
                 292
                            293
                                       294
                                                  295
                                                             296
                                                                        297
                                                                                  298
            0.384676 -8.362788
                                -1.880290 -10.799672 -12.999799
                                                                   3.225858 1.256145
        0
        1 -10.516349
                       6.040723 30.476707
                                             3.976890 -28.254610 12.613432 -7.770673
            8.663028 -0.654124 16.220601 -2.719094 10.485332 -1.103132 -7.290877
        3 -0.478381
                     -1.930166
                                  9.336016
                                             2.574459
                                                        4.803863 -1.182989 -2.962115
        4 18.828458 -40.357679 -10.336167 15.294630 -0.989347 -9.072091 -8.194567
                 299
          16.807275
        1 31.456654
        2 19.314250
            3.225704
        3
        4 23.847560
         [5 rows x 300 columns]
In [39]: # Questions 2 tfidf weighted word2vec
        df3_q2.head()
Out [39]:
                 0
                                       2
                                                 3
                                                            4
                                                                      5
                            1
            0.398579 13.991607 -0.504564 9.254431 13.906436 -4.777694 -5.274421
        0
            4.649688
                       9.974928 20.330103 -0.440372 -18.128566 -1.984671
                                                                            4.906458
        2 -17.305105 17.355614 -9.135664 -6.038550 -1.831651 4.547895
                                                                          17.935764
            3.897911
                       2.545857 -2.053792 3.385450
                                                      3.424216 -2.282545 -11.763825
        4 -5.391206
                       1.767221
                                  1.810128 -4.097073 -3.623262 8.417368 -25.246265
                 7
                            8
                                        9
                                                              290
                                                                        291 \
        0 -0.201208
                       4.940558 134.735950
                                                       -17.810438 7.231024
        1 -27.797837 -21.262646
                                  96.965297
                                                        23.015827 3.435464
        2 -4.799029
                       3.100311
                                                       -24.310109 -1.216773
                                  99.380095
                       5.797674
            6.692485
                                  94.978085
                                                        -5.435584 1.672591
        3
                                               . . .
            7.473430 -2.789541
                                  89.594627
                                                       -10.407441 -8.444207
                                               . . .
                 292
                            293
                                       294
                                                  295
                                                             296
                                                                        297 \
```

```
0 1.531186 -7.528823 0.473802 -11.864658 -11.293788
                                                                                                                                                                  1.866265
                     1 -5.169600 7.102491 34.516881 6.177686 -27.770856 12.926435
                     2 11.909693 9.591573 11.846737 1.397859
                                                                                                                                        6.454157 -0.271460
                     3 -0.863278 -2.906553 -3.466688 -3.867892 -4.249463 -12.551012
                     4 -14.450059 -12.709382 -4.449050 12.563987 -11.721362 -16.459300
                                          298
                                                                     299
                              3.616046 11.971096
                     1 -4.564559 33.919834
                     2 -12.500337 27.634567
                     3 4.494087 -6.223341
                              3.626297 -9.790615
                      [5 rows x 300 columns]
In [40]: 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: 300
Number of features in question2 w2v dataframe: 300
Number of features in final dataframe : 629
In [41]: # 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')
```