Quora Question Pair Similarity

Part 3: Featurizing Data

Featurizing Text Data with TFIDF weighted Word2Vec

In [1]:

```
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 [2]:

In [3]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
# 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.

In [13]:

```
# en_vectors_web_lg, which includes over 1 million unique vectors.
nlp = spacy.load('en_core_web_sm')
vecs1 = []
for qu1 in tqdm(list(df['question1'])):
    doc1 = nlp(qu1)
    # 96 is the number of dimensions of vectors
    mean_vec1 = np.zeros([len(doc1), 96])
    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)
  0%||
```

```
0%||
| 742/404290 [00:10<2:14:13, 50.11it/s]
```

In [16]:

```
vec1 = word1.vector
len(vec1)
```

Out[16]:

96

In [45]:

```
# en vectors web lq, which includes over 1 million unique vectors.
nlp = spacy.load('en_core_web_sm')
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), len(doc1[0].vector)])
   for word1 in doc1:
       # word2vec
       vec1 = word1.vector
       # fetch df score
           idf = word2tfidf[str(word1)]
       except:
           idf = 0
       # compute final vec
       temp = vec1*idf
       mean_vec1 += temp
   mean_vec1 = mean_vec1.mean(axis=0)
   vecs1.append(mean vec1)
df['q1_feats_m'] = list(vecs1)
 0% l
| 0/404290 [00:00<?, ?it/s]
______
ValueError
                                       Traceback (most recent call last)
<ipython-input-45-e6212d65c433> in <module>
    19
              # compute final vec
    20
              temp = vec1*idf
```

```
---> 21
                mean_vec1 += temp
     22
           mean_vec1 = mean_vec1.mean(axis=0)
     23
           vecs1.append(mean_vec1)
ValueError: operands could not be broadcast together with shapes (15,384) (9
6,) (15,384)
```

In [37]:

```
vecs2 = []
for qu2 in tqdm(list(df['question2'])):
    doc2 = nlp(qu2)
    mean_vec2 = np.zeros([len(doc1), len(doc1[0].vector)])
    for word2 in doc2:
        # word2vec
        vec2 = word2.vector
        # fetch df score
        try:
            idf = word2tfidf[str(word2)]
        except:
            #print word
            idf = 0
        # compute final vec
        temp = vec2 * idf
        mean vec1 += temp
    mean_vec2 = mean_vec2.mean(axis=0)
    vecs2.append(mean vec2)
df['q2_feats_m'] = list(vecs2)
```

100%

404290/404290 [2:00:16<00:00, 56.81it/s]

In [38]:

```
#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 notebook")
```

In [39]:

```
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 [40]:

```
# dataframe of nlp features
df1.head()
```

Out[40]:

	id	is_duplicate	cwc_min	cwc_max	csc_min	csc_max	ctc_min	ctc_max	last_word_eq
0	0	0	0.999980	0.833319	0.999983	0.999983	0.916659	0.785709	0.0
1	1	0	0.799984	0.399996	0.749981	0.599988	0.699993	0.466664	0.0
2	2	0	0.399992	0.333328	0.399992	0.249997	0.399996	0.285712	0.0
3	3	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0
4	4	0	0.399992	0.199998	0.999950	0.666644	0.571420	0.307690	0.0

In [41]:

data before preprocessing
df2.head()

Out[41]:

	id	freq_qid1	freq_qid2	q1len	q2len	q1_n_words	q2_n_words	word_Common	word_Tota
0	0	1	1	66	57	14	12	10.0	23.0
1	1	4	1	51	88	8	13	4.0	20.0
2	2	1	1	73	59	14	10	4.0	24.0
3	3	1	1	50	65	11	9	0.0	19.0
4	4	3	1	76	39	13	7	2.0	20.0
4									•

In [42]:

Questions 1 tfidf weighted word2vec
df3_q1.head()

Out[42]:

	0	1	2	3	4	5	6	
0	211.129864	-144.683059	-68.811247	-153.662141	-89.931593	2.311301	136.743747	50
1	144.124685	-114.012484	-111.716694	-104.885038	-88.238478	16.441834	58.238013	102.
2	81.757898	-142.184507	0.559867	-104.660084	-84.156631	22.515110	115.521661	50.
3	-126.651922	-59.747160	-67.763201	-138.114731	-101.038699	88.148523	-22.912261	85.
4	299 444044	-188 632001	-22 946291	-273 683355	-188 480395	107 123044	174 946302	-72

5 rows × 96 columns

localhost:8888/notebooks/Desktop/Applied ML Course/Case Studies/Quora/3. Quora - Featurizing Data.ipynb

In [43]:

```
# Questions 2 tfidf weighted word2vec
df3_q2.head()
```

Out[43]:

```
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```

5 rows × 96 columns

In [44]:

```
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]
```

Number of features in nlp dataframe : 17 Number of features in preprocessed dataframe : 12 Number of features in question1 w2v dataframe : 96 Number of features in question2 w2v dataframe : 96 Number of features in final dataframe : 221

In [0]:

```
# 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')
```