# 3.6 Featurizing text data with tfidf weighted word-vectors

In [78]:

```
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
from scipy.sparse import hstack
# exctract word2vec vectors
# https://github.com/explosion/spaCy/issues/1721
# http://landinghub.visualstudio.com/visual-cpp-build-tools
import spacy
from sklearn.model_selection import train test split
```

In [79]:

In [80]:

```
#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')
    dfnlp=dfnlp[:100000]
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')
    dfppro=dfppro[:100000]
else:
    print("download df_fe_without_preprocessing_train.csv from drive or run previous notebook")
```

In [81]:

```
data=pd.concat([df,dfnlp,dfppro],axis=1)
print(data.shape)
#https://stackoverflow.com/questions/14984119/python-pandas-remove-duplicate-columns
data = data.loc[:,~data.columns.duplicated()]
print(data.shape)

(100000, 44)
(100000, 32)
```

```
In [82]:
y true = data['is duplicate']
data.drop(['is duplicate'], axis=1, inplace=True)
In [83]:
X tr,X test, y tr, y test = train test split(data, y true, stratify=y true, test size=0.3)
In [84]:
print("Number of data points in train data :", X tr.shape)
print("Number of data points in test data :", X test.shape)
Number of data points in train data: (70000, 31)
Number of data points in test data: (30000, 31)
In [85]:
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import CountVectorizer
# merge texts
questions = list(X tr['question1']) + list(X tr['question2']) + list(X test['question1']) + list(X
test['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 [86]:
# en_vectors_web_lg, which includes over 1 million unique vectors.
nlp = spacy.load('en core web sm')
def vectoriser(source df,dest df,source column,dest column):
    vecs1 = []
       # https://github.com/noamraph/tqdm
        # tqdm is used to print the progrsource columness bar
    for qu1 in tqdm(list(source_df[source_column])):
        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
            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)
    dest df[dest column] = list(vecs1)
In [87]:
vectoriser(X_tr,X_tr,'question1','q1_feats_m')
vectoriser(X_tr, X_tr, 'question2', 'q2_feats_m')
vectoriser(X_test, X_test, 'question1', 'q1_feats_m')
vectoriser(X test, X test, 'question2', 'q2 feats m')
100%|
                                                                                | 70000/70000 [30
:46<00:00, 37.90it/s]
100%|
                                                                          | 70000/70000 [31
:56<00:00, 36.53it/s]
100%|
                                                                          30000/30000 [09
:49<00:00, 50.86it/s]
```

1 30000/30000 [10

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:02<00:00, 50.54it/s]
```

#### In [88]:

```
print("Number of data points in train data :",X_tr.shape)
print("Number of data points in test data :",X_test.shape)
```

Number of data points in train data: (70000, 33) Number of data points in test data: (30000, 33)

#### In [89]:

```
#https://stackoverflow.com/questions/40924332/splitting-a-list-in-a-pandas-cell-into-multiple-colu
X_trl=pd.concat([X_tr,pd.DataFrame(X_tr['ql_feats_m'].values.tolist(),index=X_tr.index,columns=['0]
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'16_1','17_1','18_1','19_1','20_1','21_1','22_1','23_1','24_1','25_1','26_1','27_1','28_1','29_1','
30_1','31_1','32_1','33_1','34_1','35_1','36_1','37_1','38_1','39_1','40_1','41_1','42_1','43_1','4
4_1','45_1','46_1','47_1','48_1','49_1','50_1','51_1','52_1','53_1','54_1','55_1','56_1','57_1','58
   \overline{1}','59\overline{1}','60\overline{1}','61\overline{1}','62\overline{1}','63\overline{1}','64\overline{1}','65\overline{1}','66\overline{1}','67\overline{1}','68\overline{1}','69\overline{1}','70\overline{1}','71\overline{1}','72
\bar{1','73_\bar{1','74_\bar{1','75_\bar{1','77_\bar{1','78_\bar{1','79_\bar{1','80_\bar{1','82_\bar{1','83_\bar{1','84_\bar{1','85_\bar{1','86_\bar{1}}}}
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 1','235_1','236_1','237_1','238_1','239_1','240_1','241_1','242_1','243_1','244_1','245_1','246_1'
 ,'247_1','248_1','249_1','250_1','251_1','252_1','253_1','254_1','255_1','256_1','257_1','258_1','
 259 1','260 1,'261 1,'262 1','263 1,'264 1,'265 1','265 1','266 1,'267 1,'268 1','269 1,'270 1
 1_1,'272_1,'273_1,'274_1,'275_1,'276_1,'276_1,'277_1,'278_1,'279_1,'280_1,'280_1,'281_1,'282_1,'283_
 1, '284 1, '285 1, '286 1, '287 1, '288 1, '289 1, '290 1, '291 1, '292 1, '293 1, '293 1, '294 1, '295 1,
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 308 1, '309 1, '310 1, '311 1, '312 1, '313 1, '314 1, '315 1, '316 1, '317 1, '318 1, '319 1, '32
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1','333 1','334 1','335 1','336 1','337_1','338_1','339_1','340_1','341_1','342_1','343_1','344_1'
 ,'345 1','346 1','347 1','348 1','349 1','350 1','351 1','352_1','353_1','354_1','355_1','356_1','
 357 1, 358 1, 359 1, 360 1, 361 1, 361 1, 362 1, 363 1, 364 1, 365 1, 366 1, 367 1, 367 1, 368 1, 36
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 1','382_1','383_1'])], axis=1)
X_test1=pd.concat([X_test1,pd.DataFrame(X_test['q1_feats_m'].values.tolist(),index=X_test.index,col
umns=['0_2','1_2','2_2','3_2','4_2','5_2','6_2','7_2','8_2','9_2','10_2','11_2','12_2','13_2','14_2
 ','15_2','16_2','17_2','18_2','19_2','20_2','21_2','22_2','23_2','24_2','25_2','26_2','27_2','28_2'
   '29_2','30_2','31_2','32_2','33_2','34_2','35_2','36_2','37_2','38_2','39_2','40_2','41_2','42_2',
 '43_2','44_2','45_2','46_2','47_2','48_2','49_2','50_2','51_2','52_2','53_2','54_2','55_2','56_2','
    7_2','58_2','59_2','60_2','61_2','62_2','63_2','64_2','65_2','66_2','67_2','68_2','69_2','70_2','7
_2','72_2','73_2','74_2','75_2','76_2','77_2','78_2','79_2','80_2','81_2','82_2','83_2','84_2','85
_2','86_2','87_2','88_2','89_2','90_2','91_2','92_2','93_2','94_2','95_2','96_2','97_2','98_2','99_2','100_2','101_2','102_2','103_2','104_2','105_2','106_2','107_2','108_2','109_2','110_2','111_2'
  '112 2','113 2','114 2','115 2','116 2','117 2','118 2','119 2','120 2','121 2','121 2','123 2','
124_2, '125_2, '126_2, '127_2, '128_2, '129_2, '130_2, '131_2, '132_2, '133_2, '133_2, '134_2, '135_2, '13
 6 2','137 2','138 2','139 2','140 2','141 2','142 2','143 2','144 2','145 2','146 2','147 2','148
    ','149_2','150_2','151_2','152_2','153_2','154_2','155_2','156_2','157_2','158_2','159_2','160_2'
 ,'161 2','162 2','163 2<sup>'</sup>,'164 2<sup>'</sup>,'165 2<sup>'</sup>,'166 2<sup>'</sup>,'167 2<sup>'</sup>,'168 2<sup>'</sup>,'169 2<sup>'</sup>,'170 2<sup>'</sup>,'171 2<sup>'</sup>,'172 2<sup>'</sup>,'
 173 2','174 2','175 2','176 2','177 2','178 2','179 2','180 2','181 2','182 2','183 2','184 2','18
 5_2','186_2','187_2','188_2','189_2','190_2','191_2','192_2','193_2','194_2','195_2','196_2','197
 2','198_2','199_2','200_2','201_2','202_2','203_2','204_2','205_2','206_2','207_2','208_2','209_2'
   '210 2<sup>-</sup>,'211 2<sup>-</sup>,'212 2<sup>-</sup>,'213 2<sup>-</sup>,'214 2<sup>-</sup>,'215 2<sup>-</sup>,'216 2<sup>-</sup>,'217 2<sup>-</sup>,'218 2<sup>-</sup>,'219 2<sup>-</sup>,'220 2<sup>-</sup>,'221
222_2<sup>-</sup>,'223_2<sup>-</sup>,'224_2<sup>-</sup>,'225_2<sup>-</sup>,'226_2<sup>-</sup>,'227_2<sup>-</sup>,'228_2<sup>-</sup>,'229_2<sup>-</sup>,'230_2<sup>-</sup>,'231_2<sup>-</sup>,'232_2<sup>-</sup>,'233_2<sup>-</sup>,'23
 4 2','235 2','236 2','237 2','238 2','239 2','240 2','241 2','242 2','243 2','244 2','245 2','246
 2<sup>-</sup>, '247 2<sup>-</sup>, '248 2<sup>-</sup>, '249 2<sup>-</sup>, '250 2<sup>-</sup>, '251 2<sup>-</sup>, '252 2<sup>-</sup>, '253 2<sup>-</sup>, '254 2<sup>-</sup>, '255 2<sup>-</sup>, '256 2<sup>-</sup>, '257 2<sup>-</sup>, '258 2<sup>-</sup>
 ,'259 2','260 2','261 2','262 2','263 2','264 2','265 2','266 2','267 2','268 2','269 2','270 2','
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 2','296 2','297 2','298 2','299 2','300 2','301 2','302 2','303 2','304 2','305 2','306 2','307 2'
   '308 2','309 2','310 2','311 2','312 2','313 2','314 2','315 2','316 2','317 2','318 2','319 2','
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 ,'357_2<sup>T</sup>,'358_2<sup>T</sup>,'359_2<sup>T</sup>,'360_2<sup>T</sup>,'361_2<sup>T</sup>,'362_2<sup>T</sup>,'363_2<sup>T</sup>,'364_2<sup>T</sup>,'365_2<sup>T</sup>,'366_2<sup>T</sup>,'367_2<sup>T</sup>,'368_2<sup>T</sup>,'
 369 2','370 2<sup>1</sup>,'371 2<sup>1</sup>,'372 2','373 2<sup>1</sup>,'374 2<sup>1</sup>,'375 2','376 2<sup>1</sup>,'377 2<sup>1</sup>,'378 2','379 2<sup>1</sup>,'380 2<sup>1</sup>,'38
 1 2','382 2','383_2'])], axis=1)
 4
```

## In [90]:

```
print("Number of data points in train data :",X_tr.shape)
print("Number of data points in test data :",X_test.shape)
print("Number of data points in train data :",X_tr1.shape)
print("Number of data points in test data :",X_test1.shape)
```

```
Number of data points in train data: (70000, 33)
Number of data points in test data: (30000, 33)
Number of data points in train data: (70000, 801)
Number of data points in test data: (30000, 801)
In [92]:
X tr.drop(['q1 feats m','q2 feats m'], axis=1,inplace=True)
X_test.drop(['q1_feats_m','q2_feats_m'], axis=1,inplace=True)
X tr1.drop(['qid1','qid2','question1','question2','q1 feats m','q2 feats m'], axis=1,inplace=True)
X test1.drop(['qid1','qid2','question1','question2','q1 feats m','q2 feats m'],
axis=1,inplace=True)
In [93]:
print("Number of data points in train data :",X tr.shape)
print("Number of data points in test data :", X test.shape)
print("Number of data points in train data :",X_trl.shape)
print("Number of data points in test data :",X test1.shape)
Number of data points in train data: (70000, 31)
Number of data points in test data: (30000, 31)
Number of data points in train data : (70000, 795)
Number of data points in test data: (30000, 795)
In [94]:
from sqlalchemy import create engine
engine = create engine('sqlite:///w2v data.db')
\#X \text{ tr } q1 = X \text{ tr } q1.\text{merge}(X \text{ tr } q2, \text{ on='id',how='left'})
#w2v tr = df1 tr.merge(X tr q1, on='id',how='left')
print(X_tr1.shape)
X_tr1.to_sql('X_tr', engine, if_exists='replace')
y_tr.to_sql('y_tr', engine, if exists='replace')
#X test q1 = X test q1.merge(X test q2, on='id',how='left')
#w2v_test = df1_test.merge(X_test_q1, on='id',how='left')
print(X test1.shape)
X_test1.to_sql('X_test', engine, if_exists='replace')
y_test.to_sql('y_test', engine, if_exists='replace')
(70000, 795)
(30000, 795)
In [96]:
print("Number of data points in train data :", X_tr.shape)
print("Number of data points in test data :",X test.shape)
Number of data points in train data : (70000, 31)
Number of data points in test data: (30000, 31)
In [97]:
#https://stackoverflow.com/questions/45961747/append-tfidf-to-pandas-dataframe
from sklearn.feature extraction.text import TfidfVectorizer
tfidf = TfidfVectorizer(min df=10)
tfidf ques1 tr= tfidf.fit transform(X tr['question1'])
tfidf_ques1_test= tfidf.transform(X_test['question1'])
X_q1_tr = pd.DataFrame(tfidf_ques1_tr.toarray(), columns=tfidf.get_feature_names(), index=
X tr.index)
X_q1_test = pd.DataFrame(tfidf_ques1_test.toarray(), columns=tfidf.get_feature_names(), index=
X test.index)
#X tr['q1 feats m'] = list(tfidf ques1 tr.toarray())
#X test['q1 feats m'] = list(tfidf ques1 test.toarray())
#X tr['q2 feats m'] = list(tfidf ques2 tr.toarray())
#X test['q2 feats m'] = list(tfidf ques2 test.toarray())
```

- 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 [99]:

```
#X_q1_tr = pd.DataFrame(tfidf_ques1_tr.toarray(), index= X_tr.index)

tfidf = TfidfVectorizer(min_df=10)
    tfidf_ques2_tr= tfidf.fit_transform(X_tr['question2'])

tfidf_ques2_test= tfidf.transform(X_test['question2'])

X_q2_tr = pd.DataFrame(tfidf_ques2_tr.toarray(), columns=tfidf.get_feature_names(), index=
    X_tr.index)

X_q2_test = pd.DataFrame(tfidf_ques2_test.toarray(), columns=tfidf.get_feature_names(), index=
    X_test.index)
```

### In [101]:

```
print("Number of data points in train data :",X_tr.shape)
print("Number of data points in test data :",X_test.shape)

Number of data points in train data : (70000, 31)
Number of data points in test data : (30000, 31)
```

#### In [102]:

```
X_tr.drop(['qid1','qid2','question1','question2'],axis=1,inplace=True)
X_test.drop(['qid1','qid2','question1','question2'],axis=1,inplace=True)
```

## In [106]:

```
X_tr=pd.concat([X_tr,X_q1_tr,X_q2_tr], axis=1)
X_test=pd.concat([X_test,X_q1_test,X_q2_test], axis=1)
```

### In [108]:

```
print("Number of data points in train data :",X_tr.shape)
print("Number of data points in test data :",X_test.shape)
```

Number of data points in train data : (70000, 10431) Number of data points in test data : (30000, 10431)

# In [109]:

```
#X_q1_tr = X_q1_tr.merge(X_q2_tr, on='id',how='left')
#tfidf_tr = df2_tr.merge(X_q1_tr, on='id',how='left')
print(X_tr.shape)
X_tr.to_csv('X_tr.csv',index='id')
#X_q1_test = X_q1_test.merge(X_q2_test, on='id',how='left')
#tfidf_test = df2_test.merge(X_q1_test, on='id',how='left')
print(X_test.shape)
X_test.to_csv('X_test.csv',index='id')
```

(70000, 10431) (30000, 10431)