Project_2_Quara Question Pairs Similarity

Dataset Link - https://www.kaggle.com/c/quora-question-pairs)

Dataset Description

The goal of this competition is to predict which of the provided pairs of questions contain two questions with the same meaning. The ground truth is the set of labels that have been supplied by human experts. The ground truth labels are inherently subjective, as the true meaning of sentences can never be known with certainty. Human labeling is also a 'noisy' process, and reasonable people will disagree. As a result, the ground truth labels on this dataset should be taken to be 'informed' but not 100% accurate, and may include incorrect labeling. We believe the labels, on the whole, to represent a reasonable consensus, but this may often not be true on a case by case basis for individual items in the dataset.

Please note:

as an anti-cheating measure, Kaggle has supplemented the test set with computer-generated question pairs. Those rows do not come from Quora, and are not counted in the scoring. All of the questions in the training set are genuine examples from Quora.

Data fields

id - the id of a training set question pair qid1, qid2 - unique ids of each question (only available in train.csv) question1, question2 - the full text of each question is_duplicate - the target variable, set to 1 if question1 and question2 have essentially the same meaning, and 0 otherwise.

```
In [1]:
            import numpy as np
          2 import pandas as pd
          3 import nltk
          4 | from nltk.sentiment.vader import SentimentIntensityAnalyzer
            import re
          6 from textblob import TextBlob
          7 from wordcloud import WordCloud
          8 import seaborn as sns
          9 import matplotlib.pyplot as plt
         10 import cufflinks as cf
         11 %matplotlib inline
         12 from plotly.offline import init notebook mode, iplot
         13 init_notebook_mode(connected = True)
         14 cf.go_offline();
         15 import plotly.graph_objs as go
         16 | from plotly.subplots import make_subplots
         17
         18
         19 pd.set_option('display.max_columns', None)
         20
         21
         22 from nltk.corpus import stopwords
         23 from nltk.stem import SnowballStemmer
         24 from sklearn.feature extraction.text import CountVectorizer
         25
         26 from collections import Counter
         27
            from numpy import where
         28
         29
         30 from sklearn.decomposition import PCA
         31
         32 from sklearn.preprocessing import OneHotEncoder
         33 from sklearn.preprocessing import StandardScaler
         34
         35 from sklearn.model_selection import train_test_split
         36 from sklearn.linear model import LogisticRegression
         37 from sklearn.metrics import accuracy score
         38 from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
         39
            from sklearn import metrics
         40
         41
            from scipy.sparse import hstack, vstack
         42
         43 from prettytable import PrettyTable
         44
            from scipy.stats import loguniform # Log-uniform is useful for searching
            from sklearn.model_selection import RepeatedStratifiedKFold, RandomizedSe
         45
         46
            import warnings
         47
            warnings.filterwarnings('ignore')
         48
```

C:\ProgramData\anaconda3\Lib\site-packages\paramiko\transport.py:219: Crypto
graphyDeprecationWarning:

Blowfish has been deprecated

Out[2]: (404290, 6)

In [3]: 1 df.head()

Out[3]:

	id	qid1	qid2	question1	question2	is_duplicate
0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0
1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0
3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24}[/math] i	0
4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0

In [4]: 1 df.sample(10)

Out[4]:

	id	qid1	qid2	question1	question2	is_duplicate
368038	368038	498371	498372	How realistic are Tyler Durden's plans in "Fig	[SPOILER ALERT] At the end of Fight Club, why	0
16732	16732	31869	31870	If goods are being transported from one state	If I want to sell goods from one state to anot	0
82545	82545	139936	98268	What are some things new employees should know	What are some things new employees should know	0
253486	253486	368039	368040	Where does GoodRec.com get its data from?	Where does www.magnetic.com get its data from?	0
225230	225230	333516	333517	What was your latest discovery?	What is the latest discovery in finance?	0
357058	357058	9642	154412	How can you make a friend?	How do I make more friends? How do I?	1
11808	11808	22786	22787	What makes you proud to be Canadian?	How do I make perfect CV for job banks for Can	0
300133	300133	361686	422891	Has India been succesful to divert world atten	Why is India not conducting a surgical strike	0
88130	88130	148290	148291	How did Aditi Saini, once a careless girl (as	What are the perks of being an IES officer?	0
233509	233509	343774	343775	What is the use of shell scripting?	Where can I get Shell Script programs?	0

In [5]: 1 df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 404290 entries, 0 to 404289
Data columns (total 6 columns):

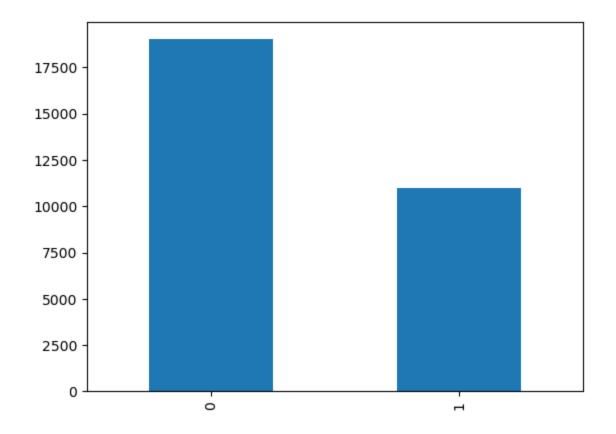
#	Column	Non-Null Count	Dtype
0	id	404290 non-null	int64
1	qid1	404290 non-null	int64
2	qid2	404290 non-null	int64
3	question1	404289 non-null	object
4	question2	404288 non-null	object
5	is_duplicate	404290 non-null	int64

dtypes: int64(4), object(2)
memory usage: 18.5+ MB

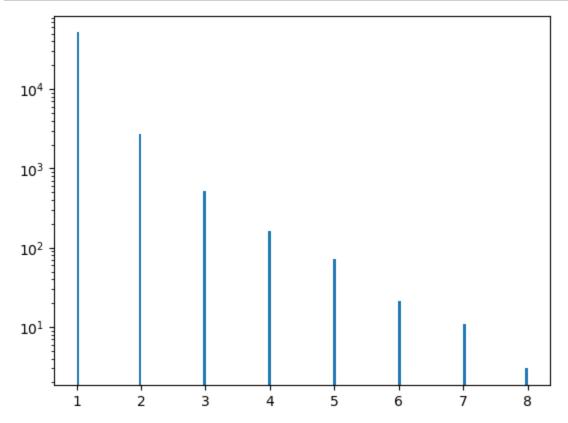
In [6]: 1 df = df.sample(30000,random_state=2)

```
In [7]:
          1 # missinig values
          2 df.isnull().sum()
Out[7]: id
                        0
        qid1
                        0
        qid2
                        0
        question1
        question2
        is_duplicate
        dtype: int64
In [8]:
          1 # duplicate raws
          2 df.duplicated().sum()
Out[8]: 0
In [9]:
          1 # Diistribution of duplicate and non-duplicate questions
          2 print(df['is_duplicate'].value_counts())
          3 print(df['is_duplicate'].value_counts()/df['is_duplicate'].count()*100)
          4 df['is_duplicate'].value_counts().plot(kind = 'bar')
             19013
        1
             10987
        Name: is_duplicate, dtype: int64
             63.376667
        1
             36.623333
        Name: is_duplicate, dtype: float64
```

Out[9]: <Axes: >



Number of unique questions 55299 Number of questions getting repeated 3480



In [13]:

1 df.head()

Out[13]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len
398782	398782	496695	532029	What is the best marketing automation tool for	What is the best marketing automation tool for	1	76	77
115086	115086	187729	187730	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh	0	49	57
327711	327711	454161	454162	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers	0	105	120
367788	367788	498109	491396	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts	0	59	146
151235	151235	237843	50930	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag	0	35	50

Out[14]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nu
398782	398782	496695	532029	What is the best marketing automation tool for	What is the best marketing automation tool for	1	76	77	
115086	115086	187729	187730	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh	0	49	57	
327711	327711	454161	454162	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers	0	105	120	
367788	367788	498109	491396	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts	0	59	146	
151235	151235	237843	50930	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag	0	35	50	
4									

In [15]:

```
def common_words(row):
    w1 = set(map(lambda word: word.lower().strip(), str(row['question1'])
    w2 = set(map(lambda word: word.lower().strip(), str(row['question2'])
    return len(w1 & w2)
```

Out[16]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nu
398782	398782	496695	532029	What is the best marketing automation tool for	What is the best marketing automation tool for	1	76	77	
115086	115086	187729	187730	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh	0	49	57	
327711	327711	454161	454162	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers	0	105	120	
367788	367788	498109	491396	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts	0	59	146	
151235	151235	237843	50930	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag	0	35	50	
4									•

```
In [17]:
```

```
def total_words(row):
    w1 = set(map(lambda word: word.lower().strip(), str(row['question1'])
    w2 = set(map(lambda word: word.lower().strip(), str(row['question2'])
    return (len(w1) + len(w2))
```

Out[18]:

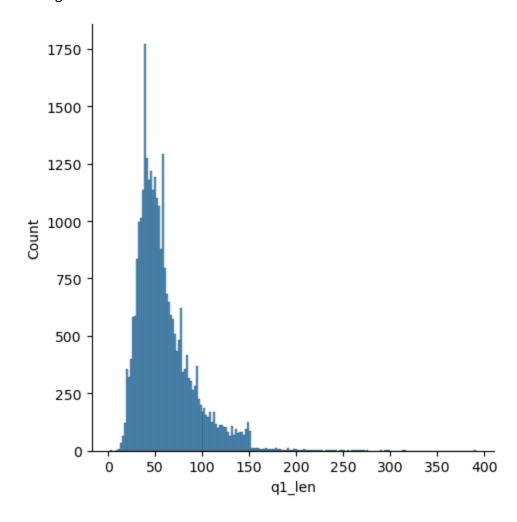
	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nu
398782	398782	496695	532029	What is the best marketing automation tool for	What is the best marketing automation tool for	1	76	77	
115086	115086	187729	187730	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh	0	49	57	
327711	327711	454161	454162	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers	0	105	120	
367788	367788	498109	491396	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts	0	59	146	
151235	151235	237843	50930	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag	0	35	50	
4									•

Out[19]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nu
398782	398782	496695	532029	What is the best marketing automation tool for	What is the best marketing automation tool for	1	76	77	
115086	115086	187729	187730	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh	0	49	57	
327711	327711	454161	454162	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers	0	105	120	
367788	367788	498109	491396	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts	0	59	146	
151235	151235	237843	50930	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag	0	35	50	
4									•

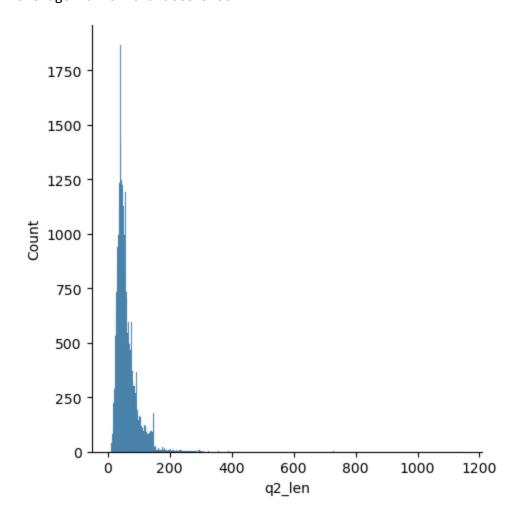
```
In [20]: 1 # Analysis of features
2 sns.displot(df['q1_len'])
3 print('minimum characters',df['q1_len'].min())
4 print('maximum characters',df['q1_len'].max())
5 print('average num of characters',int(df['q1_len'].mean()))
```

minimum characters 2 maximum characters 391 average num of characters 59

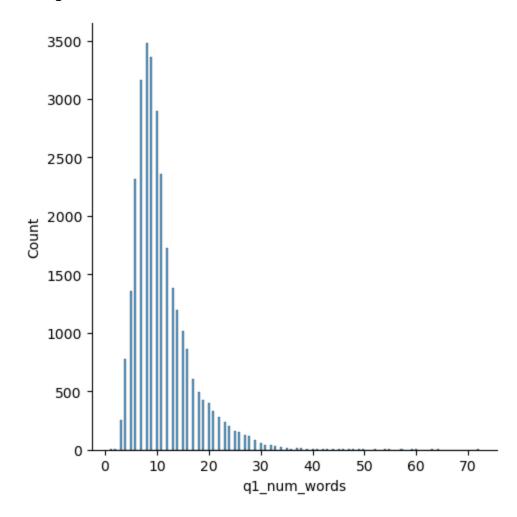


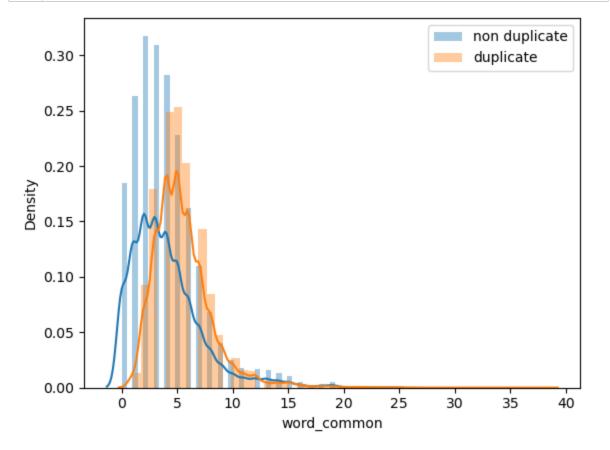
```
In [21]: 1 sns.displot(df['q2_len'])
2 print('minimum characters',df['q2_len'].min())
3 print('maximum characters',df['q2_len'].max())
4 print('average num of characters',int(df['q2_len'].mean()))
```

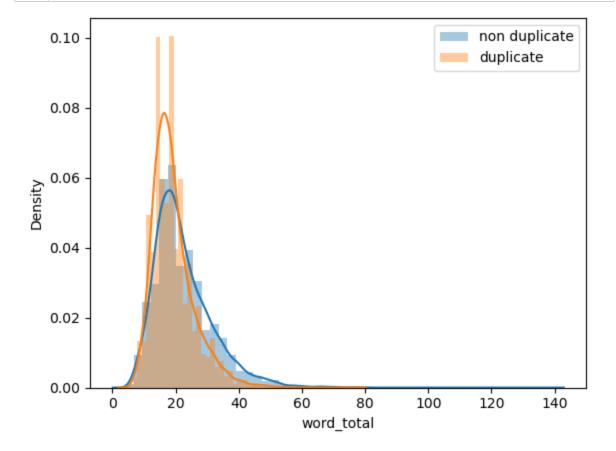
minimum characters 6 maximum characters 1151 average num of characters 60

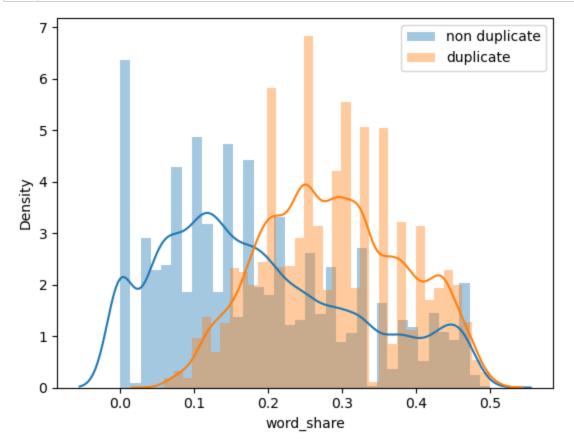


minimum words 1 maximum words 72 average num of words 10









```
In [26]: 1 ques_df = df[['question1','question2']]
2 ques_df.head()
```

Out[26]:

	question1	question2
398782	What is the best marketing automation tool for	What is the best marketing automation tool for
115086	I am poor but I want to invest. What should I do?	I am quite poor and I want to be very rich. Wh
327711	I am from India and live abroad. I met a guy f	T.I.E.T to Thapar University to Thapar Univers
367788	Why do so many people in the U.S. hate the sou	My boyfriend doesnt feel guilty when he hurts
151235	Consequences of Bhopal gas tragedy?	What was the reason behind the Bhopal gas trag

```
In [27]: 1 final_df = df.drop(columns=['id','qid1','qid2','question1','question2'])
2 print(final_df.shape)
3 final_df.head()
```

(30000, 8)

Out[27]:

	is_duplicate	q1_len	q2_len	q1_num_words	q2_num_words	word_common	word_total
398782	1	76	77	12	12	11	24
115086	0	49	57	12	15	7	23
327711	0	105	120	25	17	2	34
367788	0	59	146	12	30	0	32
151235	0	35	50	5	9	3	13

In [28]: 1 from sklearn.feature_extraction.text import CountVectorizer
2 # merge texts
3 questions = list(ques_df['question1']) + list(ques_df['question2'])
4 cv = CountVectorizer(max_features=3000)
6 q1_arr, q2_arr = np.vsplit(cv.fit_transform(questions).toarray(),2)

```
In [29]: 1 temp_df1 = pd.DataFrame(q1_arr, index= ques_df.index)
    temp_df2 = pd.DataFrame(q2_arr, index= ques_df.index)
    temp_df = pd.concat([temp_df1, temp_df2], axis=1)
    temp_df.shape
```

Out[29]: (30000, 6000)

```
In [30]: 1 final_df = pd.concat([final_df, temp_df], axis=1)
    print(final_df.shape)
    final_df.head()
```

(30000, 6008)

Out[30]:

	is_duplicate	q1_len	q2_len	q1_num_words	q2_num_words	word_common	word_total
398782	1	76	77	12	12	11	24
115086	0	49	57	12	15	7	23
327711	0	105	120	25	17	2	34
367788	0	59	146	12	30	0	32
151235	0	35	50	5	9	3	13
4							•

```
In [31]: 1     from sklearn.model_selection import train_test_split
2     X_train,X_test,y_train,y_test = train_test_split(final_df.iloc[:,1:].valu
3
```

RandomForestClassifier

```
In [33]: 1  from sklearn.ensemble import RandomForestClassifier
2  from sklearn.metrics import accuracy_score
3  rf = RandomForestClassifier()
4  rf.fit(X_train,y_train)
5  y_pred = rf.predict(X_test)
6  accuracy_score(y_test,y_pred)
```

Out[33]: 0.771666666666666

XGBClassifier

```
In [34]: 1  from xgboost import XGBClassifier
2  xgb = XGBClassifier()
3  xgb.fit(X_train,y_train)
4  y_pred = xgb.predict(X_test)
5  accuracy_score(y_test,y_pred)
```

Out[34]: 0.766166666666667

Advanced Features

```
In [35]: 1 import re
2 from bs4 import BeautifulSoup
```

```
In [36]:
               def preprocess(q):
            2
            3
                   q = str(q).lower().strip()
            4
            5
                   # Replace certain special characters with their string equivalents
                   q = q.replace('%', ' percent')
q = q.replace('$', ' dollar ')
q = q.replace('₹', ' rupee ')
q = q.replace('€', ' euro ')
            6
            7
            8
            9
           10
                   q = q.replace('@', ' at ')
           11
           12
                   # The pattern '[math]' appears around 900 times in the whole dataset.
           13
                   q = q.replace('[math]', '')
           14
           15
                   # Replacing some numbers with string equivalents (not perfect, can be
                   q = q.replace(',000,000,000 ', 'b ')
           16
           17
                   q = q.replace(',000,000 ', 'm ')
                   q = q.replace(',000 ', 'k ')
           18
                   q = re.sub(r'([0-9]+)000000000', r'\1b', q)
           19
           20
                   q = re.sub(r'([0-9]+)000000', r'\1m', q)
           21
                   q = re.sub(r'([0-9]+)000', r'\1k', q)
           22
           23
                   # Decontracting words
           24
                   # https://en.wikipedia.org/wiki/Wikipedia%3aList_of_English_contracti
           25
                   # https://stackoverflow.com/a/19794953
           26
                    contractions = {
           27
                    "ain't": "am not",
                    "aren't": "are not",
           28
                    "can't": "can not",
           29
           30
                   "can't've": "can not have",
                   "'cause": "because",
           31
           32
                   "could've": "could have",
                    "couldn't": "could not",
           33
                    "couldn't've": "could not have",
           34
           35
                   "didn't": "did not",
                   "doesn't": "does not",
           36
                   "don't": "do not",
           37
                    "hadn't": "had not"
           38
           39
                    "hadn't've": "had not have",
                   "hasn't": "has not",
           40
           41
                   "haven't": "have not",
           42
                   "he'd": "he would",
                    "he'd've": "he would have",
           43
                   "he'll": "he will",
           44
           45
                    "he'll've": "he will have",
                    "he's": "he is",
           46
           47
                    "how'd": "how did",
                    "how'd'y": "how do you",
           48
                   "how'll": "how will",
           49
                   "how's": "how is",
           50
                   "i'd": "i would",
           51
           52
                   "i'd've": "i would have",
                    "i'll": "i will",
           53
                   "i'll've": "i will have",
           54
                   "i'm": "i am",
           55
                   "i've": "i have",
           56
                    "isn't": "is not",
           57
```

```
58
         "it'd": "it would",
         "it'd've": "it would have",
 59
         "it'll": "it will",
 60
         "it'll've": "it will have",
61
         "it's": "it is",
 62
 63
         "let's": "let us",
 64
         "ma'am": "madam",
 65
         "mayn't": "may not",
         "might've": "might have",
 66
 67
         "mightn't": "might not",
         "mightn't've": "might not have",
 68
 69
         "must've": "must have",
         "mustn't": "must not",
70
         "mustn't've": "must not have",
71
         "needn't": "need not",
72
73
         "needn't've": "need not have",
74
         "o'clock": "of the clock",
75
         "oughtn't": "ought not",
         "oughtn't've": "ought not have",
76
         "shan't": "shall not",
77
         "sha'n't": "shall not",
78
79
         "shan't've": "shall not have",
         "she'd": "she would",
 80
81
         "she'd've": "she would have",
         "she'll": "she will",
 82
         "she'll've": "she will have",
 83
         "she's": "she is",
 84
85
         "should've": "should have",
         "shouldn't": "should not",
 86
 87
         "shouldn't've": "should not have",
         "so've": "so have",
 88
         "so's": "so as",
 89
90
         "that'd": "that would",
91
         "that'd've": "that would have",
92
         "that's": "that is",
         "there'd": "there would",
93
 94
         "there'd've": "there would have",
         "there's": "there is",
95
         "they'd": "they would",
96
         "they'd've": "they would have",
97
98
         "they'll": "they will",
99
         "they'll've": "they will have",
         "they're": "they are",
100
         "they've": "they have",
101
         "to've": "to have",
102
         "wasn't": "was not",
103
         "we'd": "we would",
104
105
         "we'd've": "we would have",
         "we'll": "we will",
106
         "we'll've": "we will have",
107
         "we're": "we are",
108
         "we've": "we have",
109
110
         "weren't": "were not"
         "what'll": "what will",
111
         "what'll've": "what will have",
112
         "what're": "what are",
113
         "what's": "what is",
114
```

```
"what've": "what have",
115
         "when's": "when is",
116
117
         "when've": "when have",
         "where'd": "where did",
118
         "where's": "where is",
119
120
         "where've": "where have",
         "who'll": "who will",
121
122
         "who'll've": "who will have",
         "who's": "who is",
123
         "who've": "who have",
124
         "why's": "why is",
125
         "why've": "why have",
126
127
         "will've": "will have",
         "won't": "will not",
128
         "won't've": "will not have",
129
130
         "would've": "would have",
         "wouldn't": "would not",
131
132
         "wouldn't've": "would not have",
         "y'all": "you all",
133
         "y'all'd": "you all would",
134
         "y'all'd've": "you all would have",
135
136
         "y'all're": "you all are",
137
         "y'all've": "you all have",
         "you'd": "you would",
138
         "you'd've": "you would have",
139
         "you'll": "you will",
140
         "you'll've": "you will have",
141
142
         "you're": "you are",
         "you've": "you have"
143
144
         }
145
146
         q_decontracted = []
147
148
         for word in q.split():
149
             if word in contractions:
150
                 word = contractions[word]
151
152
             q_decontracted.append(word)
153
         q = ' '.join(q_decontracted)
154
         q = q.replace("'ve", " have")
q = q.replace("n't", " not")
155
156
         q = q.replace("'re", " are")
157
         q = q.replace("'ll", " will")
158
159
160
         # Removing HTML tags
         q = BeautifulSoup(q)
161
162
         q = q.get_text()
163
164
         # Remove punctuations
165
         pattern = re.compile('\W')
166
         q = re.sub(pattern, ' ', q).strip()
167
168
169
         return q
```

```
preprocess("I've already! wasn't <b>done</b>?")
In [37]:
Out[37]:
           'i have already was not done'
                 df['question1'] = df['question1'].apply(preprocess)
In [38]:
                df['question2'] = df['question2'].apply(preprocess)
In [39]:
                df.head()
Out[39]:
                         id
                                qid1
                                        qid2
                                                 question1
                                                            question2 is_duplicate q1_len q2_len q1_nur
                                                 what is the
                                                            what is the
                                                      best
                                                                  best
            398782 398782 496695
                                     532029
                                                  marketing
                                                             marketing
                                                                                  1
                                                                                        76
                                                                                                77
                                                 automation
                                                            automation
                                                   tool for...
                                                              tool for...
                                                             i am quite
                                              i am poor but i
                                                             poor and i
                                              want to invest
             115086 115086 187729 187730
                                                             want to be
                                                                                 0
                                                                                        49
                                                                                                57
                                               what should i
                                                              very rich
                                                        do
                                                                 wh...
                                                               t i e t to
                                                  i am from
                                                                thapar
                                               india and live
             327711 327711 454161 454162
                                                                                 0
                                                                                       105
                                                                                               120
                                                             university
                                              abroad i met a
                                                              to thapar
                                                    guy f...
                                                              univers...
                                                                   my
                                                 why do so
                                                              boyfriend
                                               many people
                                                            doesnt feel
            367788 367788 498109 491396
                                                                                 0
                                                                                        59
                                                                                                146
                                              in the u s hate
                                                                 guilty
                                                  the sou...
                                                              when he
                                                               hurts ...
                                                             what was
                                              consequences
                                                            the reason
            151235 151235 237843
                                      50930
                                                                                 0
                                                                                        35
                                                                                                 50
                                              of bhopal gas
                                                             behind the
                                                    tragedy
                                                            bhopal gas
                                                               tragedy
In [40]:
                 df['q1_len'] = df['question1'].str.len()
                df['q2_len'] = df['question2'].str.len()
```

```
In [41]:
             df['q1_num_words'] = df['question1'].apply(lambda row: len(row.split(" ")
           2 df['q2_num_words'] = df['question2'].apply(lambda row: len(row.split(" ")
           3 df.head()
```

Out[41]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

In [42]:

```
1
  def common_words(row):
2
      w1 = set(map(lambda word: word.lower().strip(), row['question1'].spli
3
      w2 = set(map(lambda word: word.lower().strip(), row['question2'].spli
       return len(w1 & w2)
4
```

Out[43]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur	
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76		
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56		
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119		
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145		
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49		
4									•	
<pre>1 def total words(row):</pre>										

In [44]:

```
def total_words(row):
    w1 = set(map(lambda word: word.lower().strip(), row['question1'].spli
    w2 = set(map(lambda word: word.lower().strip(), row['question2'].spli
    return (len(w1) + len(w2))
```

Out[45]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

Out[46]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

```
In [47]:
              # Advanced Features
           1
              from nltk.corpus import stopwords
           2
           3
           4
              def fetch_token_features(row):
           5
           6
                  q1 = row['question1']
           7
                  q2 = row['question2']
           8
           9
                  SAFE DIV = 0.0001
          10
          11
                  STOP WORDS = stopwords.words("english")
          12
          13
                  token_features = [0.0]*8
          14
          15
                  # Converting the Sentence into Tokens:
          16
                  q1_tokens = q1.split()
          17
                  q2\_tokens = q2.split()
          18
          19
                  if len(q1_tokens) == 0 or len(q2_tokens) == 0:
          20
                      return token_features
          21
          22
                  # Get the non-stopwords in Questions
          23
                  q1_words = set([word for word in q1_tokens if word not in STOP_WORDS]
          24
                  q2 words = set([word for word in q2 tokens if word not in STOP WORDS]
          25
          26
                  #Get the stopwords in Questions
          27
                  q1_stops = set([word for word in q1_tokens if word in STOP_WORDS])
          28
                  q2_stops = set([word for word in q2_tokens if word in STOP_WORDS])
          29
          30
                  # Get the common non-stopwords from Question pair
          31
                  common_word_count = len(q1_words.intersection(q2_words))
          32
          33
                  # Get the common stopwords from Question pair
          34
                  common_stop_count = len(q1_stops.intersection(q2_stops))
          35
          36
                  # Get the common Tokens from Question pair
          37
                  common_token_count = len(set(q1_tokens).intersection(set(q2_tokens)))
          38
          39
          40
                  token_features[0] = common_word_count / (min(len(q1_words), len(q2_words), len(q2_words))
                  token_features[1] = common_word_count / (max(len(q1_words), len(q2_wo
          41
          42
                  token_features[2] = common_stop_count / (min(len(q1_stops), len(q2_st
          43
                  token_features[3] = common_stop_count / (max(len(q1_stops), len(q2_st
          44
                  token_features[4] = common_token_count / (min(len(q1_tokens), len(q2_
          45
                  token_features[5] = common_token_count / (max(len(q1_tokens), len(q2_
          46
          47
                  # Last word of both question is same or not
          48
                  token_features[6] = int(q1_tokens[-1] == q2_tokens[-1])
          49
                  # First word of both question is same or not
          50
          51
                  token_features[7] = int(q1_tokens[0] == q2_tokens[0])
          52
          53
                  return token features
```

```
In [48]:
             token_features = df.apply(fetch_token_features, axis=1)
           1
           2
           3
             df["cwc_min"]
                                  = list(map(lambda x: x[0], token_features))
           4
             df["cwc_max"]
                                  = list(map(lambda x: x[1], token_features))
                                  = list(map(lambda x: x[2], token_features))
             df["csc_min"]
           5
                                  = list(map(lambda x: x[3], token_features))
             df["csc_max"]
                                  = list(map(lambda x: x[4], token_features))
           7
             df["ctc_min"]
                                  = list(map(lambda x: x[5], token_features))
           8
             df["ctc_max"]
             df["last_word_eq"] = list(map(lambda x: x[6], token_features))
          10 | df["first_word_eq"] = list(map(lambda x: x[7], token_features))
In [49]:
             df.head()
```

Out[49]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

```
In [50]:
           1
              import distance
           2
           3
              def fetch_length_features(row):
           4
           5
                  q1 = row['question1']
                  q2 = row['question2']
           6
           7
                  length features = [0.0]*3
           8
           9
          10
                  # Converting the Sentence into Tokens:
          11
                  q1 tokens = q1.split()
          12
                  q2\_tokens = q2.split()
          13
          14
                  if len(q1_tokens) == 0 or len(q2_tokens) == 0:
          15
                      return length_features
          16
          17
                  # Absolute length features
          18
                  length_features[0] = abs(len(q1_tokens) - len(q2_tokens))
          19
          20
                  #Average Token Length of both Questions
          21
                  length_features[1] = (len(q1_tokens) + len(q2_tokens))/2
          22
          23
                  strs = list(distance.lcsubstrings(q1, q2))
          24
                  length_features[2] = len(strs[0]) / (min(len(q1), len(q2)) + 1)
          25
          26
                  return length_features
```

In [52]: 1 df.head()

Out[52]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

```
In [53]:
              # Fuzzy Features
           1
           2
              from fuzzywuzzy import fuzz
           3
           4
              def fetch_fuzzy_features(row):
           5
                  q1 = row['question1']
           6
           7
                  q2 = row['question2']
           8
           9
                  fuzzy_features = [0.0]*4
          10
                  # fuzz ratio
          11
          12
                  fuzzy_features[0] = fuzz.QRatio(q1, q2)
          13
                  # fuzz_partial_ratio
          14
          15
                  fuzzy_features[1] = fuzz.partial_ratio(q1, q2)
          16
          17
                  # token_sort_ratio
                  fuzzy_features[2] = fuzz.token_sort_ratio(q1, q2)
          18
          19
          20
                  # token_set_ratio
          21
                  fuzzy_features[3] = fuzz.token_set_ratio(q1, q2)
          22
          23
                  return fuzzy_features
```

In [55]: 1 print(df.shape)
2 df.head()

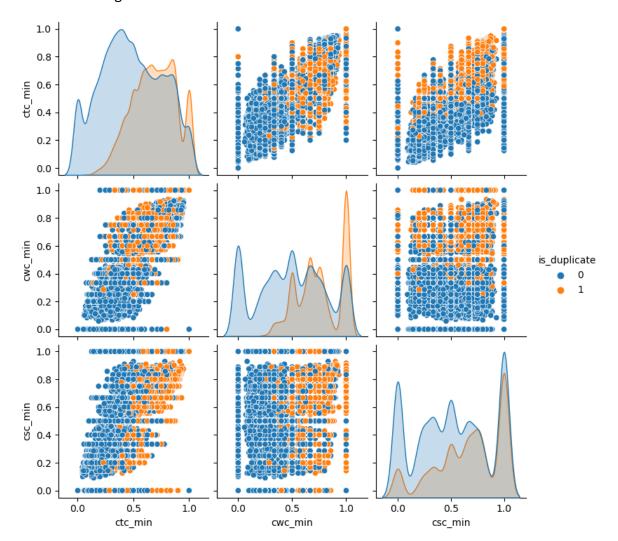
(30000, 28)

Out[55]:

	id	qid1	qid2	question1	question2	is_duplicate	q1_len	q2_len	q1_nur
398782	398782	496695	532029	what is the best marketing automation tool for	what is the best marketing automation tool for	1	75	76	
115086	115086	187729	187730	i am poor but i want to invest what should i do	i am quite poor and i want to be very rich wh	0	48	56	
327711	327711	454161	454162	i am from india and live abroad i met a guy f	t i e t to thapar university to thapar univers	0	104	119	
367788	367788	498109	491396	why do so many people in the u s hate the sou	my boyfriend doesnt feel guilty when he hurts	0	58	145	
151235	151235	237843	50930	consequences of bhopal gas tragedy	what was the reason behind the bhopal gas tragedy	0	34	49	
4									•

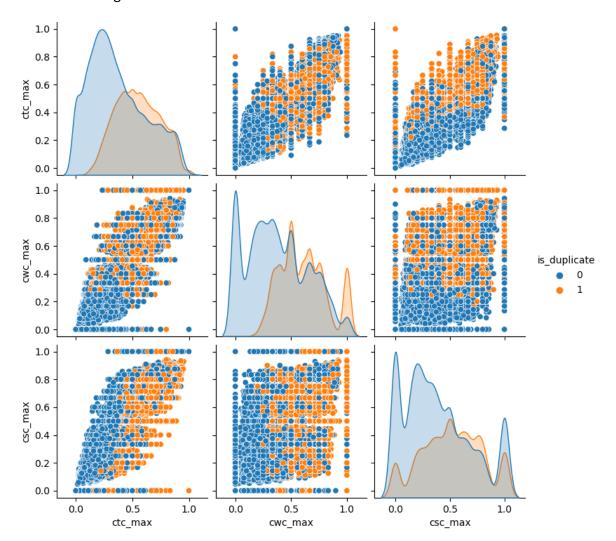
In [56]: 1 sns.pairplot(df[['ctc_min', 'cwc_min', 'csc_min', 'is_duplicate']],hue='i

Out[56]: <seaborn.axisgrid.PairGrid at 0x2fb06ddbb10>



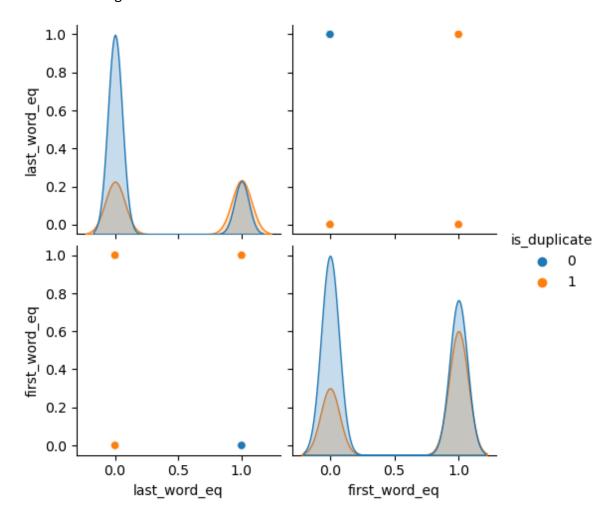
In [57]: 1 sns.pairplot(df[['ctc_max', 'cwc_max', 'csc_max', 'is_duplicate']],hue='i

Out[57]: <seaborn.axisgrid.PairGrid at 0x2fc584de490>



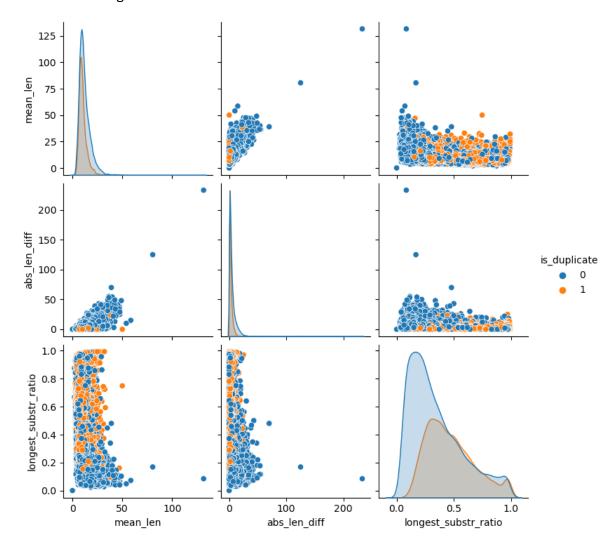
In [58]: 1 sns.pairplot(df[['last_word_eq', 'first_word_eq', 'is_duplicate']],hue='i

Out[58]: <seaborn.axisgrid.PairGrid at 0x2fb06e1a0d0>



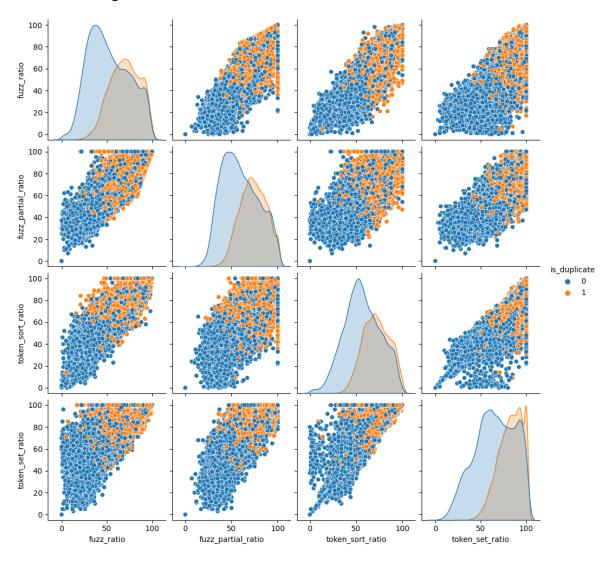
In [59]: 1 sns.pairplot(df[['mean_len', 'abs_len_diff','longest_substr_ratio', 'is_d

Out[59]: <seaborn.axisgrid.PairGrid at 0x2fb06606490>



```
In [60]: 1 sns.pairplot(df[['fuzz_ratio', 'fuzz_partial_ratio','token_sort_ratio','t
```

Out[60]: <seaborn.axisgrid.PairGrid at 0x2fc5871bd10>

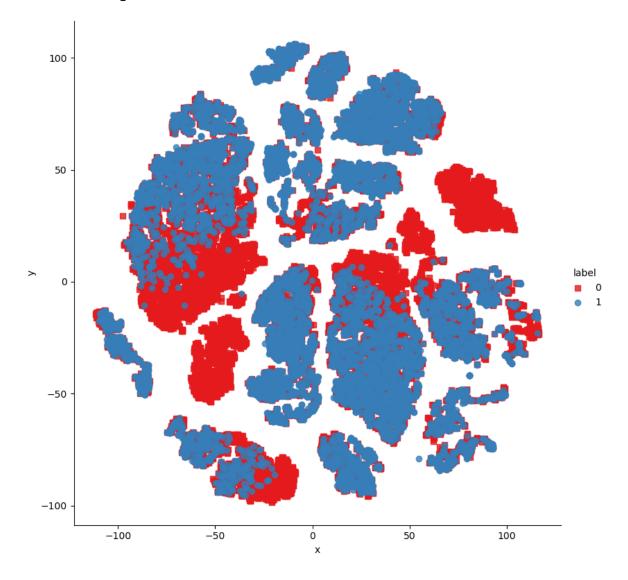


```
In [62]:
              from sklearn.manifold import TSNE
           2
           3
              tsne2d = TSNE(
           4
                  n_components=2,
                  init='random', # pca
           5
           6
                  random_state=101,
           7
                  method='barnes_hut',
           8
                  n_iter=1000,
           9
                  verbose=2,
          10
                  angle=0.5
             ).fit_transform(X)
```

```
[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 30000 samples in 0.068s...
[t-SNE] Computed neighbors for 30000 samples in 5.413s...
[t-SNE] Computed conditional probabilities for sample 1000 / 30000
[t-SNE] Computed conditional probabilities for sample 2000 / 30000
[t-SNE] Computed conditional probabilities for sample 3000 / 30000
[t-SNE] Computed conditional probabilities for sample 4000 / 30000
[t-SNE] Computed conditional probabilities for sample 5000 / 30000
[t-SNE] Computed conditional probabilities for sample 6000 / 30000
[t-SNE] Computed conditional probabilities for sample 7000 / 30000
[t-SNE] Computed conditional probabilities for sample 8000 / 30000
[t-SNE] Computed conditional probabilities for sample 9000 / 30000
[t-SNE] Computed conditional probabilities for sample 10000 / 30000
[t-SNE] Computed conditional probabilities for sample 11000 / 30000
[t-SNE] Computed conditional probabilities for sample 12000 / 30000
[t-SNE] Computed conditional probabilities for sample 13000 / 30000
[t-SNE] Computed conditional probabilities for sample 14000 / 30000
[t-SNE] Computed conditional probabilities for sample 15000 / 30000
[t-SNE] Computed conditional probabilities for sample 16000 / 30000
[t-SNE] Computed conditional probabilities for sample 17000 / 30000
[t-SNE] Computed conditional probabilities for sample 18000 / 30000
[t-SNE] Computed conditional probabilities for sample 19000 / 30000
[t-SNE] Computed conditional probabilities for sample 20000 / 30000
[t-SNE] Computed conditional probabilities for sample 21000 / 30000
[t-SNE] Computed conditional probabilities for sample 22000 / 30000
[t-SNE] Computed conditional probabilities for sample 23000 / 30000
[t-SNE] Computed conditional probabilities for sample 24000 / 30000
[t-SNE] Computed conditional probabilities for sample 25000 / 30000
[t-SNE] Computed conditional probabilities for sample 26000 / 30000
[t-SNE] Computed conditional probabilities for sample 27000 / 30000
[t-SNE] Computed conditional probabilities for sample 28000 / 30000
[t-SNE] Computed conditional probabilities for sample 29000 / 30000
[t-SNE] Computed conditional probabilities for sample 30000 / 30000
[t-SNE] Mean sigma: 0.080413
[t-SNE] Computed conditional probabilities in 0.941s
[t-SNE] Iteration 50: error = 110.0354767, gradient norm = 0.0222168 (50 ite
rations in 15.025s)
[t-SNE] Iteration 100: error = 89.8034515, gradient norm = 0.0077262 (50 ite
rations in 11.897s)
[t-SNE] Iteration 150: error = 85.8413696, gradient norm = 0.0047961 (50 ite
rations in 12.044s)
[t-SNE] Iteration 200: error = 83.9956284, gradient norm = 0.0039413 (50 ite
rations in 10.530s)
[t-SNE] Iteration 250: error = 82.8466949, gradient norm = 0.0029376 (50 ite
rations in 13.284s)
[t-SNE] KL divergence after 250 iterations with early exaggeration: 82.84669
[t-SNE] Iteration 300: error = 3.3983612, gradient norm = 0.0073742 (50 iter
ations in 9.378s)
[t-SNE] Iteration 350: error = 2.7810087, gradient norm = 0.0073581 (50 iter
ations in 9.022s)
[t-SNE] Iteration 400: error = 2.4468734, gradient norm = 0.0068702 (50 iter
ations in 7.910s)
[t-SNE] Iteration 450: error = 2.2388027, gradient norm = 0.0064638 (50 iter
ations in 7.764s)
[t-SNE] Iteration 500: error = 2.0952334, gradient norm = 0.0061122 (50 iter
ations in 7.882s)
```

```
[t-SNE] Iteration 550: error = 1.9906163, gradient norm = 0.0057656 (50 iter
ations in 9.103s)
[t-SNE] Iteration 600: error = 1.9118668, gradient norm = 0.0053742 (50 iter
ations in 8.223s)
[t-SNE] Iteration 650: error = 1.8510072, gradient norm = 0.0050041 (50 iter
ations in 7.879s)
[t-SNE] Iteration 700: error = 1.8027258, gradient norm = 0.0046457 (50 iter
ations in 7.946s)
[t-SNE] Iteration 750: error = 1.7636205, gradient norm = 0.0043452 (50 iter
ations in 8.119s)
[t-SNE] Iteration 800: error = 1.7313575, gradient norm = 0.0040350 (50 iter
ations in 8.592s)
[t-SNE] Iteration 850: error = 1.7042049, gradient norm = 0.0037814 (50 iter
ations in 8.236s)
[t-SNE] Iteration 900: error = 1.6811382, gradient norm = 0.0035771 (50 iter
ations in 8.122s)
[t-SNE] Iteration 950: error = 1.6614714, gradient norm = 0.0033116 (50 iter
ations in 8.398s)
[t-SNE] Iteration 1000: error = 1.6444725, gradient norm = 0.0031130 (50 ite
rations in 10.454s)
[t-SNE] KL divergence after 1000 iterations: 1.644472
```

Out[63]: <seaborn.axisgrid.FacetGrid at 0x2fc5cc3a810>



```
In [64]:
           1
              tsne3d = TSNE(
           2
                  n_components=3,
                  init='random', # pca
           3
           4
                  random_state=101,
           5
                  method='barnes_hut',
                  n_iter=1000,
           6
           7
                  verbose=2,
           8
                  angle=0.5
              ).fit_transform(X)
           9
```

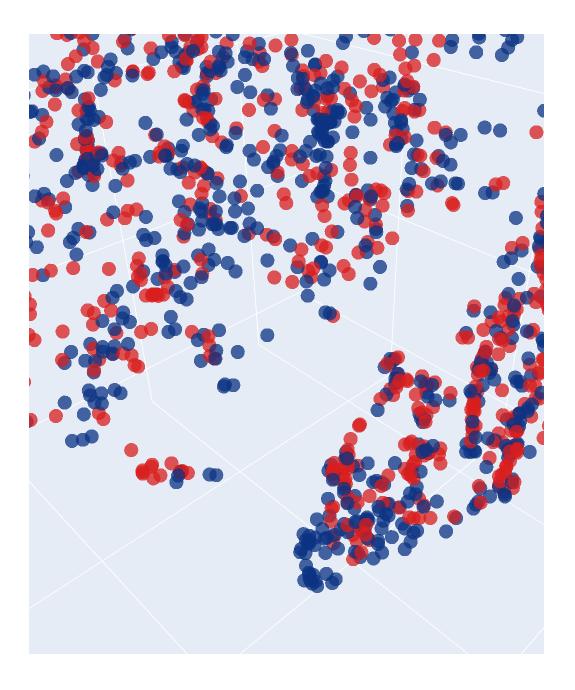
```
[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 30000 samples in 0.071s...
[t-SNE] Computed neighbors for 30000 samples in 4.859s...
[t-SNE] Computed conditional probabilities for sample 1000 / 30000
[t-SNE] Computed conditional probabilities for sample 2000 / 30000
[t-SNE] Computed conditional probabilities for sample 3000 / 30000
[t-SNE] Computed conditional probabilities for sample 4000 / 30000
[t-SNE] Computed conditional probabilities for sample 5000 / 30000
[t-SNE] Computed conditional probabilities for sample 6000 / 30000
[t-SNE] Computed conditional probabilities for sample 7000 / 30000
[t-SNE] Computed conditional probabilities for sample 8000 / 30000
[t-SNE] Computed conditional probabilities for sample 9000 / 30000
[t-SNE] Computed conditional probabilities for sample 10000 / 30000
[t-SNE] Computed conditional probabilities for sample 11000 / 30000
[t-SNE] Computed conditional probabilities for sample 12000 / 30000
[t-SNE] Computed conditional probabilities for sample 13000 / 30000
[t-SNE] Computed conditional probabilities for sample 14000 / 30000
[t-SNE] Computed conditional probabilities for sample 15000 / 30000
[t-SNE] Computed conditional probabilities for sample 16000 / 30000
[t-SNE] Computed conditional probabilities for sample 17000 / 30000
[t-SNE] Computed conditional probabilities for sample 18000 / 30000
[t-SNE] Computed conditional probabilities for sample 19000 / 30000
[t-SNE] Computed conditional probabilities for sample 20000 / 30000
[t-SNE] Computed conditional probabilities for sample 21000 / 30000
[t-SNE] Computed conditional probabilities for sample 22000 / 30000
[t-SNE] Computed conditional probabilities for sample 23000 / 30000
[t-SNE] Computed conditional probabilities for sample 24000 / 30000
[t-SNE] Computed conditional probabilities for sample 25000 / 30000
[t-SNE] Computed conditional probabilities for sample 26000 / 30000
[t-SNE] Computed conditional probabilities for sample 27000 / 30000
[t-SNE] Computed conditional probabilities for sample 28000 / 30000
[t-SNE] Computed conditional probabilities for sample 29000 / 30000
[t-SNE] Computed conditional probabilities for sample 30000 / 30000
[t-SNE] Mean sigma: 0.080413
[t-SNE] Computed conditional probabilities in 0.858s
[t-SNE] Iteration 50: error = 110.5925522, gradient norm = 0.0148754 (50 ite
rations in 33.847s)
[t-SNE] Iteration 100: error = 86.6675262, gradient norm = 0.0046031 (50 ite
rations in 24.531s)
[t-SNE] Iteration 150: error = 83.4906464, gradient norm = 0.0022670 (50 ite
rations in 24.854s)
[t-SNE] Iteration 200: error = 82.2365875, gradient norm = 0.0015197 (50 ite
rations in 25.397s)
[t-SNE] Iteration 250: error = 81.5207214, gradient norm = 0.0011329 (50 ite
rations in 20.254s)
[t-SNE] KL divergence after 250 iterations with early exaggeration: 81.52072
[t-SNE] Iteration 300: error = 2.9802334, gradient norm = 0.0038841 (50 iter
ations in 27.950s)
[t-SNE] Iteration 350: error = 2.3667295, gradient norm = 0.0030882 (50 iter
ations in 29.651s)
[t-SNE] Iteration 400: error = 2.0555992, gradient norm = 0.0024692 (50 iter
ations in 31.315s)
[t-SNE] Iteration 450: error = 1.8723609, gradient norm = 0.0020611 (50 iter
ations in 29.924s)
[t-SNE] Iteration 500: error = 1.7512760, gradient norm = 0.0017748 (50 iter
ations in 30.053s)
```

[t-SNE] Iteration 550: error = 1.6657951, gradient norm = 0.0015336 (50 iter ations in 27.281s) [t-SNE] Iteration 600: error = 1.6035148, gradient norm = 0.0013240 (50 iter ations in 28.066s) [t-SNE] Iteration 650: error = 1.5577862, gradient norm = 0.0011421 (50 iter ations in 30.451s) [t-SNE] Iteration 700: error = 1.5237987, gradient norm = 0.0009500 (50 iter ations in 28.701s) [t-SNE] Iteration 750: error = 1.4991173, gradient norm = 0.0007596 (50 iter ations in 30.606s) [t-SNE] Iteration 800: error = 1.4807075, gradient norm = 0.0006571 (50 iter ations in 33.086s) [t-SNE] Iteration 850: error = 1.4661463, gradient norm = 0.0005565 (50 iter ations in 36.494s) [t-SNE] Iteration 900: error = 1.4544582, gradient norm = 0.0004825 (50 iter ations in 38.770s) [t-SNE] Iteration 950: error = 1.4449804, gradient norm = 0.0004071 (50 iter ations in 32.307s) [t-SNE] Iteration 1000: error = 1.4379606, gradient norm = 0.0003029 (50 ite rations in 31.926s) [t-SNE] KL divergence after 1000 iterations: 1.437961

localhost:8888/notebooks/ES_PML_2303_Majedur Rahman/Project_2_Quara Question Pairs Similarity.ipynb#

```
In [65]:
              import plotly.graph_objs as go
              import plotly.tools as tls
             import plotly.offline as py
           3
             py.init_notebook_mode(connected=True)
             trace1 = go.Scatter3d(
           6
           7
                  x=tsne3d[:,0],
           8
                  y=tsne3d[:,1],
           9
                  z=tsne3d[:,2],
          10
                  mode='markers',
                  marker=dict(
          11
                      sizemode='diameter',
          12
          13
                      color = y,
                      colorscale = 'Portland',
          14
                      colorbar = dict(title = 'duplicate'),
          15
          16
                      line=dict(color='rgb(255, 255, 255)'),
          17
                      opacity=0.75
          18
                  )
          19
              )
          20
          21 data=[trace1]
          22 layout=dict(height=800, width=800, title='3d embedding with engineered fe
          23 fig=dict(data=data, layout=layout)
          24 py.iplot(fig, filename='3DBubble')
```

3d embedding with engineered features



Out[66]:

question2	question1							
what is the best marketing automation tool for	what is the best marketing automation tool for	398782						
i am quite poor and i want to be very rich wh	i am poor but i want to invest what should i do	115086						
t i e t to thapar university to thapar univers	i am from india and live abroad i met a guy f	327711						
my boyfriend doesnt feel guilty when he hurts	why do so many people in the u s hate the sou	367788						
what was the reason behind the bhopal gas tragedy	consequences of bhopal gas tragedy	151235						
d1','qid2','question1','question2'])	ual df = df dron(columns=['id'.'aid	1 fin						
az , ques cioniz , quescioniz]/		1)						

```
In [67]: 1 final_df = df.drop(columns=['id','qid1','qid2','question1','question2'])
2 print(final_df.shape)
3 final_df.head()
```

(30000, 23)

Out[67]:

	is_duplicate	q1_len	q2_len	q1_num_words	q2_num_words	word_common	word_total
398782	1	75	76	13	13	12	26
115086	0	48	56	13	16	8	24
327711	0	104	119	28	21	4	38
367788	0	58	145	14	32	1	34
151235	0	34	49	5	9	3	13

```
In [68]: 1 from sklearn.feature_extraction.text import CountVectorizer
```

2 # merge texts
3 questions = list(ques_df['question1']) + list(ques_df['question2'])

5 cv = CountVectorizer(max_features=3000)

q1_arr, q2_arr = np.vsplit(cv.fit_transform(questions).toarray(),2)

```
In [69]: 1 temp_df1 = pd.DataFrame(q1_arr, index= ques_df.index)
```

2 temp_df2 = pd.DataFrame(q2_arr, index= ques_df.index)

3 temp_df = pd.concat([temp_df1, temp_df2], axis=1)

4 temp_df.shape

Out[69]: (30000, 6000)

(30000, 6023)

Out[70]:

	is_duplicate	q1_len	q2_len	q1_num_words	q2_num_words	word_common	word_total
398782	1	75	76	13	13	12	26
115086	0	48	56	13	16	8	24
327711	0	104	119	28	21	4	38
367788	0	58	145	14	32	1	34
151235	0	34	49	5	9	3	13
4							•

In	[7	1]	:

```
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(final_df.iloc[:,1:].valu
```

KNN

```
In [72]: 1  from sklearn.neighbors import KNeighborsClassifier
2  from sklearn.metrics import accuracy_score
3  knn_classifier = KNeighborsClassifier(n_neighbors=3)
4  knn_classifier.fit(X_train, y_train)
5  KNeighborsClassifier(n_neighbors=3)
9  y_pred = knn_classifier.predict(X_test)
7  accuracy = accuracy_score(y_test, y_pred)
8  accuracy
```

Out[72]: 0.695666666666667

Decision Tree

```
In [73]: 1  from sklearn.tree import DecisionTreeClassifier
2  clf = DecisionTreeClassifier()
3  clf.fit(X_train, y_train)
4  DecisionTreeClassifier()
5  y_pred = clf.predict(X_test)
6  accuracy = metrics.accuracy_score(y_test, y_pred)
7  accuracy
```

Out[73]: 0.73133333333333333

```
In [ ]: 1
```

Random Forest

```
In [74]: 1  from sklearn.ensemble import RandomForestClassifier
2  from sklearn.metrics import accuracy_score
3  rf = RandomForestClassifier()
4  rf.fit(X_train,y_train)
5  y_pred = rf.predict(X_test)
6  accuracy_score(y_test,y_pred)
```

Out[74]: 0.7838333333333334

XGBoost (Extreme Gradient Boosting)

```
In [75]: 1 from xgboost import XGBClassifier
2 xgb = XGBClassifier()
3 xgb.fit(X_train,y_train)
4 y_pred1 = xgb.predict(X_test)
5 accuracy_score(y_test,y_pred1)
```

Out[75]: 0.794666666666666

Boosting Algorithm

```
In [79]: 1  from sklearn.ensemble import AdaBoostClassifier
2  base_classifier = DecisionTreeClassifier(max_depth=3)
3  adaboost_classifier = AdaBoostClassifier(base_classifier,
4  n_estimators=50, random_state=42)
5  adaboost_classifier.fit(X_train, y_train)
6  AdaBoostClassifier(estimator=DecisionTreeClassifier(max_depth=1),
7  random_state=42)
8  y_pred = adaboost_classifier.predict(X_test)
9  accuracy = metrics.accuracy_score(y_test, y_pred)
10  accuracy
```

Out[79]: 0.7725

Logistic Regression

```
In [80]:
           1 from sklearn import linear_model
           2 lrg = linear_model.LogisticRegression()
           3 lrg.fit(X_train, y_train)
           4 LogisticRegression()
           5 lrg.score(X_test, y_test)
Out[80]: 0.715166666666666
In [84]:
             from sklearn.metrics import confusion_matrix
           1 # for random forest model
In [85]:
           2 confusion_matrix(y_test,y_pred)
Out[85]: array([[3191, 621],
                [ 744, 1444]], dtype=int64)
In [86]:
           1 # for xgboost model
           2 confusion_matrix(y_test,y_pred1)
Out[86]: array([[3237, 575],
                [ 657, 1531]], dtype=int64)
```