

# **Quora Question Pairs**

# 1. Business Problem

# 1.1 Description

Quora is a place to gain and share knowledge—about anything. It's a platform to ask questions and connect with people who contribute unique insights and quality answers. This empowers people to learn from each other and to better understand the world.

Over 100 million people visit Quora every month, so it's no surprise that many people ask similarly worded questions. Multiple questions with the same intent can cause seekers to spend more time finding the best answer to their question, and make writers feel they need to answer multiple versions of the same question. Quora values canonical questions because they provide a better experience to active seekers and writers, and offer more value to both of these groups in the long term.

Credits: Kaggle

#### **Problem Statement**

- Identify which questions asked on Quora are duplicates of questions that have already been asked.
- This could be useful to instantly provide answers to questions that have already been answered.
- We are tasked with predicting whether a pair of questions are duplicates or not.

### 1.2 Sources/Useful Links

• Source : <a href="https://www.kaggle.com/c/quora-question-pairs">https://www.kaggle.com/c/quora-question-pairs</a>

#### **Useful Links**

- Discussions : <a href="https://www.kaggle.com/anokas/data-analysis-xgboost-starter-0-35460-lb/comments">https://www.kaggle.com/anokas/data-analysis-xgboost-starter-0-35460-lb/comments</a>
- Blog 1: <a href="https://engineering.quora.com/Semantic-Question-Matching-with-Deep-Learning">https://engineering.quora.com/Semantic-Question-Matching-with-Deep-Learning</a>
- Blog 2: <a href="https://towardsdatascience.com/identifying-duplicate-questions-on-quora-top-12-on-kaggle-4c1cf93f1c30">https://towardsdatascience.com/identifying-duplicate-questions-on-quora-top-12-on-kaggle-4c1cf93f1c30</a>

# 1.3 Real world/Business Objectives and Constraints

- 1. The cost of a mis-classification can be very high.
- 2. You would want a probability of a pair of questions to be duplicates so that you can choose any threshold of choice.
- 3. No strict latency concerns.
- 4. Interpretability is partially important.

# 2. Machine Learning Probelm

# 2.1 Data

### 2.1.1 Data Overview

- Data will be in a file Train.csv
- Train.csv contains 5 columns : qid1, qid2, question1, question2, is duplicate
- Size of Train.csv 60MB

### 2.1.2 Example Data point

"id","qid1","qid2","question1","question2","is\_duplica
te"

"0","1","2","What is the step by step guide to invest
in share market in india?","What is the step by step g
uide to invest in share market?","0"

"1","3","4","What is the story of Kohinoor (Koh-i-Noo
r) Diamond?","What would happen if the Indian governme
nt stole the Kohinoor (Koh-i-Noor) diamond back?","0"

"7","15","16","How can I be a good geologist?","What s
hould I do to be a great geologist?","1"

"11","23","24","How do I read and find my YouTube comm
ents?","How can I see all my Youtube comments?","1"

# 2.2 Mapping the real world problem to an ML problem

## 2.2.1 Type of Machine Leaning Problem

It is a binary classification problem, for a given pair of questions we need to predict if they are duplicate or not.

#### 2.2.2 Performance Metric

Source: <a href="https://www.kaggle.com/c/quora-question-pairs#evaluation">https://www.kaggle.com/c/quora-question-pairs#evaluation</a>

### Metric(s):

- log-loss : https://www.kaggle.com/wiki/LogarithmicLoss
- · Binary Confusion Matrix

## 2.3 Train and Test Construction

We build train and test by randomly splitting in the ratio of 70:30 or 80:20 whatever we choose as we have sufficient points to work with.

# 3. Exploratory Data Analysis

```
In [1]: import numpy as np
        import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        from subprocess import check_output
        %matplotlib inline
        import plotly.offline as py
        py.init_notebook_mode(connected=True)
        import plotly.graph_objs as go
        import plotly.tools as tls
        import os
        import gc
        import re
        from nltk.corpus import stopwords
        import distance
        from nltk.stem import PorterStemmer
        from bs4 import BeautifulSoup
```

# 3.1 Reading data and basic stats

### Out[5]:

	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 [6]: df.info()

We are given a minimal number of data fields here, consisting of:

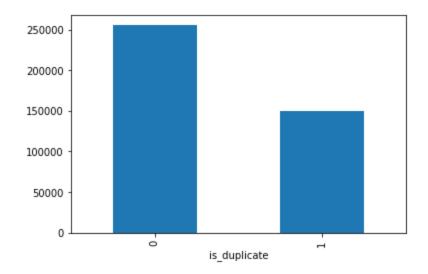
- id: Looks like a simple rowID
- qid{1, 2}: The unique ID of each question in the pair
- question{1, 2}: The actual textual contents of the questions.
- is\_duplicate: The label that we are trying to predict whether the two questions are duplicates of each other.

## 3.2.1 Distribution of data points among output classes

• Number of duplicate(smilar) and non-duplicate(non similar) questions

```
In [7]: df.groupby("is_duplicate")['id'].count().plot.bar()
```

Out[7]: <matplotlib.axes.\_subplots.AxesSubplot at 0xa0072431d0>



~> Total number of question pairs for training:
 404290

```
In [9]: print('~> Question pairs are not Similar (is_duplicate =
    0):\n {}%'.format(100 - round(df['is_duplicate'].mean()*10
    0, 2)))
    print('\n~> Question pairs are Similar (is_duplicate = 1):\n
    {}%'.format(round(df['is_duplicate'].mean()*100, 2)))
```

- ~> Question pairs are not Similar (is\_duplicate = 0):
   63.08%
- ~> Question pairs are Similar (is\_duplicate = 1):
   36.92%

# 3.2.2 Number of unique questions

```
In [10]: qids = pd.Series(df['qid1'].tolist() + df['qid2'].tolist())
    unique_qs = len(np.unique(qids))
    qs_morethan_onetime = np.sum(qids.value_counts() > 1)
    print ('Total number of Unique Questions are: {}\n'.format(u
        nique_qs))
    #print Len(np.unique(qids))

print ('Number of unique questions that appear more than one
    time: {} ({}%)\n'.format(qs_morethan_onetime,qs_morethan_onet
    ime/unique_qs*100))

print ('Max number of times a single question is repeated: {}
    \n'.format(max(qids.value_counts())))

q_vals=qids.value_counts()

q_vals=q_vals.values
```

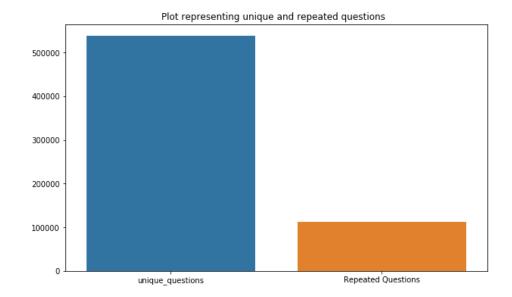
Total number of Unique Questions are: 537933

Number of unique questions that appear more than one time: 1 11780 (20.77953945937505%)

Max number of times a single question is repeated: 157

```
In [11]: x = ["unique_questions" , "Repeated Questions"]
y = [unique_qs , qs_morethan_onetime]

plt.figure(figsize=(10, 6))
plt.title ("Plot representing unique and repeated questions")
sns.barplot(x,y)
plt.show()
```



# 3.2.3 Checking for Duplicates

```
In [12]: #checking whether there are any repeated pair of questions

pair_duplicates = df[['qid1','qid2','is_duplicate']].groupby
   (['qid1','qid2']).count().reset_index()

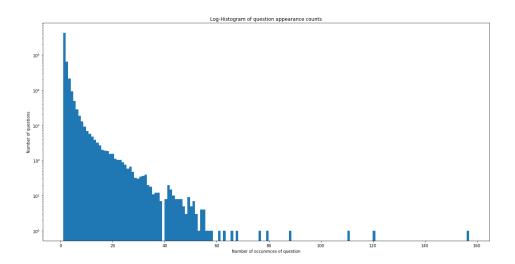
print ("Number of duplicate questions",(pair_duplicates).shap
   e[0] - df.shape[0])
```

Number of duplicate questions 0

# 3.2.4 Number of occurrences of each question

```
In [13]: plt.figure(figsize=(20, 10))
    plt.hist(qids.value_counts(), bins=160)
    plt.yscale('log', nonposy='clip')
    plt.title('Log-Histogram of question appearance counts')
    plt.xlabel('Number of occurences of question')
    plt.ylabel('Number of questions')
    print ('Maximum number of times a single question is repeate d: {}\n'.format(max(qids.value_counts())))
```

Maximum number of times a single question is repeated: 157



# 3.2.5 Checking for NULL values

```
In [14]: #Checking whether there are any rows with null values
    nan_rows = df[df.isnull().any(1)]
    print (nan_rows)
```

	id	qid1	qid2	quest
ion1 \				
105780	105780	174363	174364	How can I develop android
app?				
201841	201841	303951	174364	How can I create an Android
app?				
363362	363362	493340	493341	
NaN				

question2 i
s\_duplicate
105780 NaN
0
201841 NaN
0
363362 My Chinese name is Haichao Yu. What English na...

• There are two rows with null values in question2

```
In [15]: # Filling the null values with ' '
    df = df.fillna('')
    nan_rows = df[df.isnull().any(1)]
    print (nan_rows)

Empty DataFrame
    Columns: [id, qid1, qid2, question1, question2, is_duplicat
    e]
```

# 3.3 Basic Feature Extraction (before cleaning)

Let us now construct a few features like:

- freq\_qid1 = Frequency of qid1's
- **freq\_qid2** = Frequency of qid2's
- q1len = Length of q1

Index: []

- q2len = Length of q2
- q1\_n\_words = Number of words in Question 1
- q2\_n\_words = Number of words in Question 2
- word\_Common = (Number of common unique words in Question 1 and Question 2)
- word\_Total =(Total num of words in Question 1 + Total num of words in Question 2)
- word\_share = (word\_common)/(word\_Total)
- freq\_q1+freq\_q2 = sum total of frequency of qid1 and qid2
- freq\_q1-freq\_q2 = absolute difference of frequency of qid1 and qid2

```
In [16]: if os.path.isfile('df fe without preprocessing train.csv'):
             df = pd.read_csv("df_fe_without_preprocessing_train.csv",
         encoding='latin-1')
             df['freq qid1'] = df.groupby('qid1')['qid1'].transform('c
         ount')
             df['freq_qid2'] = df.groupby('qid2')['qid2'].transform('c
         ount')
             df['q1len'] = df['question1'].str.len()
             df['q2len'] = df['question2'].str.len()
             df['q1_n_words'] = df['question1'].apply(lambda row: len
         (row.split(" ")))
             df['q2_n_words'] = df['question2'].apply(lambda row: len
         (row.split(" ")))
             def normalized_word_Common(row):
                 w1 = set(map(lambda word: word.lower().strip(), row
         ['question1'].split(" ")))
                 w2 = set(map(lambda word: word.lower().strip(), row
         ['question2'].split(" ")))
                 return 1.0 * len(w1 & w2)
             df['word_Common'] = df.apply(normalized_word_Common, axis
         =1)
             def normalized_word_Total(row):
                 w1 = set(map(lambda word: word.lower().strip(), row
         ['question1'].split(" ")))
                 w2 = set(map(lambda word: word.lower().strip(), row
         ['question2'].split(" ")))
                 return 1.0 * (len(w1) + len(w2))
             df['word_Total'] = df.apply(normalized_word_Total, axis=
         1)
             def normalized_word_share(row):
                 w1 = set(map(lambda word: word.lower().strip(), row
         ['question1'].split(" ")))
                 w2 = set(map(lambda word: word.lower().strip(), row
         ['question2'].split(" ")))
                 return 1.0 * len(w1 & w2)/(len(w1) + len(w2))
             df['word share'] = df.apply(normalized word share, axis=
         1)
             df['freq_q1+q2'] = df['freq_qid1']+df['freq_qid2']
             df['freq_q1-q2'] = abs(df['freq_qid1']-df['freq_qid2'])
             df.to csv("df fe without preprocessing train.csv", index=
         False)
         df.head()
```

### Out[16]:

	id	qid1	qid2	question1	question2	is_duplicate	freq_qid1	freq_qid
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	1	3	4	What is the story of Kohinoor (Koh-i- Noor) Dia	What would happen if the Indian government sto	0	4	
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0	1	
3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24} [/math] i	0	1	
4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0	3	
4								<b>+</b>

# 3.3.1 Analysis of some of the extracted features

• Here are some questions have only one single words.

```
In [17]: print ("Minimum length of the questions in question1 : " , mi
    n(df['q1_n_words']))

print ("Minimum length of the questions in question2 : " , mi
    n(df['q2_n_words']))

print ("Number of Questions with minimum length [question1]
    :", df[df['q1_n_words']== 1].shape[0])
print ("Number of Questions with minimum length [question2]
    :", df[df['q2_n_words']== 1].shape[0])
```

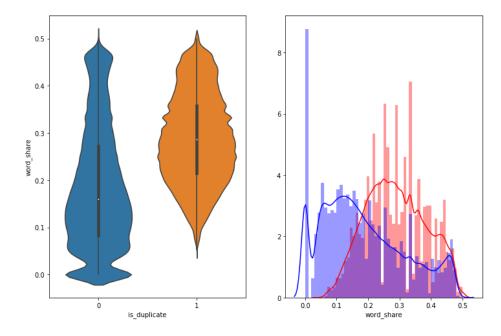
Minimum length of the questions in question1 : 1
Minimum length of the questions in question2 : 1
Number of Questions with minimum length [question1] : 67
Number of Questions with minimum length [question2] : 24

#### 3.3.1.1 Feature: word share

```
In [18]: plt.figure(figsize=(12, 8))

plt.subplot(1,2,1)
sns.violinplot(x = 'is_duplicate', y = 'word_share', data = d
f[0:])

plt.subplot(1,2,2)
sns.distplot(df[df['is_duplicate'] == 1.0]['word_share'][0:]
, label = "1", color = 'red')
sns.distplot(df[df['is_duplicate'] == 0.0]['word_share'][0:]
, label = "0" , color = 'blue' )
plt.show()
```



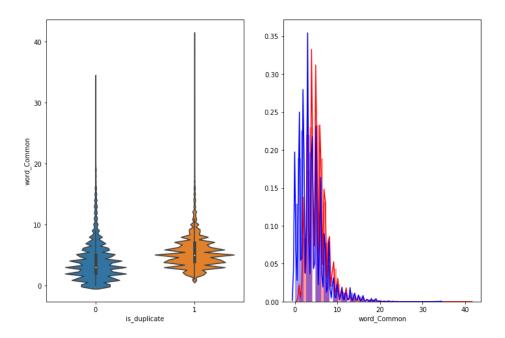
- The distributions for normalized word\_share have some overlap on the far right-hand side, i.e., there are quite a lot of questions with high word similarity
- The average word share and Common no. of words of qid1 and qid2 is more when they are duplicate(Similar)

### 3.3.1.2 Feature: word\_Common

```
In [19]: plt.figure(figsize=(12, 8))

plt.subplot(1,2,1)
sns.violinplot(x = 'is_duplicate', y = 'word_Common', data =
    df[0:])

plt.subplot(1,2,2)
sns.distplot(df[df['is_duplicate'] == 1.0]['word_Common'][0:]
, label = "1", color = 'red')
sns.distplot(df[df['is_duplicate'] == 0.0]['word_Common'][0:]
, label = "0" , color = 'blue' )
plt.show()
```



The distributions of the word\_Common feature in similar and non-similar questions are highly overlapping