# **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 : https://www.kaggle.com/c/quora-question-pairs

#### **Useful Links**

- Discussions: https://www.kaggle.com/anokas/data-analysis-xgboost-starter-0-35460-lb/comments
- Kaggle Winning Solution and other approaches: https://www.dropbox.com/sh/93968nfnrzh8bp5/AACZdtsApc1QSTQc7X0H3QZ5a?dl=0
- Blog 1 : https://engineering.quora.com/Semantic-Question-Matching-with-Deep-Learning
- Blog 2: https://towardsdatascience.com/identifying-duplicate-questions-on-quora-top-12-on-kaggle-4c1cf93f1c30

# 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
- Number of rows in Train.csv = 404.290

### 2.1.2 Example Data point

```
"id", "qid1", "qid2", "question1", "question2", "is_duplicate"
"0", "1", "2", "What is the step by step guide to invest in share market in india?", "What is the st ep by step guide to invest in share market?", "0"
"1", "3", "4", "What is the story of Kohinoor (Koh-i-Noor) Diamond?", "What would happen if the Indi an government stole the Kohinoor (Koh-i-Noor) diamond back?", "0"
"7", "15", "16", "How can I be a good geologist?", "What should I do to be a great geologist?", "1"
"11", "23", "24", "How do I read and find my YouTube comments?", "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: https://www.kaggle.com/c/quora-question-pairs#evaluation

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

#### In [2]:

```
df = pd.read_csv("train.csv")
print("Number of data points:",df.shape[0])
```

Number of data points: 404290

#### In [3]:

```
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]:

```
df.info()
```

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

- id: Looks like a simple rowlD
- 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.

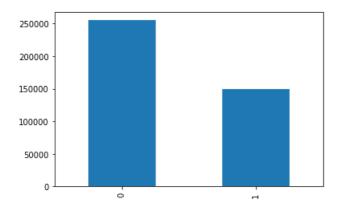
### 3.2.1 Distribution of data points among output classes

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

### In [10]:

```
df['is_duplicate'].value_counts().plot.bar()
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x1d5776f34e0>



#### In [11]:

```
print('~> Total number of question pairs for training:\n {}'.format(len(df)))
```

~> Total number of question pairs for training: 404290

#### In [16]:

- ~> 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 [21]:

```
pd.Series(df['qid1'].tolist() + df['qid2'].tolist()).value_counts().values
```

#### Out[21]:

```
array([157, 120, 111, ..., 1, 1], dtype=int64)
```

### In [22]:

```
# Creating all qid's value into the single column
# Ref : https://www.geeksforgeeks.org/python-pandas-series/
qids = pd.Series(df['qid1'].tolist() + df['qid2'].tolist())

# Finding the number of unique qid after creating into single column
unique_qs = len(np.unique(qids))

# Finding the number of qid which appear more than one (which is not unique)
qs_morethan_onetime = np.sum(qids.value_counts() > 1)

print ('Total number of Unique Questions are: {}\n'.format(unique_qs))

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

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

```
# Store the list of occurrence on each qid q_vals=qids.value_counts().values
```

Total number of Unique Questions are: 537933

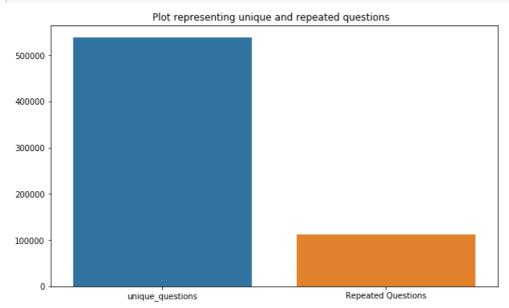
Number of unique questions that appear more than one time: 111780 (20.77953945937505%)

Max number of times a single question is repeated: 157

#### In [23]:

```
# Let plot the bar graph between unique and repeating questions occurrance appear
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 [24]:

```
# 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).shape[0] - df.shape[0])
```

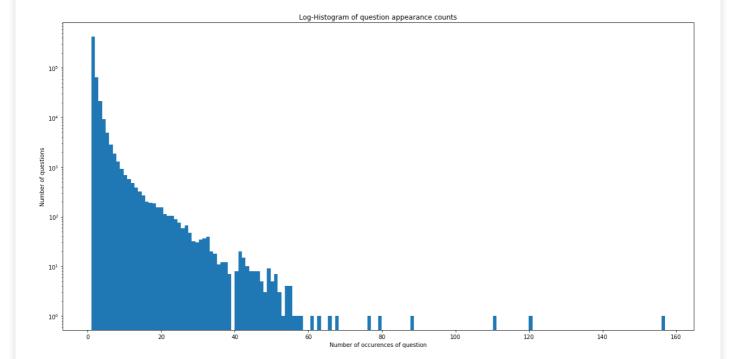
Number of duplicate questions 0

### 3.2.4 Number of occurrences of each question

#### In [27]:

```
plt.figure(figsize=(20, 10))
plt.hist(q_vals, 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 repeated: {}\n'.format(max(qids.value_counts())))
```

Maximum number of times a single question is repeated: 157



### **Understanding Observation**

From the above graph,

The last bin, its value is 157 as the number of occurrence of question and on y axis, its value is 1. That is, there is one question present where it occurrs 157 times. The first bin, its value is 1 as the number of occurrence of question ( which is basically mean unique) and on y axis, its value is above 10\daggers. So that mean, there are above 10\daggers questions that appear occurrence only once.

### 3.2.5 Checking for NULL values

```
In [30]:
# Checking whether there are any rows with null values
# We have observe info on this dataset that there are 3 data missing values
nan_rows = df[df.isnull().any(1)]
print(nan rows)
            id
                 qid1
                         qid2
                                                      question1
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
                                               question2 is duplicate
105780
                                                     NaN
                                                                     0
201841
                                                     NaN
                                                                     0
363362
      My Chinese name is Haichao Yu. What English na...
```

- There are one row with null values in question1
- There are two rows with null values in question2

#### In [31]:

```
# 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_duplicate]
```

## 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
- 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 [49]:
```

```
if os.path.isfile('df basicfe train.csv'):
   df = pd.read csv("df basicfe train.csv", encoding='latin-1')
else:
   df['freq_qid1'] = df.groupby('qid1')['qid1'].transform('count')
   df['freq qid2'] = df.groupby('qid2')['qid2'].transform('count')
   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 basicfe train.csv", index=False)
df.head()
```

Out[49]:

id qid1 qid2 question1 question2 is\_duplicate freq\_qid1 freq\_qid2 q1len q2len q1\_n\_words q2\_n\_words word\_Common

J	id	qid1	qid2	guide to <b>quasticn</b> 11 sh	guide to inv <b>questidm2</b>	is_duplicate	freq_qid1	freq_qid2	q1len	q2len	q1_n_words	q2_n_words	word_Common
1	1	3	4	What is the story of Kohinoor (Koh-i- Noor) Dia	What would happen if the Indian government sto	0	4	1	51	88	8	13	4.0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0	1	1	73	59	14	10	4.0
3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24} [/math] i	0	1	1	50	65	11	9	0.0
4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0	3	1	76	39	13	7	2.0
4													<b>F</b>

### 3.3.1 Analysis of some of the extracted features

• Here are some questions have only one single words.

### In [50]:

```
print ("Minimum length of the questions in question1 : " , min(df['q1_n_words']))
print ("Minimum length of the questions in question2 : " , min(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

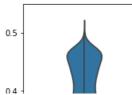
### 3.3.1.1 Feature: word\_share

#### In [55]:

```
plt.figure(figsize=(12, 8))

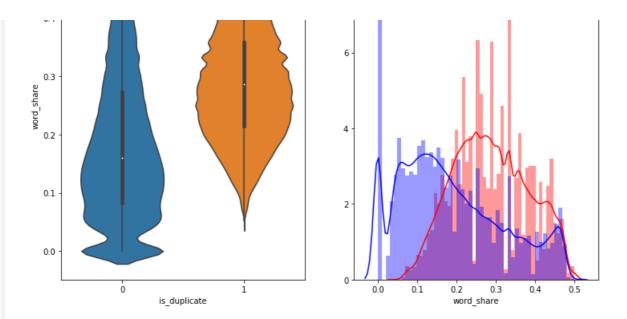
plt.subplot(1,2,1)
sns.violinplot(x = 'is_duplicate', y = 'word_share', data = df[0:])

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()
```









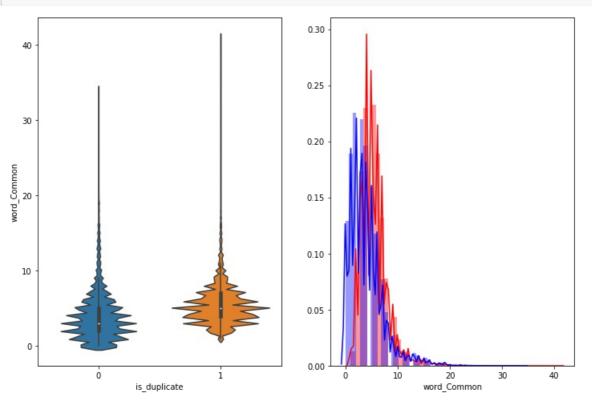
- The distributions for normalized word\_share have some overlap on the far right-hand side, i.e., there are quite a lot of questions with high word similarity
- The average word share and Common no. of words of gid1 and gid2 is more when they are duplicate(Similar)

### 3.3.1.2 Feature: word\_Common

#### In [56]:

```
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()
```



тне изшрицонь от тье мого\_сонной театите изыннагани пон-энниа диезцонь аге нідніў оченаррінд