

wrangle_act

September 17, 2020

1 Data Wrangling and Visualization - WeRateDogs Twitter account

1.1 Import the necessary libraries

```
In [5]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import requests
from bs4 import BeautifulSoup
import os
import tweepy
import json
%matplotlib inline
```

1.2 Gathering Data

Data is gathered from three different sources: Twitter archive (downloading the csv file manually), tweet image predictions (downloading the tsv file programmatically using Python's Requests library) and data scraped using the Twitter API using Python's Tweepy library . Each data is in a different file format. Each piece of data is imported into a separate pandas DataFrame at first.

1.2.1 i) Data from Twitter archive - manually downloading csv file

```
In [6]: df_archive = pd.read_csv('twitter-archive-enhanced.csv')
```

```
In [7]: df_archive.head()
```

```
Out[7]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	

	timestamp	\
0	2017-08-01 16:23:56 +0000	
1	2017-08-01 00:17:27 +0000	

```

2 2017-07-31 00:18:03 +0000
3 2017-07-30 15:58:51 +0000
4 2017-07-29 16:00:24 +0000

```

```

                                source \
0 <a href="http://twitter.com/download/iphone" r...
1 <a href="http://twitter.com/download/iphone" r...
2 <a href="http://twitter.com/download/iphone" r...
3 <a href="http://twitter.com/download/iphone" r...
4 <a href="http://twitter.com/download/iphone" r...

```

```

                                text  retweeted_status_id \
0 This is Phineas. He's a mystical boy. Only eve...      NaN
1 This is Tilly. She's just checking pup on you...      NaN
2 This is Archie. He is a rare Norwegian Pouncin...      NaN
3 This is Darla. She commenced a snooze mid meal...      NaN
4 This is Franklin. He would like you to stop ca...      NaN

```

```

retweeted_status_user_id retweeted_status_timestamp \
0                          NaN                      NaN
1                          NaN                      NaN
2                          NaN                      NaN
3                          NaN                      NaN
4                          NaN                      NaN

```

```

                                expanded_urls  rating_numerator \
0 https://twitter.com/dog_rates/status/892420643...      13
1 https://twitter.com/dog_rates/status/892177421...      13
2 https://twitter.com/dog_rates/status/891815181...      12
3 https://twitter.com/dog_rates/status/891689557...      13
4 https://twitter.com/dog_rates/status/891327558...      12

```

```

rating_denominator  name doggo floofer pupper puppo
0                10  Phineas  None    None  None  None
1                10   Tilly  None    None  None  None
2                10  Archie  None    None  None  None
3                10   Darla  None    None  None  None
4                10 Franklin  None    None  None  None

```

1.2.2 ii) Data from twitt image predictions file - programmatically downloading tsv file

```

In [8]: # Make directory if it doesn't already exist
        folder_name = 'image_predictions'
        if not os.path.exists(folder_name):
            os.makedirs(folder_name)

```

```

In [9]: # Getting the data from the URL
        url = 'https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image-predictions'

```

```
response = requests.get(url)
response
```

```
Out[9]: <Response [200]>
```

```
In [10]: # Writing the data to a file
         with open(os.path.join(folder_name,url.split('/')[-1]), mode='wb') as file:
             file.write(response.content)
```

```
In [11]: os.listdir('image_predictions')
```

```
Out[11]: ['image-predictions.tsv']
```

```
In [12]: df_predictions = pd.read_csv('image_predictions/image-predictions.tsv',sep='\t')
```

```
In [13]: df_predictions.head()
```

```
Out[13]:
```

	tweet_id	jpg_url	
0	666020888022790149	https://pbs.twimg.com/media/CT4udnOWwAA0aMy.jpg	
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg	
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	

	img_num	p1	p1_conf	p1_dog	p2
0	1	Welsh_springer_spaniel	0.465074	True	collie
1	1	redbone	0.506826	True	miniature_pinscher
2	1	German_shepherd	0.596461	True	malinois
3	1	Rhodesian_ridgeback	0.408143	True	redbone
4	1	miniature_pinscher	0.560311	True	Rottweiler

	p2_conf	p2_dog	p3	p3_conf	p3_dog
0	0.156665	True	Shetland_sheepdog	0.061428	True
1	0.074192	True	Rhodesian_ridgeback	0.072010	True
2	0.138584	True	bloodhound	0.116197	True
3	0.360687	True	miniature_pinscher	0.222752	True
4	0.243682	True	Doberman	0.154629	True

1.2.3 iii) Data scraped using the Twitter API - using Python's Tweepy library

```
In [104]: consumer_key = 'XXXXXXXXXX'
          consumer_secret = 'XXXXXXXXXX'
          access_token = 'XXXXXXXXXX'
          access_secret = 'XXXXXXXXXX'

          auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
          auth.set_access_token(access_token, access_secret)

          api = tweepy.API(auth,wait_on_rate_limit = True, wait_on_rate_limit_notify = True)
```

```
In [15]: tweet_id = df_archive.tweet_id  
        len(tweet_id)
```

```
Out[15]: 2356
```

```
In [17]: working_ids=[]  
        nonworking_ids=[]
```

```
In [18]: with open ('tweet_json.txt','w') as outfile:  
        for tweet_id in tweet_id:  
            try:  
                tweet = api.get_status(tweet_id, tweet_mode='extended')  
                json.dump(tweet._json,outfile)  
                outfile.write('\n')  
                working_ids.append(tweet_id)  
            except:  
                print ("Error for: " + str(tweet_id))  
                nonworking_ids.append(tweet_id)
```

```
Error for: 888202515573088257  
Error for: 873697596434513921  
Error for: 872668790621863937  
Error for: 869988702071779329  
Error for: 866816280283807744  
Error for: 861769973181624320  
Error for: 845459076796616705  
Error for: 842892208864923648  
Error for: 837012587749474308  
Error for: 827228250799742977  
Error for: 812747805718642688  
Error for: 802247111496568832  
Error for: 775096608509886464  
Error for: 770743923962707968
```

```
Rate limit reached. Sleeping for: 541
```

```
Error for: 758740312047005698  
Error for: 754011816964026368  
Error for: 680055455951884288
```

```
Rate limit reached. Sleeping for: 541
```

```
Error for: 676957860086095872
```

```
In [138]: working_ids
```

```
Out[138]: [892420643555336193,  
            892177421306343426,  
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            883360690899218434,  
            883117836046086144,  
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```

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747219827526344708,  
747204161125646336,  
747103485104099331,  
...]
```

```
In [139]: len(working_ids)
```

```
Out[139]: 2338
```

```
In [90]: nonworking_ids
```

```
Out[90]: [888202515573088257,  
873697596434513921,
```

```

872668790621863937,
869988702071779329,
866816280283807744,
861769973181624320,
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770743923962707968,
758740312047005698,
754011816964026368,
680055455951884288,
676957860086095872]

```

```
In [19]: len(nonworking_ids)
```

```
Out[19]: 18
```

```
In [20]: df_tweetdata = pd.read_json('tweet_json.txt',lines=True)
```

```
In [21]: df_tweetdata.head()
```

```

Out[21]:   contributors  coordinates  created_at display_text_range \
0          NaN           NaN  2017-08-01 16:23:56          [0, 85]
1          NaN           NaN  2017-08-01 00:17:27          [0, 138]
2          NaN           NaN  2017-07-31 00:18:03          [0, 121]
3          NaN           NaN  2017-07-30 15:58:51          [0, 79]
4          NaN           NaN  2017-07-29 16:00:24          [0, 138]

                                entities \
0  {'hashtags': [], 'symbols': [], 'user_mentions...
1  {'hashtags': [], 'symbols': [], 'user_mentions...
2  {'hashtags': [], 'symbols': [], 'user_mentions...
3  {'hashtags': [], 'symbols': [], 'user_mentions...
4  {'hashtags': [{'text': 'BarkWeek', 'indices': ...

                                extended_entities  favorite_count \
0  {'media': [{'id': 892420639486877696, 'id_str'...          37950
1  {'media': [{'id': 892177413194625024, 'id_str'...          32588
2  {'media': [{'id': 891815175371796480, 'id_str'...          24536
3  {'media': [{'id': 891689552724799489, 'id_str'...          41291
4  {'media': [{'id': 891327551943041024, 'id_str'...          39478

    favorited  full_text  geo \
0      False  This is Phineas. He's a mystical boy. Only eve...  NaN
1      False  This is Tilly. She's just checking pup on you...  NaN

```

```

2      False  This is Archie. He is a rare Norwegian Pouncin...  NaN
3      False  This is Darla. She commenced a snooze mid meal...  NaN
4      False  This is Franklin. He would like you to stop ca...  NaN

... quoted_status \
0      ...      NaN
1      ...      NaN
2      ...      NaN
3      ...      NaN
4      ...      NaN

quoted_status_id quoted_status_id_str quoted_status_permalink \
0      NaN      NaN      NaN
1      NaN      NaN      NaN
2      NaN      NaN      NaN
3      NaN      NaN      NaN
4      NaN      NaN      NaN

retweet_count  retweeted  retweeted_status \
0      8291      False      NaN
1      6123      False      NaN
2      4054      False      NaN
3      8428      False      NaN
4      9130      False      NaN

source truncated \
0  <a href="http://twitter.com/download/iphone" r...  False
1  <a href="http://twitter.com/download/iphone" r...  False
2  <a href="http://twitter.com/download/iphone" r...  False
3  <a href="http://twitter.com/download/iphone" r...  False
4  <a href="http://twitter.com/download/iphone" r...  False

user
0  {'id': 4196983835, 'id_str': '4196983835', 'na...
1  {'id': 4196983835, 'id_str': '4196983835', 'na...
2  {'id': 4196983835, 'id_str': '4196983835', 'na...
3  {'id': 4196983835, 'id_str': '4196983835', 'na...
4  {'id': 4196983835, 'id_str': '4196983835', 'na...

[5 rows x 32 columns]

```

1.3 Assessing Data

After gathering the data, the next step is to assess the data visually and programmatically for quality and tidiness issues. We assess each of the three dataframes: `df_archive`, `df_predictions` and `df_tweetdata`, detect the issues and document them.

```
In [22]: # display the df_archive dataframe
df_archive
```

```
Out[22]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	
5	891087950875897856	NaN	NaN	
6	890971913173991426	NaN	NaN	
7	890729181411237888	NaN	NaN	
8	890609185150312448	NaN	NaN	
9	890240255349198849	NaN	NaN	
10	890006608113172480	NaN	NaN	
11	889880896479866881	NaN	NaN	
12	889665388333682689	NaN	NaN	
13	889638837579907072	NaN	NaN	
14	889531135344209921	NaN	NaN	
15	889278841981685760	NaN	NaN	
16	888917238123831296	NaN	NaN	
17	888804989199671297	NaN	NaN	
18	888554962724278272	NaN	NaN	
19	888202515573088257	NaN	NaN	
20	888078434458587136	NaN	NaN	
21	887705289381826560	NaN	NaN	
22	887517139158093824	NaN	NaN	
23	887473957103951883	NaN	NaN	
24	887343217045368832	NaN	NaN	
25	887101392804085760	NaN	NaN	
26	886983233522544640	NaN	NaN	
27	886736880519319552	NaN	NaN	
28	886680336477933568	NaN	NaN	
29	886366144734445568	NaN	NaN	
...	
2326	666411507551481857	NaN	NaN	
2327	666407126856765440	NaN	NaN	
2328	666396247373291520	NaN	NaN	
2329	666373753744588802	NaN	NaN	
2330	666362758909284353	NaN	NaN	
2331	666353288456101888	NaN	NaN	
2332	666345417576210432	NaN	NaN	
2333	666337882303524864	NaN	NaN	
2334	666293911632134144	NaN	NaN	
2335	666287406224695296	NaN	NaN	
2336	666273097616637952	NaN	NaN	
2337	666268910803644416	NaN	NaN	
2338	666104133288665088	NaN	NaN	

2339	666102155909144576	NaN	NaN
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2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2343	666073100786774016	NaN	NaN
2344	666071193221509120	NaN	NaN
2345	666063827256086533	NaN	NaN
2346	666058600524156928	NaN	NaN
2347	666057090499244032	NaN	NaN
2348	666055525042405380	NaN	NaN
2349	666051853826850816	NaN	NaN
2350	666050758794694657	NaN	NaN
2351	666049248165822465	NaN	NaN
2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp \
0	2017-08-01 16:23:56 +0000
1	2017-08-01 00:17:27 +0000
2	2017-07-31 00:18:03 +0000
3	2017-07-30 15:58:51 +0000
4	2017-07-29 16:00:24 +0000
5	2017-07-29 00:08:17 +0000
6	2017-07-28 16:27:12 +0000
7	2017-07-28 00:22:40 +0000
8	2017-07-27 16:25:51 +0000
9	2017-07-26 15:59:51 +0000
10	2017-07-26 00:31:25 +0000
11	2017-07-25 16:11:53 +0000
12	2017-07-25 01:55:32 +0000
13	2017-07-25 00:10:02 +0000
14	2017-07-24 17:02:04 +0000
15	2017-07-24 00:19:32 +0000
16	2017-07-23 00:22:39 +0000
17	2017-07-22 16:56:37 +0000
18	2017-07-22 00:23:06 +0000
19	2017-07-21 01:02:36 +0000
20	2017-07-20 16:49:33 +0000
21	2017-07-19 16:06:48 +0000
22	2017-07-19 03:39:09 +0000
23	2017-07-19 00:47:34 +0000
24	2017-07-18 16:08:03 +0000
25	2017-07-18 00:07:08 +0000
26	2017-07-17 16:17:36 +0000
27	2017-07-16 23:58:41 +0000
28	2017-07-16 20:14:00 +0000

```

29      2017-07-15 23:25:31 +0000
...
2326    2015-11-17 00:24:19 +0000
2327    2015-11-17 00:06:54 +0000
2328    2015-11-16 23:23:41 +0000
2329    2015-11-16 21:54:18 +0000
2330    2015-11-16 21:10:36 +0000
2331    2015-11-16 20:32:58 +0000
2332    2015-11-16 20:01:42 +0000
2333    2015-11-16 19:31:45 +0000
2334    2015-11-16 16:37:02 +0000
2335    2015-11-16 16:11:11 +0000
2336    2015-11-16 15:14:19 +0000
2337    2015-11-16 14:57:41 +0000
2338    2015-11-16 04:02:55 +0000
2339    2015-11-16 03:55:04 +0000
2340    2015-11-16 03:44:34 +0000
2341    2015-11-16 03:22:39 +0000
2342    2015-11-16 02:38:37 +0000
2343    2015-11-16 01:59:36 +0000
2344    2015-11-16 01:52:02 +0000
2345    2015-11-16 01:22:45 +0000
2346    2015-11-16 01:01:59 +0000
2347    2015-11-16 00:55:59 +0000
2348    2015-11-16 00:49:46 +0000
2349    2015-11-16 00:35:11 +0000
2350    2015-11-16 00:30:50 +0000
2351    2015-11-16 00:24:50 +0000
2352    2015-11-16 00:04:52 +0000
2353    2015-11-15 23:21:54 +0000
2354    2015-11-15 23:05:30 +0000
2355    2015-11-15 22:32:08 +0000

```

```

source \
0      <a href="http://twitter.com/download/iphone" r...
1      <a href="http://twitter.com/download/iphone" r...
2      <a href="http://twitter.com/download/iphone" r...
3      <a href="http://twitter.com/download/iphone" r...
4      <a href="http://twitter.com/download/iphone" r...
5      <a href="http://twitter.com/download/iphone" r...
6      <a href="http://twitter.com/download/iphone" r...
7      <a href="http://twitter.com/download/iphone" r...
8      <a href="http://twitter.com/download/iphone" r...
9      <a href="http://twitter.com/download/iphone" r...
10     <a href="http://twitter.com/download/iphone" r...
11     <a href="http://twitter.com/download/iphone" r...
12     <a href="http://twitter.com/download/iphone" r...
13     <a href="http://twitter.com/download/iphone" r...

```

[illegible]

	text	retweeted_status_id \
0	This is Phineas. He's a mystical boy. Only eve...	NaN
1	This is Tilly. She's just checking pup on you...	NaN
2	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	This is Darla. She commenced a snooze mid meal...	NaN
4	This is Franklin. He would like you to stop ca...	NaN
5	Here we have a majestic great white breaching ...	NaN
6	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	When you watch your owner call another dog a g...	NaN
8	This is Zoey. She doesn't want to be one of th...	NaN
9	This is Cassie. She is a college pup. Studying...	NaN
10	This is Koda. He is a South Australian decksha...	NaN
11	This is Bruno. He is a service shark. Only get...	NaN
12	Here's a puppo that seems to be on the fence a...	NaN
13	This is Ted. He does his best. Sometimes that'...	NaN
14	This is Stuart. He's sporting his favorite fan...	NaN
15	This is Oliver. You're witnessing one of his m...	NaN
16	This is Jim. He found a fren. Taught him how t...	NaN
17	This is Zeke. He has a new stick. Very proud o...	NaN
18	This is Ralphus. He's powering up. Attempting ...	NaN
19	RT @dog_rates: This is Canela. She attempted s...	8.874740e+17
20	This is Gerald. He was just told he didn't get...	NaN
21	This is Jeffrey. He has a monopoly on the pool...	NaN
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
23	This is Canela. She attempted some fancy porch...	NaN
24	You may not have known you needed to see this ...	NaN
25	This... is a Jubilant Antarctic House Bear. We...	NaN
26	This is Maya. She's very shy. Rarely leaves he...	NaN
27	This is Mingus. He's a wonderful father to his...	NaN
28	This is Derek. He's late for a dog meeting. 13...	NaN
29	This is Roscoe. Another pupper fallen victim t...	NaN
...
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN

2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2345	This is the happiest dog you will ever see. Ve...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2348	Here is a Siberian heavily armored polar bear ...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2350	This is a truly beautiful English Wilson Staff...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	4.196984e+09	2017-07-19 00:47:34 +0000
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	NaN	NaN
26	NaN	NaN
27	NaN	NaN
28	NaN	NaN
29	NaN	NaN
...
2326	NaN	NaN

2327	NaN	NaN
2328	NaN	NaN
2329	NaN	NaN
2330	NaN	NaN
2331	NaN	NaN
2332	NaN	NaN
2333	NaN	NaN
2334	NaN	NaN
2335	NaN	NaN
2336	NaN	NaN
2337	NaN	NaN
2338	NaN	NaN
2339	NaN	NaN
2340	NaN	NaN
2341	NaN	NaN
2342	NaN	NaN
2343	NaN	NaN
2344	NaN	NaN
2345	NaN	NaN
2346	NaN	NaN
2347	NaN	NaN
2348	NaN	NaN
2349	NaN	NaN
2350	NaN	NaN
2351	NaN	NaN
2352	NaN	NaN
2353	NaN	NaN
2354	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator \
0	https://twitter.com/dog_rates/status/892420643...	13
1	https://twitter.com/dog_rates/status/892177421...	13
2	https://twitter.com/dog_rates/status/891815181...	12
3	https://twitter.com/dog_rates/status/891689557...	13
4	https://twitter.com/dog_rates/status/891327558...	12
5	https://twitter.com/dog_rates/status/891087950...	13
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	13
7	https://twitter.com/dog_rates/status/890729181...	13
8	https://twitter.com/dog_rates/status/890609185...	13
9	https://twitter.com/dog_rates/status/890240255...	14
10	https://twitter.com/dog_rates/status/890006608...	13
11	https://twitter.com/dog_rates/status/889880896...	13
12	https://twitter.com/dog_rates/status/889665388...	13
13	https://twitter.com/dog_rates/status/889638837...	12
14	https://twitter.com/dog_rates/status/889531135...	13
15	https://twitter.com/dog_rates/status/889278841...	13
16	https://twitter.com/dog_rates/status/888917238...	12

17	https://twitter.com/dog_rates/status/888804989...	13
18	https://twitter.com/dog_rates/status/888554962...	13
19	https://twitter.com/dog_rates/status/887473957...	13
20	https://twitter.com/dog_rates/status/888078434...	12
21	https://twitter.com/dog_rates/status/887705289...	13
22	https://twitter.com/dog_rates/status/887517139...	14
23	https://twitter.com/dog_rates/status/887473957...	13
24	https://twitter.com/dog_rates/status/887343217...	13
25	https://twitter.com/dog_rates/status/887101392...	12
26	https://twitter.com/dog_rates/status/886983233...	13
27	https://www.gofundme.com/mingusneedsus , https://...	13
28	https://twitter.com/dog_rates/status/886680336...	13
29	https://twitter.com/dog_rates/status/886366144...	12
...
2326	https://twitter.com/dog_rates/status/666411507...	2
2327	https://twitter.com/dog_rates/status/666407126...	7
2328	https://twitter.com/dog_rates/status/666396247...	9
2329	https://twitter.com/dog_rates/status/666373753...	11
2330	https://twitter.com/dog_rates/status/666362758...	6
2331	https://twitter.com/dog_rates/status/666353288...	8
2332	https://twitter.com/dog_rates/status/666345417...	10
2333	https://twitter.com/dog_rates/status/666337882...	9
2334	https://twitter.com/dog_rates/status/666293911...	3
2335	https://twitter.com/dog_rates/status/666287406...	1
2336	https://twitter.com/dog_rates/status/666273097...	11
2337	https://twitter.com/dog_rates/status/666268910...	10
2338	https://twitter.com/dog_rates/status/666104133...	1
2339	https://twitter.com/dog_rates/status/666102155...	11
2340	https://twitter.com/dog_rates/status/666099513...	8
2341	https://twitter.com/dog_rates/status/666094000...	9
2342	https://twitter.com/dog_rates/status/666082916...	6
2343	https://twitter.com/dog_rates/status/666073100...	10
2344	https://twitter.com/dog_rates/status/666071193...	9
2345	https://twitter.com/dog_rates/status/666063827...	10
2346	https://twitter.com/dog_rates/status/666058600...	8
2347	https://twitter.com/dog_rates/status/666057090...	9
2348	https://twitter.com/dog_rates/status/666055525...	10
2349	https://twitter.com/dog_rates/status/666051853...	2
2350	https://twitter.com/dog_rates/status/666050758...	10
2351	https://twitter.com/dog_rates/status/666049248...	5
2352	https://twitter.com/dog_rates/status/666044226...	6
2353	https://twitter.com/dog_rates/status/666033412...	9
2354	https://twitter.com/dog_rates/status/666029285...	7
2355	https://twitter.com/dog_rates/status/666020888...	8

	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None
1	10	Tilly	None	None	None	None

2	10	Archie	None	None	None	None
3	10	Darla	None	None	None	None
4	10	Franklin	None	None	None	None
5	10	None	None	None	None	None
6	10	Jax	None	None	None	None
7	10	None	None	None	None	None
8	10	Zoey	None	None	None	None
9	10	Cassie	doggo	None	None	None
10	10	Koda	None	None	None	None
11	10	Bruno	None	None	None	None
12	10	None	None	None	None	puppo
13	10	Ted	None	None	None	None
14	10	Stuart	None	None	None	puppo
15	10	Oliver	None	None	None	None
16	10	Jim	None	None	None	None
17	10	Zeke	None	None	None	None
18	10	Ralphus	None	None	None	None
19	10	Canela	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None
28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None

2345	10	the	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2348	10	a	None	None	None	None
2349	10	an	None	None	None	None
2350	10	a	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[2356 rows x 17 columns]

In [23]: df_archive.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2356 non-null int64
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2356 non-null object
source                  2356 non-null object
text                    2356 non-null object
retweeted_status_id     181 non-null float64
retweeted_status_user_id 181 non-null float64
retweeted_status_timestamp 181 non-null object
expanded_urls           2297 non-null object
rating_numerator        2356 non-null int64
rating_denominator      2356 non-null int64
name                    2356 non-null object
doggo                   2356 non-null object
floofer                 2356 non-null object
pupper                 2356 non-null object
puppo                   2356 non-null object
dtypes: float64(4), int64(3), object(10)
memory usage: 313.0+ KB
```

In [24]: df_archive.describe()

```
Out[24]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
count	2.356000e+03	7.800000e+01	7.800000e+01	
mean	7.427716e+17	7.455079e+17	2.014171e+16	
std	6.856705e+16	7.582492e+16	1.252797e+17	
min	6.660209e+17	6.658147e+17	1.185634e+07	
25%	6.783989e+17	6.757419e+17	3.086374e+08	
50%	7.196279e+17	7.038708e+17	4.196984e+09	

75%	7.993373e+17	8.257804e+17	4.196984e+09
max	8.924206e+17	8.862664e+17	8.405479e+17

	retweeted_status_id	retweeted_status_user_id	rating_numerator \
count	1.810000e+02	1.810000e+02	2356.000000
mean	7.720400e+17	1.241698e+16	13.126486
std	6.236928e+16	9.599254e+16	45.876648
min	6.661041e+17	7.832140e+05	0.000000
25%	7.186315e+17	4.196984e+09	10.000000
50%	7.804657e+17	4.196984e+09	11.000000
75%	8.203146e+17	4.196984e+09	12.000000
max	8.874740e+17	7.874618e+17	1776.000000

	rating_denominator
count	2356.000000
mean	10.455433
std	6.745237
min	0.000000
25%	10.000000
50%	10.000000
75%	10.000000
max	170.000000

```
In [97]: # display the df_predictions dataframe
df_predictions
```

```
Out[97]:
```

	tweet_id	jpg_url \
0	666020888022790149	https://pbs.twimg.com/media/CT4udnOWwAA0aMy.jpg
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg
5	666050758794694657	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg
6	666051853826850816	https://pbs.twimg.com/media/CT5KoJ1WoAAJash.jpg
7	666055525042405380	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg
8	666057090499244032	https://pbs.twimg.com/media/CT5PY90WoAAQGLo.jpg
9	666058600524156928	https://pbs.twimg.com/media/CT5Qw94XAAA-2dP.jpg
10	666063827256086533	https://pbs.twimg.com/media/CT5Vg_wXIAAXfnj.jpg
11	666071193221509120	https://pbs.twimg.com/media/CT5cN_3WEAA10oZ.jpg
12	666073100786774016	https://pbs.twimg.com/media/CT5d9DZXAAALcwe.jpg
13	666082916733198337	https://pbs.twimg.com/media/CT5m4VGWEAAAtKc8.jpg
14	666094000022159362	https://pbs.twimg.com/media/CT5w9gUW4AAsBNN.jpg
15	666099513787052032	https://pbs.twimg.com/media/CT51-JJUEAA6hV8.jpg
16	666102155909144576	https://pbs.twimg.com/media/CT54YGiwUAENZnoK.jpg
17	666104133288665088	https://pbs.twimg.com/media/CT56LSZWAAALJj2.jpg
18	666268910803644416	https://pbs.twimg.com/media/CT8QCd1WEAADXws.jpg
19	666273097616637952	https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg
20	666287406224695296	https://pbs.twimg.com/media/CT8g3BpUEAAuFjg.jpg

21	666293911632134144	https://pbs.twimg.com/media/CT8mx7KW4AEQu8N.jpg
22	666337882303524864	https://pbs.twimg.com/media/CT90wFIWEAMuRje.jpg
23	666345417576210432	https://pbs.twimg.com/media/CT9Vn7PW0AA_ZCM.jpg
24	666353288456101888	https://pbs.twimg.com/media/CT9cx0tUEAAhNN_.jpg
25	666362758909284353	https://pbs.twimg.com/media/CT9lXGsUcAAyUft.jpg
26	666373753744588802	https://pbs.twimg.com/media/CT9vZEYUWAA1Z05.jpg
27	666396247373291520	https://pbs.twimg.com/media/CT-D2ZHWIAA3gK1.jpg
28	666407126856765440	https://pbs.twimg.com/media/CT-NvwmW4AAAugGZ.jpg
29	666411507551481857	https://pbs.twimg.com/media/CT-RugiWIAELEaq.jpg
...
2045	886366144734445568	https://pbs.twimg.com/media/DE0BTnQUwAApKEH.jpg
2046	886680336477933568	https://pbs.twimg.com/media/DE4fEDzWAAAYHMM.jpg
2047	886736880519319552	https://pbs.twimg.com/media/DE5Se8FXcAAJFx4.jpg
2048	886983233522544640	https://pbs.twimg.com/media/DE8yicJW0AAAABJ.jpg
2049	887101392804085760	https://pbs.twimg.com/media/DE-eAq6UwAA-jaE.jpg
2050	887343217045368832	https://pbs.twimg.com/ext_tw_video_thumb/88734...
2051	887473957103951883	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2052	887517139158093824	https://pbs.twimg.com/ext_tw_video_thumb/88751...
2053	887705289381826560	https://pbs.twimg.com/media/DFHDQBbXgAEqY7t.jpg
2054	888078434458587136	https://pbs.twimg.com/media/DFMWn56WsAAkA7B.jpg
2055	888202515573088257	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2056	888554962724278272	https://pbs.twimg.com/media/DFTH_0-UQAAACu20.jpg
2057	888804989199671297	https://pbs.twimg.com/media/DFWra-3VYAA2piG.jpg
2058	888917238123831296	https://pbs.twimg.com/media/DFYRgsOUQAARGh0.jpg
2059	889278841981685760	https://pbs.twimg.com/ext_tw_video_thumb/88927...
2060	889531135344209921	https://pbs.twimg.com/media/DFg_2PVW0AEHN3p.jpg
2061	889638837579907072	https://pbs.twimg.com/media/DFihzFfXsAYGDPR.jpg
2062	889665388333682689	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg
2063	889880896479866881	https://pbs.twimg.com/media/DF199B1WsAITKsg.jpg
2064	890006608113172480	https://pbs.twimg.com/media/DFnwsY4WAAAMliS.jpg
2065	890240255349198849	https://pbs.twimg.com/media/DFrEyVuW0AA03t9.jpg
2066	890609185150312448	https://pbs.twimg.com/media/DFwUU__XcAEpyXI.jpg
2067	890729181411237888	https://pbs.twimg.com/media/DFyBahAVwAAhUTd.jpg
2068	890971913173991426	https://pbs.twimg.com/media/DF1eOmZXUAAALucq.jpg
2069	891087950875897856	https://pbs.twimg.com/media/DF3HwyEWsAABqE6.jpg
2070	891327558926688256	https://pbs.twimg.com/media/DF6hr6BUMAAZgT.jpg
2071	891689557279858688	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg
2072	891815181378084864	https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg
2073	892177421306343426	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg
2074	892420643555336193	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg

	img_num		p1	p1_conf	p1_dog \
0	1	Welsh_springer_spaniel	0.465074	True	
1	1	redbone	0.506826	True	
2	1	German_shepherd	0.596461	True	
3	1	Rhodesian_ridgeback	0.408143	True	
4	1	miniature_pinscher	0.560311	True	
5	1	Bernese_mountain_dog	0.651137	True	

6	1	box_turtle	0.933012	False
7	1	chow	0.692517	True
8	1	shopping_cart	0.962465	False
9	1	miniature_poodle	0.201493	True
10	1	golden_retriever	0.775930	True
11	1	Gordon_setter	0.503672	True
12	1	Walker_hound	0.260857	True
13	1	pug	0.489814	True
14	1	bloodhound	0.195217	True
15	1	Lhasa	0.582330	True
16	1	English_setter	0.298617	True
17	1	hen	0.965932	False
18	1	desktop_computer	0.086502	False
19	1	Italian_greyhound	0.176053	True
20	1	Maltese_dog	0.857531	True
21	1	three-toed_sloth	0.914671	False
22	1	ox	0.416669	False
23	1	golden_retriever	0.858744	True
24	1	malamute	0.336874	True
25	1	guinea_pig	0.996496	False
26	1	soft-coated_wheaten_terrier	0.326467	True
27	1	Chihuahua	0.978108	True
28	1	black-and-tan_coonhound	0.529139	True
29	1	coho	0.404640	False
...
2045	1	French_bulldog	0.999201	True
2046	1	convertible	0.738995	False
2047	1	kuvasz	0.309706	True
2048	2	Chihuahua	0.793469	True
2049	1	Samoyed	0.733942	True
2050	1	Mexican_hairless	0.330741	True
2051	2	Pembroke	0.809197	True
2052	1	limousine	0.130432	False
2053	1	basset	0.821664	True
2054	1	French_bulldog	0.995026	True
2055	2	Pembroke	0.809197	True
2056	3	Siberian_husky	0.700377	True
2057	1	golden_retriever	0.469760	True
2058	1	golden_retriever	0.714719	True
2059	1	whippet	0.626152	True
2060	1	golden_retriever	0.953442	True
2061	1	French_bulldog	0.991650	True
2062	1	Pembroke	0.966327	True
2063	1	French_bulldog	0.377417	True
2064	1	Samoyed	0.957979	True
2065	1	Pembroke	0.511319	True
2066	1	Irish_terrier	0.487574	True
2067	2	Pomeranian	0.566142	True

2068	1	Appenzeller	0.341703	True
2069	1	Chesapeake_Bay_retriever	0.425595	True
2070	2	basset	0.555712	True
2071	1	paper_towel	0.170278	False
2072	1	Chihuahua	0.716012	True
2073	1	Chihuahua	0.323581	True
2074	1	orange	0.097049	False

		p2	p2_conf	p2_dog	p3 \
0		collie	0.156665	True	Shetland_sheepdog
1	miniature_pinscher		0.074192	True	Rhodesian_ridgeback
2	malinois		0.138584	True	bloodhound
3	redbone		0.360687	True	miniature_pinscher
4	Rottweiler		0.243682	True	Doberman
5	English_springer		0.263788	True	Greater_Swiss_Mountain_dog
6	mud_turtle		0.045885	False	terrapi
7	Tibetan_mastiff		0.058279	True	fur_coat
8	shopping_basket		0.014594	False	golden_retriever
9	komondor		0.192305	True	soft-coated_wheaten_terrier
10	Tibetan_mastiff		0.093718	True	Labrador_retriever
11	Yorkshire_terrier		0.174201	True	Pekinese
12	English_foxhound		0.175382	True	Ibizan_hound
13	bull_mastiff		0.404722	True	French_bulldog
14	German_shepherd		0.078260	True	malinois
15	Shih-Tzu		0.166192	True	Dandie_Dinmont
16	Newfoundland		0.149842	True	borzoi
17	cock		0.033919	False	partridge
18	desk		0.085547	False	bookcase
19	toy_terrier		0.111884	True	basenji
20	toy_poodle		0.063064	True	miniature_poodle
21	otter		0.015250	False	great_grey_owl
22	Newfoundland		0.278407	True	groenendael
23	Chesapeake_Bay_retriever		0.054787	True	Labrador_retriever
24	Siberian_husky		0.147655	True	Eskimo_dog
25	skunk		0.002402	False	hamster
26	Afghan_hound		0.259551	True	briard
27	toy_terrier		0.009397	True	papillon
28	bloodhound		0.244220	True	flat-coated_retriever
29	barracouta		0.271485	False	gar
...
2045	Chihuahua		0.000361	True	Boston_bull
2046	sports_car		0.139952	False	car_wheel
2047	Great_Pyrenees		0.186136	True	Dandie_Dinmont
2048	toy_terrier		0.143528	True	can_opener
2049	Eskimo_dog		0.035029	True	Staffordshire_bullterrier
2050	sea_lion		0.275645	False	Weimaraner
2051	Rhodesian_ridgeback		0.054950	True	beagle
2052	tow_truck		0.029175	False	shopping_cart

2053	redbone	0.087582	True	Weimaraner
2054	pug	0.000932	True	bull_mastiff
2055	Rhodesian_ridgeback	0.054950	True	beagle
2056	Eskimo_dog	0.166511	True	malamute
2057	Labrador_retriever	0.184172	True	English_setter
2058	Tibetan_mastiff	0.120184	True	Labrador_retriever
2059	borzoi	0.194742	True	Saluki
2060	Labrador_retriever	0.013834	True	redbone
2061	boxer	0.002129	True	Staffordshire_bullterrier
2062	Cardigan	0.027356	True	basenji
2063	Labrador_retriever	0.151317	True	muzzle
2064	Pomeranian	0.013884	True	chow
2065	Cardigan	0.451038	True	Chihuahua
2066	Irish_setter	0.193054	True	Chesapeake_Bay_retriever
2067	Eskimo_dog	0.178406	True	Pembroke
2068	Border_collie	0.199287	True	ice_lolly
2069	Irish_terrier	0.116317	True	Indian_elephant
2070	English_springer	0.225770	True	German_short-haired_pointer
2071	Labrador_retriever	0.168086	True	spatula
2072	malamute	0.078253	True	kelpie
2073	Pekinese	0.090647	True	papillon
2074	bagel	0.085851	False	banana

	p3_conf	p3_dog
0	0.061428	True
1	0.072010	True
2	0.116197	True
3	0.222752	True
4	0.154629	True
5	0.016199	True
6	0.017885	False
7	0.054449	False
8	0.007959	True
9	0.082086	True
10	0.072427	True
11	0.109454	True
12	0.097471	True
13	0.048960	True
14	0.075628	True
15	0.089688	True
16	0.133649	True
17	0.000052	False
18	0.079480	False
19	0.111152	True
20	0.025581	True
21	0.013207	False
22	0.102643	True
23	0.014241	True

24	0.093412	True
25	0.000461	False
26	0.206803	True
27	0.004577	True
28	0.173810	True
29	0.189945	False
...
2045	0.000076	True
2046	0.044173	False
2047	0.086346	True
2048	0.032253	False
2049	0.029705	True
2050	0.134203	True
2051	0.038915	True
2052	0.026321	False
2053	0.026236	True
2054	0.000903	True
2055	0.038915	True
2056	0.111411	True
2057	0.073482	True
2058	0.105506	True
2059	0.027351	True
2060	0.007958	True
2061	0.001498	True
2062	0.004633	True
2063	0.082981	False
2064	0.008167	True
2065	0.029248	True
2066	0.118184	True
2067	0.076507	True
2068	0.193548	False
2069	0.076902	False
2070	0.175219	True
2071	0.040836	False
2072	0.031379	True
2073	0.068957	True
2074	0.076110	False

[2075 rows x 12 columns]

```
In [25]: df_predictions.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2075 entries, 0 to 2074
Data columns (total 12 columns):
tweet_id    2075 non-null int64
jpg_url     2075 non-null object
img_num     2075 non-null int64
```

```

p1          2075 non-null object
p1_conf     2075 non-null float64
p1_dog      2075 non-null bool
p2          2075 non-null object
p2_conf     2075 non-null float64
p2_dog      2075 non-null bool
p3          2075 non-null object
p3_conf     2075 non-null float64
p3_dog      2075 non-null bool
dtypes: bool(3), float64(3), int64(2), object(4)
memory usage: 152.1+ KB

```

```
In [26]: df_predictions.describe()
```

```

Out[26]:
```

	tweet_id	img_num	p1_conf	p2_conf	p3_conf
count	2.075000e+03	2075.000000	2075.000000	2.075000e+03	2.075000e+03
mean	7.384514e+17	1.203855	0.594548	1.345886e-01	6.032417e-02
std	6.785203e+16	0.561875	0.271174	1.006657e-01	5.090593e-02
min	6.660209e+17	1.000000	0.044333	1.011300e-08	1.740170e-10
25%	6.764835e+17	1.000000	0.364412	5.388625e-02	1.622240e-02
50%	7.119988e+17	1.000000	0.588230	1.181810e-01	4.944380e-02
75%	7.932034e+17	1.000000	0.843855	1.955655e-01	9.180755e-02
max	8.924206e+17	4.000000	1.000000	4.880140e-01	2.734190e-01

```
In [27]: # display the df_tweetdata dataframe
df_tweetdata
```

```

Out[27]:
```

	contributors	coordinates	created_at	display_text_range \
0	NaN	NaN	2017-08-01 16:23:56	[0, 85]
1	NaN	NaN	2017-08-01 00:17:27	[0, 138]
2	NaN	NaN	2017-07-31 00:18:03	[0, 121]
3	NaN	NaN	2017-07-30 15:58:51	[0, 79]
4	NaN	NaN	2017-07-29 16:00:24	[0, 138]
5	NaN	NaN	2017-07-29 00:08:17	[0, 138]
6	NaN	NaN	2017-07-28 16:27:12	[0, 140]
7	NaN	NaN	2017-07-28 00:22:40	[0, 118]
8	NaN	NaN	2017-07-27 16:25:51	[0, 122]
9	NaN	NaN	2017-07-26 15:59:51	[0, 133]
10	NaN	NaN	2017-07-26 00:31:25	[0, 130]
11	NaN	NaN	2017-07-25 16:11:53	[0, 107]
12	NaN	NaN	2017-07-25 01:55:32	[0, 106]
13	NaN	NaN	2017-07-25 00:10:02	[0, 91]
14	NaN	NaN	2017-07-24 17:02:04	[0, 118]
15	NaN	NaN	2017-07-24 00:19:32	[0, 138]
16	NaN	NaN	2017-07-23 00:22:39	[0, 86]
17	NaN	NaN	2017-07-22 16:56:37	[0, 128]
18	NaN	NaN	2017-07-22 00:23:06	[0, 87]
19	NaN	NaN	2017-07-20 16:49:33	[0, 127]

20	NaN	NaN	2017-07-19 16:06:48	[0, 127]
21	NaN	NaN	2017-07-19 03:39:09	[0, 108]
22	NaN	NaN	2017-07-19 00:47:34	[0, 99]
23	NaN	NaN	2017-07-18 16:08:03	[0, 88]
24	NaN	NaN	2017-07-18 00:07:08	[0, 129]
25	NaN	NaN	2017-07-17 16:17:36	[0, 101]
26	NaN	NaN	2017-07-16 23:58:41	[0, 121]
27	NaN	NaN	2017-07-16 20:14:00	[0, 71]
28	NaN	NaN	2017-07-15 23:25:31	[0, 131]
29	NaN	NaN	2017-07-15 16:51:35	[27, 105]
...
2308	NaN	NaN	2015-11-17 00:24:19	[0, 140]
2309	NaN	NaN	2015-11-17 00:06:54	[0, 139]
2310	NaN	NaN	2015-11-16 23:23:41	[0, 137]
2311	NaN	NaN	2015-11-16 21:54:18	[0, 81]
2312	NaN	NaN	2015-11-16 21:10:36	[0, 140]
2313	NaN	NaN	2015-11-16 20:32:58	[0, 135]
2314	NaN	NaN	2015-11-16 20:01:42	[0, 112]
2315	NaN	NaN	2015-11-16 19:31:45	[0, 139]
2316	NaN	NaN	2015-11-16 16:37:02	[0, 138]
2317	NaN	NaN	2015-11-16 16:11:11	[0, 136]
2318	NaN	NaN	2015-11-16 15:14:19	[0, 46]
2319	NaN	NaN	2015-11-16 14:57:41	[0, 82]
2320	NaN	NaN	2015-11-16 04:02:55	[0, 134]
2321	NaN	NaN	2015-11-16 03:55:04	[0, 128]
2322	NaN	NaN	2015-11-16 03:44:34	[0, 140]
2323	NaN	NaN	2015-11-16 03:22:39	[0, 132]
2324	NaN	NaN	2015-11-16 02:38:37	[0, 125]
2325	NaN	NaN	2015-11-16 01:59:36	[0, 137]
2326	NaN	NaN	2015-11-16 01:52:02	[0, 137]
2327	NaN	NaN	2015-11-16 01:22:45	[0, 107]
2328	NaN	NaN	2015-11-16 01:01:59	[0, 135]
2329	NaN	NaN	2015-11-16 00:55:59	[0, 124]
2330	NaN	NaN	2015-11-16 00:49:46	[0, 140]
2331	NaN	NaN	2015-11-16 00:35:11	[0, 138]
2332	NaN	NaN	2015-11-16 00:30:50	[0, 140]
2333	NaN	NaN	2015-11-16 00:24:50	[0, 120]
2334	NaN	NaN	2015-11-16 00:04:52	[0, 137]
2335	NaN	NaN	2015-11-15 23:21:54	[0, 130]
2336	NaN	NaN	2015-11-15 23:05:30	[0, 139]
2337	NaN	NaN	2015-11-15 22:32:08	[0, 131]

```

                                entities \
0    {'hashtags': [], 'symbols': [], 'user_mentions...
1    {'hashtags': [], 'symbols': [], 'user_mentions...
2    {'hashtags': [], 'symbols': [], 'user_mentions...
3    {'hashtags': [], 'symbols': [], 'user_mentions...
4    {'hashtags': [{'text': 'BarkWeek', 'indices': ...

```



```

2330 {'hashtags': [], 'symbols': [], 'user_mentions...
2331 {'hashtags': [], 'symbols': [], 'user_mentions...
2332 {'hashtags': [], 'symbols': [], 'user_mentions...
2333 {'hashtags': [], 'symbols': [], 'user_mentions...
2334 {'hashtags': [], 'symbols': [], 'user_mentions...
2335 {'hashtags': [], 'symbols': [], 'user_mentions...
2336 {'hashtags': [], 'symbols': [], 'user_mentions...
2337 {'hashtags': [], 'symbols': [], 'user_mentions...

```

		extended_entities	favorite_count \
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1	{'media': [{'id': 892177413194625024, 'id_str'...		32588
2	{'media': [{'id': 891815175371796480, 'id_str'...		24536
3	{'media': [{'id': 891689552724799489, 'id_str'...		41291
4	{'media': [{'id': 891327551943041024, 'id_str'...		39478
5	{'media': [{'id': 891087942176911360, 'id_str'...		19839
6	{'media': [{'id': 890971906207338496, 'id_str'...		11599
7	{'media': [{'id': 890729118844600320, 'id_str'...		64022
8	{'media': [{'id': 890609177319665665, 'id_str'...		27250
9	{'media': [{'id': 890240245463175168, 'id_str'...		31269
10	{'media': [{'id': 890006600089468928, 'id_str'...		30042
11	{'media': [{'id': 889880888800096258, 'id_str'...		27249
12	{'media': [{'id': 889665366129029120, 'id_str'...		47146
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14	{'media': [{'id': 889531127467266049, 'id_str'...		14810
15	{'media': [{'id': 889278779352338437, 'id_str'...		24738
16	{'media': [{'id': 888917229776945152, 'id_str'...		28532
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22	{'media': [{'id': 887473949361045505, 'id_str'...		67599
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28	{'media': [{'id': 886366138128449536, 'id_str'...		20766
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2317	{'media': [{'id': 666287399580733440, 'id_str'...	144
2318	{'media': [{'id': 666273081518768128, 'id_str'...	171
2319	{'media': [{'id': 666268904428277760, 'id_str'...	100
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2321	{'media': [{'id': 666102150364286977, 'id_str'...	78
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2323	{'media': [{'id': 666093996847063040, 'id_str'...	162
2324	{'media': [{'id': 666082912819875840, 'id_str'...	114
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2332	{'media': [{'id': 666050754986266625, 'id_str'...	130
2333	{'media': [{'id': 666049244999131136, 'id_str'...	106
2334	{'media': [{'id': 666044217047650304, 'id_str'...	293
2335	{'media': [{'id': 666033409081393153, 'id_str'...	123
2336	{'media': [{'id': 666029276303482880, 'id_str'...	125
2337	{'media': [{'id': 666020881337073664, 'id_str'...	2532

	favorited	full_text	geo \
0	False	This is Phineas. He's a mystical boy. Only eve...	NaN
1	False	This is Tilly. She's just checking pup on you...	NaN
2	False	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	False	This is Darla. She commenced a snooze mid meal...	NaN
4	False	This is Franklin. He would like you to stop ca...	NaN
5	False	Here we have a majestic great white breaching ...	NaN
6	False	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	False	When you watch your owner call another dog a g...	NaN
8	False	This is Zoey. She doesn't want to be one of th...	NaN
9	False	This is Cassie. She is a college pup. Studying...	NaN
10	False	This is Koda. He is a South Australian decksha...	NaN
11	False	This is Bruno. He is a service shark. Only get...	NaN
12	False	Here's a puppo that seems to be on the fence a...	NaN
13	False	This is Ted. He does his best. Sometimes that'...	NaN
14	False	This is Stuart. He's sporting his favorite fan...	NaN
15	False	This is Oliver. You're witnessing one of his m...	NaN
16	False	This is Jim. He found a fren. Taught him how t...	NaN
17	False	This is Zeke. He has a new stick. Very proud o...	NaN
18	False	This is Ralphus. He's powering up. Attempting ...	NaN
19	False	This is Gerald. He was just told he didn't get...	NaN
20	False	This is Jeffrey. He has a monopoly on the pool...	NaN
21	False	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
22	False	This is Canela. She attempted some fancy porch...	NaN

23	False	You may not have known you needed to see this ...	NaN
24	False	This... is a Jubilant Antarctic House Bear. We...	NaN
25	False	This is Maya. She's very shy. Rarely leaves he...	NaN
26	False	This is Mingus. He's a wonderful father to his...	NaN
27	False	This is Derek. He's late for a dog meeting. 13...	NaN
28	False	This is Roscoe. Another pupper fallen victim t...	NaN
29	False	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
...
2308	False	This is quite the dog. Gets really excited whe...	NaN
2309	False	This is a southern Vesuvius bumblegruff. Can d...	NaN
2310	False	Oh goodness. A super rare northeast Qdoba kang...	NaN
2311	False	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2312	False	Unique dog here. Very small. Lives in containe...	NaN
2313	False	Here we have a mixed Asiago from the Galápagos...	NaN
2314	False	Look at this jokester thinking seat belt laws ...	NaN
2315	False	This is an extremely rare horned Parthenon. No...	NaN
2316	False	This is a funny dog. Weird toes. Won't come do...	NaN
2317	False	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2318	False	Can take selfies 11/10 https://t.co/ws2AMaWpW	NaN
2319	False	Very concerned about fellow dog trapped in com...	NaN
2320	False	Not familiar with this breed. No tail (weird)...	NaN
2321	False	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2322	False	Can stand on stump for what seems like a while...	NaN
2323	False	This appears to be a Mongolian Presbyterian mi...	NaN
2324	False	Here we have a well-established sunblockerspan...	NaN
2325	False	Let's hope this flight isn't Malaysian (lol). ...	NaN
2326	False	Here we have a northern speckled Rhododendron...	NaN
2327	False	This is the happiest dog you will ever see. Ve...	NaN
2328	False	Here is the Rand Paul of retrievers folks! He'...	NaN
2329	False	My oh my. This is a rare blond Canadian terrie...	NaN
2330	False	Here is a Siberian heavily armored polar bear ...	NaN
2331	False	This is an odd dog. Hard on the outside but lo...	NaN
2332	False	This is a truly beautiful English Wilson Staff...	NaN
2333	False	Here we have a 1949 1st generation vulpix. Enj...	NaN
2334	False	This is a purebred Piers Morgan. Loves to Netf...	NaN
2335	False	Here is a very happy pup. Big fan of well-main...	NaN
2336	False	This is a western brown Mitsubishi terrier. Up...	NaN
2337	False	Here we have a Japanese Irish Setter. Lost eye...	NaN

	...	quoted_status \
0	...	NaN
1	...	NaN
2	...	NaN
3	...	NaN
4	...	NaN
5	...	NaN
6	...	NaN
7	...	NaN

8	...	NaN
9	...	NaN
10	...	NaN
11	...	NaN
12	...	NaN
13	...	NaN
14	...	NaN
15	...	NaN
16	...	NaN
17	...	NaN
18	...	NaN
19	...	NaN
20	...	NaN
21	...	NaN
22	...	NaN
23	...	NaN
24	...	NaN
25	...	NaN
26	...	NaN
27	...	NaN
28	...	NaN
29	...	NaN
...
2308	...	NaN
2309	...	NaN
2310	...	NaN
2311	...	NaN
2312	...	NaN
2313	...	NaN
2314	...	NaN
2315	...	NaN
2316	...	NaN
2317	...	NaN
2318	...	NaN
2319	...	NaN
2320	...	NaN
2321	...	NaN
2322	...	NaN
2323	...	NaN
2324	...	NaN
2325	...	NaN
2326	...	NaN
2327	...	NaN
2328	...	NaN
2329	...	NaN
2330	...	NaN
2331	...	NaN
2332	...	NaN

2333	...	NaN
2334	...	NaN
2335	...	NaN
2336	...	NaN
2337	...	NaN

	quoted_status_id	quoted_status_id_str	quoted_status_permalink	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	
5	NaN	NaN	NaN	
6	NaN	NaN	NaN	
7	NaN	NaN	NaN	
8	NaN	NaN	NaN	
9	NaN	NaN	NaN	
10	NaN	NaN	NaN	
11	NaN	NaN	NaN	
12	NaN	NaN	NaN	
13	NaN	NaN	NaN	
14	NaN	NaN	NaN	
15	NaN	NaN	NaN	
16	NaN	NaN	NaN	
17	NaN	NaN	NaN	
18	NaN	NaN	NaN	
19	NaN	NaN	NaN	
20	NaN	NaN	NaN	
21	NaN	NaN	NaN	
22	NaN	NaN	NaN	
23	NaN	NaN	NaN	
24	NaN	NaN	NaN	
25	NaN	NaN	NaN	
26	NaN	NaN	NaN	
27	NaN	NaN	NaN	
28	NaN	NaN	NaN	
29	NaN	NaN	NaN	
...	
2308	NaN	NaN	NaN	
2309	NaN	NaN	NaN	
2310	NaN	NaN	NaN	
2311	NaN	NaN	NaN	
2312	NaN	NaN	NaN	
2313	NaN	NaN	NaN	
2314	NaN	NaN	NaN	
2315	NaN	NaN	NaN	
2316	NaN	NaN	NaN	
2317	NaN	NaN	NaN	

2318	NaN	NaN	NaN
2319	NaN	NaN	NaN
2320	NaN	NaN	NaN
2321	NaN	NaN	NaN
2322	NaN	NaN	NaN
2323	NaN	NaN	NaN
2324	NaN	NaN	NaN
2325	NaN	NaN	NaN
2326	NaN	NaN	NaN
2327	NaN	NaN	NaN
2328	NaN	NaN	NaN
2329	NaN	NaN	NaN
2330	NaN	NaN	NaN
2331	NaN	NaN	NaN
2332	NaN	NaN	NaN
2333	NaN	NaN	NaN
2334	NaN	NaN	NaN
2335	NaN	NaN	NaN
2336	NaN	NaN	NaN
2337	NaN	NaN	NaN

	retweet_count	retweeted	retweeted_status \
0	8291	False	NaN
1	6123	False	NaN
2	4054	False	NaN
3	8428	False	NaN
4	9130	False	NaN
5	3039	False	NaN
6	2010	False	NaN
7	18389	False	NaN
8	4176	False	NaN
9	7192	False	NaN
10	7147	False	NaN
11	4853	False	NaN
12	9787	False	NaN
13	4413	False	NaN
14	2193	False	NaN
15	5234	False	NaN
16	4393	False	NaN
17	4178	False	NaN
18	3456	False	NaN
19	3407	False	NaN
20	5258	False	NaN
21	11399	False	NaN
22	17726	False	NaN
23	10157	False	NaN
24	5814	False	NaN
25	7582	False	NaN

26	3197	False	NaN
27	4354	False	NaN
28	3126	False	NaN
29	4	False	NaN
...
2308	322	False	NaN
2309	40	False	NaN
2310	83	False	NaN
2311	89	False	NaN
2312	561	False	NaN
2313	71	False	NaN
2314	132	False	NaN
2315	90	False	NaN
2316	344	False	NaN
2317	64	False	NaN
2318	76	False	NaN
2319	35	False	NaN
2320	6398	False	NaN
2321	11	False	NaN
2322	66	False	NaN
2323	72	False	NaN
2324	44	False	NaN
2325	159	False	NaN
2326	58	False	NaN
2327	213	False	NaN
2328	57	False	NaN
2329	138	False	NaN
2330	235	False	NaN
2331	831	False	NaN
2332	59	False	NaN
2333	43	False	NaN
2334	138	False	NaN
2335	43	False	NaN
2336	47	False	NaN
2337	499	False	NaN

		source truncated \
0	<a href="http://twitter.com/download/iphone" r...	False
1	<a href="http://twitter.com/download/iphone" r...	False
2	<a href="http://twitter.com/download/iphone" r...	False
3	<a href="http://twitter.com/download/iphone" r...	False
4	<a href="http://twitter.com/download/iphone" r...	False
5	<a href="http://twitter.com/download/iphone" r...	False
6	<a href="http://twitter.com/download/iphone" r...	False
7	<a href="http://twitter.com/download/iphone" r...	False
8	<a href="http://twitter.com/download/iphone" r...	False
9	<a href="http://twitter.com/download/iphone" r...	False
10	<a href="http://twitter.com/download/iphone" r...	False

[illegible]

2336	<a href="http://twitter.com/download/iphone" r...	False
2337	<a href="http://twitter.com/download/iphone" r...	False

		user
0	{'id': 4196983835, 'id_str': '4196983835', 'na...	
1	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2	{'id': 4196983835, 'id_str': '4196983835', 'na...	
3	{'id': 4196983835, 'id_str': '4196983835', 'na...	
4	{'id': 4196983835, 'id_str': '4196983835', 'na...	
5	{'id': 4196983835, 'id_str': '4196983835', 'na...	
6	{'id': 4196983835, 'id_str': '4196983835', 'na...	
7	{'id': 4196983835, 'id_str': '4196983835', 'na...	
8	{'id': 4196983835, 'id_str': '4196983835', 'na...	
9	{'id': 4196983835, 'id_str': '4196983835', 'na...	
10	{'id': 4196983835, 'id_str': '4196983835', 'na...	
11	{'id': 4196983835, 'id_str': '4196983835', 'na...	
12	{'id': 4196983835, 'id_str': '4196983835', 'na...	
13	{'id': 4196983835, 'id_str': '4196983835', 'na...	
14	{'id': 4196983835, 'id_str': '4196983835', 'na...	
15	{'id': 4196983835, 'id_str': '4196983835', 'na...	
16	{'id': 4196983835, 'id_str': '4196983835', 'na...	
17	{'id': 4196983835, 'id_str': '4196983835', 'na...	
18	{'id': 4196983835, 'id_str': '4196983835', 'na...	
19	{'id': 4196983835, 'id_str': '4196983835', 'na...	
20	{'id': 4196983835, 'id_str': '4196983835', 'na...	
21	{'id': 4196983835, 'id_str': '4196983835', 'na...	
22	{'id': 4196983835, 'id_str': '4196983835', 'na...	
23	{'id': 4196983835, 'id_str': '4196983835', 'na...	
24	{'id': 4196983835, 'id_str': '4196983835', 'na...	
25	{'id': 4196983835, 'id_str': '4196983835', 'na...	
26	{'id': 4196983835, 'id_str': '4196983835', 'na...	
27	{'id': 4196983835, 'id_str': '4196983835', 'na...	
28	{'id': 4196983835, 'id_str': '4196983835', 'na...	
29	{'id': 4196983835, 'id_str': '4196983835', 'na...	
...
2308	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2309	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2310	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2311	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2312	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2313	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2314	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2315	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2316	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2317	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2318	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2319	{'id': 4196983835, 'id_str': '4196983835', 'na...	
2320	{'id': 4196983835, 'id_str': '4196983835', 'na...	

```

2321 {'id': 4196983835, 'id_str': '4196983835', 'na...
2322 {'id': 4196983835, 'id_str': '4196983835', 'na...
2323 {'id': 4196983835, 'id_str': '4196983835', 'na...
2324 {'id': 4196983835, 'id_str': '4196983835', 'na...
2325 {'id': 4196983835, 'id_str': '4196983835', 'na...
2326 {'id': 4196983835, 'id_str': '4196983835', 'na...
2327 {'id': 4196983835, 'id_str': '4196983835', 'na...
2328 {'id': 4196983835, 'id_str': '4196983835', 'na...
2329 {'id': 4196983835, 'id_str': '4196983835', 'na...
2330 {'id': 4196983835, 'id_str': '4196983835', 'na...
2331 {'id': 4196983835, 'id_str': '4196983835', 'na...
2332 {'id': 4196983835, 'id_str': '4196983835', 'na...
2333 {'id': 4196983835, 'id_str': '4196983835', 'na...
2334 {'id': 4196983835, 'id_str': '4196983835', 'na...
2335 {'id': 4196983835, 'id_str': '4196983835', 'na...
2336 {'id': 4196983835, 'id_str': '4196983835', 'na...
2337 {'id': 4196983835, 'id_str': '4196983835', 'na...

```

[2338 rows x 32 columns]

In [28]: df_tweetdata.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2338 entries, 0 to 2337
Data columns (total 32 columns):
contributors      0 non-null float64
coordinates       0 non-null float64
created_at       2338 non-null datetime64[ns]
display_text_range 2338 non-null object
entities         2338 non-null object
extended_entities 2065 non-null object
favorite_count    2338 non-null int64
favorited        2338 non-null bool
full_text        2338 non-null object
geo              0 non-null float64
id               2338 non-null int64
id_str           2338 non-null int64
in_reply_to_screen_name 77 non-null object
in_reply_to_status_id 77 non-null float64
in_reply_to_status_id_str 77 non-null float64
in_reply_to_user_id 77 non-null float64
in_reply_to_user_id_str 77 non-null float64
is_quote_status   2338 non-null bool
lang              2338 non-null object
place            1 non-null object
possibly_sensitive 2203 non-null float64
possibly_sensitive_appealable 2203 non-null float64
quoted_status     24 non-null object

```



```

quoted_status_id          26 non-null float64
quoted_status_id_str      26 non-null float64
quoted_status_permalink   26 non-null object
retweet_count             2338 non-null int64
retweeted                 2338 non-null bool
retweeted_status          167 non-null object
source                   2338 non-null object
truncated                 2338 non-null bool
user                     2338 non-null object
dtypes: bool(4), datetime64[ns](1), float64(11), int64(4), object(12)
memory usage: 520.6+ KB

```

```

In [29]: columns_of_interest = ['id', 'retweet_count', 'retweeted', 'retweeted_status', 'favorite_count']
df_tweetdata = df_tweetdata[columns_of_interest]

```

```

In [30]: df_tweetdata.head()

```

```

Out[30]:
   id  retweet_count  retweeted  retweeted_status  \
0  892420643555336193      8291      False          NaN
1  892177421306343426      6123      False          NaN
2  891815181378084864      4054      False          NaN
3  891689557279858688      8428      False          NaN
4  891327558926688256      9130      False          NaN

   favorite_count
0           37950
1           32588
2           24536
3           41291
4           39478

```

```

In [31]: df_tweetdata.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2338 entries, 0 to 2337
Data columns (total 5 columns):
id                2338 non-null int64
retweet_count     2338 non-null int64
retweeted         2338 non-null bool
retweeted_status  167 non-null object
favorite_count    2338 non-null int64
dtypes: bool(1), int64(3), object(1)
memory usage: 75.4+ KB

```

```

In [32]: df_tweetdata.describe()

```

```

Out[32]:
   id  retweet_count  favorite_count
count  2.338000e+03   2338.000000   2338.000000

```

mean	7.422385e+17	2920.497861	7946.362703
std	6.834068e+16	4919.812857	12308.340561
min	6.660209e+17	0.000000	0.000000
25%	6.783509e+17	584.250000	1368.750000
50%	7.186224e+17	1362.000000	3453.500000
75%	7.986971e+17	3401.250000	9721.000000
max	8.924206e+17	83352.000000	163892.000000

```
In [33]: # check the 'name' column in the df_archive table for any 'None' values.
df_archive[df_archive['name']=='None']
```

```
Out[33]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
5	891087950875897856	NaN	NaN	
7	890729181411237888	NaN	NaN	
12	889665388333682689	NaN	NaN	
24	887343217045368832	NaN	NaN	
25	887101392804085760	NaN	NaN	
30	886267009285017600	8.862664e+17	2.281182e+09	
32	886054160059072513	NaN	NaN	
35	885518971528720385	NaN	NaN	
37	885167619883638784	NaN	NaN	
41	884441805382717440	NaN	NaN	
42	884247878851493888	NaN	NaN	
47	883117836046086144	NaN	NaN	
55	881633300179243008	8.816070e+17	4.738443e+07	
59	880872448815771648	NaN	NaN	
62	880095782870896641	NaN	NaN	
64	879674319642796034	8.795538e+17	3.105441e+09	
72	878604707211726852	NaN	NaN	
78	877611172832227328	NaN	NaN	
83	876537666061221889	NaN	NaN	
88	875097192612077568	NaN	NaN	
89	875021211251597312	NaN	NaN	
93	874057562936811520	NaN	NaN	
96	873580283840344065	NaN	NaN	
99	872967104147763200	NaN	NaN	
100	872820683541237760	NaN	NaN	
101	872668790621863937	NaN	NaN	
103	872486979161796608	NaN	NaN	
110	871102520638267392	NaN	NaN	
112	870804317367881728	NaN	NaN	
113	870726314365509632	8.707262e+17	1.648776e+07	
...	
2294	667138269671505920	NaN	NaN	
2298	667070482143944705	6.670655e+17	4.196984e+09	
2299	667065535570550784	NaN	NaN	
2301	667044094246576128	NaN	NaN	
2305	666837028449972224	NaN	NaN	

2306	666835007768551424	NaN	NaN
2307	666826780179869698	NaN	NaN
2310	666786068205871104	NaN	NaN
2316	666649482315059201	NaN	NaN
2320	666437273139982337	NaN	NaN
2321	666435652385423360	NaN	NaN
2322	666430724426358785	NaN	NaN
2323	666428276349472768	NaN	NaN
2324	666421158376562688	NaN	NaN
2328	666396247373291520	NaN	NaN
2329	666373753744588802	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2332	666345417576210432	NaN	NaN
2336	666273097616637952	NaN	NaN
2337	666268910803644416	NaN	NaN
2338	666104133288665088	NaN	NaN
2339	666102155909144576	NaN	NaN
2340	666099513787052032	NaN	NaN
2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2343	666073100786774016	NaN	NaN
2344	666071193221509120	NaN	NaN
2351	666049248165822465	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp \
5	2017-07-29 00:08:17 +0000
7	2017-07-28 00:22:40 +0000
12	2017-07-25 01:55:32 +0000
24	2017-07-18 16:08:03 +0000
25	2017-07-18 00:07:08 +0000
30	2017-07-15 16:51:35 +0000
32	2017-07-15 02:45:48 +0000
35	2017-07-13 15:19:09 +0000
37	2017-07-12 16:03:00 +0000
41	2017-07-10 15:58:53 +0000
42	2017-07-10 03:08:17 +0000
47	2017-07-07 00:17:54 +0000
55	2017-07-02 21:58:53 +0000
59	2017-06-30 19:35:32 +0000
62	2017-06-28 16:09:20 +0000
64	2017-06-27 12:14:36 +0000
72	2017-06-24 13:24:20 +0000
78	2017-06-21 19:36:23 +0000
83	2017-06-18 20:30:39 +0000
88	2017-06-14 21:06:43 +0000
89	2017-06-14 16:04:48 +0000

```

93      2017-06-12 00:15:36 +0000
96      2017-06-10 16:39:04 +0000
99      2017-06-09 00:02:31 +0000
100     2017-06-08 14:20:41 +0000
101     2017-06-08 04:17:07 +0000
103     2017-06-07 16:14:40 +0000
110     2017-06-03 20:33:19 +0000
112     2017-06-03 00:48:22 +0000
113     2017-06-02 19:38:25 +0000
...
2294    2015-11-19 00:32:12 +0000
2298    2015-11-18 20:02:51 +0000
2299    2015-11-18 19:43:11 +0000
2301    2015-11-18 18:17:59 +0000
2305    2015-11-18 04:35:11 +0000
2306    2015-11-18 04:27:09 +0000
2307    2015-11-18 03:54:28 +0000
2310    2015-11-18 01:12:41 +0000
2316    2015-11-17 16:09:56 +0000
2320    2015-11-17 02:06:42 +0000
2321    2015-11-17 02:00:15 +0000
2322    2015-11-17 01:40:41 +0000
2323    2015-11-17 01:30:57 +0000
2324    2015-11-17 01:02:40 +0000
2328    2015-11-16 23:23:41 +0000
2329    2015-11-16 21:54:18 +0000
2330    2015-11-16 21:10:36 +0000
2331    2015-11-16 20:32:58 +0000
2332    2015-11-16 20:01:42 +0000
2336    2015-11-16 15:14:19 +0000
2337    2015-11-16 14:57:41 +0000
2338    2015-11-16 04:02:55 +0000
2339    2015-11-16 03:55:04 +0000
2340    2015-11-16 03:44:34 +0000
2341    2015-11-16 03:22:39 +0000
2342    2015-11-16 02:38:37 +0000
2343    2015-11-16 01:59:36 +0000
2344    2015-11-16 01:52:02 +0000
2351    2015-11-16 00:24:50 +0000
2355    2015-11-15 22:32:08 +0000

```

```

source \
5      <a href="http://twitter.com/download/iphone" r...
7      <a href="http://twitter.com/download/iphone" r...
12     <a href="http://twitter.com/download/iphone" r...
24     <a href="http://twitter.com/download/iphone" r...
25     <a href="http://twitter.com/download/iphone" r...
30     <a href="http://twitter.com/download/iphone" r...

```

[illegible]

2340 <a href="http://twitter.com/download/iphone" r...
 2341 <a href="http://twitter.com/download/iphone" r...
 2342 <a href="http://twitter.com/download/iphone" r...
 2343 <a href="http://twitter.com/download/iphone" r...
 2344 <a href="http://twitter.com/download/iphone" r...
 2351 <a href="http://twitter.com/download/iphone" r...
 2355 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
5	Here we have a majestic great white breaching ...	NaN
7	When you watch your owner call another dog a g...	NaN
12	Here's a puppo that seems to be on the fence a...	NaN
24	You may not have known you needed to see this ...	NaN
25	This... is a Jubilant Antarctic House Bear. We...	NaN
30	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
32	RT @Athletics: 12/10 #BATP https://t.co/WxwJmv...	8.860537e+17
35	I have a new hero and his name is Howard. 14/1...	NaN
37	Here we have a corgi undercover as a malamute...	NaN
41	I present to you, Pup in Hat. Pup in Hat is gr...	NaN
42	OMG HE DIDN'T MEAN TO HE WAS JUST TRYING A LIT...	NaN
47	Please only send dogs. We don't rate mechanics...	NaN
55	@roushfenway These are good dogs but 17/10 is ...	NaN
59	Ugh not again. We only rate dogs. Please don't...	NaN
62	Please don't send in photos without dogs in th...	NaN
64	@RealKentMurphy 14/10 confirmed	NaN
72	Martha is stunning how h*ckin dare you. 13/10 ...	NaN
78	RT @rachel2195: @dog_rates the boyfriend and h...	8.768508e+17
83	I can say with the pupmost confidence that the...	NaN
88	You'll get your package when that precious man...	NaN
89	Guys please stop sending pictures without any ...	NaN
93	I can't believe this keeps happening. This, is...	NaN
96	We usually don't rate Deck-bound Saskatoon Bla...	NaN
99	Here's a very large dog. He has a date later. ...	NaN
100	Here are my favorite #dogsatpollingstations \n...	NaN
101	RT @loganamnosis: Penelope here is doing me qu...	8.726576e+17
103	We. Only. Rate. Dogs. Do not send in other thi...	NaN
110	Never doubt a doggo 14/10 https://t.co/AbBLh2FZCH	NaN
112	Real funny guys. Sending in a pic without a do...	NaN
113	@ComplicitOwl @ShopWeRateDogs >10/10 is res...	NaN
...
2294	Extremely intelligent dog here. Has learned to...	NaN
2298	After much debate this dog is being upgraded t...	NaN
2299	Here we have a Hufflepuff. Loves vest. Eyes wi...	NaN
2301	12/10 gimme now https://t.co/QZAnwgnOMB	NaN
2305	My goodness. Very rare dog here. Large. Tail d...	NaN
2306	These are Peruvian Feldspars. Their names are ...	NaN
2307	12/10 simply brilliant pup https://t.co/V6ZzG4...	NaN
2310	Unfamiliar with this breed. Ears pointy af. Wo...	NaN

2316	Cool dog. Enjoys couch. Low monotone bark. Ver...	NaN
2320	Here we see a lone northeastern Cumberbatch. H...	NaN
2321	"Can you behave? You're ruining my wedding day...	NaN
2322	Oh boy what a pup! Sunglasses take this one to...	NaN
2323	Here we have an Austrian Pulitzer. Collectors ...	NaN
2324	*internally screaming* 12/10 https://t.co/YMcr...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp	\
5	NaN	NaN	
7	NaN	NaN	
12	NaN	NaN	
24	NaN	NaN	
25	NaN	NaN	
30	NaN	NaN	
32	19607400.0	2017-07-15 02:44:07	+0000
35	NaN	NaN	
37	NaN	NaN	
41	NaN	NaN	
42	NaN	NaN	
47	NaN	NaN	
55	NaN	NaN	
59	NaN	NaN	
62	NaN	NaN	
64	NaN	NaN	
72	NaN	NaN	
78	512804507.0	2017-06-19 17:14:49	+0000
83	NaN	NaN	
88	NaN	NaN	
89	NaN	NaN	
93	NaN	NaN	
96	NaN	NaN	
99	NaN	NaN	

100	NaN	NaN
101	154767397.0	2017-06-08 03:32:35 +0000
103	NaN	NaN
110	NaN	NaN
112	NaN	NaN
113	NaN	NaN
...
2294	NaN	NaN
2298	NaN	NaN
2299	NaN	NaN
2301	NaN	NaN
2305	NaN	NaN
2306	NaN	NaN
2307	NaN	NaN
2310	NaN	NaN
2316	NaN	NaN
2320	NaN	NaN
2321	NaN	NaN
2322	NaN	NaN
2323	NaN	NaN
2324	NaN	NaN
2328	NaN	NaN
2329	NaN	NaN
2330	NaN	NaN
2331	NaN	NaN
2332	NaN	NaN
2336	NaN	NaN
2337	NaN	NaN
2338	NaN	NaN
2339	NaN	NaN
2340	NaN	NaN
2341	NaN	NaN
2342	NaN	NaN
2343	NaN	NaN
2344	NaN	NaN
2351	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator \
5	https://twitter.com/dog_rates/status/891087950...	13
7	https://twitter.com/dog_rates/status/890729181...	13
12	https://twitter.com/dog_rates/status/889665388...	13
24	https://twitter.com/dog_rates/status/887343217...	13
25	https://twitter.com/dog_rates/status/887101392...	12
30	NaN	12
32	https://twitter.com/dog_rates/status/886053434...	12
35	https://twitter.com/4bonds2carbon/status/88551...	14
37	https://twitter.com/dog_rates/status/885167619...	13

41	https://twitter.com/dog_rates/status/884441805...	14
42	https://twitter.com/kaijohnson_19/status/88396...	13
47	https://twitter.com/dog_rates/status/883117836...	13
55	NaN	17
59	https://twitter.com/dog_rates/status/880872448...	12
62	https://twitter.com/dog_rates/status/880095782...	11
64	NaN	14
72	https://twitter.com/bbcworld/status/8785998685...	13
78	https://twitter.com/rachel2195/status/87685077...	14
83	https://twitter.com/mpstowerham/status/8761629...	14
88	https://twitter.com/drboondoc/status/874413398...	13
89	https://twitter.com/dog_rates/status/875021211...	12
93	https://twitter.com/dog_rates/status/874057562...	12
96	https://twitter.com/dog_rates/status/873580283...	13
99	https://twitter.com/dog_rates/status/872967104...	12
100	https://twitter.com/dog_rates/status/872820683...	13
101	https://twitter.com/loganamnosis/status/872657...	14
103	https://twitter.com/dog_rates/status/872486979...	12
110	https://twitter.com/animalcog/status/871075758...	14
112	https://twitter.com/dog_rates/status/870804317...	11
113	NaN	10
...
2294	https://twitter.com/dog_rates/status/667138269...	10
2298	NaN	10
2299	https://twitter.com/dog_rates/status/667065535...	8
2301	https://twitter.com/dog_rates/status/667044094...	12
2305	https://twitter.com/dog_rates/status/666837028...	3
2306	https://twitter.com/dog_rates/status/666835007...	10
2307	https://twitter.com/dog_rates/status/666826780...	12
2310	https://twitter.com/dog_rates/status/666786068...	2
2316	https://twitter.com/dog_rates/status/666649482...	4
2320	https://twitter.com/dog_rates/status/666437273...	7
2321	https://twitter.com/dog_rates/status/666435652...	10
2322	https://twitter.com/dog_rates/status/666430724...	6
2323	https://twitter.com/dog_rates/status/666428276...	7
2324	https://twitter.com/dog_rates/status/666421158...	12
2328	https://twitter.com/dog_rates/status/666396247...	9
2329	https://twitter.com/dog_rates/status/666373753...	11
2330	https://twitter.com/dog_rates/status/666362758...	6
2331	https://twitter.com/dog_rates/status/666353288...	8
2332	https://twitter.com/dog_rates/status/666345417...	10
2336	https://twitter.com/dog_rates/status/666273097...	11
2337	https://twitter.com/dog_rates/status/666268910...	10
2338	https://twitter.com/dog_rates/status/666104133...	1
2339	https://twitter.com/dog_rates/status/666102155...	11
2340	https://twitter.com/dog_rates/status/666099513...	8
2341	https://twitter.com/dog_rates/status/666094000...	9
2342	https://twitter.com/dog_rates/status/666082916...	6

2343	https://twitter.com/dog_rates/status/666073100...	10
2344	https://twitter.com/dog_rates/status/666071193...	9
2351	https://twitter.com/dog_rates/status/666049248...	5
2355	https://twitter.com/dog_rates/status/666020888...	8

	rating_denominator	name	doggo	floofer	pupper	puppo
5	10	None	None	None	None	None
7	10	None	None	None	None	None
12	10	None	None	None	None	puppo
24	10	None	None	None	None	None
25	10	None	None	None	None	None
30	10	None	None	None	None	None
32	10	None	None	None	None	None
35	10	None	None	None	None	None
37	10	None	None	None	None	None
41	10	None	None	None	None	None
42	10	None	None	None	None	None
47	10	None	None	None	None	None
55	10	None	None	None	None	None
59	10	None	None	None	None	None
62	10	None	None	None	None	None
64	10	None	None	None	None	None
72	10	None	None	None	None	None
78	10	None	None	None	pupper	None
83	10	None	None	None	None	None
88	10	None	None	None	None	None
89	10	None	None	None	None	None
93	10	None	None	None	None	None
96	10	None	None	None	None	None
99	10	None	doggo	None	None	None
100	10	None	None	None	None	None
101	10	None	None	None	None	None
103	10	None	None	None	None	None
110	10	None	doggo	None	None	None
112	10	None	None	None	None	None
113	10	None	None	None	None	None
...
2294	10	None	None	None	None	None
2298	10	None	None	None	None	None
2299	10	None	None	None	None	None
2301	10	None	None	None	None	None
2305	10	None	None	None	None	None
2306	10	None	None	None	None	None
2307	10	None	None	None	None	None
2310	10	None	None	None	None	None
2316	10	None	None	None	None	None
2320	10	None	None	None	None	None
2321	10	None	None	None	None	None

2322	10	None	None	None	None	None
2323	10	None	None	None	None	None
2324	10	None	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2351	10	None	None	None	None	None
2355	10	None	None	None	None	None

[745 rows x 7 columns]

In [34]: *# check the 'expanded_urls' column in the df_archive table for any 'None' values to identify*
df_archive[df_archive['expanded_urls'].isnull()]

Out[34]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
30	886267009285017600	8.862664e+17	2.281182e+09	
55	881633300179243008	8.816070e+17	4.738443e+07	
64	879674319642796034	8.795538e+17	3.105441e+09	
113	870726314365509632	8.707262e+17	1.648776e+07	
148	863427515083354112	8.634256e+17	7.759620e+07	
179	857214891891077121	8.571567e+17	1.806710e+08	
185	856330835276025856	NaN	NaN	
186	856288084350160898	8.562860e+17	2.792810e+08	
188	855862651834028034	8.558616e+17	1.943518e+08	
189	855860136149123072	8.558585e+17	1.361572e+07	
218	850333567704068097	8.503288e+17	2.195506e+07	
228	848213670039564288	8.482121e+17	4.196984e+09	
234	847617282490613760	8.476062e+17	4.196984e+09	
274	840698636975636481	8.406983e+17	8.405479e+17	
290	838150277551247360	8.381455e+17	2.195506e+07	
291	838085839343206401	8.380855e+17	2.894131e+09	
313	835246439529840640	8.352460e+17	2.625958e+07	
342	832088576586297345	8.320875e+17	3.058208e+07	
346	831926988323639298	8.319030e+17	2.068372e+07	
375	828361771580813312	NaN	NaN	
387	826598799820865537	8.265984e+17	4.196984e+09	
409	823333489516937216	8.233264e+17	1.582854e+09	

427	821153421864615936	8.211526e+17	1.132119e+08
498	813130366689148928	8.131273e+17	4.196984e+09
513	811647686436880384	8.116272e+17	4.196984e+09
570	801854953262350336	8.018543e+17	1.185634e+07
576	800859414831898624	8.008580e+17	2.918590e+08
611	797165961484890113	7.971238e+17	2.916630e+07
701	786051337297522688	7.727430e+17	7.305050e+17
707	785515384317313025	NaN	NaN
843	766714921925144576	7.667118e+17	4.196984e+09
857	763956972077010945	7.638652e+17	1.584641e+07
967	750381685133418496	7.501805e+17	4.717297e+09
1005	747651430853525504	7.476487e+17	4.196984e+09
1080	738891149612572673	7.384119e+17	3.589728e+08
1295	707983188426153984	7.079801e+17	2.319108e+09
1345	704491224099647488	7.044857e+17	2.878549e+07
1445	696518437233913856	NaN	NaN
1446	696490539101908992	6.964887e+17	4.196984e+09
1474	693644216740769793	6.936422e+17	4.196984e+09
1479	693582294167244802	6.935722e+17	1.198989e+09
1497	692423280028966913	6.924173e+17	4.196984e+09
1523	690607260360429569	6.903413e+17	4.670367e+08
1598	686035780142297088	6.860340e+17	4.196984e+09
1605	685681090388975616	6.855479e+17	4.196984e+09
1618	684969860808454144	6.849598e+17	4.196984e+09
1663	682808988178739200	6.827884e+17	4.196984e+09
1689	681340665377193984	6.813394e+17	4.196984e+09
1774	678023323247357953	6.780211e+17	4.196984e+09
1819	676590572941893632	6.765883e+17	4.196984e+09
1844	675849018447167488	6.758457e+17	4.196984e+09
1895	674742531037511680	6.747400e+17	4.196984e+09
1905	674606911342424069	6.744689e+17	4.196984e+09
1914	674330906434379776	6.658147e+17	1.637468e+07
1940	673716320723169284	6.737159e+17	4.196984e+09
2038	671550332464455680	6.715449e+17	4.196984e+09
2149	669684865554620416	6.693544e+17	4.196984e+09
2189	668967877119254528	6.689207e+17	2.143566e+07
2298	667070482143944705	6.670655e+17	4.196984e+09

	timestamp \
30	2017-07-15 16:51:35 +0000
55	2017-07-02 21:58:53 +0000
64	2017-06-27 12:14:36 +0000
113	2017-06-02 19:38:25 +0000
148	2017-05-13 16:15:35 +0000
179	2017-04-26 12:48:51 +0000
185	2017-04-24 02:15:55 +0000
186	2017-04-23 23:26:03 +0000
188	2017-04-22 19:15:32 +0000

189	2017-04-22	19:05:32	+0000
218	2017-04-07	13:04:55	+0000
228	2017-04-01	16:41:12	+0000
234	2017-03-31	01:11:22	+0000
274	2017-03-11	22:59:09	+0000
290	2017-03-04	22:12:52	+0000
291	2017-03-04	17:56:49	+0000
313	2017-02-24	21:54:03	+0000
342	2017-02-16	04:45:50	+0000
346	2017-02-15	18:03:45	+0000
375	2017-02-05	21:56:51	+0000
387	2017-02-01	01:11:25	+0000
409	2017-01-23	00:56:15	+0000
427	2017-01-17	00:33:26	+0000
498	2016-12-25	21:12:41	+0000
513	2016-12-21	19:01:02	+0000
570	2016-11-24	18:28:13	+0000
576	2016-11-22	00:32:18	+0000
611	2016-11-11	19:55:50	+0000
701	2016-10-12	03:50:17	+0000
707	2016-10-10	16:20:36	+0000
843	2016-08-19	19:14:16	+0000
857	2016-08-12	04:35:10	+0000
967	2016-07-05	17:31:49	+0000
1005	2016-06-28	04:42:46	+0000
1080	2016-06-04	00:32:32	+0000
1295	2016-03-10	17:35:20	+0000
1345	2016-03-01	02:19:31	+0000
1445	2016-02-08	02:18:30	+0000
1446	2016-02-08	00:27:39	+0000
1474	2016-01-31	03:57:23	+0000
1479	2016-01-30	23:51:19	+0000
1497	2016-01-27	19:05:49	+0000
1523	2016-01-22	18:49:36	+0000
1598	2016-01-10	04:04:10	+0000
1605	2016-01-09	04:34:45	+0000
1618	2016-01-07	05:28:35	+0000
1663	2016-01-01	06:22:03	+0000
1689	2015-12-28	05:07:27	+0000
1774	2015-12-19	01:25:31	+0000
1819	2015-12-15	02:32:17	+0000
1844	2015-12-13	01:25:37	+0000
1895	2015-12-10	00:08:50	+0000
1905	2015-12-09	15:09:55	+0000
1914	2015-12-08	20:53:11	+0000
1940	2015-12-07	04:11:02	+0000
2038	2015-12-01	04:44:10	+0000
2149	2015-11-26	01:11:28	+0000

2189 2015-11-24 01:42:25 +0000
2298 2015-11-18 20:02:51 +0000

```
source \
30  <a href="http://twitter.com/download/iphone" r...
55  <a href="http://twitter.com/download/iphone" r...
64  <a href="http://twitter.com/download/iphone" r...
113 <a href="http://twitter.com/download/iphone" r...
148 <a href="http://twitter.com/download/iphone" r...
179 <a href="http://twitter.com/download/iphone" r...
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291 <a href="http://twitter.com/download/iphone" r...
313 <a href="http://twitter.com/download/iphone" r...
342 <a href="http://twitter.com/download/iphone" r...
346 <a href="http://twitter.com/download/iphone" r...
375 <a href="http://twitter.com" rel="nofollow">Tw...
387 <a href="http://twitter.com/download/iphone" r...
409 <a href="http://twitter.com/download/iphone" r...
427 <a href="http://twitter.com/download/iphone" r...
498 <a href="http://twitter.com/download/iphone" r...
513 <a href="http://twitter.com/download/iphone" r...
570 <a href="http://twitter.com/download/iphone" r...
576 <a href="http://twitter.com/download/iphone" r...
611 <a href="http://twitter.com/download/iphone" r...
701 <a href="http://twitter.com/download/iphone" r...
707 <a href="http://twitter.com/download/iphone" r...
843 <a href="http://twitter.com/download/iphone" r...
857 <a href="http://twitter.com/download/iphone" r...
967 <a href="http://twitter.com/download/iphone" r...
1005 <a href="http://twitter.com/download/iphone" r...
1080 <a href="http://twitter.com/download/iphone" r...
1295 <a href="http://twitter.com/download/iphone" r...
1345 <a href="http://twitter.com/download/iphone" r...
1445 <a href="http://twitter.com/download/iphone" r...
1446 <a href="http://twitter.com/download/iphone" r...
1474 <a href="http://twitter.com/download/iphone" r...
1479 <a href="http://twitter.com/download/iphone" r...
1497 <a href="http://twitter.com/download/iphone" r...
1523 <a href="http://twitter.com/download/iphone" r...
1598 <a href="http://twitter.com/download/iphone" r...
```

1605 <a href="http://twitter.com/download/iphone" r...
1618 <a href="http://twitter.com/download/iphone" r...
1663 <a href="http://twitter.com/download/iphone" r...
1689 <a href="http://twitter.com/download/iphone" r...
1774 <a href="http://twitter.com/download/iphone" r...
1819 <a href="http://twitter.com/download/iphone" r...
1844 <a href="http://twitter.com/download/iphone" r...
1895 <a href="http://twitter.com/download/iphone" r...
1905 <a href="http://twitter.com/download/iphone" r...
1914 <a href="http://twitter.com/download/iphone" r...
1940 <a href="http://twitter.com/download/iphone" r...
2038 <a href="http://twitter.com/download/iphone" r...
2149 <a href="http://twitter.com/download/iphone" r...
2189 <a href="http://twitter.com/download/iphone" r...
2298 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
30	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
55	@roushfenway These are good dogs but 17/10 is ...	NaN
64	@RealKentMurphy 14/10 confirmed	NaN
113	@ComplicitOwl @ShopWeRateDogs >10/10 is res...	NaN
148	@Jack_Septic_Eye I'd need a few more pics to p...	NaN
179	@Marc_IRL pixelated af 12/10	NaN
185	RT @Jenna_Marbles: @dog_rates Thanks for ratin...	8.563302e+17
186	@xianmcguire @Jenna_Marbles Kardashians wouldn...	NaN
188	@dhmontgomery We also gave snoop dogg a 420/10...	NaN
189	@s8n You tried very hard to portray this good ...	NaN
218	@markhoppus MARK THAT DOG HAS SEEN AND EXPERIE...	NaN
228	Jerry just apuppologized to me. He said there ...	NaN
234	.@breaannanicolee PUPDATE: Cannon has a heart ...	NaN
274	@O_kelvin_0 >10/10 is reserved for puppos s...	NaN
290	@markhoppus 182/10	NaN
291	@bragg6of8 @Andy_Pace_ we are still looking fo...	NaN
313	@jonnyusun @Lin_Manuel ok jomny I know you're e...	NaN
342	@docmisterio account started on 11/15/15	NaN
346	@UNC can confirm 12/10	NaN
375	Beebop and Doobert should start a band 12/10 w...	NaN
387	I was going to do 007/10, but the joke wasn't ...	NaN
409	@HistoryInPics 13/10	NaN
427	@imgur for a polar bear tho I'd say 13/10 is a...	NaN
498	I've been informed by multiple sources that th...	NaN
513	PUPDATE: I've been informed that Augie was act...	NaN
570	.@NBCSports OMG THE TINY HAT I'M GOING TO HAVE...	NaN
576	@SkyWilliams doggo simply protecting you from ...	NaN
611	@JODYHiGROLLER it may be an 11/10 but what do...	NaN
701	13/10 for breakdancing puppo @shibbnbot	NaN
707	Today, 10/10, should be National Dog Rates Day	NaN
843	His name is Charley and he already has a new s...	NaN

857	@TheEllenShow I'm not sure if you know this bu...	NaN
967	13/10 such a good doggo\n@spaghemily	NaN
1005	Other pupper asked not to have his identity sh...	NaN
1080	@mount_alex3 13/10	NaN
1295	@serial @MrRoles OH MY GOD I listened to all o...	NaN
1345	13/10 hero af\n@ABC	NaN
1445	Oh my god 10/10 for every little hot dog pupper	NaN
1446	After reading the comments I may have overesti...	NaN
1474	BREAKING PUPDATE: I've just been notified that...	NaN
1479	Personally I'd give him an 11/10. Not sure why...	NaN
1497	PUPDATE: just noticed this dog has some extra ...	NaN
1523	12/10 @LightningHolt	NaN
1598	Yes I do realize a rating of 4/20 would've bee...	NaN
1605	Jack deserves another round of applause. If yo...	NaN
1618	For those who claim this is a goat, u are wron...	NaN
1663	I'm aware that I could've said 20/16, but here...	NaN
1689	I've been told there's a slight possibility he...	NaN
1774	After getting lost in Reese's eyes for several...	NaN
1819	After some outrage from the crowd. Bubbles is ...	NaN
1844	This dog is being demoted to a 9/10 for not we...	NaN
1895	Some clarification is required. The dog is sin...	NaN
1905	The 13/10 also takes into account this impecca...	NaN
1914	13/10\n@ABC7	NaN
1940	The millennials have spoken and we've decided ...	NaN
2038	After 22 minutes of careful deliberation this ...	NaN
2149	After countless hours of research and hundreds...	NaN
2189	12/10 good shit Bubka\n@wanel15	NaN
2298	After much debate this dog is being upgraded t...	NaN

	retweeted_status_user_id	retweeted_status_timestamp	expanded_urls	\
30	NaN	NaN	NaN	
55	NaN	NaN	NaN	
64	NaN	NaN	NaN	
113	NaN	NaN	NaN	
148	NaN	NaN	NaN	
179	NaN	NaN	NaN	
185	66699013.0	2017-04-24 02:13:14 +0000	NaN	
186	NaN	NaN	NaN	
188	NaN	NaN	NaN	
189	NaN	NaN	NaN	
218	NaN	NaN	NaN	
228	NaN	NaN	NaN	
234	NaN	NaN	NaN	
274	NaN	NaN	NaN	
290	NaN	NaN	NaN	
291	NaN	NaN	NaN	
313	NaN	NaN	NaN	
342	NaN	NaN	NaN	

346	NaN	NaN	NaN
375	NaN	NaN	NaN
387	NaN	NaN	NaN
409	NaN	NaN	NaN
427	NaN	NaN	NaN
498	NaN	NaN	NaN
513	NaN	NaN	NaN
570	NaN	NaN	NaN
576	NaN	NaN	NaN
611	NaN	NaN	NaN
701	NaN	NaN	NaN
707	NaN	NaN	NaN
843	NaN	NaN	NaN
857	NaN	NaN	NaN
967	NaN	NaN	NaN
1005	NaN	NaN	NaN
1080	NaN	NaN	NaN
1295	NaN	NaN	NaN
1345	NaN	NaN	NaN
1445	NaN	NaN	NaN
1446	NaN	NaN	NaN
1474	NaN	NaN	NaN
1479	NaN	NaN	NaN
1497	NaN	NaN	NaN
1523	NaN	NaN	NaN
1598	NaN	NaN	NaN
1605	NaN	NaN	NaN
1618	NaN	NaN	NaN
1663	NaN	NaN	NaN
1689	NaN	NaN	NaN
1774	NaN	NaN	NaN
1819	NaN	NaN	NaN
1844	NaN	NaN	NaN
1895	NaN	NaN	NaN
1905	NaN	NaN	NaN
1914	NaN	NaN	NaN
1940	NaN	NaN	NaN
2038	NaN	NaN	NaN
2149	NaN	NaN	NaN
2189	NaN	NaN	NaN
2298	NaN	NaN	NaN

	rating_numerator	rating_denominator	name	doggo	floofer	pupper	puppo
30	12	10	None	None	None	None	None
55	17	10	None	None	None	None	None
64	14	10	None	None	None	None	None
113	10	10	None	None	None	None	None
148	12	10	None	None	None	None	None

179	12	10	None	None	None	None	None	None
185	14	10	None	None	None	None	None	None
186	14	10	None	None	None	None	None	None
188	420	10	None	None	None	None	None	None
189	666	10	None	None	None	None	None	None
218	13	10	None	None	None	None	None	None
228	11	10	None	None	None	None	None	None
234	13	10	None	None	None	None	None	None
274	10	10	None	None	None	None	None	None
290	182	10	None	None	None	None	None	None
291	15	10	None	None	None	None	None	None
313	960	0	None	None	None	None	None	None
342	11	15	None	None	None	None	None	None
346	12	10	None	None	None	None	None	None
375	12	10	None	None	None	None	None	None
387	7	10	None	None	None	None	None	None
409	13	10	None	None	None	None	None	None
427	13	10	None	None	None	None	None	None
498	12	10	None	None	None	None	None	None
513	11	10	None	None	None	None	None	None
570	11	10	None	None	None	None	None	None
576	11	10	None	doggo	None	None	None	None
611	11	10	None	None	None	None	None	None
701	13	10	None	None	None	None	None	puppo
707	10	10	None	None	None	None	None	None
843	13	10	None	None	None	None	None	None
857	12	10	None	doggo	None	None	None	None
967	13	10	None	doggo	None	None	None	None
1005	12	10	None	None	None	None	pupper	None
1080	13	10	None	None	None	None	None	None
1295	12	10	None	None	None	None	None	None
1345	13	10	None	None	None	None	None	None
1445	10	10	None	None	None	None	pupper	None
1446	1	10	None	None	None	None	None	None
1474	10	10	None	None	None	None	None	None
1479	11	10	None	None	None	None	None	None
1497	9	10	None	None	None	None	None	None
1523	12	10	None	None	None	None	None	None
1598	4	20	None	None	None	None	None	None
1605	14	10	None	None	None	None	None	None
1618	5	10	None	None	None	None	None	None
1663	20	16	None	None	None	None	None	None
1689	5	10	None	None	None	None	None	None
1774	13	10	None	None	None	None	None	None
1819	7	10	None	None	None	None	None	None
1844	9	10	None	None	None	None	None	None
1895	11	10	None	None	None	None	None	None
1905	13	10	None	None	None	None	None	None

1914	13	10	None	None	None	None	None
1940	1	10	None	None	None	None	None
2038	1	10	None	None	None	None	None
2149	11	10	None	None	None	None	None
2189	12	10	None	None	None	None	None
2298	10	10	None	None	None	None	None

```
In [35]: #checking the rating_numerator column for outliers for values less than 10 or greater t
df_archive[(df_archive.rating_numerator > 20) | (df_archive.rating_numerator < 10)]
```

```
Out[35]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id \
45	883482846933004288	NaN	NaN
188	855862651834028034	8.558616e+17	1.943518e+08
189	855860136149123072	8.558585e+17	1.361572e+07
229	848212111729840128	NaN	NaN
290	838150277551247360	8.381455e+17	2.195506e+07
313	835246439529840640	8.352460e+17	2.625958e+07
315	835152434251116546	NaN	NaN
340	832215909146226688	NaN	NaN
387	826598799820865537	8.265984e+17	4.196984e+09
433	820690176645140481	NaN	NaN
462	817502432452313088	NaN	NaN
485	814578408554463233	NaN	NaN
516	810984652412424192	NaN	NaN
599	798682547630837760	NaN	NaN
605	798576900688019456	NaN	NaN
695	786709082849828864	NaN	NaN
730	781661882474196992	NaN	NaN
745	780092040432480260	NaN	NaN
763	778027034220126208	NaN	NaN
764	777953400541634568	NaN	NaN
765	777885040357281792	NaN	NaN
784	775096608509886464	NaN	NaN
814	771014301343748096	NaN	NaN
835	768193404517830656	NaN	NaN
845	766423258543644672	NaN	NaN
859	763183847194451968	NaN	NaN
860	763167063695355904	NaN	NaN
883	760252756032651264	NaN	NaN
896	759099523532779520	NaN	NaN
902	758467244762497024	NaN	NaN
...
2312	666776908487630848	NaN	NaN
2314	666701168228331520	NaN	NaN
2315	666691418707132416	NaN	NaN
2316	666649482315059201	NaN	NaN
2317	666644823164719104	NaN	NaN
2319	666447344410484738	NaN	NaN

2320	666437273139982337	NaN	NaN
2322	666430724426358785	NaN	NaN
2323	666428276349472768	NaN	NaN
2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2328	666396247373291520	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	NaN	NaN
2338	666104133288665088	NaN	NaN
2340	666099513787052032	NaN	NaN
2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2344	666071193221509120	NaN	NaN
2346	666058600524156928	NaN	NaN
2347	666057090499244032	NaN	NaN
2349	666051853826850816	NaN	NaN
2351	666049248165822465	NaN	NaN
2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp \
45	2017-07-08 00:28:19 +0000
188	2017-04-22 19:15:32 +0000
189	2017-04-22 19:05:32 +0000
229	2017-04-01 16:35:01 +0000
290	2017-03-04 22:12:52 +0000
313	2017-02-24 21:54:03 +0000
315	2017-02-24 15:40:31 +0000
340	2017-02-16 13:11:49 +0000
387	2017-02-01 01:11:25 +0000
433	2017-01-15 17:52:40 +0000
462	2017-01-06 22:45:43 +0000
485	2016-12-29 21:06:41 +0000
516	2016-12-19 23:06:23 +0000
599	2016-11-16 00:22:12 +0000
605	2016-11-15 17:22:24 +0000
695	2016-10-13 23:23:56 +0000
730	2016-09-30 01:08:10 +0000
745	2016-09-25 17:10:10 +0000
763	2016-09-20 00:24:34 +0000
764	2016-09-19 19:31:59 +0000
765	2016-09-19 15:00:20 +0000
784	2016-09-11 22:20:06 +0000

814 2016-08-31 15:58:28 +0000
 835 2016-08-23 21:09:14 +0000
 845 2016-08-18 23:55:18 +0000
 859 2016-08-10 01:23:03 +0000
 860 2016-08-10 00:16:21 +0000
 883 2016-08-01 23:15:56 +0000
 896 2016-07-29 18:53:24 +0000
 902 2016-07-28 01:00:57 +0000

 2312 2015-11-18 00:36:17 +0000
 2314 2015-11-17 19:35:19 +0000
 2315 2015-11-17 18:56:35 +0000
 2316 2015-11-17 16:09:56 +0000
 2317 2015-11-17 15:51:26 +0000
 2319 2015-11-17 02:46:43 +0000
 2320 2015-11-17 02:06:42 +0000
 2322 2015-11-17 01:40:41 +0000
 2323 2015-11-17 01:30:57 +0000
 2326 2015-11-17 00:24:19 +0000
 2327 2015-11-17 00:06:54 +0000
 2328 2015-11-16 23:23:41 +0000
 2330 2015-11-16 21:10:36 +0000
 2331 2015-11-16 20:32:58 +0000
 2333 2015-11-16 19:31:45 +0000
 2334 2015-11-16 16:37:02 +0000
 2335 2015-11-16 16:11:11 +0000
 2338 2015-11-16 04:02:55 +0000
 2340 2015-11-16 03:44:34 +0000
 2341 2015-11-16 03:22:39 +0000
 2342 2015-11-16 02:38:37 +0000
 2344 2015-11-16 01:52:02 +0000
 2346 2015-11-16 01:01:59 +0000
 2347 2015-11-16 00:55:59 +0000
 2349 2015-11-16 00:35:11 +0000
 2351 2015-11-16 00:24:50 +0000
 2352 2015-11-16 00:04:52 +0000
 2353 2015-11-15 23:21:54 +0000
 2354 2015-11-15 23:05:30 +0000
 2355 2015-11-15 22:32:08 +0000

source \
 45 <a href="http://twitter.com/download/iphone" r...
 188 <a href="http://twitter.com/download/iphone" r...
 189 <a href="http://twitter.com/download/iphone" r...
 229 <a href="http://twitter.com/download/iphone" r...
 290 <a href="http://twitter.com/download/iphone" r...
 313 <a href="http://twitter.com/download/iphone" r...
 315 <a href="http://twitter.com/download/iphone" r...

[illegible]

2349 <a href="http://twitter.com/download/iphone" r...
 2351 <a href="http://twitter.com/download/iphone" r...
 2352 <a href="http://twitter.com/download/iphone" r...
 2353 <a href="http://twitter.com/download/iphone" r...
 2354 <a href="http://twitter.com/download/iphone" r...
 2355 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
45	This is Bella. She hopes her smile made you sm...	NaN
188	@dhmontgomery We also gave snoop dogg a 420/10...	NaN
189	@s8n You tried very hard to portray this good ...	NaN
229	This is Jerry. He's doing a distinguished tong...	NaN
290	@markhoppus 182/10	NaN
313	@jonnyusun @Lin_Manuel ok jomny I know you're e...	NaN
315	When you're so blinded by your systematic plag...	NaN
340	RT @dog_rates: This is Logan, the Chow who liv...	7.867091e+17
387	I was going to do 007/10, but the joke wasn't ...	NaN
433	The floofs have been released I repeat the flo...	NaN
462	RT @dog_rates: Meet Herschel. He's slightly bi...	6.924173e+17
485	RT @dog_rates: Meet Beau & Wilbur. Wilbur ...	6.981954e+17
516	Meet Sam. She smiles 24/7 & secretly aspir...	NaN
599	RT @dog_rates: Here we see a rare pouched pupp...	6.769365e+17
605	RT @dog_rates: Not familiar with this breed. N...	6.661041e+17
695	This is Logan, the Chow who lived. He solemnly...	NaN
730	Who keeps sending in pictures without dogs in ...	NaN
745	RT @dog_rates: This is Hank. He's mischievous ...	7.533757e+17
763	This is Sophie. She's a Jubilant Bush Pupper. ...	NaN
764	RT @dog_rates: Meet Gerald. He's a fairly exot...	7.681934e+17
765	This is Wesley. He's clearly trespassing. Seem...	NaN
784	RT @dog_rates: After so many requests, this is...	7.403732e+17
814	Another pic without a dog in it? What am I sup...	NaN
835	Meet Gerald. He's a fairly exotic doggo. Floof...	NaN
845	This is Shadoe. Her tongue flies out of her mo...	NaN
859	This is Clark. He collects teddy bears. It's a...	NaN
860	RT @dog_rates: Meet Eve. She's a raging alcoho...	6.732953e+17
883	This is Fido. He can tell the weather. Not goo...	NaN
896	Meet Toby. He has a drinking problem. Inflatab...	NaN
902	Why does this never happen at my front door...	NaN
...
2312	This is Josep. He is a Rye Manganese mix. Can ...	NaN
2314	This is a golden Buckminsterfullerene named Jo...	NaN
2315	This is Christoper. He is a spotted Penne. Can...	NaN
2316	Cool dog. Enjoys couch. Low monotone bark. Ver...	NaN
2317	This is Jimothy. He is a Botwanian Gouda. Can ...	NaN
2319	This is Scout. She is a black Downton Abbey. I...	NaN
2320	Here we see a lone northeastern Cumberbatch. H...	NaN
2322	Oh boy what a pup! Sunglasses take this one to...	NaN
2323	Here we have an Austrian Pulitzer. Collectors ...	NaN

2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp	\
45	NaN	NaN	
188	NaN	NaN	
189	NaN	NaN	
229	NaN	NaN	
290	NaN	NaN	
313	NaN	NaN	
315	NaN	NaN	
340	4.196984e+09	2016-10-13 23:23:56 +0000	
387	NaN	NaN	
433	NaN	NaN	
462	4.196984e+09	2016-01-27 18:42:06 +0000	
485	4.196984e+09	2016-02-12 17:22:12 +0000	
516	NaN	NaN	
599	4.196984e+09	2015-12-16 01:27:03 +0000	
605	4.196984e+09	2015-11-16 04:02:55 +0000	
695	NaN	NaN	
730	NaN	NaN	
745	4.196984e+09	2016-07-13 23:48:51 +0000	
763	NaN	NaN	
764	4.196984e+09	2016-08-23 21:09:14 +0000	
765	NaN	NaN	
784	4.196984e+09	2016-06-08 02:41:38 +0000	
814	NaN	NaN	
835	NaN	NaN	
845	NaN	NaN	

859	NaN	NaN
860	4.196984e+09	2015-12-06 00:17:55 +0000
883	NaN	NaN
896	NaN	NaN
902	NaN	NaN
...
2312	NaN	NaN
2314	NaN	NaN
2315	NaN	NaN
2316	NaN	NaN
2317	NaN	NaN
2319	NaN	NaN
2320	NaN	NaN
2322	NaN	NaN
2323	NaN	NaN
2326	NaN	NaN
2327	NaN	NaN
2328	NaN	NaN
2330	NaN	NaN
2331	NaN	NaN
2333	NaN	NaN
2334	NaN	NaN
2335	NaN	NaN
2338	NaN	NaN
2340	NaN	NaN
2341	NaN	NaN
2342	NaN	NaN
2344	NaN	NaN
2346	NaN	NaN
2347	NaN	NaN
2349	NaN	NaN
2351	NaN	NaN
2352	NaN	NaN
2353	NaN	NaN
2354	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator \
45	https://twitter.com/dog_rates/status/883482846...	5
188	NaN	420
189	NaN	666
229	https://twitter.com/dog_rates/status/848212111...	6
290	NaN	182
313	NaN	960
315	https://twitter.com/dog_rates/status/835152434...	0
340	https://twitter.com/dog_rates/status/786709082...	75
387	NaN	7
433	https://twitter.com/dog_rates/status/820690176...	84

462	https://twitter.com/dog_rates/status/692417313...	7
485	https://twitter.com/dog_rates/status/698195409...	9
516	https://www.gofundme.com/sams-smile , https://tw...	24
599	https://twitter.com/dog_rates/status/676936541...	8
605	https://twitter.com/dog_rates/status/666104133...	1
695	https://twitter.com/dog_rates/status/786709082...	75
730	https://twitter.com/dog_rates/status/781661882...	5
745	https://twitter.com/dog_rates/status/753375668...	8
763	https://twitter.com/dog_rates/status/778027034...	27
764	https://twitter.com/dog_rates/status/768193404...	8
765	https://twitter.com/dog_rates/status/777885040...	3
784	https://twitter.com/dog_rates/status/740373189...	9
814	https://twitter.com/dog_rates/status/771014301...	7
835	https://twitter.com/dog_rates/status/768193404...	8
845	https://twitter.com/dog_rates/status/766423258...	9
859	https://twitter.com/dog_rates/status/763183847...	8
860	https://twitter.com/dog_rates/status/673295268...	8
883	https://twitter.com/dog_rates/status/760252756...	4
896	https://twitter.com/dog_rates/status/759099523...	7
902	https://twitter.com/dog_rates/status/758467244...	165
...
2312	https://twitter.com/dog_rates/status/666776908...	5
2314	https://twitter.com/dog_rates/status/666701168...	8
2315	https://twitter.com/dog_rates/status/666691418...	8
2316	https://twitter.com/dog_rates/status/666649482...	4
2317	https://twitter.com/dog_rates/status/666644823...	9
2319	https://twitter.com/dog_rates/status/666447344...	9
2320	https://twitter.com/dog_rates/status/666437273...	7
2322	https://twitter.com/dog_rates/status/666430724...	6
2323	https://twitter.com/dog_rates/status/666428276...	7
2326	https://twitter.com/dog_rates/status/666411507...	2
2327	https://twitter.com/dog_rates/status/666407126...	7
2328	https://twitter.com/dog_rates/status/666396247...	9
2330	https://twitter.com/dog_rates/status/666362758...	6
2331	https://twitter.com/dog_rates/status/666353288...	8
2333	https://twitter.com/dog_rates/status/666337882...	9
2334	https://twitter.com/dog_rates/status/666293911...	3
2335	https://twitter.com/dog_rates/status/666287406...	1
2338	https://twitter.com/dog_rates/status/666104133...	1
2340	https://twitter.com/dog_rates/status/666099513...	8
2341	https://twitter.com/dog_rates/status/666094000...	9
2342	https://twitter.com/dog_rates/status/666082916...	6
2344	https://twitter.com/dog_rates/status/666071193...	9
2346	https://twitter.com/dog_rates/status/666058600...	8
2347	https://twitter.com/dog_rates/status/666057090...	9
2349	https://twitter.com/dog_rates/status/666051853...	2
2351	https://twitter.com/dog_rates/status/666049248...	5
2352	https://twitter.com/dog_rates/status/666044226...	6

2353	https://twitter.com/dog_rates/status/666033412...	9
2354	https://twitter.com/dog_rates/status/666029285...	7
2355	https://twitter.com/dog_rates/status/666020888...	8

	rating_denominator	name	doggo	floofer	pupper	puppo
45	10	Bella	None	None	None	None
188	10	None	None	None	None	None
189	10	None	None	None	None	None
229	10	Jerry	None	None	None	None
290	10	None	None	None	None	None
313	0	None	None	None	None	None
315	10	None	None	None	None	None
340	10	Logan	None	None	None	None
387	10	None	None	None	None	None
433	70	None	None	None	None	None
462	10	Herschel	None	None	pupper	None
485	10	Beau	None	None	None	None
516	7	Sam	None	None	None	None
599	10	None	None	None	pupper	None
605	10	None	None	None	None	None
695	10	Logan	None	None	None	None
730	10	None	None	None	None	None
745	10	Hank	None	None	None	None
763	10	Sophie	None	None	pupper	None
764	10	Gerald	doggo	None	None	None
765	10	Wesley	None	None	None	None
784	11	None	None	None	None	None
814	10	None	None	None	None	None
835	10	Gerald	doggo	None	None	None
845	10	Shadoc	None	None	None	None
859	10	Clark	None	None	None	None
860	10	Eve	None	None	pupper	None
883	10	Fido	None	None	None	None
896	10	Toby	None	None	None	None
902	150	None	None	None	None	None
...
2312	10	Josep	None	None	None	None
2314	10	a	None	None	None	None
2315	10	Christopher	None	None	None	None
2316	10	None	None	None	None	None
2317	10	Jimothy	None	None	None	None
2319	10	Scout	None	None	None	None
2320	10	None	None	None	None	None
2322	10	None	None	None	None	None
2323	10	None	None	None	None	None
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None

2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2338	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2344	10	None	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2349	10	an	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[464 rows x 17 columns]

In [36]: *#checking the rating_denominator column for outliers with values less than 5 or greater*
df_archive[(df_archive.rating_denominator > 15) | (df_archive.rating_denominator < 5)]

Out[36]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
313	835246439529840640	8.352460e+17	2.625958e+07	
433	820690176645140481	NaN	NaN	
902	758467244762497024	NaN	NaN	
1120	731156023742988288	NaN	NaN	
1165	722974582966214656	NaN	NaN	
1202	716439118184652801	NaN	NaN	
1228	713900603437621249	NaN	NaN	
1254	710658690886586372	NaN	NaN	
1274	709198395643068416	NaN	NaN	
1351	704054845121142784	NaN	NaN	
1433	697463031882764288	NaN	NaN	
1598	686035780142297088	6.860340e+17	4.196984e+09	
1634	684225744407494656	6.842229e+17	4.196984e+09	
1635	684222868335505415	NaN	NaN	
1663	682808988178739200	6.827884e+17	4.196984e+09	
1779	677716515794329600	NaN	NaN	
1843	675853064436391936	NaN	NaN	
2335	666287406224695296	NaN	NaN	

	timestamp	\
313	2017-02-24 21:54:03 +0000	
433	2017-01-15 17:52:40 +0000	
902	2016-07-28 01:00:57 +0000	

1120	2016-05-13	16:15:54	+0000
1165	2016-04-21	02:25:47	+0000
1202	2016-04-03	01:36:11	+0000
1228	2016-03-27	01:29:02	+0000
1254	2016-03-18	02:46:49	+0000
1274	2016-03-14	02:04:08	+0000
1351	2016-02-28	21:25:30	+0000
1433	2016-02-10	16:51:59	+0000
1598	2016-01-10	04:04:10	+0000
1634	2016-01-05	04:11:44	+0000
1635	2016-01-05	04:00:18	+0000
1663	2016-01-01	06:22:03	+0000
1779	2015-12-18	05:06:23	+0000
1843	2015-12-13	01:41:41	+0000
2335	2015-11-16	16:11:11	+0000

	source \
313	<a href="http://twitter.com/download/iphone" r...
433	<a href="http://twitter.com/download/iphone" r...
902	<a href="http://twitter.com/download/iphone" r...
1120	<a href="http://twitter.com/download/iphone" r...
1165	<a href="http://twitter.com/download/iphone" r...
1202	<a href="http://twitter.com/download/iphone" r...
1228	<a href="http://twitter.com/download/iphone" r...
1254	<a href="http://twitter.com/download/iphone" r...
1274	<a href="http://twitter.com/download/iphone" r...
1351	<a href="http://twitter.com/download/iphone" r...
1433	<a href="http://twitter.com/download/iphone" r...
1598	<a href="http://twitter.com/download/iphone" r...
1634	<a href="http://twitter.com/download/iphone" r...
1635	<a href="http://twitter.com/download/iphone" r...
1663	<a href="http://twitter.com/download/iphone" r...
1779	<a href="http://twitter.com/download/iphone" r...
1843	<a href="http://twitter.com/download/iphone" r...
2335	<a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
313	@jonnysun @Lin_Manuel ok jomny I know you're e...	NaN
433	The floofs have been released I repeat the flo...	NaN
902	Why does this never happen at my front door...	NaN
1120	Say hello to this unbelievably well behaved sq...	NaN
1165	Happy 4/20 from the squad! 13/10 for all https...	NaN
1202	This is Bluebert. He just saw that both #Final...	NaN
1228	Happy Saturday here's 9 puppers on a bench. 99...	NaN
1254	Here's a brigade of puppers. All look very pre...	NaN
1274	From left to right:\nCletus, Jerome, Alejandro...	NaN
1351	Here is a whole flock of puppers. 60/50 I'll ...	NaN
1433	Happy Wednesday here's a bucket of pups. 44/40...	NaN

1598	Yes I do realize a rating of 4/20 would've bee...	NaN
1634	Two sneaky puppers were not initially seen, mo...	NaN
1635	Someone help the girl is being mugged. Several...	NaN
1663	I'm aware that I could've said 20/16, but here...	NaN
1779	IT'S PUPPERGEDDON. Total of 144/120 ...I think...	NaN
1843	Here we have an entire platoon of puppers. Tot...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN

	retweeted_status_user_id	retweeted_status_timestamp	\
313	NaN	NaN	
433	NaN	NaN	
902	NaN	NaN	
1120	NaN	NaN	
1165	NaN	NaN	
1202	NaN	NaN	
1228	NaN	NaN	
1254	NaN	NaN	
1274	NaN	NaN	
1351	NaN	NaN	
1433	NaN	NaN	
1598	NaN	NaN	
1634	NaN	NaN	
1635	NaN	NaN	
1663	NaN	NaN	
1779	NaN	NaN	
1843	NaN	NaN	
2335	NaN	NaN	

	expanded_urls	rating_numerator	\
313	NaN	960	
433	https://twitter.com/dog_rates/status/820690176...	84	
902	https://twitter.com/dog_rates/status/758467244...	165	
1120	https://twitter.com/dog_rates/status/731156023...	204	
1165	https://twitter.com/dog_rates/status/722974582...	4	
1202	https://twitter.com/dog_rates/status/716439118...	50	
1228	https://twitter.com/dog_rates/status/713900603...	99	
1254	https://twitter.com/dog_rates/status/710658690...	80	
1274	https://twitter.com/dog_rates/status/709198395...	45	
1351	https://twitter.com/dog_rates/status/704054845...	60	
1433	https://twitter.com/dog_rates/status/697463031...	44	
1598	NaN	4	
1634	https://twitter.com/dog_rates/status/684225744...	143	
1635	https://twitter.com/dog_rates/status/684222868...	121	
1663	NaN	20	
1779	https://twitter.com/dog_rates/status/677716515...	144	
1843	https://twitter.com/dog_rates/status/675853064...	88	
2335	https://twitter.com/dog_rates/status/666287406...	1	

	rating_denominator	name	doggo	floofer	pupper	puppo
313	0	None	None	None	None	None
433	70	None	None	None	None	None
902	150	None	None	None	None	None
1120	170	this	None	None	None	None
1165	20	None	None	None	None	None
1202	50	Bluebert	None	None	None	None
1228	90	None	None	None	None	None
1254	80	None	None	None	None	None
1274	50	None	None	None	None	None
1351	50	a	None	None	None	None
1433	40	None	None	None	None	None
1598	20	None	None	None	None	None
1634	130	None	None	None	None	None
1635	110	None	None	None	None	None
1663	16	None	None	None	None	None
1779	120	None	None	None	None	None
1843	80	None	None	None	None	None
2335	2	an	None	None	None	None

```
In [37]: #identify retweets from df_archive
df_archive[df_archive['retweeted_status_id'].notnull()]
```

```
Out[37]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
19	888202515573088257	NaN	NaN	
32	886054160059072513	NaN	NaN	
36	885311592912609280	NaN	NaN	
68	879130579576475649	NaN	NaN	
73	878404777348136964	NaN	NaN	
74	878316110768087041	NaN	NaN	
78	877611172832227328	NaN	NaN	
91	874434818259525634	NaN	NaN	
95	873697596434513921	NaN	NaN	
97	873337748698140672	NaN	NaN	
101	872668790621863937	NaN	NaN	
109	871166179821445120	NaN	NaN	
118	869988702071779329	NaN	NaN	
124	868639477480148993	NaN	NaN	
130	867072653475098625	NaN	NaN	
132	866816280283807744	NaN	NaN	
137	866094527597207552	NaN	NaN	
146	863471782782697472	NaN	NaN	
155	861769973181624320	NaN	NaN	
159	860981674716409858	NaN	NaN	
160	860924035999428608	NaN	NaN	
165	860177593139703809	NaN	NaN	
171	858860390427611136	NaN	NaN	
180	857062103051644929	NaN	NaN	

182	856602993587888130	NaN	NaN
185	856330835276025856	NaN	NaN
194	855245323840757760	NaN	NaN
195	855138241867124737	NaN	NaN
204	852936405516943360	NaN	NaN
211	851953902622658560	NaN	NaN
...
784	775096608509886464	NaN	NaN
794	773336787167145985	NaN	NaN
800	772615324260794368	NaN	NaN
811	771171053431250945	NaN	NaN
815	771004394259247104	NaN	NaN
818	770743923962707968	NaN	NaN
822	770093767776997377	NaN	NaN
826	769335591808995329	NaN	NaN
829	768909767477751808	NaN	NaN
833	768554158521745409	NaN	NaN
841	766864461642756096	NaN	NaN
847	766078092750233600	NaN	NaN
860	763167063695355904	NaN	NaN
868	761750502866649088	NaN	NaN
872	761371037149827077	NaN	NaN
885	760153949710192640	NaN	NaN
890	759566828574212096	NaN	NaN
895	759159934323924993	NaN	NaN
908	757729163776290825	NaN	NaN
911	757597904299253760	NaN	NaN
926	754874841593970688	NaN	NaN
937	753298634498793472	NaN	NaN
943	752701944171524096	NaN	NaN
949	752309394570878976	NaN	NaN
1012	747242308580548608	NaN	NaN
1023	746521445350707200	NaN	NaN
1043	743835915802583040	NaN	NaN
1242	711998809858043904	NaN	NaN
2259	667550904950915073	NaN	NaN
2260	667550882905632768	NaN	NaN

	timestamp \
19	2017-07-21 01:02:36 +0000
32	2017-07-15 02:45:48 +0000
36	2017-07-13 01:35:06 +0000
68	2017-06-26 00:13:58 +0000
73	2017-06-24 00:09:53 +0000
74	2017-06-23 18:17:33 +0000
78	2017-06-21 19:36:23 +0000
91	2017-06-13 01:14:41 +0000
95	2017-06-11 00:25:14 +0000

97	2017-06-10	00:35:19	+0000
101	2017-06-08	04:17:07	+0000
109	2017-06-04	00:46:17	+0000
118	2017-05-31	18:47:24	+0000
124	2017-05-28	01:26:04	+0000
130	2017-05-23	17:40:04	+0000
132	2017-05-23	00:41:20	+0000
137	2017-05-21	00:53:21	+0000
146	2017-05-13	19:11:30	+0000
155	2017-05-09	02:29:07	+0000
159	2017-05-06	22:16:42	+0000
160	2017-05-06	18:27:40	+0000
165	2017-05-04	17:01:34	+0000
171	2017-05-01	01:47:28	+0000
180	2017-04-26	02:41:43	+0000
182	2017-04-24	20:17:23	+0000
185	2017-04-24	02:15:55	+0000
194	2017-04-21	02:22:29	+0000
195	2017-04-20	19:16:59	+0000
204	2017-04-14	17:27:40	+0000
211	2017-04-12	00:23:33	+0000
...			...
784	2016-09-11	22:20:06	+0000
794	2016-09-07	01:47:12	+0000
800	2016-09-05	02:00:22	+0000
811	2016-09-01	02:21:21	+0000
815	2016-08-31	15:19:06	+0000
818	2016-08-30	22:04:05	+0000
822	2016-08-29	03:00:36	+0000
826	2016-08-27	00:47:53	+0000
829	2016-08-25	20:35:48	+0000
833	2016-08-24	21:02:45	+0000
841	2016-08-20	05:08:29	+0000
847	2016-08-18	01:03:45	+0000
860	2016-08-10	00:16:21	+0000
868	2016-08-06	02:27:27	+0000
872	2016-08-05	01:19:35	+0000
885	2016-08-01	16:43:19	+0000
890	2016-07-31	01:50:18	+0000
895	2016-07-29	22:53:27	+0000
908	2016-07-26	00:08:05	+0000
911	2016-07-25	15:26:30	+0000
926	2016-07-18	03:06:01	+0000
937	2016-07-13	18:42:44	+0000
943	2016-07-12	03:11:42	+0000
949	2016-07-11	01:11:51	+0000
1012	2016-06-27	01:37:04	+0000
1023	2016-06-25	01:52:36	+0000

1043 2016-06-17 16:01:16 +0000
 1242 2016-03-21 19:31:59 +0000
 2259 2015-11-20 03:51:52 +0000
 2260 2015-11-20 03:51:47 +0000

```

                                source \
19  <a href="http://twitter.com/download/iphone" r...
32  <a href="http://twitter.com/download/iphone" r...
36  <a href="http://twitter.com/download/iphone" r...
68  <a href="http://twitter.com/download/iphone" r...
73  <a href="http://twitter.com/download/iphone" r...
74  <a href="http://twitter.com/download/iphone" r...
78  <a href="http://twitter.com/download/iphone" r...
91  <a href="http://twitter.com/download/iphone" r...
95  <a href="http://twitter.com/download/iphone" r...
97  <a href="http://twitter.com/download/iphone" r...
101 <a href="http://twitter.com/download/iphone" r...
109 <a href="http://twitter.com/download/iphone" r...
118 <a href="http://twitter.com/download/iphone" r...
124 <a href="http://twitter.com/download/iphone" r...
130 <a href="http://twitter.com/download/iphone" r...
132 <a href="http://twitter.com/download/iphone" r...
137 <a href="http://twitter.com/download/iphone" r...
146 <a href="http://twitter.com/download/iphone" r...
155 <a href="http://twitter.com/download/iphone" r...
159 <a href="http://twitter.com/download/iphone" r...
160 <a href="http://twitter.com/download/iphone" r...
165 <a href="http://twitter.com/download/iphone" r...
171 <a href="http://twitter.com/download/iphone" r...
180 <a href="http://twitter.com/download/iphone" r...
182 <a href="http://twitter.com/download/iphone" r...
185 <a href="http://twitter.com/download/iphone" r...
194 <a href="http://twitter.com/download/iphone" r...
195 <a href="http://twitter.com/download/iphone" r...
204 <a href="http://twitter.com/download/iphone" r...
211 <a href="http://twitter.com/download/iphone" r...
...
784 <a href="http://twitter.com/download/iphone" r...
794 <a href="http://twitter.com/download/iphone" r...
800 <a href="http://twitter.com/download/iphone" r...
811 <a href="http://twitter.com/download/iphone" r...
815 <a href="http://twitter.com/download/iphone" r...
818 <a href="http://twitter.com/download/iphone" r...
822 <a href="http://twitter.com/download/iphone" r...
826 <a href="http://twitter.com/download/iphone" r...
829 <a href="http://twitter.com/download/iphone" r...
833 <a href="http://twitter.com/download/iphone" r...
841 <a href="http://twitter.com/download/iphone" r...

```

847 <a href="http://twitter.com/download/iphone" r...
 860 <a href="http://twitter.com/download/iphone" r...
 868 <a href="http://twitter.com/download/iphone" r...
 872 <a href="http://twitter.com/download/iphone" r...
 885 <a href="http://twitter.com/download/iphone" r...
 890 <a href="http://twitter.com/download/iphone" r...
 895 <a href="http://twitter.com/download/iphone" r...
 908 <a href="http://twitter.com/download/iphone" r...
 911 <a href="http://twitter.com/download/iphone" r...
 926 <a href="http://twitter.com/download/iphone" r...
 937 <a href="http://twitter.com/download/iphone" r...
 943 <a href="http://twitter.com/download/iphone" r...
 949 <a href="http://twitter.com/download/iphone" r...
 1012 <a href="http://twitter.com/download/iphone" r...
 1023 <a href="http://twitter.com/download/iphone" r...
 1043 <a href="http://twitter.com/download/iphone" r...
 1242 <a href="http://twitter.com/download/iphone" r...
 2259 Tw...
 2260 Tw...

	text	retweeted_status_id \
19	RT @dog_rates: This is Canela. She attempted s...	8.874740e+17
32	RT @Athletics: 12/10 #BATP https://t.co/WxwJmv...	8.860537e+17
36	RT @dog_rates: This is Lilly. She just paralle...	8.305833e+17
68	RT @dog_rates: This is Emmy. She was adopted t...	8.780576e+17
73	RT @dog_rates: Meet Shadow. In an attempt to r...	8.782815e+17
74	RT @dog_rates: Meet Terrance. He's being yelle...	6.690004e+17
78	RT @rachel2195: @dog_rates the boyfriend and h...	8.768508e+17
91	RT @dog_rates: This is Coco. At first I though...	8.663350e+17
95	RT @dog_rates: This is Walter. He won't start ...	8.688804e+17
97	RT @dog_rates: This is Sierra. She's one preci...	8.732138e+17
101	RT @loganamnosis: Penelope here is doing me qu...	8.726576e+17
109	RT @dog_rates: This is Dawn. She's just checki...	8.410770e+17
118	RT @dog_rates: We only rate dogs. This is quit...	8.591970e+17
124	RT @dog_rates: Say hello to Cooper. His expres...	8.685523e+17
130	RT @rachaeleasler: these @dog_rates hats are 1...	8.650134e+17
132	RT @dog_rates: This is Jamesy. He gives a kiss...	8.664507e+17
137	RT @dog_rates: Here's a pupper before and afte...	8.378202e+17
146	RT @dog_rates: Say hello to Quinn. She's quite...	8.630625e+17
155	RT @dog_rates: "Good afternoon class today we'...	8.066291e+17
159	RT @dog_rates: Meet Lorenzo. He's an avid nift...	8.605638e+17
160	RT @tallylott: h*ckin adorable promposal. 13/1...	8.609145e+17
165	RT @dog_rates: Ohboyohboyohboyohboyohboyohboyo...	7.616730e+17
171	RT @dog_rates: Meet Winston. He knows he's a l...	8.395493e+17
180	RT @AaronChewning: First time wearing my @dog_...	8.570611e+17
182	RT @dog_rates: This is Luna. It's her first ti...	8.447048e+17
185	RT @Jenna_Marbles: @dog_rates Thanks for ratin...	8.563302e+17
194	RT @dog_rates: Meet George. He looks slightly ...	8.421635e+17

195	RT @frasercampbell_: oh my... what's that... b...	8.551225e+17
204	RT @dog_rates: I usually only share these on F...	8.316501e+17
211	RT @dog_rates: This is Astrid. She's a guide d...	8.293743e+17
...
784	RT @dog_rates: After so many requests, this is...	7.403732e+17
794	RT @dog_rates: Meet Fizz. She thinks love is a...	7.713808e+17
800	RT @dog_rates: This is Gromit. He's pupset bec...	7.652221e+17
811	RT @dog_rates: This is Frankie. He's wearing b...	6.733201e+17
815	RT @katieornah: @dog_rates learning a lot at c...	7.710021e+17
818	RT @dog_rates: Here's a doggo blowing bubbles...	7.392382e+17
822	RT @dog_rates: This is just downright precious...	7.410673e+17
826	RT @dog_rates: Ever seen a dog pet another dog...	7.069045e+17
829	RT @dog_rates: When it's Janet from accounting...	7.001438e+17
833	RT @dog_rates: This is Nollie. She's waving at...	7.399792e+17
841	RT @dog_rates: We only rate dogs... this is a ...	7.599238e+17
847	RT @dog_rates: This is Colby. He's currently r...	7.258423e+17
860	RT @dog_rates: Meet Eve. She's a raging alcoho...	6.732953e+17
868	RT @dog_rates: "Tristan do not speak to me wit...	6.853251e+17
872	RT @dog_rates: Oh. My. God. 13/10 magical af h...	7.116948e+17
885	RT @hownottodraw: The story/person behind @dog...	7.601538e+17
890	RT @dog_rates: This... is a Tyrannosaurus rex...	7.395441e+17
895	RT @dog_rates: AT DAWN...\nWE RIDE\n\n11/10 ht...	6.703191e+17
908	RT @dog_rates: This is Chompsky. He lives up t...	6.790626e+17
911	RT @jon_hill987: @dog_rates There is a cunning...	7.575971e+17
926	RT @dog_rates: This is Rubio. He has too much ...	6.791584e+17
937	RT @dog_rates: This is Carly. She's actually 2...	6.815232e+17
943	RT @dog_rates: HEY PUP WHAT'S THE PART OF THE ...	6.835159e+17
949	RT @dog_rates: Everyone needs to watch this. 1...	6.753544e+17
1012	RT @dog_rates: This pupper killed this great w...	7.047611e+17
1023	RT @dog_rates: This is Shaggy. He knows exactl...	6.678667e+17
1043	RT @dog_rates: Extremely intelligent dog here...	6.671383e+17
1242	RT @twitter: @dog_rates Awesome Tweet! 12/10. ...	7.119983e+17
2259	RT @dogratingrating: Exceptional talent. Origi...	6.675487e+17
2260	RT @dogratingrating: Unoriginal idea. Blatant ...	6.675484e+17

	retweeted_status_user_id	retweeted_status_timestamp	\
19	4.196984e+09	2017-07-19 00:47:34 +0000	
32	1.960740e+07	2017-07-15 02:44:07 +0000	
36	4.196984e+09	2017-02-12 01:04:29 +0000	
68	4.196984e+09	2017-06-23 01:10:23 +0000	
73	4.196984e+09	2017-06-23 16:00:04 +0000	
74	4.196984e+09	2015-11-24 03:51:38 +0000	
78	5.128045e+08	2017-06-19 17:14:49 +0000	
91	4.196984e+09	2017-05-21 16:48:45 +0000	
95	4.196984e+09	2017-05-28 17:23:24 +0000	
97	4.196984e+09	2017-06-09 16:22:42 +0000	
101	1.547674e+08	2017-06-08 03:32:35 +0000	
109	4.196984e+09	2017-03-13 00:02:39 +0000	

118	4.196984e+09	2017-05-02 00:04:57 +0000
124	4.196984e+09	2017-05-27 19:39:34 +0000
130	7.874618e+17	2017-05-18 01:17:25 +0000
132	4.196984e+09	2017-05-22 00:28:40 +0000
137	4.196984e+09	2017-03-04 00:21:08 +0000
146	4.196984e+09	2017-05-12 16:05:02 +0000
155	4.196984e+09	2016-12-07 22:38:52 +0000
159	4.196984e+09	2017-05-05 18:36:06 +0000
160	3.638908e+08	2017-05-06 17:49:42 +0000
165	4.196984e+09	2016-08-05 21:19:27 +0000
171	4.196984e+09	2017-03-08 18:52:12 +0000
180	5.870972e+07	2017-04-26 02:37:47 +0000
182	4.196984e+09	2017-03-23 00:18:10 +0000
185	6.669901e+07	2017-04-24 02:13:14 +0000
194	4.196984e+09	2017-03-16 00:00:07 +0000
195	7.475543e+17	2017-04-20 18:14:33 +0000
204	4.196984e+09	2017-02-14 23:43:18 +0000
211	4.196984e+09	2017-02-08 17:00:26 +0000
...
784	4.196984e+09	2016-06-08 02:41:38 +0000
794	4.196984e+09	2016-09-01 16:14:48 +0000
800	4.196984e+09	2016-08-15 16:22:20 +0000
811	4.196984e+09	2015-12-06 01:56:44 +0000
815	1.732729e+09	2016-08-31 15:10:07 +0000
818	4.196984e+09	2016-06-04 23:31:25 +0000
822	4.196984e+09	2016-06-10 00:39:48 +0000
826	4.196984e+09	2016-03-07 18:09:06 +0000
829	4.196984e+09	2016-02-18 02:24:13 +0000
833	4.196984e+09	2016-06-07 00:36:02 +0000
841	4.196984e+09	2016-08-01 01:28:46 +0000
847	4.196984e+09	2016-04-29 00:21:01 +0000
860	4.196984e+09	2015-12-06 00:17:55 +0000
868	4.196984e+09	2016-01-08 05:00:14 +0000
872	4.196984e+09	2016-03-20 23:23:54 +0000
885	1.950368e+08	2016-08-01 16:42:51 +0000
890	4.196984e+09	2016-06-05 19:47:03 +0000
895	4.196984e+09	2015-11-27 19:11:49 +0000
908	4.196984e+09	2015-12-21 22:15:18 +0000
911	2.804798e+08	2016-07-25 15:23:28 +0000
926	4.196984e+09	2015-12-22 04:35:49 +0000
937	4.196984e+09	2015-12-28 17:12:42 +0000
943	4.196984e+09	2016-01-03 05:11:12 +0000
949	4.196984e+09	2015-12-11 16:40:19 +0000
1012	4.196984e+09	2016-03-01 20:11:59 +0000
1023	4.196984e+09	2015-11-21 00:46:50 +0000
1043	4.196984e+09	2015-11-19 00:32:12 +0000
1242	7.832140e+05	2016-03-21 19:29:52 +0000
2259	4.296832e+09	2015-11-20 03:43:06 +0000

	expanded_urls	rating_numerator \
19	https://twitter.com/dog_rates/status/887473957...	13
32	https://twitter.com/dog_rates/status/886053434...	12
36	https://twitter.com/dog_rates/status/830583320...	13
68	https://twitter.com/dog_rates/status/878057613...	14
73	https://www.gofundme.com/3yd6y1c,https://twitt...	13
74	https://twitter.com/dog_rates/status/669000397...	11
78	https://twitter.com/rachel2195/status/87685077...	14
91	https://twitter.com/dog_rates/status/866334964...	12
95	https://twitter.com/dog_rates/status/868880397...	14
97	https://www.gofundme.com/help-my-baby-sierra-g...	12
101	https://twitter.com/loganamnosis/status/872657...	14
109	https://twitter.com/dog_rates/status/841077006...	12
118	https://twitter.com/dog_rates/status/859196978...	12
124	https://www.gofundme.com/3ti3nps,https://twitt...	12
130	https://twitter.com/rachaeleasler/status/86501...	13
132	https://twitter.com/dog_rates/status/866450705...	13
137	https://twitter.com/dog_rates/status/837820167...	12
146	https://www.gofundme.com/helpquinny,https://tw...	13
155	https://twitter.com/dog_rates/status/806629075...	13
159	https://www.gofundme.com/help-lorenzo-beat-can...	13
160	https://twitter.com/tallylott/status/860914485...	13
165	https://twitter.com/dog_rates/status/761672994...	10
171	https://twitter.com/dog_rates/status/839549326...	12
180	https://twitter.com/AaronChewning/status/85706...	13
182	https://twitter.com/dog_rates/status/844704788...	13
185	NaN	14
194	https://twitter.com/dog_rates/status/842163532...	12
195	https://twitter.com/frasercampbell_/status/855...	14
204	http://www.gofundme.com/bluethewhitehusky,http...	13
211	https://twitter.com/dog_rates/status/829374341...	13
...
784	https://twitter.com/dog_rates/status/740373189...	9
794	https://twitter.com/dog_rates/status/771380798...	11
800	https://twitter.com/dog_rates/status/765222098...	10
811	https://twitter.com/dog_rates/status/673320132...	11
815	https://twitter.com/katieornah/status/77100213...	12
818	https://twitter.com/dog_rates/status/739238157...	13
822	https://twitter.com/dog_rates/status/741067306...	12
826	https://vine.co/v/iXQAm5Lrgrh,https://vine.co/...	13
829	https://twitter.com/dog_rates/status/700143752...	10
833	https://twitter.com/dog_rates/status/739979191...	12
841	https://twitter.com/dog_rates/status/759923798...	10
847	https://twitter.com/dog_rates/status/725842289...	12
860	https://twitter.com/dog_rates/status/673295268...	8
868	https://twitter.com/dog_rates/status/685325112...	10

872	https://twitter.com/dog_rates/status/711694788...	13
885	https://weratedogs.com/pages/about-us,https://...	11
890	https://twitter.com/dog_rates/status/739544079...	10
895	https://twitter.com/dog_rates/status/670319130...	11
908	https://twitter.com/dog_rates/status/679062614...	11
911	https://twitter.com/jon_hill1987/status/7575971...	11
926	https://twitter.com/dog_rates/status/679158373...	11
937	https://twitter.com/dog_rates/status/681523177...	12
943	https://vine.co/v/ibvnzrauFuV,https://vine.co/...	11
949	https://twitter.com/dog_rates/status/675354435...	13
1012	https://twitter.com/dog_rates/status/704761120...	13
1023	https://twitter.com/dog_rates/status/667866724...	10
1043	https://twitter.com/dog_rates/status/667138269...	10
1242	https://twitter.com/twitter/status/71199827977...	12
2259	https://twitter.com/dogratingrating/status/667...	12
2260	https://twitter.com/dogratingrating/status/667...	5

	rating_denominator	name	doggo	floofer	pupper	puppo
19	10	Canela	None	None	None	None
32	10	None	None	None	None	None
36	10	Lilly	None	None	None	None
68	10	Emmy	None	None	None	None
73	10	Shadow	None	None	None	None
74	10	Terrance	None	None	None	None
78	10	None	None	None	pupper	None
91	10	Coco	None	None	None	None
95	10	Walter	None	None	None	None
97	10	Sierra	None	None	pupper	None
101	10	None	None	None	None	None
109	10	Dawn	None	None	None	None
118	10	quite	None	None	None	None
124	10	Cooper	None	None	None	None
130	10	None	None	None	None	None
132	10	Jamesy	None	None	pupper	None
137	10	None	None	None	pupper	None
146	10	Quinn	None	None	None	None
155	10	None	None	None	None	None
159	10	Lorenzo	None	None	None	None
160	10	None	None	None	None	None
165	10	None	None	None	None	None
171	10	Winston	None	None	None	None
180	10	None	None	None	None	None
182	10	Luna	None	None	None	None
185	10	None	None	None	None	None
194	10	George	None	None	None	None
195	10	None	None	None	None	None
204	10	None	None	None	None	None
211	10	Astrid	doggo	None	None	None

...
784	11	None	None	None	None	None
794	10	Fizz	None	None	None	None
800	10	Gromit	None	None	None	None
811	10	Frankie	None	None	None	None
815	10	None	None	None	pupper	None
818	10	None	doggo	None	None	None
822	10	just	doggo	None	pupper	None
826	10	None	None	None	None	None
829	10	None	None	None	pupper	None
833	10	Nollie	None	None	None	None
841	10	None	None	None	None	None
847	10	Colby	None	None	None	None
860	10	Eve	None	None	pupper	None
868	10	None	None	None	None	None
872	10	None	None	None	None	None
885	10	None	None	None	None	None
890	10	None	None	None	None	None
895	10	None	None	None	None	None
908	10	Chompsky	None	None	None	None
911	10	None	None	None	pupper	None
926	10	Rubio	None	None	None	None
937	10	Carly	None	None	None	None
943	10	None	None	None	None	None
949	10	None	None	None	None	None
1012	10	None	None	None	pupper	None
1023	10	Shaggy	None	None	None	None
1043	10	None	None	None	None	None
1242	10	None	None	None	None	None
2259	10	None	None	None	None	None
2260	10	None	None	None	None	None

[181 rows x 17 columns]

In [38]: *#identify reply tweets from df_archive*

`df_archive[df_archive['in_reply_to_status_id'].notnull()]`

Out[38]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
30	886267009285017600	8.862664e+17	2.281182e+09	
55	881633300179243008	8.816070e+17	4.738443e+07	
64	879674319642796034	8.795538e+17	3.105441e+09	
113	870726314365509632	8.707262e+17	1.648776e+07	
148	863427515083354112	8.634256e+17	7.759620e+07	
149	863079547188785154	6.671522e+17	4.196984e+09	
179	857214891891077121	8.571567e+17	1.806710e+08	
184	856526610513747968	8.558181e+17	4.196984e+09	
186	856288084350160898	8.562860e+17	2.792810e+08	
188	855862651834028034	8.558616e+17	1.943518e+08	

189	855860136149123072	8.558585e+17	1.361572e+07
218	850333567704068097	8.503288e+17	2.195506e+07
228	848213670039564288	8.482121e+17	4.196984e+09
234	847617282490613760	8.476062e+17	4.196984e+09
251	844979544864018432	7.590995e+17	4.196984e+09
274	840698636975636481	8.406983e+17	8.405479e+17
290	838150277551247360	8.381455e+17	2.195506e+07
291	838085839343206401	8.380855e+17	2.894131e+09
313	835246439529840640	8.352460e+17	2.625958e+07
342	832088576586297345	8.320875e+17	3.058208e+07
346	831926988323639298	8.319030e+17	2.068372e+07
387	826598799820865537	8.265984e+17	4.196984e+09
409	823333489516937216	8.233264e+17	1.582854e+09
427	821153421864615936	8.211526e+17	1.132119e+08
498	813130366689148928	8.131273e+17	4.196984e+09
513	811647686436880384	8.116272e+17	4.196984e+09
565	802265048156610565	7.331095e+17	4.196984e+09
570	801854953262350336	8.018543e+17	1.185634e+07
576	800859414831898624	8.008580e+17	2.918590e+08
611	797165961484890113	7.971238e+17	2.916630e+07
...
1479	693582294167244802	6.935722e+17	1.198989e+09
1497	692423280028966913	6.924173e+17	4.196984e+09
1501	692142790915014657	6.920419e+17	4.196984e+09
1523	690607260360429569	6.903413e+17	4.670367e+08
1598	686035780142297088	6.860340e+17	4.196984e+09
1605	685681090388975616	6.855479e+17	4.196984e+09
1618	684969860808454144	6.849598e+17	4.196984e+09
1630	684538444857667585	6.844811e+17	4.196984e+09
1634	684225744407494656	6.842229e+17	4.196984e+09
1663	682808988178739200	6.827884e+17	4.196984e+09
1689	681340665377193984	6.813394e+17	4.196984e+09
1774	678023323247357953	6.780211e+17	4.196984e+09
1819	676590572941893632	6.765883e+17	4.196984e+09
1842	675870721063669760	6.757073e+17	4.196984e+09
1844	675849018447167488	6.758457e+17	4.196984e+09
1852	675707330206547968	6.754971e+17	4.196984e+09
1866	675349384339542016	6.749998e+17	4.196984e+09
1882	674999807681908736	6.747934e+17	4.196984e+09
1885	674793399141146624	6.717299e+17	4.196984e+09
1892	674754018082705410	6.747522e+17	4.196984e+09
1895	674742531037511680	6.747400e+17	4.196984e+09
1905	674606911342424069	6.744689e+17	4.196984e+09
1914	674330906434379776	6.658147e+17	1.637468e+07
1940	673716320723169284	6.737159e+17	4.196984e+09
2036	671729906628341761	6.715610e+17	4.196984e+09
2038	671550332464455680	6.715449e+17	4.196984e+09
2149	669684865554620416	6.693544e+17	4.196984e+09

2169	669353438988365824	6.678065e+17	4.196984e+09
2189	668967877119254528	6.689207e+17	2.143566e+07
2298	667070482143944705	6.670655e+17	4.196984e+09

	timestamp \
30	2017-07-15 16:51:35 +0000
55	2017-07-02 21:58:53 +0000
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113	2017-06-02 19:38:25 +0000
148	2017-05-13 16:15:35 +0000
149	2017-05-12 17:12:53 +0000
179	2017-04-26 12:48:51 +0000
184	2017-04-24 15:13:52 +0000
186	2017-04-23 23:26:03 +0000
188	2017-04-22 19:15:32 +0000
189	2017-04-22 19:05:32 +0000
218	2017-04-07 13:04:55 +0000
228	2017-04-01 16:41:12 +0000
234	2017-03-31 01:11:22 +0000
251	2017-03-23 18:29:57 +0000
274	2017-03-11 22:59:09 +0000
290	2017-03-04 22:12:52 +0000
291	2017-03-04 17:56:49 +0000
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346	2017-02-15 18:03:45 +0000
387	2017-02-01 01:11:25 +0000
409	2017-01-23 00:56:15 +0000
427	2017-01-17 00:33:26 +0000
498	2016-12-25 21:12:41 +0000
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565	2016-11-25 21:37:47 +0000
570	2016-11-24 18:28:13 +0000
576	2016-11-22 00:32:18 +0000
611	2016-11-11 19:55:50 +0000
...	...
1479	2016-01-30 23:51:19 +0000
1497	2016-01-27 19:05:49 +0000
1501	2016-01-27 00:31:15 +0000
1523	2016-01-22 18:49:36 +0000
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1618	2016-01-07 05:28:35 +0000
1630	2016-01-06 00:54:18 +0000
1634	2016-01-05 04:11:44 +0000
1663	2016-01-01 06:22:03 +0000
1689	2015-12-28 05:07:27 +0000
1774	2015-12-19 01:25:31 +0000

1819 2015-12-15 02:32:17 +0000
 1842 2015-12-13 02:51:51 +0000
 1844 2015-12-13 01:25:37 +0000
 1852 2015-12-12 16:02:36 +0000
 1866 2015-12-11 16:20:15 +0000
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 1885 2015-12-10 03:30:58 +0000
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 2038 2015-12-01 04:44:10 +0000
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 2169 2015-11-25 03:14:30 +0000
 2189 2015-11-24 01:42:25 +0000
 2298 2015-11-18 20:02:51 +0000

source \

30 <a href="http://twitter.com/download/iphone" r...
 55 <a href="http://twitter.com/download/iphone" r...
 64 <a href="http://twitter.com/download/iphone" r...
 113 <a href="http://twitter.com/download/iphone" r...
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 513 <a href="http://twitter.com/download/iphone" r...
 565 <a href="http://twitter.com/download/iphone" r...
 570 <a href="http://twitter.com/download/iphone" r...

576 <a href="http://twitter.com/download/iphone" r...
 611 <a href="http://twitter.com/download/iphone" r...
 ...
 1479 <a href="http://twitter.com/download/iphone" r...
 1497 <a href="http://twitter.com/download/iphone" r...
 1501 <a href="http://twitter.com/download/iphone" r...
 1523 <a href="http://twitter.com/download/iphone" r...
 1598 <a href="http://twitter.com/download/iphone" r...
 1605 <a href="http://twitter.com/download/iphone" r...
 1618 <a href="http://twitter.com/download/iphone" r...
 1630 <a href="http://twitter.com/download/iphone" r...
 1634 <a href="http://twitter.com/download/iphone" r...
 1663 <a href="http://twitter.com/download/iphone" r...
 1689 <a href="http://twitter.com/download/iphone" r...
 1774 <a href="http://twitter.com/download/iphone" r...
 1819 <a href="http://twitter.com/download/iphone" r...
 1842 <a href="http://twitter.com/download/iphone" r...
 1844 <a href="http://twitter.com/download/iphone" r...
 1852 <a href="http://twitter.com/download/iphone" r...
 1866 <a href="http://twitter.com/download/iphone" r...
 1882 <a href="http://twitter.com/download/iphone" r...
 1885 <a href="http://twitter.com/download/iphone" r...
 1892 <a href="http://twitter.com/download/iphone" r...
 1895 <a href="http://twitter.com/download/iphone" r...
 1905 <a href="http://twitter.com/download/iphone" r...
 1914 <a href="http://twitter.com/download/iphone" r...
 1940 <a href="http://twitter.com/download/iphone" r...
 2036 <a href="http://twitter.com/download/iphone" r...
 2038 <a href="http://twitter.com/download/iphone" r...
 2149 <a href="http://twitter.com/download/iphone" r...
 2169 <a href="http://twitter.com/download/iphone" r...
 2189 <a href="http://twitter.com/download/iphone" r...
 2298 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
30	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
55	@roushfenway These are good dogs but 17/10 is ...	NaN
64	@RealKentMurphy 14/10 confirmed	NaN
113	@ComplicitOwl @ShopWeRateDogs >10/10 is res...	NaN
148	@Jack_Septic_Eye I'd need a few more pics to p...	NaN
149	Ladies and gentlemen... I found Pipsy. He may ...	NaN
179	@Marc_IRL pixelated af 12/10	NaN
184	THIS IS CHARLIE, MARK. HE DID JUST WANT TO SAY...	NaN
186	@xianmcguire @Jenna_Marbles Kardashians wouldn...	NaN
188	@dhmontgomery We also gave snoop dogg a 420/10...	NaN
189	@s8n You tried very hard to portray this good ...	NaN
218	@markhoppus MARK THAT DOG HAS SEEN AND EXPERIE...	NaN
228	Jerry just apuppologized to me. He said there ...	NaN

234	.@breaannanicolee PUPDATE: Cannon has a heart ...	NaN
251	PUPDATE: I'm proud to announce that Toby is 23...	NaN
274	@O_kelvin_0 >10/10 is reserved for puppos s...	NaN
290	@markhoppus 182/10	NaN
291	@bragg6of8 @Andy_Pace_ we are still looking fo...	NaN
313	@jonnyysun @Lin_Manuel ok jomny I know you're e...	NaN
342	@docmisterio account started on 11/15/15	NaN
346	@UNC can confirm 12/10	NaN
387	I was going to do 007/10, but the joke wasn't ...	NaN
409	@HistoryInPics 13/10	NaN
427	@imgur for a polar bear tho I'd say 13/10 is a...	NaN
498	I've been informed by multiple sources that th...	NaN
513	PUPDATE: I've been informed that Augie was act...	NaN
565	Like doggo, like pupper version 2. Both 11/10 ...	NaN
570	.@NBCSports OMG THE TINY HAT I'M GOING TO HAVE...	NaN
576	@SkyWilliams doggo simply protecting you from ...	NaN
611	@JODYHiGHIROLLER it may be an 11/10 but what do...	NaN
...
1479	Personally I'd give him an 11/10. Not sure why...	NaN
1497	PUPDATE: just noticed this dog has some extra ...	NaN
1501	These are some pictures of Teddy that further ...	NaN
1523	12/10 @LightningHolt	NaN
1598	Yes I do realize a rating of 4/20 would've bee...	NaN
1605	Jack deserves another round of applause. If yo...	NaN
1618	For those who claim this is a goat, u are wron...	NaN
1630	After watching this video, we've determined th...	NaN
1634	Two sneaky puppers were not initially seen, mo...	NaN
1663	I'm aware that I could've said 20/16, but here...	NaN
1689	I've been told there's a slight possibility he...	NaN
1774	After getting lost in Reese's eyes for several...	NaN
1819	After some outrage from the crowd. Bubbles is ...	NaN
1842	& this is Yoshi. Another world record cont...	NaN
1844	This dog is being demoted to a 9/10 for not we...	NaN
1852	We've got ourselves a battle here. Watch out R...	NaN
1866	Yea I lied. Here's more. All 13/10 https://t.c...	NaN
1882	Ok last one of these. I may try to make some m...	NaN
1885	I have found another. 13/10 https://t.co/HwroP...	NaN
1892	Just received another perfect photo of dogs an...	NaN
1895	Some clarification is required. The dog is sin...	NaN
1905	The 13/10 also takes into account this impecca...	NaN
1914	13/10\n@ABC7	NaN
1940	The millennials have spoken and we've decided ...	NaN
2036	I'm just going to leave this one here as well...	NaN
2038	After 22 minutes of careful deliberation this ...	NaN
2149	After countless hours of research and hundreds...	NaN
2169	This is Tessa. She is also very pleased after ...	NaN
2189	12/10 good shit Bubka\n@wane15	NaN
2298	After much debate this dog is being upgraded t...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
30	NaN	NaN
55	NaN	NaN
64	NaN	NaN
113	NaN	NaN
148	NaN	NaN
149	NaN	NaN
179	NaN	NaN
184	NaN	NaN
186	NaN	NaN
188	NaN	NaN
189	NaN	NaN
218	NaN	NaN
228	NaN	NaN
234	NaN	NaN
251	NaN	NaN
274	NaN	NaN
290	NaN	NaN
291	NaN	NaN
313	NaN	NaN
342	NaN	NaN
346	NaN	NaN
387	NaN	NaN
409	NaN	NaN
427	NaN	NaN
498	NaN	NaN
513	NaN	NaN
565	NaN	NaN
570	NaN	NaN
576	NaN	NaN
611	NaN	NaN
...
1479	NaN	NaN
1497	NaN	NaN
1501	NaN	NaN
1523	NaN	NaN
1598	NaN	NaN
1605	NaN	NaN
1618	NaN	NaN
1630	NaN	NaN
1634	NaN	NaN
1663	NaN	NaN
1689	NaN	NaN
1774	NaN	NaN
1819	NaN	NaN
1842	NaN	NaN
1844	NaN	NaN

1852	NaN	NaN
1866	NaN	NaN
1882	NaN	NaN
1885	NaN	NaN
1892	NaN	NaN
1895	NaN	NaN
1905	NaN	NaN
1914	NaN	NaN
1940	NaN	NaN
2036	NaN	NaN
2038	NaN	NaN
2149	NaN	NaN
2169	NaN	NaN
2189	NaN	NaN
2298	NaN	NaN

	expanded_urls	rating_numerator \
30	NaN	12
55	NaN	17
64	NaN	14
113	NaN	10
148	NaN	12
149	https://twitter.com/dog_rates/status/863079547...	14
179	NaN	12
184	https://twitter.com/dog_rates/status/856526610...	14
186	NaN	14
188	NaN	420
189	NaN	666
218	NaN	13
228	NaN	11
234	NaN	13
251	https://twitter.com/dog_rates/status/844979544...	13
274	NaN	10
290	NaN	182
291	NaN	15
313	NaN	960
342	NaN	11
346	NaN	12
387	NaN	7
409	NaN	13
427	NaN	13
498	NaN	12
513	NaN	11
565	https://twitter.com/dog_rates/status/802265048...	11
570	NaN	11
576	NaN	11
611	NaN	11
...

1479	NaN	11
1497	NaN	9
1501	https://twitter.com/dog_rates/status/692142790...	13
1523	NaN	12
1598	NaN	4
1605	NaN	14
1618	NaN	5
1630	https://twitter.com/dog_rates/status/684538444...	12
1634	https://twitter.com/dog_rates/status/684225744...	143
1663	NaN	20
1689	NaN	5
1774	NaN	13
1819	NaN	7
1842	https://twitter.com/dog_rates/status/675870721...	11
1844	NaN	9
1852	https://twitter.com/dog_rates/status/675707330...	11
1866	https://twitter.com/dog_rates/status/675349384...	13
1882	https://twitter.com/dog_rates/status/674999807...	13
1885	https://twitter.com/dog_rates/status/674793399...	13
1892	https://twitter.com/dog_rates/status/674754018...	12
1895	NaN	11
1905	NaN	13
1914	NaN	13
1940	NaN	1
2036	https://twitter.com/dog_rates/status/671729906...	13
2038	NaN	1
2149	NaN	11
2169	https://twitter.com/dog_rates/status/669353438...	10
2189	NaN	12
2298	NaN	10

	rating_denominator	name	doggo	floofer	pupper	puppo
30	10	None	None	None	None	None
55	10	None	None	None	None	None
64	10	None	None	None	None	None
113	10	None	None	None	None	None
148	10	None	None	None	None	None
149	10	None	None	None	None	None
179	10	None	None	None	None	None
184	10	None	None	None	None	None
186	10	None	None	None	None	None
188	10	None	None	None	None	None
189	10	None	None	None	None	None
218	10	None	None	None	None	None
228	10	None	None	None	None	None
234	10	None	None	None	None	None
251	10	None	None	None	None	None
274	10	None	None	None	None	None

290	10	None	None	None	None	None
291	10	None	None	None	None	None
313	0	None	None	None	None	None
342	15	None	None	None	None	None
346	10	None	None	None	None	None
387	10	None	None	None	None	None
409	10	None	None	None	None	None
427	10	None	None	None	None	None
498	10	None	None	None	None	None
513	10	None	None	None	None	None
565	10	None	doggo	None	pupper	None
570	10	None	None	None	None	None
576	10	None	doggo	None	None	None
611	10	None	None	None	None	None
...
1479	10	None	None	None	None	None
1497	10	None	None	None	None	None
1501	10	None	None	None	None	None
1523	10	None	None	None	None	None
1598	20	None	None	None	None	None
1605	10	None	None	None	None	None
1618	10	None	None	None	None	None
1630	10	None	None	None	None	None
1634	130	None	None	None	None	None
1663	16	None	None	None	None	None
1689	10	None	None	None	None	None
1774	10	None	None	None	None	None
1819	10	None	None	None	None	None
1842	10	None	None	None	None	None
1844	10	None	None	None	None	None
1852	10	None	None	None	None	None
1866	10	None	None	None	None	None
1882	10	None	None	None	None	None
1885	10	None	None	None	None	None
1892	10	None	None	None	None	None
1895	10	None	None	None	None	None
1905	10	None	None	None	None	None
1914	10	None	None	None	None	None
1940	10	None	None	None	None	None
2036	10	None	None	None	None	None
2038	10	None	None	None	None	None
2149	10	None	None	None	None	None
2169	10	Tessa	None	None	None	None
2189	10	None	None	None	None	None
2298	10	None	None	None	None	None

[78 rows x 17 columns]

```
In [39]: #check the values in the source column
df_archive.source.value_counts()
```

```
Out[39]: <a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhone</a>
<a href="http://vine.co" rel="nofollow">Vine - Make a Scene</a>
<a href="http://twitter.com" rel="nofollow">Twitter Web Client</a>
<a href="https://about.twitter.com/products/tweetdeck" rel="nofollow">TweetDeck</a>
Name: source, dtype: int64
```

```
In [40]: df_archive.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2356 non-null int64
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2356 non-null object
source                  2356 non-null object
text                    2356 non-null object
retweeted_status_id      181 non-null float64
retweeted_status_user_id 181 non-null float64
retweeted_status_timestamp 181 non-null object
expanded_urls           2297 non-null object
rating_numerator         2356 non-null int64
rating_denominator       2356 non-null int64
name                    2356 non-null object
doggo                   2356 non-null object
floofer                 2356 non-null object
pupper                  2356 non-null object
puppo                   2356 non-null object
dtypes: float64(4), int64(3), object(10)
memory usage: 313.0+ KB
```

1.3.1 Quality

1. The following columns have incorrect data types: df_archive table: tweet_id, in_reply_to_status_id, in_reply_to_user_id, retweeted_status_id, retweeted_status_user_id. These have numeric values(int/float) instead of string type. The 'timestamp' and 'retweeted_status_timestamp' columns have datatype 'string' instead of 'datetime'. df_predictions table: incorrect data types: tweet_id (int instead of string) df_tweetdata table: incorrect data types: tweet_id (int instead of string)
2. The date and time information is contained in a single column.
3. The 'name' column in df_archive table has several 'None' or incorrect values.
4. The 'expanded_urls' column in df_archive table has several 'None' values (no images present).

5. The 'rating_denominator' column in df_archive table has several incorrect values, including 0.
6. The df_archive and df_tweetdata tables contain retweets and replies.
7. The 'source' column in df_archive table has information that is difficult to read and comprehend.
8. The 'text' column in df_archive is truncated, so the full information in this column cannot be read.
9. The name of the 'id' column in the df_tweetdata table is inconsistent with the column name 'tweet_id' in the other two dataframes.

1.3.2 Tidiness

1. There are 4 different columns for dog stage instead of a single column.
2. The columns not relevant for our analysis may be removed.
3. The three datasets can be merged into a single dataset according to the rules of tidy data.

1.4 Cleaning Data

```
In [41]: # Make copies of the original datasets
df_archive_clean = df_archive.copy()
df_predictions_clean = df_predictions.copy()
df_tweetdata_clean = df_tweetdata.copy()
```

We now start cleaning the data to address the quality and tidiness issues identified above.

Define Rename the 'id' column in the df_tweetdata_clean to 'tweet_id' for consistency with the other two dataframes

Code

```
In [42]: # rename column 'id' to 'tweet_id' in df_tweetdata_clean
df_tweetdata_clean.rename(columns={'id': 'tweet_id'}, inplace=True)
```

Test

```
In [43]: df_tweetdata_clean.head()
```

```
Out[43]:
```

	tweet_id	retweet_count	retweeted	retweeted_status	\
0	892420643555336193	8291	False	NaN	
1	892177421306343426	6123	False	NaN	
2	891815181378084864	4054	False	NaN	
3	891689557279858688	8428	False	NaN	
4	891327558926688256	9130	False	NaN	

favorite_count

0	37950
1	32588
2	24536
3	41291
4	39478

Define Merge the three datasets into a single dataset according to the rules of tidy data.

Code

```
In [44]: df_final = pd.merge(df_archive_clean, df_predictions_clean, on = 'tweet_id', how = 'inner')
df_final = pd.merge(df_final, df_tweetdata_clean, on = 'tweet_id', how = 'inner')
```

Test

```
In [45]: df_final.head()
```

```
Out[45]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	

	timestamp	\
0	2017-08-01 16:23:56 +0000	
1	2017-08-01 00:17:27 +0000	
2	2017-07-31 00:18:03 +0000	
3	2017-07-30 15:58:51 +0000	
4	2017-07-29 16:00:24 +0000	

	source	\
0	<a href="http://twitter.com/download/iphone" r...	
1	<a href="http://twitter.com/download/iphone" r...	
2	<a href="http://twitter.com/download/iphone" r...	
3	<a href="http://twitter.com/download/iphone" r...	
4	<a href="http://twitter.com/download/iphone" r...	

	text	retweeted_status_id	\
0	This is Phineas. He's a mystical boy. Only eve...	NaN	
1	This is Tilly. She's just checking pup on you...	NaN	
2	This is Archie. He is a rare Norwegian Pouncin...	NaN	
3	This is Darla. She commenced a snooze mid meal...	NaN	
4	This is Franklin. He would like you to stop ca...	NaN	

	retweeted_status_user_id	retweeted_status_timestamp	\
0	NaN	NaN	
1	NaN	NaN	

2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

	expanded_urls	...	\
0	https://twitter.com/dog_rates/status/892420643...	...	
1	https://twitter.com/dog_rates/status/892177421...	...	
2	https://twitter.com/dog_rates/status/891815181...	...	
3	https://twitter.com/dog_rates/status/891689557...	...	
4	https://twitter.com/dog_rates/status/891327558...	...	

	p2	p2_conf	p2_dog		p3	p3_conf	\
0	bagel	0.085851	False		banana	0.076110	
1	Pekinese	0.090647	True		papillon	0.068957	
2	malamute	0.078253	True		kelpie	0.031379	
3	Labrador_retriever	0.168086	True		spatula	0.040836	
4	English_springer	0.225770	True	German_short-haired_pointer		0.175219	

	p3_dog	retweet_count	retweeted	retweeted_status	favorite_count
0	False	8291	False	NaN	37950
1	True	6123	False	NaN	32588
2	True	4054	False	NaN	24536
3	False	8428	False	NaN	41291
4	True	9130	False	NaN	39478

[5 rows x 32 columns]

In [46]: df_final.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2065 entries, 0 to 2064
Data columns (total 32 columns):
tweet_id                2065 non-null int64
in_reply_to_status_id   23 non-null float64
in_reply_to_user_id     23 non-null float64
timestamp                2065 non-null object
source                  2065 non-null object
text                    2065 non-null object
retweeted_status_id      75 non-null float64
retweeted_status_user_id 75 non-null float64
retweeted_status_timestamp 75 non-null object
expanded_urls            2065 non-null object
rating_numerator         2065 non-null int64
rating_denominator       2065 non-null int64
name                    2065 non-null object
doggo                   2065 non-null object
floofer                 2065 non-null object
pupper                  2065 non-null object
```

```

puppo                2065 non-null object
jpg_url              2065 non-null object
img_num              2065 non-null int64
p1                   2065 non-null object
p1_conf              2065 non-null float64
p1_dog               2065 non-null bool
p2                   2065 non-null object
p2_conf              2065 non-null float64
p2_dog               2065 non-null bool
p3                   2065 non-null object
p3_conf              2065 non-null float64
p3_dog               2065 non-null bool
retweet_count        2065 non-null int64
retweeted             2065 non-null bool
retweeted_status      75 non-null object
favorite_count        2065 non-null int64
dtypes: bool(4), float64(7), int64(6), object(15)
memory usage: 475.9+ KB

```

Define Merge the 4 different columns for dog stage to a single column.

Code

```

In [47]: # merge the columns 'doggo', 'floofer', 'pupper', 'puppo' into a single column called 'stage'
df_final['stage'] = df_final[df_final.columns[13:17]].apply(
    lambda x: ','.join(x.astype(str)),
    axis=1
)

```

```

In [48]: df_final['stage'].value_counts()

```

```

Out[48]: None, None, None, None      1744
None, None, pupper, None             211
doggo, None, None, None               67
None, None, None, puppo               23
doggo, None, pupper, None             11
None, floofer, None, None              7
doggo, None, None, puppo               1
doggo, floofer, None, None             1
Name: stage, dtype: int64

```

```

In [49]: #replace the column values as seen from the above value_counts results by removing multiple
df_final['stage']
replace_dict = dict([
    ('None, None, None, None', 'None'), ('None, None, pupper, None', 'pupper'), ('doggo, None, None, None', 'doggo'),
    ('None, None, None, puppo', 'puppo'), ('doggo, None, pupper, None', 'doggo, pupper'), ('doggo, floofer, None, None', 'floofer'),
    ('doggo, None, None, puppo', 'doggo, puppo'), ('doggo, floofer, None, None', 'floofer')
])

```

```

    ])

df_final=df_final.replace(replace_dict, regex=True)

# delete the separate dog stage columns from the dataset
columns_to_drop = ['doggo', 'floofer', 'pupper', 'puppo']
df_final = df_final.drop(columns_to_drop, axis=1)

#set the datatype of the 'stage' column as category
df_final['stage'] = df_final.stage.astype('category')

```

Test

```
In [50]: df_final.sample(5)
```

```

Out[50]:
      tweet_id  in_reply_to_status_id  in_reply_to_user_id \
722    759197388317847553             NaN                 NaN
1028   712065007010385924             NaN                 NaN
330    822462944365645825             NaN                 NaN
1480   678991772295516161             NaN                 NaN
1628   674410619106390016             NaN                 NaN

      timestamp \
722  2016-07-30 01:22:17 +0000
1028 2016-03-21 23:55:01 +0000
330   2017-01-20 15:17:01 +0000
1480 2015-12-21 17:33:48 +0000
1628 2015-12-09 02:09:56 +0000

      source \
722  <a href="http://twitter.com/download/iphone" r...
1028 <a href="http://twitter.com/download/iphone" r...
330   <a href="http://twitter.com/download/iphone" r...
1480 <a href="http://twitter.com/download/iphone" r...
1628 <a href="http://twitter.com/download/iphone" r...

      text  retweeted_status_id \
722  This is Luna. She's just heckin precious af I ...      NaN
1028 This is Chester. He's clearly in charge of the...      NaN
330   This is Gabe. He was the unequivocal embodimen...      NaN
1480 If your Monday isn't going so well just take a...      NaN
1628 This is Lenny. He wants to be a sprinkler. 10/...      NaN

      retweeted_status_user_id  retweeted_status_timestamp \
722                          NaN                          NaN
1028                         NaN                          NaN
330                          NaN                          NaN
1480                         NaN                          NaN

```

1628		NaN		NaN		
------	--	-----	--	-----	--	--

				expanded_urls	...	p2_conf \
722				https://twitter.com/dog_rates/status/759197388...	...	0.076899
1028				https://twitter.com/dog_rates/status/712065007...	...	0.084253
330				https://twitter.com/dog_rates/status/822462944...	...	0.023056
1480				https://twitter.com/dog_rates/status/678991772...	...	0.187003
1628				https://twitter.com/dog_rates/status/674410619...	...	0.046475

	p2_dog		p3	p3_conf	p3_dog	retweet_count	retweeted \
722	True	white_wolf		0.063269	False	2097	False
1028	False		pizza	0.080168	False	662	False
330	True	Maltese_dog		0.008945	True	15884	False
1480	True	Chihuahua		0.101420	True	1251	False
1628	False		beagle	0.019427	True	482	False

		retweeted_status	favorite_count	stage
722		NaN	6383	None
1028		NaN	2353	None
330		NaN	30137	pupper
1480		NaN	2420	None
1628		NaN	1225	None

[5 rows x 29 columns]

In [51]: df_final.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2065 entries, 0 to 2064
Data columns (total 29 columns):
tweet_id                2065 non-null int64
in_reply_to_status_id    23 non-null float64
in_reply_to_user_id      23 non-null float64
timestamp               2065 non-null object
source                  2065 non-null object
text                    2065 non-null object
retweeted_status_id       75 non-null float64
retweeted_status_user_id  75 non-null float64
retweeted_status_timestamp 75 non-null object
expanded_urls            2065 non-null object
rating_numerator          2065 non-null int64
rating_denominator        2065 non-null int64
name                     2065 non-null object
jpg_url                  2065 non-null object
img_num                  2065 non-null int64
p1                       2065 non-null object
p1_conf                  2065 non-null float64
p1_dog                   2065 non-null bool
```



```

p2                2065 non-null object
p2_conf           2065 non-null float64
p2_dog            2065 non-null bool
p3                2065 non-null object
p3_conf           2065 non-null float64
p3_dog            2065 non-null bool
retweet_count     2065 non-null int64
retweeted         2065 non-null bool
retweeted_status   75 non-null object
favorite_count    2065 non-null int64
stage             2065 non-null category
dtypes: bool(4), category(1), float64(7), int64(6), object(11)
memory usage: 413.8+ KB

```

Define Convert the incorrect datatypes into strings for the following columns

Code

```

In [52]: df_final['tweet_id']=df_final['tweet_id'].astype(str)
         df_final['in_reply_to_status_id']=df_final['in_reply_to_status_id'].astype(str)
         df_final['in_reply_to_user_id']=df_final['in_reply_to_user_id'].astype(str)
         df_final['retweeted_status_id']=df_final['retweeted_status_id'].astype(str)
         df_final['retweeted_status_user_id']=df_final['retweeted_status_user_id'].astype(str)

```

Test

```

In [53]: df_final.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2065 entries, 0 to 2064
Data columns (total 29 columns):
tweet_id                2065 non-null object
in_reply_to_status_id   2065 non-null object
in_reply_to_user_id     2065 non-null object
timestamp               2065 non-null object
source                 2065 non-null object
text                   2065 non-null object
retweeted_status_id     2065 non-null object
retweeted_status_user_id 2065 non-null object
retweeted_status_timestamp 75 non-null object
expanded_urls           2065 non-null object
rating_numerator        2065 non-null int64
rating_denominator      2065 non-null int64
name                   2065 non-null object
jpg_url                2065 non-null object
img_num                2065 non-null int64
p1                     2065 non-null object

```

```

p1_conf                2065 non-null float64
p1_dog                 2065 non-null bool
p2                     2065 non-null object
p2_conf                2065 non-null float64
p2_dog                 2065 non-null bool
p3                     2065 non-null object
p3_conf                2065 non-null float64
p3_dog                 2065 non-null bool
retweet_count          2065 non-null int64
retweeted              2065 non-null bool
retweeted_status        75 non-null object
favorite_count         2065 non-null int64
stage                  2065 non-null category
dtypes: bool(4), category(1), float64(3), int64(5), object(16)
memory usage: 413.8+ KB

```

Define Convert the 'timestamp' and 'retweeted_status_timestamp' columns into datetime datatypes, and split into two separate columns for date and time.

Code

```

In [54]: from datetime import datetime

#convert datatypes to datetime
date_time = pd.to_datetime(df_final.timestamp, errors='coerce')
retweeted_date_time = pd.to_datetime(df_finalretweeted_status_timestamp, errors='coerce')

#split each column into separate date and time columns
df_final = df_final.assign(tweet_date=date_time.dt.date, tweet_time=date_time.dt.time)
df_final = df_final.assign(retweeted_status_date=retweeted_date_time.dt.date, retweeted_s

#delete the original columns
datecolumns_to_drop=['timestamp', 'retweeted_status_timestamp']
df_final = df_final.drop(datecolumns_to_drop, axis=1)

```

Test

```
In [56]: df_final.sample(5)
```

```

Out[56]:
      tweet_id  in_reply_to_status_id  in_reply_to_user_id \
573    783347506784731136                nan            nan
162    852912242202992640                nan            nan
1210   696754882863349760                nan            nan
224    839239871831150596                nan            nan
1119   704761120771465216                nan            nan

```

source \

```

573 <a href="http://twitter.com/download/iphone" r...
162 <a href="http://twitter.com/download/iphone" r...
1210 <a href="http://twitter.com/download/iphone" r...
224 <a href="http://twitter.com/download/iphone" r...
1119 <a href="http://twitter.com/download/iphone" r...

```

```

text \
573 RT @dog_rates: This is Kenny. He just wants to...
162 Meet Benny. He likes being adorable and making...
1210 This is Reptar. He specifically asked for his ...
224 This is Odie. He's big. 13/10 would attempt to...
1119 This pupper killed this great white in an epic...

```

```

retweeted_status_id retweeted_status_user_id \
573 6.742918370630533e+17 4196983835.0
162 nan nan
1210 nan nan
224 nan nan
1119 nan nan

```

```

expanded_urls rating_numerator \
573 https://twitter.com/dog_rates/status/674291837... 11
162 https://www.gofundme.com/bennys-medical-bills,... 12
1210 https://twitter.com/dog_rates/status/696754882... 10
224 https://twitter.com/dog_rates/status/839239871... 13
1119 https://twitter.com/dog_rates/status/704761120... 13

```

```

rating_denominator ... p3_dog retweet_count \
573 10 ... True 6102
162 10 ... True 1908
1210 10 ... True 377
224 10 ... True 6969
1119 10 ... True 3074

```

```

retweeted retweeted_status \
573 False {'created_at': 'Tue Dec 08 18:17:56 +0000 2015...
162 False NaN
1210 False NaN
224 False NaN
1119 False NaN

```

```

favorite_count stage tweet_date tweet_time retweeted_status_date \
573 0 None 2016-10-04 16:46:14 2015-12-08
162 9271 None 2017-04-14 15:51:39 NaT
1210 1543 None 2016-02-08 17:58:03 NaT
224 28400 None 2017-03-07 22:22:32 NaT
1119 6913 pupper 2016-03-01 20:11:59 NaT

```

	retweeted_status_time
573	18:17:56
162	NaT
1210	NaT
224	NaT
1119	NaT

[5 rows x 31 columns]

Define Remove the retweets from the dataset

Code

```
In [57]: df_final = df_final[df_final['retweeted_status_id']=='nan']
```

Test

```
In [58]: df_final['retweeted_status_id'].value_counts()
```

```
Out[58]: nan    1990
         Name: retweeted_status_id, dtype: int64
```

Define Remove the reply tweets from the dataset

Code

```
In [59]: df_final = df_final[df_final['in_reply_to_status_id']=='nan']
```

Test

```
In [60]: df_final['in_reply_to_status_id'].value_counts()
```

```
Out[60]: nan    1967
         Name: in_reply_to_status_id, dtype: int64
```

```
In [473]: df_final.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1967 entries, 0 to 2064
Data columns (total 31 columns):
tweet_id                1967 non-null object
in_reply_to_status_id   1967 non-null object
in_reply_to_user_id     1967 non-null object
source                  1967 non-null object
text                    1967 non-null object
retweeted_status_id     1967 non-null object
retweeted_status_user_id 1967 non-null object
expanded_urls           1967 non-null object
rating_numerator        1967 non-null int64
```

```

rating_denominator      1967 non-null int64
name                    1967 non-null object
jpg_url                 1967 non-null object
img_num                 1967 non-null int64
p1                      1967 non-null object
p1_conf                 1967 non-null float64
p1_dog                  1967 non-null bool
p2                      1967 non-null object
p2_conf                 1967 non-null float64
p2_dog                  1967 non-null bool
p3                      1967 non-null object
p3_conf                 1967 non-null float64
p3_dog                  1967 non-null bool
retweet_count           1967 non-null int64
retweeted                1967 non-null bool
retweeted_status         0 non-null object
favorite_count           1967 non-null int64
stage                   1967 non-null object
tweet_date              1967 non-null object
tweet_time              1967 non-null object
retweeted_status_date    0 non-null object
retweeted_status_time    0 non-null object
dtypes: bool(4), float64(3), int64(5), object(19)
memory usage: 438.0+ KB

```

Define Check and remove the following columns as they do not contain any relevant information for our analysis: `in_reply_to_status_id`, `in_reply_to_user_id`, `retweeted_status_id`, `retweeted_status_user_id`, `retweeted_status`, `retweeted_status_date`, `retweeted_status_time`

Code

```
In [61]: df_final['in_reply_to_status_id'].value_counts()
```

```
Out[61]: nan      1967
         Name: in_reply_to_status_id, dtype: int64
```

```
In [62]: df_final['in_reply_to_user_id'].value_counts()
```

```
Out[62]: nan      1967
         Name: in_reply_to_user_id, dtype: int64
```

```
In [484]: df_final['retweeted'].value_counts()
```

```
Out[484]: False      1967
         Name: retweeted, dtype: int64
```

```
In [477]: df_final['retweeted_status_id'].value_counts()
```

```

Out[477]: nan      1967
          Name: retweeted_status_id, dtype: int64

In [478]: df_final['retweeted_status_user_id'].value_counts()

Out[478]: nan      1967
          Name: retweeted_status_user_id, dtype: int64

In [479]: df_final['retweeted_status'].value_counts()

Out[479]: Series([], Name: retweeted_status, dtype: int64)

In [480]: df_final['retweeted_status_date'].value_counts()

Out[480]: Series([], Name: retweeted_status_date, dtype: int64)

In [481]: df_final['retweeted_status_time'].value_counts()

Out[481]: Series([], Name: retweeted_status_time, dtype: int64)

In [63]: columns_to_be_dropped = ['in_reply_to_status_id', 'in_reply_to_user_id', 'retweeted', 'ret
df_final = df_final.drop(columns_to_be_dropped, axis=1)

```

Test

```

In [64]: df_final.sample(5)

Out[64]:
          tweet_id                                     source \
182  848324959059550208  <a href="http://twitter.com/download/iphone" r...
1719 672272411274932228  <a href="http://twitter.com/download/iphone" r...
1982 667491009379606528  <a href="http://twitter.com" rel="nofollow">Tw...
1840 670319130621435904  <a href="http://twitter.com/download/iphone" r...
90    871762521631449091  <a href="http://twitter.com/download/iphone" r...

          text \
182  Meet Odin. He's supposed to be giving directio...
1719 This is Pippin. He is terrified of his new lit...
1982 Two dogs in this one. Both are rare Jujitsu Py...
1840 AT DAWN...\nWE RIDE\n\n11/10 https://t.co/Qnf0...
90   This is Rover. As part of pupper protocol he h...

          expanded_urls  rating_numerator \
182  https://twitter.com/dog_rates/status/848324959...      12
1719 https://twitter.com/dog_rates/status/672272411...      11
1982 https://twitter.com/dog_rates/status/667491009...       7
1840 https://twitter.com/dog_rates/status/670319130...      11
90   https://twitter.com/dog_rates/status/871762521...      12

          rating_denominator  name \
182                10      Odin

```

1719	10	Pippin					
1982	10	None					
1840	10	None					
90	10	Rover					

	jpg_url	p1	\
182	https://pbs.twimg.com/media/C8XbDR1WAAAxND8.jpg	malamute	
1719	https://pbs.twimg.com/media/CVRkLuJWUAAhhYp.jpg	pug	
1982	https://pbs.twimg.com/media/CUNniSlUYAEj1Jl.jpg	borzoi	
1840	https://pbs.twimg.com/media/CU1zsMSUAAASOqW.jpg	Irish_terrier	
90	https://pbs.twimg.com/media/DBkfy58XcAEdzZy.jpg	Labrador_retriever	

	p1_conf	...	p2_conf	p2_dog	p3	\
182	0.544576	...	0.290268	True	Eskimo_dog	
1719	0.914685	...	0.014982	True	Siamese_cat	
1982	0.852088	...	0.132264	False	weasel	
1840	0.254856	...	0.227716	True	soft-coated_wheaten_terrier	
90	0.921393	...	0.064608	True	bloodhound	

	p3_conf	p3_dog	retweet_count	favorite_count	stage	tweet_date	\
182	0.154421	True	4824	25537	None	2017-04-02	
1719	0.009221	False	3478	6538	None	2015-12-03	
1982	0.005730	False	231	533	None	2015-11-19	
1840	0.223263	True	1247	3905	None	2015-11-27	
90	0.003383	True	3438	19985	pupper	2017-06-05	

	tweet_time
182	00:03:26
1719	04:33:27
1982	23:53:52
1840	19:11:49
90	16:15:56

[5 rows x 22 columns]

Define Convert the source column into readable text

```
In [65]: df_final['source'].value_counts()
```

```
Out[65]: <a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhone</a>
<a href="http://twitter.com" rel="nofollow">Twitter Web Client</a>
<a href="https://about.twitter.com/products/tweetdeck" rel="nofollow">TweetDeck</a>
Name: source, dtype: int64
```

Code

```
In [66]: replace_source_dict = dict([
        ('<a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhon
```

```

    ])
df_final=df_final.replace(replace_source_dict)

```

Test

```
In [67]: df_final.sample(5)
```

```

Out[67]:
      tweet_id      source \
48  881906580714921986  Twitter for iPhone
361  817827839487737858  Twitter for iPhone
1448  680221482581123072  Twitter for iPhone
1284  690005060500217858  Twitter for iPhone
2045  666273097616637952  Twitter for iPhone

      text \
48  This is Rey. He's a Benebop Cumberfloof. 12/10...
361  This is Buddy. He ran into a glass door once. ...
1448  This is CeCe. She's patiently waiting for Sant...
1284  "I'm the only one that ever does anything in t...
2045  Can take selfies 11/10 https://t.co/ws2AMaWpPW

      expanded_urls  rating_numerator \
48  https://twitter.com/dog_rates/status/881906580...      12
361  https://twitter.com/dog_rates/status/817827839...      13
1448  https://twitter.com/dog_rates/status/680221482...      10
1284  https://twitter.com/dog_rates/status/690005060...      10
2045  https://twitter.com/dog_rates/status/666273097...      11

      rating_denominator  name \
48              10      Rey
361              10    Buddy
1448              10     CeCe
1284              10     None
2045              10     None

      jpg_url      p1 \
48  https://pbs.twimg.com/media/DDOpWm9XcAAeSBL.jpg      Weimaraner
361  https://pbs.twimg.com/ext_tw_video_thumb/81782...      cocker_spaniel
1448  https://pbs.twimg.com/media/CXCh0QZW8AALdXm.jpg      bubble
1284  https://pbs.twimg.com/media/CZNj8N-WQAMXASZ.jpg      Samoyed
2045  https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg  Italian_greyhound

      p1_conf  ...      p2_conf  p2_dog      p3      p3_conf  p3_dog \
48  0.291539  ...      0.278966  True      koala  0.127017  False
361  0.387608  ...      0.264844  True  Pekinese  0.122123   True
1448  0.240173  ...      0.146222  False    abaya  0.139342  False
1284  0.270287  ...      0.114027  True     teddy  0.072475  False
2045  0.176053  ...      0.111884  True   basenji  0.111152   True

```


	retweet_count	favorite_count	stage	tweet_date	tweet_time
48	3315	23822	None	2017-07-03	16:04:48
361	29567	55035	None	2017-01-07	20:18:46
1448	306	1519	None	2015-12-25	03:00:14
1284	1794	3838	None	2016-01-21	02:56:40
2045	76	171	None	2015-11-16	15:14:19

[5 rows x 22 columns]

Define Rearrange the columns according to our requirement

Code

```
In [68]: cols = list(df_final.columns.values)
        cols
```

```
Out[68]: ['tweet_id',
          'source',
          'text',
          'expanded_urls',
          'rating_numerator',
          'rating_denominator',
          'name',
          'jpg_url',
          'p1',
          'p1_conf',
          'p1_dog',
          'p2',
          'p2_conf',
          'p2_dog',
          'p3',
          'p3_conf',
          'p3_dog',
          'retweet_count',
          'favorite_count',
          'stage',
          'tweet_date',
          'tweet_time']
```

```
In [69]: df_final = df_final[['tweet_id', 'tweet_date', 'tweet_time', 'name', 'stage', 'retweet_count',
                              'rating_denominator', 'source', 'expanded_urls', 'jpg_url', \
                              'p1', 'p1_conf', 'p1_dog', 'p2', 'p2_conf', 'p2_dog', 'p3', 'p3_conf', 'p3_dog']
```

Test

```
In [70]: df_final.sample(5)
```

```

Out[70]:
      tweet_id  tweet_date  tweet_time      name stage  retweet_count \
1761  671511350426865664  2015-12-01  02:09:16  Hammond  None          741
1667  673662677122719744  2015-12-07  00:37:52  Kendall  None          370
46    882268110199369728  2017-07-04  16:01:23    Alfy  None       11327
1309  688547210804498433  2016-01-17  02:23:42   Frönq  None          744
1825  670435821946826752  2015-11-28  02:55:30    None  None          529

      favorite_count  rating_numerator  rating_denominator \
1761             1629                8                10
1667             1274               12                10
46             44025               13                10
1309             2737                9                10
1825             1066               10                10

      source
1761  Twitter for iPhone
1667  Twitter for iPhone
46    Twitter for iPhone
1309  Twitter for iPhone
1825  Twitter for iPhone

      p1  p1_conf  p1_dog      p2  p2_conf \
1761  hermit_crab  0.625409  False      tick  0.127333
1667  Labrador_retriever  0.957670  True      beagle  0.012413
46    golden_retriever  0.762211  True  Labrador_retriever  0.098985
1309  papillon  0.531279  True    Blenheim_spaniel  0.214197
1825  sorrel  0.460370  False      basenji  0.135767

      p2_dog      p3  p3_conf  p3_dog \
1761  False      snail  0.097916  False
1667  True  golden_retriever  0.005689  True
46    True  cocker_spaniel  0.017199  True
1309  True  Border_collie  0.053840  True
1825  True  Cardigan  0.099174  True

      text
1761  Say hello to Hammond. He's just a wee lil pup...
1667  This is Kendall. 12/10 would cuddle the hell o...
46    This is Alfy. You're witnessing his first wate...
1309  This is Frönq. He got caught stealing a waffle...
1825  Two unbelievably athletic dogs here. Great for...

[5 rows x 22 columns]

```

Define Remove invalid names and replace by 'Not known'. We check the names that start with a lower case letter, these names can be assumed to be invalid and are replaced by 'Not known'. We also replace the names 'A' and 'None', by 'Not known'.

Code

```
In [71]: # identify the names starting with lowercase letter and convert them to a list.  
invalid_names = list(df_final[df_final['name'].str[0].str.islower()]['name'])  
invalid_names
```

```
Out[71]: ['such',  
          'a',  
          'quite',  
          'quite',  
          'one',  
          'incredibly',  
          'a',  
          'a',  
          'very',  
          'my',  
          'one',  
          'not',  
          'his',  
          'one',  
          'a',  
          'a',  
          'a',  
          'an',  
          'very',  
          'a',  
          'just',  
          'getting',  
          'very',  
          'this',  
          'unacceptable',  
          'all',  
          'a',  
          'a',  
          'infuriating',  
          'a',  
          'a',  
          'a',  
          'an',  
          'a',  
          'a',  
          'very',  
          'getting',  
          'just',  
          'a',  
          'the',  
          'the',  
          'actually',  
          'by',
```

'a',
'officially',
'a',
'the',
'the',
'a',
'a',
'a',
'a',
'a',
'one',
'a',
'a',
'a',
'light',
'just',
'space',
'a',
'the',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'an',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'a',
'quite',
'a',
'an',
'a',
'an',
'the',
'the',

```
'a',  
'a',  
'an',  
'a',  
'a',  
'a',  
'a']
```

```
In [72]: df_final['name'] = df_final['name'].replace(invalid_names, 'Not known')  
df_final['name'] = df_final['name'].replace('A', 'Not known')  
df_final['name'] = df_final['name'].replace('None', 'Not known')
```

Test

```
In [73]: df_final['name'].value_counts()
```

```
Out[73]: Not known      621  
Cooper      10  
Oliver      10  
Charlie     10  
Lucy        10  
Penny       9  
Tucker      9  
Winston     8  
Sadie       8  
Lola         7  
Toby        7  
Daisy       7  
Koda        6  
Bo          6  
Stanley     6  
Bella       6  
Jax         6  
Milo        5  
Buddy       5  
Chester     5  
Bailey      5  
Leo         5  
Louis       5  
Scout       5  
Oscar       5  
Rusty       5  
Dave        5  
Brody       4  
Finn        4  
Oakley      4  
...  
Bell        1
```

```

Sora          1
Murphy        1
Moofasa       1
Amélie        1
Fiji          1
Samsom        1
Clybe         1
Perry         1
Flash         1
Travis        1
Ron           1
Darby         1
Dale          1
Mack          1
Marvin        1
Sage          1
Emanuel       1
Crumpet       1
Nida          1
Joey          1
Rumpole       1
Traviss       1
Duddles       1
Ralphie       1
Snicku        1
Finnegus      1
Severus       1
Jiminus       1
Blakely       1
Name: name, Length: 912, dtype: int64

```

Define Format the 'text' column in the dataframe so that the information in this column can be read in full without being truncated.

Code

```
In [74]: #check the display width of the column and increase it accordingly so that the full text
pd.options.display.max_colwidth
```

```
Out[74]: 50
```

```
In [75]: pd.options.display.max_colwidth = 200
```

Test

```
In [76]: df_final
```

```
Out[76]:
```

	tweet_id	tweet_date	tweet_time	name	stage	\
0	892420643555336193	2017-08-01	16:23:56	Phineas	None	

1	892177421306343426	2017-08-01	00:17:27	Tilly	None
2	891815181378084864	2017-07-31	00:18:03	Archie	None
3	891689557279858688	2017-07-30	15:58:51	Darla	None
4	891327558926688256	2017-07-29	16:00:24	Franklin	None
5	891087950875897856	2017-07-29	00:08:17	Not known	None
6	890971913173991426	2017-07-28	16:27:12	Jax	None
7	890729181411237888	2017-07-28	00:22:40	Not known	None
8	890609185150312448	2017-07-27	16:25:51	Zoey	None
9	890240255349198849	2017-07-26	15:59:51	Cassie	doggo
10	890006608113172480	2017-07-26	00:31:25	Koda	None
11	889880896479866881	2017-07-25	16:11:53	Bruno	None
12	889665388333682689	2017-07-25	01:55:32	Not known	puppo
13	889638837579907072	2017-07-25	00:10:02	Ted	None
14	889531135344209921	2017-07-24	17:02:04	Stuart	puppo
15	889278841981685760	2017-07-24	00:19:32	Oliver	None
16	888917238123831296	2017-07-23	00:22:39	Jim	None
17	888804989199671297	2017-07-22	16:56:37	Zeke	None
18	888554962724278272	2017-07-22	00:23:06	Ralphus	None
19	888078434458587136	2017-07-20	16:49:33	Gerald	None
20	887705289381826560	2017-07-19	16:06:48	Jeffrey	None
21	887517139158093824	2017-07-19	03:39:09	Not known	None
22	887473957103951883	2017-07-19	00:47:34	Canela	None
23	887343217045368832	2017-07-18	16:08:03	Not known	None
24	887101392804085760	2017-07-18	00:07:08	Not known	None
25	886983233522544640	2017-07-17	16:17:36	Maya	None
26	886736880519319552	2017-07-16	23:58:41	Mingus	None
27	886680336477933568	2017-07-16	20:14:00	Derek	None
28	886366144734445568	2017-07-15	23:25:31	Roscoe	pupper
29	886258384151887873	2017-07-15	16:17:19	Waffles	None
...
2035	666411507551481857	2015-11-17	00:24:19	Not known	None
2036	666407126856765440	2015-11-17	00:06:54	Not known	None
2037	666396247373291520	2015-11-16	23:23:41	Not known	None
2038	666373753744588802	2015-11-16	21:54:18	Not known	None
2039	666362758909284353	2015-11-16	21:10:36	Not known	None
2040	666353288456101888	2015-11-16	20:32:58	Not known	None
2041	666345417576210432	2015-11-16	20:01:42	Not known	None
2042	666337882303524864	2015-11-16	19:31:45	Not known	None
2043	666293911632134144	2015-11-16	16:37:02	Not known	None
2044	666287406224695296	2015-11-16	16:11:11	Not known	None
2045	666273097616637952	2015-11-16	15:14:19	Not known	None
2046	666268910803644416	2015-11-16	14:57:41	Not known	None
2047	666104133288665088	2015-11-16	04:02:55	Not known	None
2048	666102155909144576	2015-11-16	03:55:04	Not known	None
2049	666099513787052032	2015-11-16	03:44:34	Not known	None
2050	666094000022159362	2015-11-16	03:22:39	Not known	None
2051	666082916733198337	2015-11-16	02:38:37	Not known	None
2052	666073100786774016	2015-11-16	01:59:36	Not known	None

2053	666071193221509120	2015-11-16	01:52:02	Not known	None
2054	666063827256086533	2015-11-16	01:22:45	Not known	None
2055	666058600524156928	2015-11-16	01:01:59	Not known	None
2056	666057090499244032	2015-11-16	00:55:59	Not known	None
2057	666055525042405380	2015-11-16	00:49:46	Not known	None
2058	666051853826850816	2015-11-16	00:35:11	Not known	None
2059	666050758794694657	2015-11-16	00:30:50	Not known	None
2060	666049248165822465	2015-11-16	00:24:50	Not known	None
2061	666044226329800704	2015-11-16	00:04:52	Not known	None
2062	666033412701032449	2015-11-15	23:21:54	Not known	None
2063	666029285002620928	2015-11-15	23:05:30	Not known	None
2064	666020888022790149	2015-11-15	22:32:08	Not known	None

	retweet_count	favorite_count	rating_numerator	rating_denominator	\
0	8291	37950	13	10	
1	6123	32588	13	10	
2	4054	24536	12	10	
3	8428	41291	13	10	
4	9130	39478	12	10	
5	3039	19839	13	10	
6	2010	11599	13	10	
7	18389	64022	13	10	
8	4176	27250	13	10	
9	7192	31269	14	10	
10	7147	30042	13	10	
11	4853	27249	13	10	
12	9787	47146	13	10	
13	4413	26608	12	10	
14	2193	14810	13	10	
15	5234	24738	13	10	
16	4393	28532	12	10	
17	4178	25053	13	10	
18	3456	19420	13	10	
19	3407	21313	12	10	
20	5258	29565	13	10	
21	11399	45347	14	10	
22	17726	67599	13	10	
23	10157	33001	13	10	
24	5814	29961	12	10	
25	7582	34432	13	10	
26	3197	11817	13	10	
27	4354	21978	13	10	
28	3126	20766	12	10	
29	6135	27428	13	10	
...	
2035	322	435	2	10	
2036	40	105	7	10	
2037	83	162	9	10	

2038	89	182	11	10
2039	561	760	6	10
2040	71	215	8	10
2041	132	286	10	10
2042	90	191	9	10
2043	344	489	3	10
2044	64	144	1	2
2045	76	171	11	10
2046	35	100	10	10
2047	6398	14052	1	10
2048	11	78	11	10
2049	66	151	8	10
2050	72	162	9	10
2051	44	114	6	10
2052	159	315	10	10
2053	58	143	9	10
2054	213	466	10	10
2055	57	109	8	10
2056	138	290	9	10
2057	235	429	10	10
2058	831	1200	2	10
2059	59	130	10	10
2060	43	106	5	10
2061	138	293	6	10
2062	43	123	9	10
2063	47	125	7	10
2064	499	2532	8	10

	source \
0	Twitter for iPhone
1	Twitter for iPhone
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	p1	p1_conf	p1_dog	p2 \
0	orange	0.097049	False	bagel
1	Chihuahua	0.323581	True	Pekinese
2	Chihuahua	0.716012	True	malamute
3	paper_towel	0.170278	False	Labrador_retriever
4	basset	0.555712	True	English_springer
5	Chesapeake_Bay_retriever	0.425595	True	Irish_terrier
6	Appenzeller	0.341703	True	Border_collie
7	Pomeranian	0.566142	True	Eskimo_dog
8	Irish_terrier	0.487574	True	Irish_setter
9	Pembroke	0.511319	True	Cardigan
10	Samoyed	0.957979	True	Pomeranian
11	French_bulldog	0.377417	True	Labrador_retriever
12	Pembroke	0.966327	True	Cardigan
13	French_bulldog	0.991650	True	boxer
14	golden_retriever	0.953442	True	Labrador_retriever
15	whippet	0.626152	True	borzoi
16	golden_retriever	0.714719	True	Tibetan_mastiff
17	golden_retriever	0.469760	True	Labrador_retriever
18	Siberian_husky	0.700377	True	Eskimo_dog
19	French_bulldog	0.995026	True	pug
20	basset	0.821664	True	redbone
21	limousine	0.130432	False	tow_truck
22	Pembroke	0.809197	True	Rhodesian_ridgeback
23	Mexican_hairless	0.330741	True	sea_lion
24	Samoyed	0.733942	True	Eskimo_dog
25	Chihuahua	0.793469	True	toy_terrier
26	kuvasz	0.309706	True	Great_Pyrenees
27	convertible	0.738995	False	sports_car
28	French_bulldog	0.999201	True	Chihuahua
29	pug	0.943575	True	shower_cap
...
2035	coho	0.404640	False	barracouta
2036	black-and-tan_coonhound	0.529139	True	bloodhound
2037	Chihuahua	0.978108	True	toy_terrier
2038	soft-coated_wheaten_terrier	0.326467	True	Afghan_hound
2039	guinea_pig	0.996496	False	skunk
2040	malamute	0.336874	True	Siberian_husky

2041	golden_retriever	0.858744	True	Chesapeake_Bay_retriever
2042	ox	0.416669	False	Newfoundland
2043	three-toed_sloth	0.914671	False	otter
2044	Maltese_dog	0.857531	True	toy_poodle
2045	Italian_greyhound	0.176053	True	toy_terrier
2046	desktop_computer	0.086502	False	desk
2047	hen	0.965932	False	cock
2048	English_setter	0.298617	True	Newfoundland
2049	Lhasa	0.582330	True	Shih-Tzu
2050	bloodhound	0.195217	True	German_shepherd
2051	pug	0.489814	True	bull_mastiff
2052	Walker_hound	0.260857	True	English_foxhound
2053	Gordon_setter	0.503672	True	Yorkshire_terrier
2054	golden_retriever	0.775930	True	Tibetan_mastiff
2055	miniature_poodle	0.201493	True	komondor
2056	shopping_cart	0.962465	False	shopping_basket
2057	chow	0.692517	True	Tibetan_mastiff
2058	box_turtle	0.933012	False	mud_turtle
2059	Bernese_mountain_dog	0.651137	True	English_springer
2060	miniature_pinscher	0.560311	True	Rottweiler
2061	Rhodesian_ridgeback	0.408143	True	redbone
2062	German_shepherd	0.596461	True	malinois
2063	redbone	0.506826	True	miniature_pinscher
2064	Welsh_springer_spaniel	0.465074	True	collie

	p2_conf	p2_dog		p3	p3_conf	p3_dog	\
0	0.085851	False		banana	0.076110	False	
1	0.090647	True		papillon	0.068957	True	
2	0.078253	True		kelpie	0.031379	True	
3	0.168086	True		spatula	0.040836	False	
4	0.225770	True	German_short-haired_pointer		0.175219	True	
5	0.116317	True	Indian_elephant		0.076902	False	
6	0.199287	True	ice_lolly		0.193548	False	
7	0.178406	True	Pembroke		0.076507	True	
8	0.193054	True	Chesapeake_Bay_retriever		0.118184	True	
9	0.451038	True	Chihuahua		0.029248	True	
10	0.013884	True	chow		0.008167	True	
11	0.151317	True	muzzle		0.082981	False	
12	0.027356	True	basenji		0.004633	True	
13	0.002129	True	Staffordshire_bullterrier		0.001498	True	
14	0.013834	True	redbone		0.007958	True	
15	0.194742	True	Saluki		0.027351	True	
16	0.120184	True	Labrador_retriever		0.105506	True	
17	0.184172	True	English_setter		0.073482	True	
18	0.166511	True	malamute		0.111411	True	
19	0.000932	True	bull_mastiff		0.000903	True	
20	0.087582	True	Weimaraner		0.026236	True	
21	0.029175	False	shopping_cart		0.026321	False	

22	0.054950	True	beagle	0.038915	True
23	0.275645	False	Weimaraner	0.134203	True
24	0.035029	True	Staffordshire_bullterrier	0.029705	True
25	0.143528	True	can_opener	0.032253	False
26	0.186136	True	Dandie_Dinmont	0.086346	True
27	0.139952	False	car_wheel	0.044173	False
28	0.000361	True	Boston_bull	0.000076	True
29	0.025286	False	Siamese_cat	0.002849	False
...
2035	0.271485	False	gar	0.189945	False
2036	0.244220	True	flat-coated_retriever	0.173810	True
2037	0.009397	True	papillon	0.004577	True
2038	0.259551	True	briard	0.206803	True
2039	0.002402	False	hamster	0.000461	False
2040	0.147655	True	Eskimo_dog	0.093412	True
2041	0.054787	True	Labrador_retriever	0.014241	True
2042	0.278407	True	groenendael	0.102643	True
2043	0.015250	False	great_grey_owl	0.013207	False
2044	0.063064	True	miniature_poodle	0.025581	True
2045	0.111884	True	basenji	0.111152	True
2046	0.085547	False	bookcase	0.079480	False
2047	0.033919	False	partridge	0.000052	False
2048	0.149842	True	borzoi	0.133649	True
2049	0.166192	True	Dandie_Dinmont	0.089688	True
2050	0.078260	True	malinois	0.075628	True
2051	0.404722	True	French_bulldog	0.048960	True
2052	0.175382	True	Ibizan_hound	0.097471	True
2053	0.174201	True	Pekinese	0.109454	True
2054	0.093718	True	Labrador_retriever	0.072427	True
2055	0.192305	True	soft-coated_wheaten_terrier	0.082086	True
2056	0.014594	False	golden_retriever	0.007959	True
2057	0.058279	True	fur_coat	0.054449	False
2058	0.045885	False	terrapin	0.017885	False
2059	0.263788	True	Greater_Swiss_Mountain_dog	0.016199	True
2060	0.243682	True	Doberman	0.154629	True
2061	0.360687	True	miniature_pinscher	0.222752	True
2062	0.138584	True	bloodhound	0.116197	True
2063	0.074192	True	Rhodesian_ridgeback	0.072010	True
2064	0.156665	True	Shetland_sheepdog	0.061428	True

0 This is Phineas. He's a

1 This is Tilly. She's just checking pup on you. Hopes you're doing ok. If not,

2 This is Archie. He is a rare Norwegian Pouncing Corgo. Lives

3 This is Darla. She

4 This is Franklin. He would like you to stop calling him "cute." He is a very

5 Here we have a majestic great white breaching off South Africa's coast. Absol

6 Meet Jax. He enjoys ice cream so much he gets nervous around it. 13/10 help Jax e

7 When you watch your owner call another dog a good boy but
 8 This is Zoey. She doesn't want to be one of the scary sharks.
 9 This is Cassie. She is a college pup. Studying international doggo commu
 10 This is Koda. He is a South Australian deckshark. Deceptively deadly.
 11 This is Bruno. He is a service shark. Only get
 12 Here's a puppo that seems to be on the fence
 13 This is Ted. He does his best.
 14 This is Stuart. He's sporting his favorite fanny pack. Se
 15 This is Oliver. You're witnessing one of his many brutal attacks. Seems to be
 16 This is Jim. He found a f
 17 This is Zeke. He has a new stick. Very proud of it. Would like you
 18 This is Ralphus. He's powe
 19 This is Gerald. He was just told he didn't get the job he interview
 20 This is Jeffrey. He has a monopoly on the pool noodles. Currently
 21 I've yet to rate a Venezuelan Hover Wiener. Thi
 22 This is Canela. She attempted some fan
 23 You may not have known you
 24 This... is a Jubilant Antarctic House Bear. We only rate dogs. Pleas
 25 This is Maya. She's very shy. Rarely lea
 26 This is Mingus. He's a wonderful father to his smol pup. Conf
 27 This is De
 28 This is Roscoe. Another pupper fallen victim to spontaneous tongue eje
 29 This is Waffles. His doggles are
 ...
 2035 This is quite the dog. Gets really excited when not in
 2036 This is a southern Vesuvius bumblegruff. Can drive a t
 2037 Oh goodness. A super rare northeast Qdoba kangaroo m
 2038
 2039 Unique dog here. Very small. Lives in container of Fros
 2040 Here we have a mixed Asiago from the Galápagos Isl
 2041 Look at this jokester think
 2042 This is an extremely rare horned Parthenon. Not amused
 2043 This is a funny dog. Weird toes. Won't come down. Lov
 2044 This is an Albanian 3 1/2 legged Episcopalian. Lov
 2045
 2046
 2047 Not familiar with this breed. No tail (weird). On
 2048 Oh my. Here you are seeing an Adobe Setter
 2049 Can stand on stump for what seems like a while. Built t
 2050 This appears to be a Mongolian Presbyterian mix
 2051 Here we have a well-established sunblock
 2052 Let's hope this flight isn't Malaysian (lol). What a
 2053 Here we have a northern speckled Rhododendron. Much
 2054 This is the happiest d
 2055 Here is the Rand Paul of retrievers folks! He's pr
 2056 My oh my. This is a rare blond Canadian
 2057 Here is a Siberian heavily armored polar bear mix. Stro
 2058 This is an odd dog. Hard on the outside but loving on

2059	This is a truly beautiful English Wilson Staff retrieve
2060	Here we have a 1949 1st generation
2061	This is a purebred Piers Morgan. Loves to Netflix an
2062	Here is a very happy pup. Big fan of well-mai
2063	This is a western brown Mitsubishi terrier. Upset abou
2064	Here we have a Japanese Irish Setter. Lost eye

[1967 rows x 22 columns]

```
In [77]: df_final.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1967 entries, 0 to 2064
Data columns (total 22 columns):
tweet_id          1967 non-null object
tweet_date        1967 non-null object
tweet_time        1967 non-null object
name              1967 non-null object
stage             1967 non-null category
retweet_count     1967 non-null int64
favorite_count    1967 non-null int64
rating_numerator  1967 non-null int64
rating_denominator 1967 non-null int64
source            1967 non-null object
expanded_urls     1967 non-null object
jpg_url           1967 non-null object
p1                1967 non-null object
p1_conf           1967 non-null float64
p1_dog            1967 non-null bool
p2                1967 non-null object
p2_conf           1967 non-null float64
p2_dog            1967 non-null bool
p3                1967 non-null object
p3_conf           1967 non-null float64
p3_dog            1967 non-null bool
text              1967 non-null object
dtypes: bool(3), category(1), float64(3), int64(4), object(11)
memory usage: 300.0+ KB
```

1.5 Storing the cleaned dataframe

```
In [78]: # Store the clean DataFrame to a csv file
df_final.to_csv('twitter_archive_master.csv')
```

1.6 Analysis and Visualisation

Identify the 10 most common dog names


```
In [79]: #df_final['name'].value_counts()
pd.DataFrame(df_final['name'].value_counts().head(11))
```

```
Out[79]:
```

	name
	Not known 621
	Cooper 10
	Oliver 10
	Charlie 10
	Lucy 10
	Penny 9
	Tucker 9
	Winston 8
	Sadie 8
	Lola 7
	Toby 7

Statistical Summary table

Find the average dog rating, average favourite count and average retweet count

```
In [80]: df_final.describe()
```

```
Out[80]:
```

	retweet_count	favorite_count	rating_numerator	rating_denominator \
count	1967.000000	1967.000000	1967.000000	1967.000000
mean	2660.608541	8765.766650	12.225724	10.478393
std	4736.170250	12837.312318	41.676305	6.860210
min	11.000000	78.000000	0.000000	2.000000
25%	591.000000	1882.000000	10.000000	10.000000
50%	1281.000000	3951.000000	11.000000	10.000000
75%	3038.500000	10934.000000	12.000000	10.000000
max	83352.000000	163892.000000	1776.000000	170.000000

	p1_conf	p2_conf	p3_conf
count	1967.000000	1.967000e+03	1.967000e+03
mean	0.594131	1.346345e-01	6.023006e-02
std	0.272163	1.010457e-01	5.096530e-02
min	0.044333	1.011300e-08	1.740170e-10
25%	0.362880	5.339800e-02	1.608055e-02
50%	0.587372	1.174020e-01	4.947150e-02
75%	0.847139	1.955655e-01	9.158675e-02
max	1.000000	4.880140e-01	2.734190e-01

Thus the average (mean) rating across the whole dataset is found to be 12.226, average favourite count is 8765.77 and average retweet count is 2660.61

Dog stages count Plot a bar chart to show the number of dogs according to the different dog stages - Pupper, Doggo, Puppo, Floofer, Doggo-pupper and Doggo-puppo

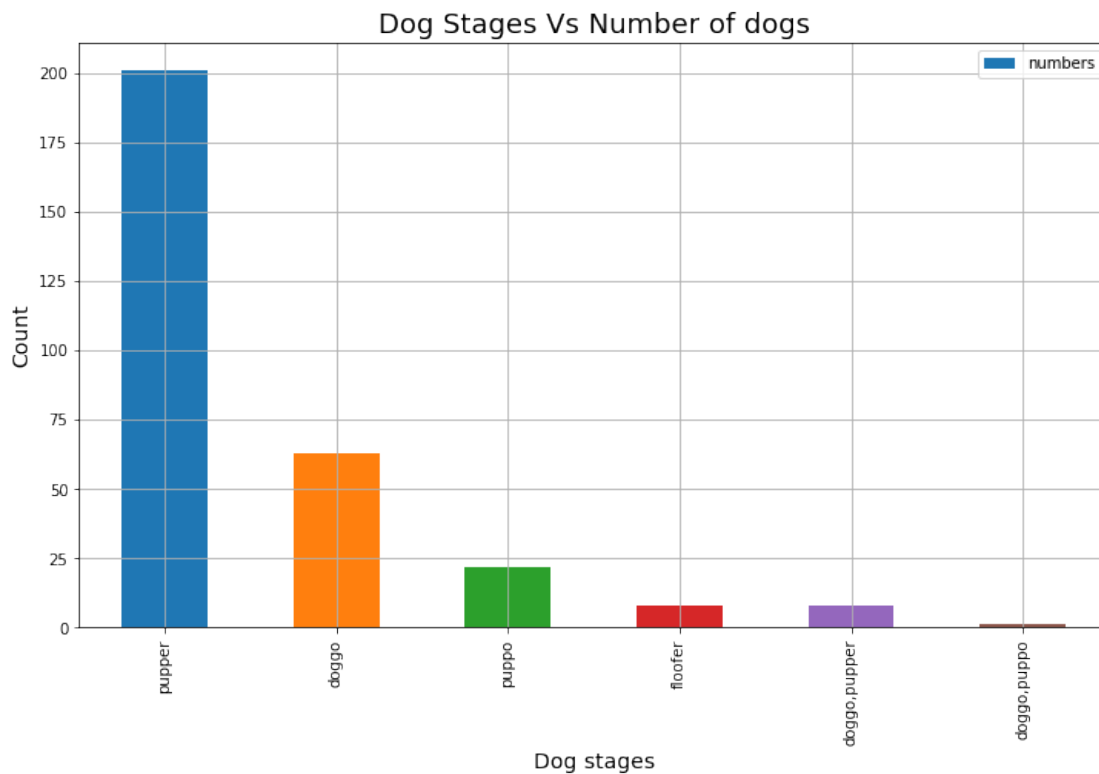
```
In [85]: count_stage = pd.DataFrame(df_final['stage'].value_counts())
count_stage = count_stage.reset_index()
```

```
count_stage.columns = ['stage', 'numbers']
count_stage
```

```
Out [85]:
```

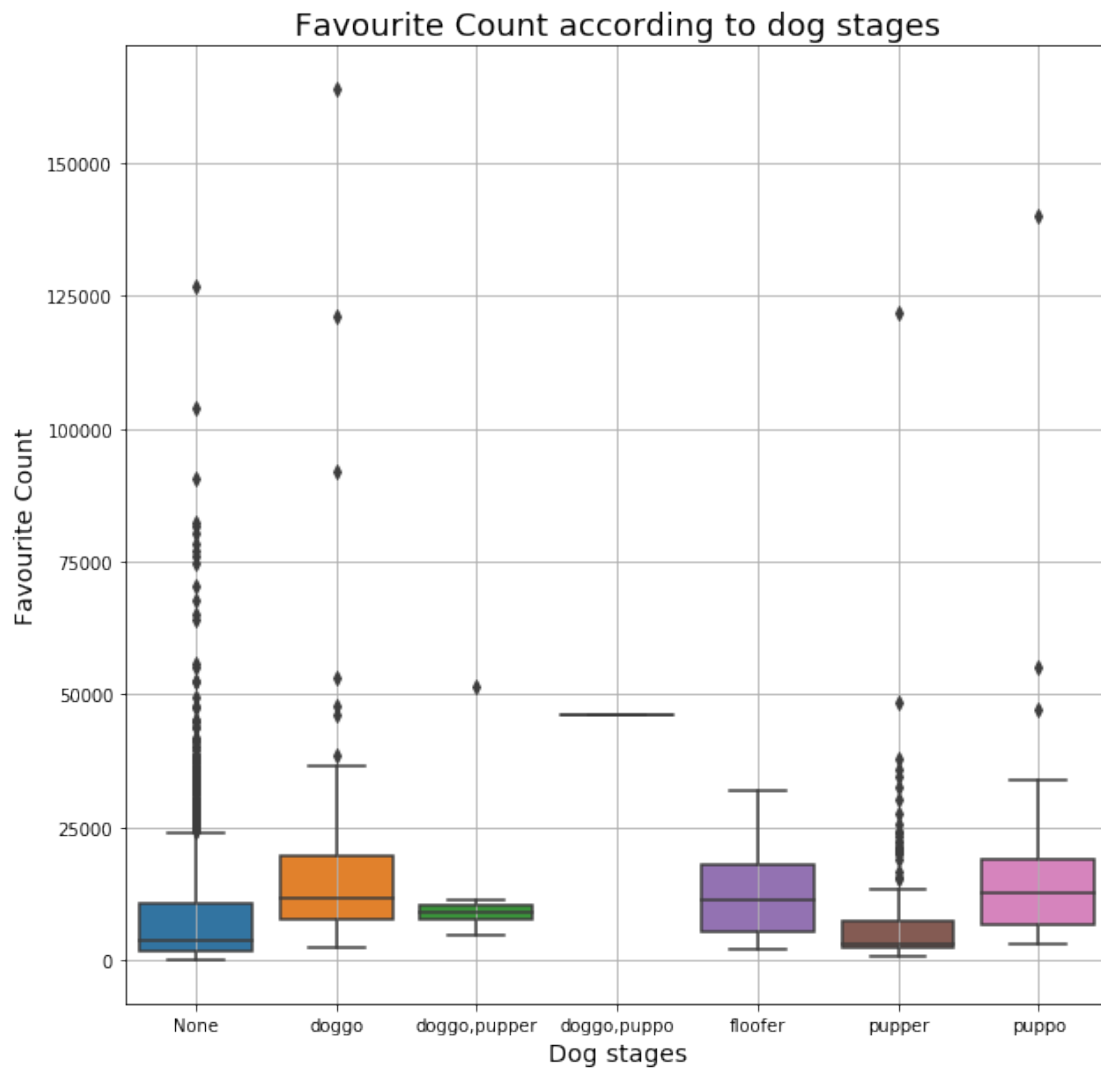
	stage	numbers
0	None	1664
1	pupper	201
2	doggo	63
3	puppo	22
4	floofer	8
5	doggo, pupper	8
6	doggo, puppo	1

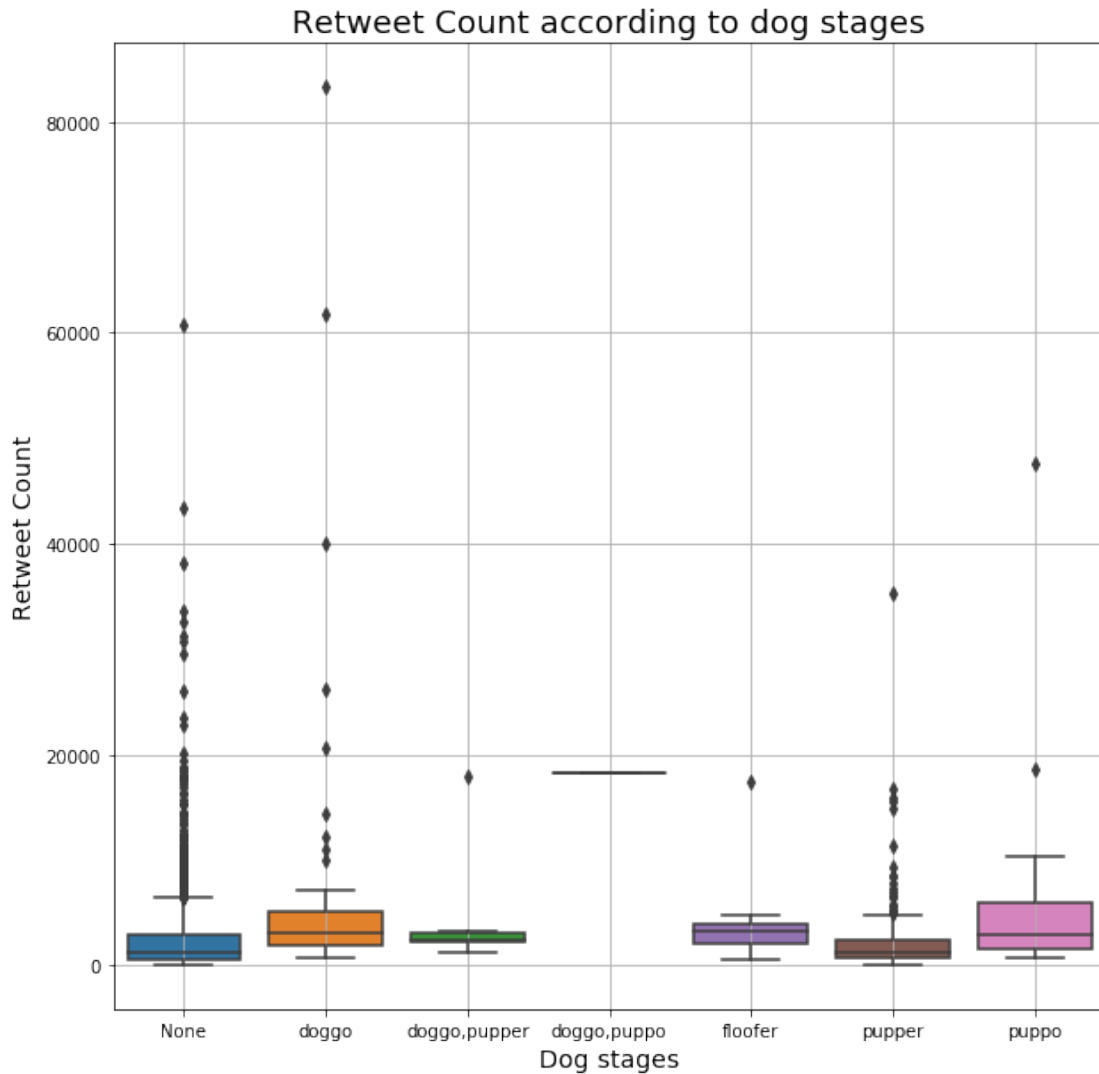
```
In [86]: count_stage = count_stage.drop(0) #remove the observations without any stages
count_stage.plot(kind='bar',x='stage',y='numbers',figsize=(12, 7))
plt.ylabel('Count',fontsize=14)
plt.xlabel('Dog stages',fontsize=14)
plt.title('Dog Stages Vs Number of dogs',fontsize=18)
plt.grid(True)
plt.show()
```



Analysis of favourite_count and retweets according to dog stages We plot a boxplot to show the favourite_count and retweets according to the different dog stages

```
In [87]: fig1=sns.boxplot(x=df_final.stage, y=df_final.favorite_count)
plt.gcf().set_size_inches(10,10)
plt.ylabel('Favourite Count',fontsize=14)
plt.xlabel('Dog stages',fontsize=14)
plt.title('Favourite Count according to dog stages',fontsize=18)
plt.grid(True)
plt.show()
fig2=sns.boxplot(x=df_final.stage, y=df_final.retweet_count)
plt.gcf().set_size_inches(10, 10)
plt.ylabel('Retweet Count',fontsize=14)
plt.xlabel('Dog stages',fontsize=14)
plt.title('Retweet Count according to dog stages',fontsize=18)
plt.grid(True)
plt.show()
```



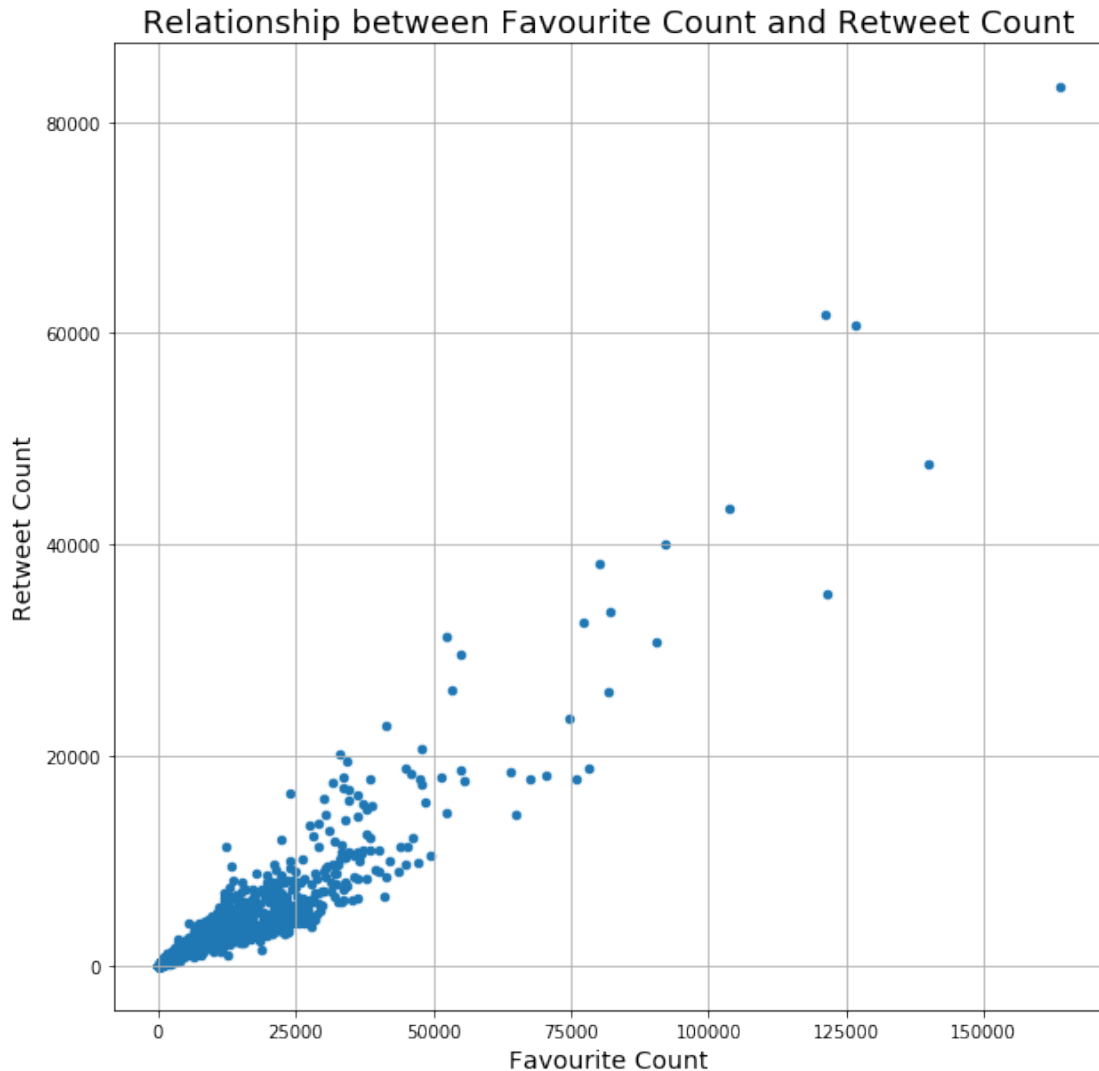


Thus it appears that although the puppers have the highest population, their average favourite_counts and retweet_counts are lower compared to the other stage groups. The average number of retweets for floofers are the highest, whereas for average favourite count, puppos and floofers lead the rankings.

Favourite Count and Retweet Count We will analyse the relation between the favourite_count and the retweet_count by plotting a scatterplot. We will then validate our findings by calculating the correlation coefficient.

```
In [88]: df_final.plot(kind='scatter',x='favorite_count',y='retweet_count',figsize=(10,10));
plt.ylabel('Retweet Count',fontsize=14);
plt.xlabel('Favourite Count',fontsize=14);
plt.title('Relationship between Favourite Count and Retweet Count',fontsize=18);
plt.grid(True)
plt.show()
```

```
correlation_coefficient = df_final['favorite_count'].corr(df_final['retweet_count'])
print('Correlation_coefficient between favorite_count and retweet_count is {}'.format(c
```



```
Correlation_coefficient between favorite_count and retweet_count is 0.9296357865320213
```

From the above scatterplot we can observe that there is a strong positive correlation between the favorite_count and the retweet_count. Thus the tweets with higher favourite count also have a higher retweet count, and vice versa. This is further validated by the value of Correlation coefficient which is found to be **0.93**

Find the dog with the highest rating From the statistical summary table, the highest rating is found to be 1776.

```
In [89]: df_final[df_final['rating_numerator']==1776]
```

```

Out[89]:          tweet_id  tweet_date  tweet_time      name stage  retweet_count  \
796  749981277374128128  2016-07-04   15:00:45  Atticus  None             2623

          favorite_count  rating_numerator  rating_denominator      source  \
796                5364                1776                10  TweetDeck

          ...
796          ...

          p1   p1_conf  p1_dog          p2   p2_conf  p2_dog          p3  \
796  bow_tie  0.533941  False  sunglasses  0.080822  False  sunglass

          p3_conf  p3_dog  \
796  0.050776  False

          text
796  This is Atticus. He's quite simply America af. 1776/10 https://t.co/GRXwMxLBkh

[1 rows x 22 columns]

```

```

In [90]: url_highest = df_final[df_final['rating_numerator']==1776]['jpg_url']
url_highest

```

```

Out[90]: 796      https://pbs.twimg.com/media/CmgBZ7kWcAAIzFD.jpg
Name: jpg_url, dtype: object

```

```

In [91]: #import necessary libraries for displaying our images
import requests
from IPython.display import Image as IPythonImage
from PIL import Image
from PIL import ImageDraw

```

```

In [92]: # Function to return the image for a given URL
def display_images(myurl):
    name='dog_image.png'
    dog_image = requests.get(myurl)
    with open(name,'wb') as image_file:
        image_file.write(dog_image.content)
    img = Image.open("dog_image.png")
    draw = ImageDraw.Draw(img)

    return img

```

```

In [93]: #display the image of the highest rated dog named Atticus
img=display_images('https://pbs.twimg.com/media/CmgBZ7kWcAAIzFD.jpg')
img = img.resize((300,400))
img.save('sample-out.png')
IPythonImage(filename='sample-out.png')

```

Out[93]:



Find the dog with the lowest rating From the table, the lowest rating is found to be 0

```
In [94]: df_final[df_final['rating_numerator']==0]
```

```
Out[94]:
```

	tweet_id	tweet_date	tweet_time	name	stage	\
245	835152434251116546	2017-02-24	15:40:31	Not known	None	
	retweet_count	favorite_count	rating_numerator	rating_denominator	\	
245	3232	23532	0	10		
	source	\				
245	Twitter for iPhone					

```

245                                     ...
                                     ...

      p1    p1_conf p1_dog                                     p2    p2_conf p2_dog  \
245  swing  0.967066  False  American_Staffordshire_terrier  0.012731    True

                                     p3    p3_conf p3_dog  \
245  Staffordshire_bullterrier  0.007039    True

245  When you're so blinded by your systematic plagiarism that you forget what day it is

[1 rows x 22 columns]

In [95]: url_lowest = df_final[df_final['rating_numerator']==0]['jpg_url']
        url_lowest

Out[95]: 245    https://pbs.twimg.com/media/C5c0tWVWMAEj05p.jpg
        Name: jpg_url, dtype: object

In [96]: #display the image of the lowest rated dog
        img=display_images('https://pbs.twimg.com/media/C5c0tWVWMAEj05p.jpg')
        img = img.resize((300,400))
        img.save('sample-out.png')
        IPythonImage(filename='sample-out.png')

Out[96]:

```




Find the dog with the highest favourite count From the statistical summary table, the highest favourite count is found to be 163917.

```
In [98]: df_final[df_final['favorite_count']==163892]
```

```
Out[98]:
```

	tweet_id	tweet_date	tweet_time	name	stage	\
845	744234799360020481	2016-06-18	18:26:18	Not known	doggo	

	retweet_count	favorite_count	rating_numerator	rating_denominator	\
845	83352	163892	13	10	

	source	\
845	Twitter for iPhone	

```

845                                     ...
                                     ...

           p1   p1_conf p1_dog           p2   p2_conf p2_dog           p3   \
845  Labrador_retriever  0.825333   True  ice_bear  0.044681  False  whippet

           p3_conf p3_dog   \
845  0.018442   True

845  Here's a doggo realizing you can stand in a pool. 13/10 enlightened af (vid by Tim

[1 rows x 22 columns]

In [99]: url_favorite = df_final[df_final['favorite_count']==163892]['jpg_url']
        url_favorite

Out[99]: 845      https://pbs.twimg.com/ext_tw_video_thumb/744234667679821824/pu/img/1GaWmtJtdqzZV
        Name: jpg_url, dtype: object

In [100]: #display the image of the dog with highest favorite count
          img=display_images('https://pbs.twimg.com/ext_tw_video_thumb/744234667679821824/pu/img
          img = img.resize((300,400))
          img.save('sample-out.png')
          IPythonImage(filename='sample-out.png')

Out[100]:

```



Find the dog with the highest retweet count From the same table, the highest retweet count is found to be 83377.

```
In [101]: url_retweet = df_final[df_final['retweet_count']==83352]['jpg_url']  
          url_retweet
```

```
Out[101]: 845      https://pbs.twimg.com/ext_tw_video_thumb/744234667679821824/pu/img/1GaWmtJtdqzZ  
          Name: jpg_url, dtype: object
```

```
In [103]: #display the image of the dog with highest retweet count  
          img=display_images('https://pbs.twimg.com/ext_tw_video_thumb/744234667679821824/pu/img  
          img = img.resize((300,400))  
          img.save('sample-out.png')  
          IPythonImage(filename='sample-out.png')
```

Out [103] :



This 'doggo' with a video image has the highest favourite count and is also found to have the highest number of retweets. We had earlier found that favourite count and retweet count have a high correlation between them.