

wrangle_act

April 24, 2018

WeRateDogs Tweets' Data Wrangling Project

```
In [2]: import pandas as pd
import numpy as np
import requests
import tweepy
import json
import re
import matplotlib.pyplot as plt
import warnings
```

```
url = 'https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image-predictio
r = requests.get(url)
```

```
In [3]: df_img_pred = pd.read_csv(url, sep = '\t')
```

```
In [4]: df_img_pred.head(10)
```

```
Out[4]:
```

	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	

	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
5	1	Bernese_mountain_dog	0.651137	True	English_springer

6	1	box_turtle	0.933012	False	mud_turtle
7	1	chow	0.692517	True	Tibetan_mastiff
8	1	shopping_cart	0.962465	False	shopping_basket
9	1	miniature_poodle	0.201493	True	komondor

	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	
5	0.263788	True	Greater_Swiss_Mountain_dog	0.016199	True	
6	0.045885	False	terrapi	0.017885	False	
7	0.058279	True	fur_coat	0.054449	False	
8	0.014594	False	golden_retriever	0.007959	True	
9	0.192305	True	soft-coated_wheaten_terrier	0.082086	True	

```
In [5]: import tweepy
```

```
consumer_key = ''
consumer_secret = ''
access_token = ''
access_secret = ''
```

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

```
api = tweepy.API(auth)
```

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

```
In [7]: # For loop which will add each available tweet to a new line of tweet_json.txt
with open('tweet_json.txt', 'a', encoding='utf8') as f:
    for tweet_id in twitter_archive['tweet_id']:
        try:
            tweet = api.get_status(tweet_id, tweet_mode='extended')
            json.dump(tweet._json, f)
            f.write('\n')
        except:
            continue
```

```
In [8]: # For loop to append each tweet into a list
tweets_data = []

tweet_file = open('tweet_json.txt', "r")

for line in tweet_file:
    try:
        tweet = json.loads(line)
```

```

        tweets_data.append(tweet)
    except:
        continue

tweet_file.close()

In [9]: # Create tweet_info DataFrame
tweet_info = pd.DataFrame()

In [10]: # Add selected variables to tweet_info DataFrame
tweet_info['id'] = list(map(lambda tweet: tweet['id'], tweets_data))
tweet_info['retweet_count'] = list(map(lambda tweet: tweet['retweet_count'], tweets_data))
tweet_info['favorite_count'] = list(map(lambda tweet: tweet['favorite_count'], tweets_data))

```

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```

In [11]: # View first 10 rows of twitter_archive DataFrame
twitter_archive.head(10)

```

```

Out[11]:
   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

   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

   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...

```

	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

	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

	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

	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

```
In [12]: # View last 5 rows of twitter_archive DataFrame
twitter_archive.tail()
```

```
Out[12]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
2351	666049248165822465	NaN	NaN	
2352	666044226329800704	NaN	NaN	
2353	666033412701032449	NaN	NaN	
2354	666029285002620928	NaN	NaN	
2355	666020888022790149	NaN	NaN	

	timestamp	\
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	\
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	\
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	\
2351	NaN	NaN	
2352	NaN	NaN	
2353	NaN	NaN	
2354	NaN	NaN	
2355	NaN	NaN	

	expanded_urls	rating_numerator	\
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
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

```
In [13]: # View twitter_archive DataFrame info
twitter_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 [14]: # View twitter_archive DataFrame descriptive statistics
twitter_archive.describe()
```

```
Out[14]:
```

	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 [15]: # View first rows of image_predictions DataFrame
df_img_pred.head()
```

```
Out[15]:
```

	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

```
In [16]: # View last 5 rows of image_predictions DataFrame
df_img_pred.tail()
```

```
Out[16]:
```

	tweet_id	jpg_url \
2070	891327558926688256	https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.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	p2	p2_conf	\
2070	2	basset	0.555712	True	English_springer	0.225770	
2071	1	paper_towel	0.170278	False	Labrador_retriever	0.168086	
2072	1	Chihuahua	0.716012	True	malamute	0.078253	
2073	1	Chihuahua	0.323581	True	Pekinese	0.090647	
2074	1	orange	0.097049	False	bagel	0.085851	

	p2_dog	p3	p3_conf	p3_dog
2070	True	German_short-haired_pointer	0.175219	True
2071	True	spatula	0.040836	False
2072	True	kelpie	0.031379	True
2073	True	papillon	0.068957	True
2074	False	banana	0.076110	False

```

In [17]: # View image_predictions DataFrame info
df_img_pred.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 [18]: # View descriptive statistics of image_predictions DataFrame
df_img_pred.describe()

```

```

Out[18]:

```

	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 [19]: # View first 5 rows of tweet_info DataFrame
tweet_info.head()
```

```
Out[19]:
```

	id	retweet_count	favorite_count
0	892420643555336193	8642	38965
1	892177421306343426	6351	33353
2	891815181378084864	4213	25139
3	891689557279858688	8759	42300
4	891327558926688256	9522	40471

```
In [20]: # View last 5 rows of tweet_info DataFrame
tweet_info.tail()
```

```
Out[20]:
```

	id	retweet_count	favorite_count
8195	759099523532779520	4681	15783
8196	759047813560868866	2246	7080
8197	758854675097526272	995	3850
8198	758828659922702336	4268	12148
8199	674019345211760640	324	1184

```
In [21]: # View info of tweet_info DataFrame
tweet_info.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8200 entries, 0 to 8199
Data columns (total 3 columns):
id                8200 non-null int64
retweet_count     8200 non-null int64
favorite_count    8200 non-null int64
dtypes: int64(3)
memory usage: 192.3 KB
```

```
In [22]: # View descriptive statistics of tweet_info DataFrame
tweet_info.describe()
```

```
Out[22]:
```

	id	retweet_count	favorite_count
count	8.200000e+03	8200.000000	8200.000000
mean	8.168399e+17	5256.821585	14543.098171
std	4.229340e+16	6093.743445	15539.191382
min	6.660209e+17	0.000000	0.000000
25%	7.869631e+17	2147.750000	5103.000000
50%	8.174156e+17	3557.500000	11184.500000
75%	8.479628e+17	6173.000000	20653.250000
max	8.924206e+17	61445.000000	143962.000000

```
In [23]: # View rows in twitter_archive which contain '&amp;' instead of '&' in 'text' column
twitter_archive[twitter_archive.text.str.contains('&amp;')]
```

```
Out[23]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id \
262	842765311967449089	NaN	NaN
273	840728873075638272	NaN	NaN
320	834458053273591808	NaN	NaN
461	817536400337801217	NaN	NaN
485	814578408554463233	NaN	NaN
516	810984652412424192	NaN	NaN
799	772826264096874500	NaN	NaN
889	759793422261743616	NaN	NaN
898	758854675097526272	NaN	NaN
976	750026558547456000	NaN	NaN
1104	735137028879360001	NaN	NaN
1179	719367763014393856	NaN	NaN
1199	716791146589110272	NaN	NaN
1222	714258258790387713	NaN	NaN
1274	709198395643068416	NaN	NaN
1366	702671118226825216	NaN	NaN
1421	698195409219559425	NaN	NaN
1465	694352839993344000	NaN	NaN
1481	693280720173801472	NaN	NaN
1508	691483041324204033	NaN	NaN
1524	690597161306841088	NaN	NaN
1538	689835978131935233	NaN	NaN
1569	687807801670897665	NaN	NaN
1593	686386521809772549	NaN	NaN
1621	684926975086034944	NaN	NaN
1646	683834909291606017	NaN	NaN
1707	680801747103793152	NaN	NaN
1763	678446151570427904	NaN	NaN
1795	677314812125323265	NaN	NaN
1812	676811746707918848	NaN	NaN
1817	676603393314578432	NaN	NaN
1842	675870721063669760	6.757073e+17	4.196984e+09
1897	674737130913071104	NaN	NaN
1899	674670581682434048	NaN	NaN
1901	674646392044941312	NaN	NaN
1913	674372068062928900	NaN	NaN
1931	674036086168010753	NaN	NaN
2031	671768281401958400	NaN	NaN
2037	671561002136281088	NaN	NaN
2064	671154572044468225	NaN	NaN
2084	670807719151067136	NaN	NaN
2096	670755717859713024	NaN	NaN
2137	670046952931721218	NaN	NaN
2177	669037058363662336	NaN	NaN

2190	668960084974809088	NaN	NaN
2196	668852170888998912	NaN	NaN
2207	668627278264475648	NaN	NaN
2210	668620235289837568	NaN	NaN
2216	668537837512433665	NaN	NaN
2232	668221241640230912	NaN	NaN
2246	667878741721415682	NaN	NaN
2268	667517642048163840	NaN	NaN
2293	667152164079423490	NaN	NaN
2306	666835007768551424	NaN	NaN

	timestamp \
262	2017-03-17 15:51:22 +0000
273	2017-03-12 00:59:17 +0000
320	2017-02-22 17:41:18 +0000
461	2017-01-07 01:00:41 +0000
485	2016-12-29 21:06:41 +0000
516	2016-12-19 23:06:23 +0000
799	2016-09-05 15:58:34 +0000
889	2016-07-31 16:50:42 +0000
898	2016-07-29 02:40:28 +0000
976	2016-07-04 18:00:41 +0000
1104	2016-05-24 15:55:00 +0000
1179	2016-04-11 03:33:34 +0000
1199	2016-04-04 00:55:01 +0000
1222	2016-03-28 01:10:13 +0000
1274	2016-03-14 02:04:08 +0000
1366	2016-02-25 01:47:04 +0000
1421	2016-02-12 17:22:12 +0000
1465	2016-02-02 02:53:12 +0000
1481	2016-01-30 03:52:58 +0000
1508	2016-01-25 04:49:38 +0000
1524	2016-01-22 18:09:28 +0000
1538	2016-01-20 15:44:48 +0000
1569	2016-01-15 01:25:33 +0000
1593	2016-01-11 03:17:53 +0000
1621	2016-01-07 02:38:10 +0000
1646	2016-01-04 02:18:42 +0000
1707	2015-12-26 17:25:59 +0000
1763	2015-12-20 05:25:42 +0000
1795	2015-12-17 02:30:09 +0000
1812	2015-12-15 17:11:09 +0000
1817	2015-12-15 03:23:14 +0000
1842	2015-12-13 02:51:51 +0000
1897	2015-12-09 23:47:22 +0000
1899	2015-12-09 19:22:56 +0000
1901	2015-12-09 17:46:48 +0000
1913	2015-12-08 23:36:44 +0000

1931 2015-12-08 01:21:40 +0000
 2031 2015-12-01 19:10:13 +0000
 2037 2015-12-01 05:26:34 +0000
 2064 2015-11-30 02:31:34 +0000
 2084 2015-11-29 03:33:17 +0000
 2096 2015-11-29 00:06:39 +0000
 2137 2015-11-27 01:10:17 +0000
 2177 2015-11-24 06:17:19 +0000
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 2196 2015-11-23 18:02:38 +0000
 2207 2015-11-23 03:09:00 +0000
 2210 2015-11-23 02:41:01 +0000
 2216 2015-11-22 21:13:35 +0000
 2232 2015-11-22 00:15:33 +0000
 2246 2015-11-21 01:34:35 +0000
 2268 2015-11-20 01:39:42 +0000
 2293 2015-11-19 01:27:25 +0000
 2306 2015-11-18 04:27:09 +0000

source \

262 <a href="http://twitter.com/download/iphone" r...
 273 <a href="http://twitter.com/download/iphone" r...
 320 <a href="http://twitter.com/download/iphone" r...
 461 <a href="http://twitter.com/download/iphone" r...
 485 <a href="http://twitter.com/download/iphone" r...
 516 <a href="http://twitter.com/download/iphone" r...
 799 <a href="http://twitter.com/download/iphone" r...
 889 <a href="http://twitter.com/download/iphone" r...
 898 <a href="http://twitter.com/download/iphone" r...
 976 <a href="https://about.twitter.com/products/tw...
 1104 <a href="http://twitter.com/download/iphone" r...
 1179 <a href="http://twitter.com/download/iphone" r...
 1199 <a href="http://twitter.com/download/iphone" r...
 1222 <a href="http://twitter.com/download/iphone" r...
 1274 <a href="http://twitter.com/download/iphone" r...
 1366 <a href="http://twitter.com/download/iphone" r...
 1421 <a href="http://twitter.com/download/iphone" r...
 1465 <a href="http://twitter.com/download/iphone" r...
 1481 <a href="http://twitter.com/download/iphone" r...
 1508 <a href="http://twitter.com/download/iphone" r...
 1524 <a href="http://twitter.com/download/iphone" r...
 1538 <a href="http://twitter.com/download/iphone" r...
 1569 <a href="http://twitter.com/download/iphone" r...
 1593 <a href="http://twitter.com/download/iphone" r...
 1621 <a href="http://twitter.com/download/iphone" r...
 1646 <a href="http://twitter.com/download/iphone" r...
 1707 <a href="http://twitter.com/download/iphone" r...
 1763 <a href="http://twitter.com/download/iphone" r...

1795 <a href="http://twitter.com/download/iphone" r...
 1812 <a href="http://twitter.com/download/iphone" r...
 1817 <a href="http://twitter.com/download/iphone" r...
 1842 <a href="http://twitter.com/download/iphone" r...
 1897 <a href="http://twitter.com/download/iphone" r...
 1899 <a href="http://twitter.com/download/iphone" r...
 1901 <a href="http://twitter.com/download/iphone" r...
 1913 <a href="http://twitter.com/download/iphone" r...
 1931 <a href="http://twitter.com/download/iphone" r...
 2031 <a href="http://twitter.com/download/iphone" r...
 2037 <a href="http://twitter.com/download/iphone" r...
 2064 <a href="http://twitter.com/download/iphone" r...
 2084 <a href="http://twitter.com/download/iphone" r...
 2096 <a href="http://twitter.com/download/iphone" r...
 2137 <a href="http://twitter.com/download/iphone" r...
 2177 <a href="http://twitter.com/download/iphone" r...
 2190 <a href="http://twitter.com/download/iphone" r...
 2196 <a href="http://twitter.com/download/iphone" r...
 2207 <a href="http://twitter.com/download/iphone" r...
 2210 <a href="http://twitter.com/download/iphone" r...
 2216 <a href="http://twitter.com/download/iphone" r...
 2232 <a href="http://twitter.com/download/iphone" r...
 2246 <a href="http://twitter.com/download/iphone" r...
 2268 Tw...
 2293 <a href="http://twitter.com/download/iphone" r...
 2306 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
262	Meet Indie. She's not a fan of baths but she's...	NaN
273	RT @dog_rates: This is Pipsy. He is a fluffbal...	6.671522e+17
320	Meet Chester (bottom) & Harold (top). They...	NaN
461	Say hello to Eugene & Patti Melt. No matte...	NaN
485	RT @dog_rates: Meet Beau & Wilbur. Wilbur ...	6.981954e+17
516	Meet Sam. She smiles 24/7 & secretly aspir...	NaN
799	Meet Roosevelt. He's preparing for takeoff. Ma...	NaN
889	Meet Maggie & Lila. Maggie is the doggo, L...	NaN
898	This is Lilli Bee & Honey Bear. Unfortunat...	NaN
976	Meet Jax & Jil. Jil is yelling the pledge ...	NaN
1104	Meet Buckley. His family & some neighbors ...	NaN
1179	Meet Sid & Murphy. Murphy floats alongside...	NaN
1199	Meet Jennifur. She's supposed to be navigating...	NaN
1222	Meet Travis and Flurp. Travis is pretty chill ...	NaN
1274	From left to right:\nCletus, Jerome, Alejandro...	NaN
1366	Meet Rambo & Kiwi. Rambo's the pup with th...	NaN
1421	Meet Beau & Wilbur. Wilbur stole Beau's be...	NaN
1465	Meet Oliiviér. He takes killer selfies. Has a d...	NaN
1481	This is Sadie and her 2 pups Shebang & Ruf...	NaN
1508	When bae says they can't go out but you see th...	NaN

1524	This is Lolo. She's America af. Behind in scie...	NaN
1538	Meet Fynn & Taco. Fynn is an all-powerful ...	NaN
1569	Meet Trooper & Maya. Trooper protects Maya...	NaN
1593	Say hello to Crimson. He's a Speckled Winnebag...	NaN
1621	Meet Bruiser & Charlie. They are the best ...	NaN
1646	Here we see a faulty pupper. Might need to rep...	NaN
1707	Great picture here. Dog on the right panicked ...	NaN
1763	Touching scene here. Really stirs up the emoti...	NaN
1795	Meet Tassy & Bee. Tassy is pretty chill, b...	NaN
1812	Say hello to Penny & Gizmo. They are pract...	NaN
1817	This is Godzilla pupper. He had a ruff childho...	NaN
1842	& this is Yoshi. Another world record cont...	NaN
1897	Meet Rufio. He is unaware of the pink legless ...	NaN
1899	Meet Jeb & Bush. Jeb is somehow stuck in t...	NaN
1901	Two gorgeous dogs here. Little waddling dog is...	NaN
1913	Meet Chesney. On the outside he stays calm &am...	NaN
1931	Meet Daisy. She has no eyes & her face has...	NaN
2031	When you try to recreate the scene from Lady &...	NaN
2037	This is the best thing I've ever seen so sprea...	NaN
2064	Meet Holly. She's trying to teach small human...	NaN
2084	Say hello to Andy. He can balance on one foot,...	NaN
2096	Say hello to Gin & Tonic. They're having a...	NaN
2137	This is Ben & Carson. It's impossible for ...	NaN
2177	Here we have Pancho and Peaches. Pancho is a C...	NaN
2190	Meet Jaycob. He got scared of the vacuum. Hide...	NaN
2196	Say hello to Bobb. Bobb is a Golden High Fescu...	NaN
2207	This is Timofy. He's a pilot for Southwest. It...	NaN
2210	Say hello to Kallie. There was a tornado in th...	NaN
2216	This is Spark. He's nervous. Other dog hasn't ...	NaN
2232	These two dogs are Bo & Smittens. Smittens...	NaN
2246	This is Tedrick. He lives on the edge. Needs s...	NaN
2268	This is Dook & Milo. Dook is struggling to...	NaN
2293	This is Pipsy. He is a fluffball. Enjoys trave...	NaN
2306	These are Peruvian Feldspars. Their names are ...	NaN

	retweeted_status_user_id	retweeted_status_timestamp	\
262	NaN	NaN	
273	4.196984e+09	2015-11-19 01:27:25 +0000	
320	NaN	NaN	
461	NaN	NaN	
485	4.196984e+09	2016-02-12 17:22:12 +0000	
516	NaN	NaN	
799	NaN	NaN	
889	NaN	NaN	
898	NaN	NaN	
976	NaN	NaN	
1104	NaN	NaN	
1179	NaN	NaN	

1199	NaN	NaN
1222	NaN	NaN
1274	NaN	NaN
1366	NaN	NaN
1421	NaN	NaN
1465	NaN	NaN
1481	NaN	NaN
1508	NaN	NaN
1524	NaN	NaN
1538	NaN	NaN
1569	NaN	NaN
1593	NaN	NaN
1621	NaN	NaN
1646	NaN	NaN
1707	NaN	NaN
1763	NaN	NaN
1795	NaN	NaN
1812	NaN	NaN
1817	NaN	NaN
1842	NaN	NaN
1897	NaN	NaN
1899	NaN	NaN
1901	NaN	NaN
1913	NaN	NaN
1931	NaN	NaN
2031	NaN	NaN
2037	NaN	NaN
2064	NaN	NaN
2084	NaN	NaN
2096	NaN	NaN
2137	NaN	NaN
2177	NaN	NaN
2190	NaN	NaN
2196	NaN	NaN
2207	NaN	NaN
2210	NaN	NaN
2216	NaN	NaN
2232	NaN	NaN
2246	NaN	NaN
2268	NaN	NaN
2293	NaN	NaN
2306	NaN	NaN

	expanded_urls	rating_numerator \
262	https://www.gofundme.com/get-indie-home/ , https...	12
273	https://twitter.com/dog_rates/status/667152164...	12
320	https://twitter.com/dog_rates/status/834458053...	12
461	https://twitter.com/dog_rates/status/817536400...	12

485	https://twitter.com/dog_rates/status/698195409...	9
516	https://www.gofundme.com/sams-smile , https://tw...	24
799	https://twitter.com/dog_rates/status/772826264...	11
889	https://twitter.com/dog_rates/status/759793422...	12
898	https://twitter.com/dog_rates/status/758854675...	11
976	https://twitter.com/dog_rates/status/750026558...	10
1104	https://twitter.com/dog_rates/status/735137028...	9
1179	https://twitter.com/dog_rates/status/719367763...	11
1199	https://twitter.com/dog_rates/status/716791146...	11
1222	https://twitter.com/dog_rates/status/714258258...	10
1274	https://twitter.com/dog_rates/status/709198395...	45
1366	https://twitter.com/dog_rates/status/702671118...	10
1421	https://twitter.com/dog_rates/status/698195409...	9
1465	https://twitter.com/dog_rates/status/694352839...	10
1481	https://twitter.com/dog_rates/status/693280720...	10
1508	https://twitter.com/dog_rates/status/691483041...	5
1524	https://twitter.com/dog_rates/status/690597161...	11
1538	https://twitter.com/dog_rates/status/689835978...	11
1569	https://twitter.com/dog_rates/status/687807801...	11
1593	https://twitter.com/dog_rates/status/686386521...	11
1621	https://twitter.com/dog_rates/status/684926975...	11
1646	https://twitter.com/dog_rates/status/683834909...	9
1707	https://twitter.com/dog_rates/status/680801747...	10
1763	https://twitter.com/dog_rates/status/678446151...	10
1795	https://twitter.com/dog_rates/status/677314812...	10
1812	https://twitter.com/dog_rates/status/676811746...	9
1817	https://twitter.com/dog_rates/status/676603393...	9
1842	https://twitter.com/dog_rates/status/675870721...	11
1897	https://twitter.com/dog_rates/status/674737130...	10
1899	https://twitter.com/dog_rates/status/674670581...	9
1901	https://twitter.com/dog_rates/status/674646392...	5
1913	https://twitter.com/dog_rates/status/674372068...	10
1931	https://twitter.com/dog_rates/status/674036086...	9
2031	https://twitter.com/dog_rates/status/671768281...	10
2037	https://twitter.com/dog_rates/status/671561002...	13
2064	https://twitter.com/dog_rates/status/671154572...	11
2084	https://twitter.com/dog_rates/status/670807719...	11
2096	https://twitter.com/dog_rates/status/670755717...	9
2137	https://twitter.com/dog_rates/status/670046952...	11
2177	https://twitter.com/dog_rates/status/669037058...	10
2190	https://twitter.com/dog_rates/status/668960084...	10
2196	https://twitter.com/dog_rates/status/668852170...	11
2207	https://twitter.com/dog_rates/status/668627278...	9
2210	https://twitter.com/dog_rates/status/668620235...	10
2216	https://twitter.com/dog_rates/status/668537837...	8
2232	https://twitter.com/dog_rates/status/668221241...	10
2246	https://twitter.com/dog_rates/status/667878741...	2
2268	https://twitter.com/dog_rates/status/667517642...	8

2293	https://twitter.com/dog_rates/status/667152164...	12
2306	https://twitter.com/dog_rates/status/666835007...	10

	rating_denominator	name	doggo	floofer	pupper	puppo
262	10	Indie	None	None	None	None
273	10	Pipsy	None	None	None	None
320	10	Chester	None	None	None	None
461	10	Eugene	None	None	None	None
485	10	Beau	None	None	None	None
516	7	Sam	None	None	None	None
799	10	Roosevelt	None	None	None	None
889	10	Maggie	doggo	None	pupper	None
898	10	Lilli	None	None	None	None
976	10	Jax	None	None	None	None
1104	10	Buckley	None	None	pupper	None
1179	10	Sid	None	None	None	None
1199	10	Jennifur	None	None	None	None
1222	10	Travis	None	None	None	None
1274	50	None	None	None	None	None
1366	10	Rambo	None	None	None	None
1421	10	Beau	None	None	None	None
1465	10	Olivier	None	None	None	None
1481	10	Sadie	None	None	None	None
1508	10	None	None	None	None	None
1524	10	Lolo	None	None	None	None
1538	10	Fynn	None	None	None	None
1569	10	Trooper	None	None	None	None
1593	10	Crimson	None	None	None	None
1621	10	Bruiser	None	None	None	None
1646	10	None	None	None	pupper	None
1707	10	None	None	None	None	None
1763	10	None	None	None	None	None
1795	10	Tassy	None	None	None	None
1812	10	Penny	None	None	None	None
1817	10	Godzilla	None	None	pupper	None
1842	10	None	None	None	None	None
1897	10	Rufio	None	None	pupper	None
1899	10	Jeb	None	None	None	None
1901	10	None	None	None	None	None
1913	10	Chesney	None	None	None	None
1931	10	Daisy	None	None	None	None
2031	10	None	None	None	None	None
2037	10	the	None	None	None	None
2064	10	Holly	None	None	None	None
2084	10	Andy	None	None	None	None
2096	10	Gin	None	None	None	None
2137	10	Ben	None	None	None	None
2177	10	None	None	None	None	None

2190	10	Jaycob	None	None	None	None
2196	10	Bobbb	None	None	None	None
2207	10	Timofy	None	None	None	None
2210	10	Kallie	None	None	None	None
2216	10	Spark	None	None	None	None
2232	10	None	None	None	None	None
2246	10	Tedrick	None	None	None	None
2268	10	Dook	None	None	None	None
2293	10	Pipsy	None	None	None	None
2306	10	None	None	None	None	None

```
In [24]: # Sort values of 'name' column alphabetically - lowercase values appear at the bottom
twitter_archive.name.sort_values()
```

```
Out[24]: 1035      Abby
1021      Abby
938       Ace
1933      Acro
1327      Adele
1934      Aiden
77       Aja
480      Akumi
820       Al
875      Albert
1954      Albert
144      Albus
412      Albus
1115     Aldrick
2046     Alejandro
374      Alexander
1189     Alexanderson
486      Alf
2238     Alfie
661      Alfie
367      Alfie
1616     Alfie
858      Alfie
51       Alfy
201      Alice
1701     Alice
1495     Amber
1334     Ambrose
2078      Amy
2195     Amélie
...
852      my
988      not
335      not
```

```

1747      officially
1206          old
993          one
1936          one
924          one
369          one
2326      quite
169      quite
118      quite
193      quite
2030      space
22      such
2345      the
2346      the
1603      the
2037      the
2212      the
1815      the
1797      the
1527      the
1120      this
1121      unacceptable
1031      very
773      very
1097      very
819      very
1385      very
Name: name, Length: 2356, dtype: object

```

```

In [25]: # View number of entries for each source
twitter_archive.source.value_counts()

```

```

Out[25]: <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 [26]: # View rows where the value of 'name' is lowercase - indicates that it is not an actual
twitter_archive.loc[(twitter_archive['name'].str.islower())]

```

```

Out[26]:
      tweet_id  in_reply_to_status_id  in_reply_to_user_id  \
22    887517139158093824              NaN              NaN
56    881536004380872706              NaN              NaN
118   869988702071779329              NaN              NaN
169   859196978902773760              NaN              NaN
193   855459453768019968              NaN              NaN
335   832645525019123713              NaN              NaN
369   828650029636317184              NaN              NaN

```

542	806219024703037440	NaN	NaN
649	792913359805018113	NaN	NaN
682	788552643979468800	NaN	NaN
759	778396591732486144	NaN	NaN
773	776249906839351296	NaN	NaN
801	772581559778025472	NaN	NaN
819	770655142660169732	NaN	NaN
822	770093767776997377	NaN	NaN
852	765395769549590528	NaN	NaN
924	755206590534418437	NaN	NaN
988	748977405889503236	NaN	NaN
992	748692773788876800	NaN	NaN
993	748575535303884801	NaN	NaN
1002	747885874273214464	NaN	NaN
1004	747816857231626240	NaN	NaN
1017	746872823977771008	NaN	NaN
1025	746369468511756288	NaN	NaN
1031	745422732645535745	NaN	NaN
1040	744223424764059648	NaN	NaN
1049	743222593470234624	NaN	NaN
1063	741067306818797568	NaN	NaN
1071	740214038584557568	NaN	NaN
1095	736392552031657984	NaN	NaN
...
2191	668955713004314625	NaN	NaN
2198	668815180734689280	NaN	NaN
2204	668636665813057536	NaN	NaN
2211	668614819948453888	NaN	NaN
2212	668587383441514497	NaN	NaN
2218	668507509523615744	NaN	NaN
2222	668466899341221888	NaN	NaN
2235	668171859951755264	NaN	NaN
2249	667861340749471744	NaN	NaN
2255	667773195014021121	NaN	NaN
2264	667538891197542400	NaN	NaN
2273	667470559035432960	NaN	NaN
2287	667177989038297088	NaN	NaN
2304	666983947667116034	NaN	NaN
2311	666781792255496192	NaN	NaN
2314	666701168228331520	NaN	NaN
2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	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
2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN

	timestamp \
22	2017-07-19 03:39:09 +0000
56	2017-07-02 15:32:16 +0000
118	2017-05-31 18:47:24 +0000
169	2017-05-02 00:04:57 +0000
193	2017-04-21 16:33:22 +0000
335	2017-02-17 17:38:57 +0000
369	2017-02-06 17:02:17 +0000
542	2016-12-06 19:29:28 +0000
649	2016-10-31 02:17:31 +0000
682	2016-10-19 01:29:35 +0000
759	2016-09-21 00:53:04 +0000
773	2016-09-15 02:42:54 +0000
801	2016-09-04 23:46:12 +0000
819	2016-08-30 16:11:18 +0000
822	2016-08-29 03:00:36 +0000
852	2016-08-16 03:52:26 +0000
924	2016-07-19 01:04:16 +0000
988	2016-07-01 20:31:43 +0000
992	2016-07-01 01:40:41 +0000
993	2016-06-30 17:54:50 +0000
1002	2016-06-28 20:14:22 +0000
1004	2016-06-28 15:40:07 +0000
1017	2016-06-26 01:08:52 +0000
1025	2016-06-24 15:48:42 +0000
1031	2016-06-22 01:06:43 +0000
1040	2016-06-18 17:41:06 +0000
1049	2016-06-15 23:24:09 +0000
1063	2016-06-10 00:39:48 +0000
1071	2016-06-07 16:09:13 +0000
1095	2016-05-28 03:04:00 +0000
...	...
2191	2015-11-24 00:54:05 +0000
2198	2015-11-23 15:35:39 +0000
2204	2015-11-23 03:46:18 +0000
2211	2015-11-23 02:19:29 +0000
2212	2015-11-23 00:30:28 +0000
2218	2015-11-22 19:13:05 +0000
2222	2015-11-22 16:31:42 +0000
2235	2015-11-21 20:59:20 +0000
2249	2015-11-21 00:25:26 +0000

2255 2015-11-20 18:35:10 +0000
 2264 2015-11-20 03:04:08 +0000
 2273 2015-11-19 22:32:36 +0000
 2287 2015-11-19 03:10:02 +0000
 2304 2015-11-18 14:18:59 +0000
 2311 2015-11-18 00:55:42 +0000
 2314 2015-11-17 19:35:19 +0000
 2326 2015-11-17 00:24:19 +0000
 2327 2015-11-17 00:06:54 +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
 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
 2352 2015-11-16 00:04:52 +0000
 2353 2015-11-15 23:21:54 +0000
 2354 2015-11-15 23:05:30 +0000

source \
 22 <a href="http://twitter.com/download/iphone" r...
 56 <a href="http://twitter.com/download/iphone" r...
 118 <a href="http://twitter.com/download/iphone" r...
 169 <a href="http://twitter.com/download/iphone" r...
 193 <a href="http://twitter.com/download/iphone" r...
 335 Tw...
 369 <a href="http://twitter.com/download/iphone" r...
 542 <a href="http://twitter.com/download/iphone" r...
 649 <a href="http://twitter.com/download/iphone" r...
 682 <a href="http://twitter.com/download/iphone" r...
 759 <a href="http://twitter.com/download/iphone" r...
 773 <a href="http://twitter.com/download/iphone" r...
 801 <a href="http://twitter.com/download/iphone" r...
 819 <a href="http://twitter.com/download/iphone" r...
 822 <a href="http://twitter.com/download/iphone" r...
 852 <a href="http://twitter.com/download/iphone" r...
 924 <a href="http://twitter.com/download/iphone" r...
 988 <a href="http://twitter.com/download/iphone" r...
 992 <a href="http://twitter.com/download/iphone" r...
 993 <a href="http://twitter.com/download/iphone" r...
 1002 <a href="http://twitter.com/download/iphone" r...
 1004 <a href="http://twitter.com/download/iphone" r...
 1017 <a href="http://twitter.com/download/iphone" r...
 1025 <a href="http://twitter.com/download/iphone" r...
 1031 <a href="http://twitter.com/download/iphone" r...

1040 <a href="http://twitter.com/download/iphone" r...
 1049 <a href="http://twitter.com/download/iphone" r...
 1063 <a href="http://twitter.com/download/iphone" r...
 1071 <a href="http://twitter.com/download/iphone" r...
 1095 Vine -...
 ...
 2191 <a href="http://twitter.com/download/iphone" r...
 2198 <a href="http://twitter.com/download/iphone" r...
 2204 <a href="http://twitter.com/download/iphone" r...
 2211 <a href="http://twitter.com/download/iphone" r...
 2212 Vine -...
 2218 <a href="http://twitter.com/download/iphone" r...
 2222 <a href="http://twitter.com/download/iphone" r...
 2235 <a href="http://twitter.com/download/iphone" r...
 2249 <a href="http://twitter.com/download/iphone" r...
 2255 Tw...
 2264 Tw...
 2273 Tw...
 2287 <a href="http://twitter.com/download/iphone" r...
 2304 <a href="http://twitter.com/download/iphone" r...
 2311 <a href="http://twitter.com/download/iphone" r...
 2314 <a href="http://twitter.com/download/iphone" r...
 2326 <a href="http://twitter.com/download/iphone" r...
 2327 <a href="http://twitter.com/download/iphone" r...
 2333 <a href="http://twitter.com/download/iphone" r...
 2334 <a href="http://twitter.com/download/iphone" r...
 2335 <a href="http://twitter.com/download/iphone" r...
 2345 <a href="http://twitter.com/download/iphone" r...
 2346 <a href="http://twitter.com/download/iphone" r...
 2347 <a href="http://twitter.com/download/iphone" r...
 2348 <a href="http://twitter.com/download/iphone" r...
 2349 <a href="http://twitter.com/download/iphone" r...
 2350 <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...

	text	retweeted_status_id \
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
56	Here is a pupper approaching maximum borkdrive...	NaN
118	RT @dog_rates: We only rate dogs. This is quit...	8.591970e+17
169	We only rate dogs. This is quite clearly a smo...	NaN
193	Guys, we only rate dogs. This is quite clearly...	NaN
335	There's going to be a dog terminal at JFK Airp...	NaN
369	Occasionally, we're sent fantastic stories. Th...	NaN
542	We only rate dogs. Please stop sending in non...	NaN
649	Here is a perfect example of someone who has t...	NaN
682	RT @dog_rates: Say hello to mad pupper. You kn...	7.363926e+17

759	RT @dog_rates: This is an East African Chalupa...	7.030419e+17
773	RT @dog_rates: We only rate dogs. Pls stop sen...	7.007478e+17
801	Guys this is getting so out of hand. We only r...	NaN
819	We only rate dogs. Pls stop sending in non-can...	NaN
822	RT @dog_rates: This is just downright precious...	7.410673e+17
852	This is my dog. Her name is Zoey. She knows I'...	NaN
924	This is one of the most inspirational stories ...	NaN
988	What jokester sent in a pic without a dog in i...	NaN
992	That is Quizno. This is his beach. He does not...	NaN
993	This is one of the most reckless puppies I've ...	NaN
1002	This is a mighty rare blue-tailed hammer sherk...	NaN
1004	Viewer discretion is advised. This is a terrib...	NaN
1017	This is a carrot. We only rate dogs. Please on...	NaN
1025	This is an Iraqi Speed Kangaroo. It is not a d...	NaN
1031	We only rate dogs. Pls stop sending in non-can...	NaN
1040	This is actually a pupper and I'd pet it so we...	NaN
1049	This is a very rare Great Alaskan Bush Pupper...	NaN
1063	This is just downright precious af. 12/10 for ...	NaN
1071	This is getting incredibly frustrating. This i...	NaN
1095	Say hello to mad pupper. You know what you did...	NaN
...
2191	This is a Slovakian Helter Skelter Feta named ...	NaN
2198	This is a wild Toblerone from Papua New Guinea...	NaN
2204	This is an Irish Rigatoni terrier named Berta...	NaN
2211	Here is a horned dog. Much grace. Can jump ove...	NaN
2212	Never forget this vine. You will not stop watc...	NaN
2218	This is a Birmingham Quagmire named Chuk. Love...	NaN
2222	Here is a mother dog caring for her pups. Snaz...	NaN
2235	This is a Trans Siberian Kellogg named Alfonso...	NaN
2249	This is a Shotokon Macadamia mix named Cheryl...	NaN
2255	This is a rare Hungarian Pinot named Jessiga. ...	NaN
2264	This is a southwest Coriander named Klint. Hat...	NaN
2273	This is a northern Wahoo named Kohl. He runs t...	NaN
2287	This is a Dasani Kingfisher from Maine. His na...	NaN
2304	This is a curly Ticonderoga named Pepe. No fee...	NaN
2311	This is a purebred Bacardi named Octaviath. Ca...	NaN
2314	This is a golden Buckminsterfullerene named Jo...	NaN
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	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
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

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

	retweeted_status_user_id	retweeted_status_timestamp	\
22	NaN	NaN	
56	NaN	NaN	
118	4.196984e+09	2017-05-02 00:04:57 +0000	
169	NaN	NaN	
193	NaN	NaN	
335	NaN	NaN	
369	NaN	NaN	
542	NaN	NaN	
649	NaN	NaN	
682	4.196984e+09	2016-05-28 03:04:00 +0000	
759	4.196984e+09	2016-02-26 02:20:37 +0000	
773	4.196984e+09	2016-02-19 18:24:26 +0000	
801	NaN	NaN	
819	NaN	NaN	
822	4.196984e+09	2016-06-10 00:39:48 +0000	
852	NaN	NaN	
924	NaN	NaN	
988	NaN	NaN	
992	NaN	NaN	
993	NaN	NaN	
1002	NaN	NaN	
1004	NaN	NaN	
1017	NaN	NaN	
1025	NaN	NaN	
1031	NaN	NaN	
1040	NaN	NaN	
1049	NaN	NaN	
1063	NaN	NaN	
1071	NaN	NaN	
1095	NaN	NaN	
...	
2191	NaN	NaN	
2198	NaN	NaN	
2204	NaN	NaN	
2211	NaN	NaN	
2212	NaN	NaN	
2218	NaN	NaN	
2222	NaN	NaN	
2235	NaN	NaN	
2249	NaN	NaN	
2255	NaN	NaN	
2264	NaN	NaN	
2273	NaN	NaN	

2287	NaN	NaN
2304	NaN	NaN
2311	NaN	NaN
2314	NaN	NaN
2326	NaN	NaN
2327	NaN	NaN
2333	NaN	NaN
2334	NaN	NaN
2335	NaN	NaN
2345	NaN	NaN
2346	NaN	NaN
2347	NaN	NaN
2348	NaN	NaN
2349	NaN	NaN
2350	NaN	NaN
2352	NaN	NaN
2353	NaN	NaN
2354	NaN	NaN

	expanded_urls	rating_numerator \
22	https://twitter.com/dog_rates/status/887517139...	14
56	https://twitter.com/dog_rates/status/881536004...	14
118	https://twitter.com/dog_rates/status/859196978...	12
169	https://twitter.com/dog_rates/status/859196978...	12
193	https://twitter.com/dog_rates/status/855459453...	12
335	http://us.blastingnews.com/news/2017/02/jfk-an...	10
369	https://twitter.com/dog_rates/status/828650029...	14
542	https://twitter.com/dog_rates/status/806219024...	11
649	https://twitter.com/dog_rates/status/792913359...	13
682	https://vine.co/v/iEggaEOiL03,https://vine.co/...	13
759	https://twitter.com/dog_rates/status/703041949...	10
773	https://twitter.com/dog_rates/status/700747788...	11
801	https://twitter.com/dog_rates/status/772581559...	10
819	https://twitter.com/dog_rates/status/770655142...	11
822	https://twitter.com/dog_rates/status/741067306...	12
852	https://twitter.com/dog_rates/status/765395769...	13
924	https://twitter.com/dog_rates/status/755206590...	14
988	https://twitter.com/dog_rates/status/748977405...	10
992	https://twitter.com/dog_rates/status/748692773...	10
993	https://twitter.com/dog_rates/status/748575535...	6
1002	https://twitter.com/dog_rates/status/747885874...	8
1004	https://twitter.com/dog_rates/status/747816857...	4
1017	https://twitter.com/dog_rates/status/746872823...	11
1025	https://twitter.com/dog_rates/status/746369468...	9
1031	https://twitter.com/dog_rates/status/745422732...	9
1040	https://twitter.com/strange_animals/status/672...	12
1049	https://twitter.com/dog_rates/status/743222593...	12
1063	https://twitter.com/dog_rates/status/741067306...	12

1071	https://twitter.com/dog_rates/status/740214038...	10
1095	https://vine.co/v/iEggaE0iL03	13
...
2191	https://twitter.com/dog_rates/status/668955713...	10
2198	https://twitter.com/dog_rates/status/668815180...	7
2204	https://twitter.com/dog_rates/status/668636665...	10
2211	https://twitter.com/dog_rates/status/668614819...	7
2212	https://vine.co/v/ea00wvPTx91	13
2218	https://twitter.com/dog_rates/status/668507509...	10
2222	https://twitter.com/dog_rates/status/668466899...	4
2235	https://twitter.com/dog_rates/status/668171859...	7
2249	https://twitter.com/dog_rates/status/667861340...	9
2255	https://twitter.com/dog_rates/status/667773195...	8
2264	https://twitter.com/dog_rates/status/667538891...	9
2273	https://twitter.com/dog_rates/status/667470559...	11
2287	https://twitter.com/dog_rates/status/667177989...	8
2304	https://twitter.com/dog_rates/status/666983947...	11
2311	https://twitter.com/dog_rates/status/666781792...	10
2314	https://twitter.com/dog_rates/status/666701168...	8
2326	https://twitter.com/dog_rates/status/666411507...	2
2327	https://twitter.com/dog_rates/status/666407126...	7
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
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
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

	rating_denominator	name	doggo	floofer	pupper	puppo
22	10	such	None	None	None	None
56	10	a	None	None	pupper	None
118	10	quite	None	None	None	None
169	10	quite	None	None	None	None
193	10	quite	None	None	None	None
335	10	not	None	None	None	None
369	10	one	None	None	None	None
542	10	incredibly	None	None	None	None
649	10	a	None	None	None	None
682	10	mad	None	None	pupper	None
759	10	an	None	None	None	None
773	10	very	None	None	None	None
801	10	a	None	None	None	None

819	10	very	None	None	None	None
822	10	just	doggo	None	pupper	None
852	10	my	None	None	None	None
924	10	one	doggo	None	None	None
988	10	not	None	None	None	None
992	10	his	doggo	None	None	None
993	10	one	None	None	None	None
1002	10	a	None	None	None	None
1004	10	a	None	None	None	None
1017	10	a	None	None	None	None
1025	10	an	None	None	None	None
1031	10	very	None	None	None	None
1040	10	actually	None	None	pupper	None
1049	10	a	None	None	pupper	None
1063	10	just	doggo	None	pupper	None
1071	10	getting	None	None	None	None
1095	10	mad	None	None	pupper	None
...
2191	10	a	None	None	None	None
2198	10	a	None	None	None	None
2204	10	an	None	None	None	None
2211	10	a	None	None	None	None
2212	10	the	None	None	None	None
2218	10	a	None	None	None	None
2222	10	a	None	None	None	None
2235	10	a	None	None	None	None
2249	10	a	None	None	None	None
2255	10	a	None	None	None	None
2264	10	a	None	None	None	None
2273	10	a	None	None	None	None
2287	10	a	None	None	None	None
2304	10	a	None	None	None	None
2311	10	a	None	None	None	None
2314	10	a	None	None	None	None
2326	10	quite	None	None	None	None
2327	10	a	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
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
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None

[109 rows x 17 columns]

```
In [27]: # View rows where the value of 'name' is lowercase and the word 'named' appears in the
# there is an actual dog name in the text
```

```
twitter_archive.loc[(twitter_archive['name'].str.islower()) & (twitter_archive['text'].
```

```
Out[27]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
1853	675706639471788032	NaN	NaN	
1955	673636718965334016	NaN	NaN	
2034	671743150407421952	NaN	NaN	
2066	671147085991960577	NaN	NaN	
2116	670427002554466305	NaN	NaN	
2125	670361874861563904	NaN	NaN	
2128	670303360680108032	NaN	NaN	
2146	669923323644657664	NaN	NaN	
2161	669564461267722241	NaN	NaN	
2191	668955713004314625	NaN	NaN	
2204	668636665813057536	NaN	NaN	
2218	668507509523615744	NaN	NaN	
2235	668171859951755264	NaN	NaN	
2249	667861340749471744	NaN	NaN	
2255	667773195014021121	NaN	NaN	
2264	667538891197542400	NaN	NaN	
2273	667470559035432960	NaN	NaN	
2304	666983947667116034	NaN	NaN	
2311	666781792255496192	NaN	NaN	
2314	666701168228331520	NaN	NaN	

	timestamp	\
1853	2015-12-12 15:59:51 +0000	
1955	2015-12-06 22:54:44 +0000	
2034	2015-12-01 17:30:22 +0000	
2066	2015-11-30 02:01:49 +0000	
2116	2015-11-28 02:20:27 +0000	
2125	2015-11-27 22:01:40 +0000	
2128	2015-11-27 18:09:09 +0000	
2146	2015-11-26 16:59:01 +0000	
2161	2015-11-25 17:13:02 +0000	
2191	2015-11-24 00:54:05 +0000	
2204	2015-11-23 03:46:18 +0000	
2218	2015-11-22 19:13:05 +0000	
2235	2015-11-21 20:59:20 +0000	
2249	2015-11-21 00:25:26 +0000	
2255	2015-11-20 18:35:10 +0000	
2264	2015-11-20 03:04:08 +0000	
2273	2015-11-19 22:32:36 +0000	
2304	2015-11-18 14:18:59 +0000	

2311 2015-11-18 00:55:42 +0000
 2314 2015-11-17 19:35:19 +0000

	source \
1853	<a href="http://twitter.com/download/iphone" r...
1955	<a href="http://twitter.com/download/iphone" r...
2034	<a href="http://twitter.com/download/iphone" r...
2066	<a href="http://twitter.com/download/iphone" r...
2116	<a href="http://twitter.com/download/iphone" r...
2125	<a href="http://twitter.com/download/iphone" r...
2128	<a href="http://twitter.com/download/iphone" r...
2146	<a href="http://twitter.com/download/iphone" r...
2161	<a href="http://twitter.com/download/iphone" r...
2191	<a href="http://twitter.com/download/iphone" r...
2204	<a href="http://twitter.com/download/iphone" r...
2218	<a href="http://twitter.com/download/iphone" r...
2235	<a href="http://twitter.com/download/iphone" r...
2249	<a href="http://twitter.com/download/iphone" r...
2255	Tw...
2264	Tw...
2273	Tw...
2304	<a href="http://twitter.com/download/iphone" r...
2311	<a href="http://twitter.com/download/iphone" r...
2314	<a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
1853	This is a Sizzlin Menorah spaniel from Brookly...	NaN
1955	This is a Lofted Aphrodisiac Terrier named Kip...	NaN
2034	This is a Tuscaloosa Alcatraz named Jacob (Yac...	NaN
2066	This is a Helvetica Listerine named Rufus. Thi...	NaN
2116	This is a Deciduous Trimester mix named Spork...	NaN
2125	This is a Rich Mahogany Seltzer named Cherokee...	NaN
2128	This is a Speckled Cauliflower Yosemite named ...	NaN
2146	This is a spotted Lipitor Rumpelstiltskin name...	NaN
2161	This is a Coriander Baton Rouge named Alfredo...	NaN
2191	This is a Slovakian Helter Skelter Feta named ...	NaN
2204	This is an Irish Rigatoni terrier named Berta...	NaN
2218	This is a Birmingham Quagmire named Chuk. Love...	NaN
2235	This is a Trans Siberian Kellogg named Alfonso...	NaN
2249	This is a Shotokon Macadamia mix named Cheryl...	NaN
2255	This is a rare Hungarian Pinot named Jessiga. ...	NaN
2264	This is a southwest Coriander named Klint. Hat...	NaN
2273	This is a northern Wahoo named Kohl. He runs t...	NaN
2304	This is a curly Ticonderoga named Pepe. No fee...	NaN
2311	This is a purebred Bacardi named Octaviath. Ca...	NaN
2314	This is a golden Buckminsterfullerene named Jo...	NaN

retweeted_status_user_id retweeted_status_timestamp \

1853	NaN	NaN
1955	NaN	NaN
2034	NaN	NaN
2066	NaN	NaN
2116	NaN	NaN
2125	NaN	NaN
2128	NaN	NaN
2146	NaN	NaN
2161	NaN	NaN
2191	NaN	NaN
2204	NaN	NaN
2218	NaN	NaN
2235	NaN	NaN
2249	NaN	NaN
2255	NaN	NaN
2264	NaN	NaN
2273	NaN	NaN
2304	NaN	NaN
2311	NaN	NaN
2314	NaN	NaN

	expanded_urls	rating_numerator	\
1853	https://twitter.com/dog_rates/status/675706639...	10	
1955	https://twitter.com/dog_rates/status/673636718...	10	
2034	https://twitter.com/dog_rates/status/671743150...	11	
2066	https://twitter.com/dog_rates/status/671147085...	9	
2116	https://twitter.com/dog_rates/status/670427002...	9	
2125	https://twitter.com/dog_rates/status/670361874...	9	
2128	https://twitter.com/dog_rates/status/670303360...	9	
2146	https://twitter.com/dog_rates/status/669923323...	10	
2161	https://twitter.com/dog_rates/status/669564461...	10	
2191	https://twitter.com/dog_rates/status/668955713...	10	
2204	https://twitter.com/dog_rates/status/668636665...	10	
2218	https://twitter.com/dog_rates/status/668507509...	10	
2235	https://twitter.com/dog_rates/status/668171859...	7	
2249	https://twitter.com/dog_rates/status/667861340...	9	
2255	https://twitter.com/dog_rates/status/667773195...	8	
2264	https://twitter.com/dog_rates/status/667538891...	9	
2273	https://twitter.com/dog_rates/status/667470559...	11	
2304	https://twitter.com/dog_rates/status/666983947...	11	
2311	https://twitter.com/dog_rates/status/666781792...	10	
2314	https://twitter.com/dog_rates/status/666701168...	8	

	rating_denominator	name	doggo	floofer	pupper	puppo
1853	10	a	None	None	None	None
1955	10	a	None	None	None	None
2034	10	a	None	None	None	None
2066	10	a	None	None	None	None

2116	10	a	None	None	None	None
2125	10	a	None	None	None	None
2128	10	a	None	None	None	None
2146	10	a	None	None	None	None
2161	10	a	None	None	None	None
2191	10	a	None	None	None	None
2204	10	an	None	None	None	None
2218	10	a	None	None	None	None
2235	10	a	None	None	None	None
2249	10	a	None	None	None	None
2255	10	a	None	None	None	None
2264	10	a	None	None	None	None
2273	10	a	None	None	None	None
2304	10	a	None	None	None	None
2311	10	a	None	None	None	None
2314	10	a	None	None	None	None

In [28]: *# View rows where the value of 'name' is lowercase and the words 'name is' appears in t*
there is an actual dog name in the text
 twitter_archive.loc[(twitter_archive['name'].str.islower()) & (twitter_archive['text'].

Out[28]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
852	765395769549590528		NaN	NaN
2287	667177989038297088		NaN	NaN

	timestamp	\
852	2016-08-16 03:52:26 +0000	
2287	2015-11-19 03:10:02 +0000	

	source	\
852	<a href="http://twitter.com/download/iphone" r...	
2287	<a href="http://twitter.com/download/iphone" r...	

	text	retweeted_status_id	\
852	This is my dog. Her name is Zoey. She knows I'...	NaN	
2287	This is a Dasani Kingfisher from Maine. His na...	NaN	

	retweeted_status_user_id	retweeted_status_timestamp	\
852	NaN	NaN	
2287	NaN	NaN	

	expanded_urls	rating_numerator	\
852	https://twitter.com/dog_rates/status/765395769...	13	
2287	https://twitter.com/dog_rates/status/667177989...	8	

	rating_denominator	name	doggo	floofer	pupper	puppo
852	10	my	None	None	None	None
2287	10	a	None	None	None	None


```
In [29]: # View row where dog name is '0' but we can see in the 'text' column that the dog's name is 0
twitter_archive[twitter_archive.name == "0"]
```

```
Out[29]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	source	text	retweeted_status_id	retweeted_status_user_id	retweeted_status_timestamp	expanded_urls	rating_numerator	rating_denominator	name	doggo	floofer	pupper	puppo
775	776201521193218049	NaN	NaN	2016-09-14 23:30:38 +0000	<a href="http://twitter.com/download/iphone" r...	This is 0'Malley. That is how he sleeps. Doesn...	NaN	NaN	NaN	https://twitter.com/dog_rates/status/776201521...	10	10	0	None	None	None	None

```
In [30]: #disable warnings
warnings.simplefilter('ignore')

# View rows where text column contains #.#/# indicating a decimal for the rating numerator
# however they do not appear in the 'rating_numerator' column
twitter_archive[twitter_archive.text.str.contains(r"(\d+\.\d*\//\d+)")]
```

```
Out[30]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp
45	883482846933004288	NaN	NaN	2017-07-08 00:28:19 +0000
340	832215909146226688	NaN	NaN	2017-02-16 13:11:49 +0000
695	786709082849828864	NaN	NaN	2016-10-13 23:23:56 +0000
763	778027034220126208	NaN	NaN	2016-09-20 00:24:34 +0000
1689	681340665377193984	6.813394e+17	4.196984e+09	2015-12-28 05:07:27 +0000
1712	680494726643068929	NaN	NaN	2015-12-25 21:06:00 +0000

source \

```

45    <a href="http://twitter.com/download/iphone" r...
340    <a href="http://twitter.com/download/iphone" r...
695    <a href="http://twitter.com/download/iphone" r...
763    <a href="http://twitter.com/download/iphone" r...
1689    <a href="http://twitter.com/download/iphone" r...
1712    <a href="http://twitter.com/download/iphone" r...

                                text    retweeted_status_id \
45    This is Bella. She hopes her smile made you sm...    NaN
340    RT @dog_rates: This is Logan, the Chow who liv...    7.867091e+17
695    This is Logan, the Chow who lived. He solemnly...    NaN
763    This is Sophie. She's a Jubilant Bush Pupper. ...    NaN
1689    I've been told there's a slight possibility he...    NaN
1712    Here we have uncovered an entire battalion of ...    NaN

    retweeted_status_user_id retweeted_status_timestamp \
45    NaN    NaN
340    4.196984e+09    2016-10-13 23:23:56 +0000
695    NaN    NaN
763    NaN    NaN
1689    NaN    NaN
1712    NaN    NaN

                                expanded_urls    rating_numerator \
45    https://twitter.com/dog_rates/status/883482846...    5
340    https://twitter.com/dog_rates/status/786709082...    75
695    https://twitter.com/dog_rates/status/786709082...    75
763    https://twitter.com/dog_rates/status/778027034...    27
1689    NaN    5
1712    https://twitter.com/dog_rates/status/680494726...    26

    rating_denominator    name    doggo    floofer    pupper    puppo
45    10    Bella    None    None    None    None
340    10    Logan    None    None    None    None
695    10    Logan    None    None    None    None
763    10    Sophie    None    None    pupper    None
1689    10    None    None    None    None    None
1712    10    None    None    None    None    None

```

```

In [31]: # View row of specific tweet using tweet_id of a tweet that doesn't have a rating
twitter_archive[twitter_archive.tweet_id == 810984652412424192]

```

```

Out[31]:
    tweet_id    in_reply_to_status_id    in_reply_to_user_id \
516    810984652412424192    NaN    NaN

    timestamp \
516    2016-12-19 23:06:23 +0000

```

```

source \
516 <a href="http://twitter.com/download/iphone" r...

text retweeted_status_id \
516 Meet Sam. She smiles 24/7 & secretly aspir... NaN

retweeted_status_user_id retweeted_status_timestamp \
516 NaN NaN

expanded_urls rating_numerator \
516 https://www.gofundme.com/sams-smile,https://tw... 24

rating_denominator name doggo floofer pupper puppo
516 7 Sam None None None None

```

1 Quality

- Tweets without any images
- Contents of 'text' cutoff
- Dataset contains retweets
- Incorrect dog names
- Erroneous datatypes (tweet_id, in_reply_to_status_id, in_reply_to_user_id, timestamp, source, dog stages)
- Missing values in 'name' and dog stages showing as 'None'
- Rating numerators with decimals not showing full float
- Tweet ID 810984652412424192 doesn't have a rating
- Extra characters after '&'
- Sources difficult to read
- Tweet with more than one #/# sometimes have the first occurrence erroneously used for the rating numerators and -denominators

2 Tidiness

- Dog "stage" variable in four columns: doggo, floofer, pupper, puppo
- Join 'tweet_info' and 'image_predictions' to 'twitter_archive'

3 Clean

```

In [32]: # Create copies of original DataFrames to work off of
twitter_archive_clean = twitter_archive.copy()
image_predictions_clean = df_img_pred.copy()
tweet_info_clean = tweet_info.copy()

```

Define

Create dog stage variable and remove individual dog stage columns

4 Code

```
In [33]: # Create 'dog_stage' variable which is made by extracting the dog stage variables from
         twitter_archive_clean['dog_stage'] = twitter_archive_clean['text'].str.extract('(puppo|

In [34]: # Create variable of columns that are no longer needed and drop them from the DataFrame
         columns = ['doggo', 'floofer', 'pupper', 'puppo']
         twitter_archive_clean = twitter_archive_clean.drop(columns, axis=1)
```

5 Test

```
In [35]: twitter_archive_clean.head(20)
```

```
Out[35]:
```

	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	

	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

	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...
14	<a href="http://twitter.com/download/iphone" r...
15	<a href="http://twitter.com/download/iphone" r...
16	<a href="http://twitter.com/download/iphone" r...
17	<a href="http://twitter.com/download/iphone" r...
18	<a href="http://twitter.com/download/iphone" r...
19	<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
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

	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

	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

	rating_denominator	name	dog_stage
0	10	Phineas	NaN
1	10	Tilly	NaN
2	10	Archie	NaN
3	10	Darla	NaN
4	10	Franklin	NaN
5	10	None	NaN
6	10	Jax	NaN
7	10	None	NaN
8	10	Zoey	NaN
9	10	Cassie	doggo
10	10	Koda	NaN
11	10	Bruno	NaN
12	10	None	puppo
13	10	Ted	NaN
14	10	Stuart	puppo
15	10	Oliver	NaN
16	10	Jim	NaN
17	10	Zeke	NaN
18	10	Ralphus	NaN
19	10	Canela	NaN

Define

Add tweet_info and image_predictions to twitter_archive table.

6 Code

```
In [36]: twitter_archive_clean = pd.merge(left=twitter_archive_clean, right=tweet_info_clean, le
```

```
In [37]: twitter_archive_clean = twitter_archive_clean.merge(image_predictions_clean, on='tweet_
```

```
In [38]: twitter_archive_clean = twitter_archive_clean.drop('id', axis=1)
```

7 Test

```
In [39]: twitter_archive_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 6743 entries, 0 to 6742
Data columns (total 27 columns):
tweet_id                6743 non-null int64
in_reply_to_status_id   36 non-null float64
in_reply_to_user_id     36 non-null float64
timestamp               6743 non-null object
source                  6743 non-null object
text                   6743 non-null object
```

```

retweeted_status_id      617 non-null float64
retweeted_status_user_id 617 non-null float64
retweeted_status_timestamp 617 non-null object
expanded_urls            6743 non-null object
rating_numerator         6743 non-null int64
rating_denominator       6743 non-null int64
name                     6743 non-null object
dog_stage                1129 non-null object
retweet_count            6743 non-null int64
favorite_count           6743 non-null int64
jpg_url                  6743 non-null object
img_num                  6743 non-null int64
p1                       6743 non-null object
p1_conf                  6743 non-null float64
p1_dog                   6743 non-null bool
p2                       6743 non-null object
p2_conf                  6743 non-null float64
p2_dog                   6743 non-null bool
p3                       6743 non-null object
p3_conf                  6743 non-null float64
p3_dog                   6743 non-null bool
dtypes: bool(3), float64(7), int64(6), object(11)
memory usage: 1.3+ MB

```

8 Define

Remove rows where there are no images (expanded_urls).

9 Code

```
In [40]: twitter_archive_clean = twitter_archive_clean.dropna(subset=['expanded_urls'])
```

10 Test

```
In [41]: sum(twitter_archive_clean['expanded_urls'].isnull())
```

```
Out[41]: 0
```

11 Define

Remove retweets.

12 Code

```
In [42]: # Select rows where 'retweeted_status_id' is null to save to twitter_archive_clean
        twitter_archive_clean = twitter_archive_clean[twitter_archive_clean['retweeted_status_id']
```

13 Test

```
In [43]: twitter_archive_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 6126 entries, 0 to 6742
Data columns (total 27 columns):
tweet_id                6126 non-null int64
in_reply_to_status_id   36 non-null float64
in_reply_to_user_id     36 non-null float64
timestamp               6126 non-null object
source                  6126 non-null object
text                    6126 non-null object
retweeted_status_id      0 non-null float64
retweeted_status_user_id 0 non-null float64
retweeted_status_timestamp 0 non-null object
expanded_urls           6126 non-null object
rating_numerator         6126 non-null int64
rating_denominator       6126 non-null int64
name                    6126 non-null object
dog_stage               1029 non-null object
retweet_count           6126 non-null int64
favorite_count          6126 non-null int64
jpg_url                 6126 non-null object
img_num                 6126 non-null int64
p1                      6126 non-null object
p1_conf                 6126 non-null float64
p1_dog                  6126 non-null bool
p2                      6126 non-null object
p2_conf                 6126 non-null float64
p2_dog                  6126 non-null bool
p3                      6126 non-null object
p3_conf                 6126 non-null float64
p3_dog                  6126 non-null bool
dtypes: bool(3), float64(7), int64(6), object(11)
memory usage: 1.2+ MB
```

14 Define

Remove retweeted columns.

15 Code

```
In [44]: # Update columns variable and drop columns related to retweets
        columns = ['retweeted_status_id', 'retweeted_status_user_id', 'retweeted_status_timestamp']
        twitter_archive_clean = twitter_archive_clean.drop(columns, axis=1)
```

16 Test

```
In [45]: twitter_archive_clean.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 6126 entries, 0 to 6742
Data columns (total 24 columns):
tweet_id                6126 non-null int64
in_reply_to_status_id   36 non-null float64
in_reply_to_user_id     36 non-null float64
timestamp               6126 non-null object
source                 6126 non-null object
text                   6126 non-null object
expanded_urls           6126 non-null object
rating_numerator        6126 non-null int64
rating_denominator      6126 non-null int64
name                   6126 non-null object
dog_stage              1029 non-null object
retweet_count           6126 non-null int64
favorite_count          6126 non-null int64
jpg_url                6126 non-null object
img_num                6126 non-null int64
p1                     6126 non-null object
p1_conf                6126 non-null float64
p1_dog                 6126 non-null bool
p2                     6126 non-null object
p2_conf                6126 non-null float64
p2_dog                 6126 non-null bool
p3                     6126 non-null object
p3_conf                6126 non-null float64
p3_dog                 6126 non-null bool
dtypes: bool(3), float64(5), int64(6), object(10)
memory usage: 1.0+ MB
```

17 Define

Display full content of 'text' column.

18 Code

```
In [46]: # Set column width to infinite so entire content of 'text' column is displayed
pd.set_option('display.max_colwidth', -1)
```

19 Test

```
In [47]: twitter_archive_clean.sample(5)
```

```
Out[47]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
1896	842163532590374912	NaN	NaN	
3360	816450570814898180	NaN	NaN	
951	867900495410671616	NaN	NaN	
4336	798209839306514432	NaN	NaN	
3461	814530161257443328	NaN	NaN	

	timestamp	\
1896	2017-03-16 00:00:07 +0000	
3360	2017-01-04 01:05:59 +0000	
951	2017-05-26 00:29:37 +0000	
4336	2016-11-14 17:03:50 +0000	
3461	2016-12-29 17:54:58 +0000	

	source
1896	Twitter for iPhone
3360	Twitter for iPhone
951	Twitter for iPhone
4336	Twitter for iPhone
3461	Twitter for iPhone

1896	Meet George. He looks slightly deflated but overall quite powerful. Not sure how
3360	Meet Moose. He doesn't want his friend to go back to college. 13/10 looks like yo
951	Unbelievable. We only rate dogs. Please don't send in non-canines like the "I" fr
4336	This is Cooper. His bow tie was too heavy for the front so he moved it to the sid
3461	This is Alf. Someone just rubbed a balloon on his head. He's only a little pupset

1896	https://twitter.com/dog_rates/status/842163532590374912/photo/1,https://twitter.c
3360	https://twitter.com/dog_rates/status/816450570814898180/photo/1,https://twitter.c
951	https://twitter.com/dog_rates/status/867900495410671616/photo/1
4336	https://twitter.com/dog_rates/status/798209839306514432/photo/1
3461	https://twitter.com/dog_rates/status/814530161257443328/photo/1

	rating_numerator	rating_denominator	name	...	img_num	\
1896	12	10	George	...	2	
3360	13	10	Moose	...	1	
951	12	10	None	...	1	

4336	13	10	Cooper	...	1
3461	12	10	Alf	...	1

		p1	p1_conf	p1_dog		p2	p2_conf	p2_dog	\
1896	French_bulldog		0.891227	True	soccer_ball		0.022811	False	
3360	web_site		0.352857	False	envelope		0.060107	False	
951	Labrador_retriever		0.522644	True	kuvasz		0.332461	True	
4336	Pekinese		0.524583	True	Shih-Tzu		0.102931	True	
3461	miniature_poodle		0.626913	True	toy_poodle		0.265582	True	

		p3	p3_conf	p3_dog
1896	bull_mastiff		0.012852	True
3360	nail		0.031291	False
951	dalmatian		0.032008	True
4336	Pomeranian		0.097893	True
3461	soft-coated_wheaten_terrier		0.041614	True

[5 rows x 24 columns]

Define

Change incorrect dog names.

20 Code

```
In [48]: # Save locations where 'name' column is lowercase, lowercase and 'text' column contains
# column contains the words 'name is'
named_to_replace = twitter_archive_clean.loc[(twitter_archive_clean['name'].str.islower
name_is_to_replace = twitter_archive_clean.loc[(twitter_archive_clean['name'].str.islow
not_named_to_replace = twitter_archive_clean.loc[(twitter_archive_clean['name'].str.isl

# Save these locations as lists
named_to_replace_list = named_to_replace['text'].tolist()
name_is_to_replace_list = name_is_to_replace['text'].tolist()
not_named_to_replace_list = not_named_to_replace['text'].tolist()

# For loop to iterate through locations where name is lowercase and the words 'named' a
# value to be the word that appears after 'named'
for entry in named_to_replace_list:
    mask = twitter_archive_clean.text == entry
    name_column = 'name'
    twitter_archive_clean.loc[mask, name_column] = re.findall(r"named\s(\w+)", entry)

# For loop to iterate through locations where name is lowercase and the words 'name is'
# value to be the word that appears after 'name is'
for entry in name_is_to_replace_list:
    mask = twitter_archive_clean.text == entry
    name_column = 'name'
```

```

        twitter_archive_clean.loc[mask, name_column] = re.findall(r"name is\s(\w+)", entry)

# For loop to iterate through locations where name is lowercase and replace the name va
for entry in not_named_to_replace_list:
    mask = twitter_archive_clean.text == entry
    name_column = 'name'
    twitter_archive_clean.loc[mask, name_column] = "None"

```

```

In [49]: # Replace the occurrence of "O" with "O'Malley"
        twitter_archive_clean.name = twitter_archive_clean.name.replace("O", "O'Malley")

```

21 Test

```

In [50]: twitter_archive_clean.name.sort_values()

```

```

Out[50]: 6600    Acro
        600     Aja
        602     Aja
        596     Aja
        597     Aja
        598     Aja
        599     Aja
        595     Aja
        601     Aja
        594     Aja
        3416    Akumi
        3411    Akumi
        3415    Akumi
        3419    Akumi
        3418    Akumi
        3412    Akumi
        3413    Akumi
        3417    Akumi
        3414    Akumi
        5939    Al
        5932    Al
        5933    Al
        5934    Al
        5936    Al
        5937    Al
        5931    Al
        5935    Al
        5938    Al
        6364    Albert
        6365    Albert
        ...

```

```

1332    Zeke
153     Zeke
1336    Zeke
855     Zoey
856     Zoey
858     Zoey
859     Zoey
860     Zoey
861     Zoey
862     Zoey
863     Zoey
857     Zoey
73      Zoey
79      Zoey
72      Zoey
75      Zoey
76      Zoey
78      Zoey
74      Zoey
80      Zoey
77      Zoey
1121   Zoey
1117   Zoey
1116   Zoey
1122   Zoey
1118   Zoey
1119   Zoey
1123   Zoey
1120   Zoey
1124   Zoey
Name: name, Length: 6126, dtype: object

```

```
In [51]: twitter_archive_clean.loc[(twitter_archive_clean['name'].str.islower())]
```

```

Out[51]: Empty DataFrame
Columns: [tweet_id, in_reply_to_status_id, in_reply_to_user_id, timestamp, source, text]
Index: []

[0 rows x 24 columns]

```

```
In [52]: twitter_archive_clean[twitter_archive_clean.name == "O'Malley"]
```

```

Out[52]:
   tweet_id  in_reply_to_status_id  in_reply_to_user_id \
5571  776201521193218049  NaN
5572  776201521193218049  NaN
5573  776201521193218049  NaN
5574  776201521193218049  NaN
5575  776201521193218049  NaN
5576  776201521193218049  NaN

```

5577	776201521193218049	NaN	NaN
5578	776201521193218049	NaN	NaN
5579	776201521193218049	NaN	NaN

	timestamp \
5571	2016-09-14 23:30:38 +0000
5572	2016-09-14 23:30:38 +0000
5573	2016-09-14 23:30:38 +0000
5574	2016-09-14 23:30:38 +0000
5575	2016-09-14 23:30:38 +0000
5576	2016-09-14 23:30:38 +0000
5577	2016-09-14 23:30:38 +0000
5578	2016-09-14 23:30:38 +0000
5579	2016-09-14 23:30:38 +0000

	source
5571	Twitter for iPhone
5572	Twitter for iPhone
5573	Twitter for iPhone
5574	Twitter for iPhone
5575	Twitter for iPhone
5576	Twitter for iPhone
5577	Twitter for iPhone
5578	Twitter for iPhone
5579	Twitter for iPhone

5571	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5572	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5573	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5574	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5575	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5576	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5577	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5578	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10
5579	This is O'Malley. That is how he sleeps. Doesn't care what you think about it. 10

	expanded_urls \
5571	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5572	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5573	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5574	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5575	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5576	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5577	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5578	https://twitter.com/dog_rates/status/776201521193218049/photo/1
5579	https://twitter.com/dog_rates/status/776201521193218049/photo/1

	rating_numerator	rating_denominator	name	...	img_num	\
5571	10	10	0'Malley	...	1	
5572	10	10	0'Malley	...	1	
5573	10	10	0'Malley	...	1	
5574	10	10	0'Malley	...	1	
5575	10	10	0'Malley	...	1	
5576	10	10	0'Malley	...	1	
5577	10	10	0'Malley	...	1	
5578	10	10	0'Malley	...	1	
5579	10	10	0'Malley	...	1	

	p1	p1_conf	p1_dog	p2	p2_conf	p2_dog	\
5571	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5572	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5573	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5574	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5575	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5576	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5577	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5578	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	
5579	Rottweiler	0.502228	True	black-and-tan_coonhound	0.154594	True	

	p3	p3_conf	p3_dog
5571	bloodhound	0.135176	True
5572	bloodhound	0.135176	True
5573	bloodhound	0.135176	True
5574	bloodhound	0.135176	True
5575	bloodhound	0.135176	True
5576	bloodhound	0.135176	True
5577	bloodhound	0.135176	True
5578	bloodhound	0.135176	True
5579	bloodhound	0.135176	True

[9 rows x 24 columns]

22 Define

Change missing values in 'name' from 'None' to NaN.

23 Code

```
In [53]: twitter_archive_clean['name'] = twitter_archive_clean['name'].replace('None', np.NaN)
```

24 Test

```
In [54]: twitter_archive_clean.info()
```



```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 6126 entries, 0 to 6742
Data columns (total 24 columns):
tweet_id                6126 non-null int64
in_reply_to_status_id   36 non-null float64
in_reply_to_user_id     36 non-null float64
timestamp               6126 non-null object
source                  6126 non-null object
text                    6126 non-null object
expanded_urls           6126 non-null object
rating_numerator        6126 non-null int64
rating_denominator      6126 non-null int64
name                    4888 non-null object
dog_stage               1029 non-null object
retweet_count           6126 non-null int64
favorite_count          6126 non-null int64
jpg_url                 6126 non-null object
img_num                 6126 non-null int64
p1                      6126 non-null object
p1_conf                 6126 non-null float64
p1_dog                  6126 non-null bool
p2                      6126 non-null object
p2_conf                 6126 non-null float64
p2_dog                  6126 non-null bool
p3                      6126 non-null object
p3_conf                 6126 non-null float64
p3_dog                  6126 non-null bool
dtypes: bool(3), float64(5), int64(6), object(10)
memory usage: 1.0+ MB

```

25 Define

Fix rating numerator and denominators that are not actually ratings.

26 Code

```

In [55]: # View all occurrences where there are more than one #/# in 'text' column
         twitter_archive_clean[twitter_archive_clean.text.str.contains( r"(\d+\.?\d*\/\d+\.?\d*\/

Out[55]:
         tweet_id  in_reply_to_status_id  in_reply_to_user_id \
5499  777684233540206592  NaN
5500  777684233540206592  NaN
5501  777684233540206592  NaN
5502  777684233540206592  NaN
5503  777684233540206592  NaN
5504  777684233540206592  NaN

```

5505	777684233540206592	NaN	NaN
5506	777684233540206592	NaN	NaN
5507	777684233540206592	NaN	NaN
6611	668537837512433665	NaN	NaN
6657	667544320556335104	NaN	NaN
6664	667491009379606528	NaN	NaN
6697	666835007768551424	NaN	NaN
6724	666287406224695296	NaN	NaN

	timestamp \
5499	2016-09-19 01:42:24 +0000
5500	2016-09-19 01:42:24 +0000
5501	2016-09-19 01:42:24 +0000
5502	2016-09-19 01:42:24 +0000
5503	2016-09-19 01:42:24 +0000
5504	2016-09-19 01:42:24 +0000
5505	2016-09-19 01:42:24 +0000
5506	2016-09-19 01:42:24 +0000
5507	2016-09-19 01:42:24 +0000
6611	2015-11-22 21:13:35 +0000
6657	2015-11-20 03:25:43 +0000
6664	2015-11-19 23:53:52 +0000
6697	2015-11-18 04:27:09 +0000
6724	2015-11-16 16:11:11 +0000

	source
5499	Twitter for iPhone
5500	Twitter for iPhone
5501	Twitter for iPhone
5502	Twitter for iPhone
5503	Twitter for iPhone
5504	Twitter for iPhone
5505	Twitter for iPhone
5506	Twitter for iPhone
5507	Twitter for iPhone
6611	Twitter for iPhone
6657	Twitter Web Client
6664	Twitter Web Client
6697	Twitter for iPhone
6724	Twitter for iPhone

5499	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
5500	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
5501	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
5502	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
5503	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
5504	"Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no

5505 "Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
 5506 "Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
 5507 "Yep... just as I suspected. You're not flossing." 12/10 and 11/10 for the pup no
 6611 This is Spark. He's nervous. Other dog hasn't moved in a while. Won't come when c
 6657 This is Kial. Kial is either wearing a cape, which would be rad, or flashing us,
 6664 Two dogs in this one. Both are rare Jujitsu Pythagoreans. One slightly whiter tha
 6697 These are Peruvian Feldspars. Their names are Cupit and Prencer. Both resemble Ra
 6724 This is an Albanian 3 1/2 legged Episcopalian. Loves well-polished hardwood floo

	expanded_urls \
5499	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5500	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5501	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5502	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5503	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5504	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5505	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5506	https://twitter.com/dog_rates/status/777684233540206592/photo/1
5507	https://twitter.com/dog_rates/status/777684233540206592/photo/1
6611	https://twitter.com/dog_rates/status/668537837512433665/photo/1
6657	https://twitter.com/dog_rates/status/667544320556335104/photo/1
6664	https://twitter.com/dog_rates/status/667491009379606528/photo/1
6697	https://twitter.com/dog_rates/status/666835007768551424/photo/1
6724	https://twitter.com/dog_rates/status/666287406224695296/photo/1

	rating_numerator	rating_denominator	name	...	img_num \
5499	12	10	NaN	...	1
5500	12	10	NaN	...	1
5501	12	10	NaN	...	1
5502	12	10	NaN	...	1
5503	12	10	NaN	...	1
5504	12	10	NaN	...	1
5505	12	10	NaN	...	1
5506	12	10	NaN	...	1
5507	12	10	NaN	...	1
6611	8	10	Spark	...	1
6657	10	10	Kial	...	1
6664	7	10	NaN	...	1
6697	10	10	NaN	...	1
6724	1	2	NaN	...	1

	p1	p1_conf	p1_dog	p2	p2_conf	p2_dog \
5499	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True
5500	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True
5501	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True
5502	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True
5503	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True
5504	cocker_spaniel	0.253442	True	golden_retriever	0.162850	True

		p3	p3_conf	p3_dog
5499	otterhound		0.110921	True
5500	otterhound		0.110921	True
5501	otterhound		0.110921	True
5502	otterhound		0.110921	True
5503	otterhound		0.110921	True
5504	otterhound		0.110921	True
5505	otterhound		0.110921	True
5506	otterhound		0.110921	True
5507	otterhound		0.110921	True
6611	Chihuahua		0.189737	True
6657	Chihuahua		0.071960	True
6664	weasel		0.005730	False
6697	teddy		0.110183	False
6724	miniature_poodle		0.025581	True

```
In [56]: # Save the text where the rating numerator and denominators were incorrectly extracted
ratings_to_fix = ['After so many requests, this is Bretagne. She was the last surviving
    'Happy 4/20 from the squad! 13/10 for all https://t.co/eV1diwds8a',
    'This is Bluebert. He just saw that both #FinalFur match ups are split 50/50. Amazed a
    'This is Darrel. He just robbed a 7/11 and is in a high speed police chase. Was just s
    'This is an Albanian 3 1/2 legged Episcopalian. Loves well-polished hardwood flooring

In [57]: # Loop through the list of ratings to fix and extract the second occurrence of #/ to save
# occurrences of the actual ratings in the ratings to fix list have a denominator of 10,
#entry instead of extracting it.
for entry in ratings_to_fix:
    mask = twitter_archive_clean.text == entry
    column_name1 = 'rating_numerator'
    column_name2 = 'rating_denominator'
    twitter_archive_clean.loc[mask, column_name1] = re.findall(r"\d+\.\.?*\d*/\d+\.\.?*\d*",
    twitter_archive_clean.loc[mask, column_name2] = 10
```

```
In [58]: twitter_archive_clean[twitter_archive_clean.text.isin(ratings_to_fix)]
```

```

Out[58]:          tweet_id  in_reply_to_status_id  in_reply_to_user_id  \
6724  666287406224695296  NaN                                NaN

          timestamp  \
6724  2015-11-16 16:11:11 +0000

          source
6724  <a href="http://twitter.com/download/iphone" rel="nofollow">Twitter for iPhone</a>

6724  This is an Albanian 3 1/2 legged  Episcopalian. Loves well-polished hardwood floor

          expanded_urls  \
6724  https://twitter.com/dog_rates/status/666287406224695296/photo/1

          rating_numerator  rating_denominator  name  ...  img_num  p1  \
6724  9                    10                NaN  ...    1      Maltese_dog

          p1_conf  p1_dog          p2  p2_conf  p2_dog          p3  \
6724  0.857531  True  toy_poodle  0.063064  True  miniature_poodle

          p3_conf  p3_dog
6724  0.025581  True

[1 rows x 24 columns]

```

28 Define

Fix rating numerator that have decimals.

```

In [59]: # View tweets with decimals in rating in 'text' column
twitter_archive_clean[twitter_archive_clean.text.str.contains(r"(\d+\.\d*\//\d+)")]

```

```

Out[59]:          tweet_id  in_reply_to_status_id  in_reply_to_user_id  \
360    883482846933004288  NaN                                NaN
361    883482846933004288  NaN                                NaN
362    883482846933004288  NaN                                NaN
363    883482846933004288  NaN                                NaN
364    883482846933004288  NaN                                NaN
365    883482846933004288  NaN                                NaN
366    883482846933004288  NaN                                NaN
367    883482846933004288  NaN                                NaN
368    883482846933004288  NaN                                NaN
4977   786709082849828864  NaN                                NaN
4978   786709082849828864  NaN                                NaN
4979   786709082849828864  NaN                                NaN
4980   786709082849828864  NaN                                NaN
4981   786709082849828864  NaN                                NaN

```

4982	786709082849828864	NaN	NaN
4983	786709082849828864	NaN	NaN
4984	786709082849828864	NaN	NaN
4985	786709082849828864	NaN	NaN
5481	778027034220126208	NaN	NaN
5482	778027034220126208	NaN	NaN
5483	778027034220126208	NaN	NaN
5484	778027034220126208	NaN	NaN
5485	778027034220126208	NaN	NaN
5486	778027034220126208	NaN	NaN
5487	778027034220126208	NaN	NaN
5488	778027034220126208	NaN	NaN
5489	778027034220126208	NaN	NaN

	timestamp \
360	2017-07-08 00:28:19 +0000
361	2017-07-08 00:28:19 +0000
362	2017-07-08 00:28:19 +0000
363	2017-07-08 00:28:19 +0000
364	2017-07-08 00:28:19 +0000
365	2017-07-08 00:28:19 +0000
366	2017-07-08 00:28:19 +0000
367	2017-07-08 00:28:19 +0000
368	2017-07-08 00:28:19 +0000
4977	2016-10-13 23:23:56 +0000
4978	2016-10-13 23:23:56 +0000
4979	2016-10-13 23:23:56 +0000
4980	2016-10-13 23:23:56 +0000
4981	2016-10-13 23:23:56 +0000
4982	2016-10-13 23:23:56 +0000
4983	2016-10-13 23:23:56 +0000
4984	2016-10-13 23:23:56 +0000
4985	2016-10-13 23:23:56 +0000
5481	2016-09-20 00:24:34 +0000
5482	2016-09-20 00:24:34 +0000
5483	2016-09-20 00:24:34 +0000
5484	2016-09-20 00:24:34 +0000
5485	2016-09-20 00:24:34 +0000
5486	2016-09-20 00:24:34 +0000
5487	2016-09-20 00:24:34 +0000
5488	2016-09-20 00:24:34 +0000
5489	2016-09-20 00:24:34 +0000

	source
360	Twitter for iPhone
361	Twitter for iPhone
362	Twitter for iPhone
363	Twitter for iPhone

364 Twitter for iPhone
365 Twitter for iPhone
366 Twitter for iPhone
367 Twitter for iPhone
368 Twitter for iPhone
4977 Twitter for iPhone
4978 Twitter for iPhone
4979 Twitter for iPhone
4980 Twitter for iPhone
4981 Twitter for iPhone
4982 Twitter for iPhone
4983 Twitter for iPhone
4984 Twitter for iPhone
4985 Twitter for iPhone
5481 Twitter for iPhone
5482 Twitter for iPhone
5483 Twitter for iPhone
5484 Twitter for iPhone
5485 Twitter for iPhone
5486 Twitter for iPhone
5487 Twitter for iPhone
5488 Twitter for iPhone
5489 Twitter for iPhone

360 This is Bella. She hopes her smile made you smile. If not, she is also offering y
361 This is Bella. She hopes her smile made you smile. If not, she is also offering y
362 This is Bella. She hopes her smile made you smile. If not, she is also offering y
363 This is Bella. She hopes her smile made you smile. If not, she is also offering y
364 This is Bella. She hopes her smile made you smile. If not, she is also offering y
365 This is Bella. She hopes her smile made you smile. If not, she is also offering y
366 This is Bella. She hopes her smile made you smile. If not, she is also offering y
367 This is Bella. She hopes her smile made you smile. If not, she is also offering y
368 This is Bella. She hopes her smile made you smile. If not, she is also offering y
4977 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4978 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4979 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4980 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4981 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4982 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4983 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4984 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4985 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
5481 This is Sophie. She's a Jubilant Bush Pupper. Super h*ckin rare. Appears at rando
5482 This is Sophie. She's a Jubilant Bush Pupper. Super h*ckin rare. Appears at rando
5483 This is Sophie. She's a Jubilant Bush Pupper. Super h*ckin rare. Appears at rando
5484 This is Sophie. She's a Jubilant Bush Pupper. Super h*ckin rare. Appears at rando
5485 This is Sophie. She's a Jubilant Bush Pupper. Super h*ckin rare. Appears at rando

4981	75	10	Logan	...	1
4982	75	10	Logan	...	1
4983	75	10	Logan	...	1
4984	75	10	Logan	...	1
4985	75	10	Logan	...	1
5481	27	10	Sophie	...	1
5482	27	10	Sophie	...	1
5483	27	10	Sophie	...	1
5484	27	10	Sophie	...	1
5485	27	10	Sophie	...	1
5486	27	10	Sophie	...	1
5487	27	10	Sophie	...	1
5488	27	10	Sophie	...	1
5489	27	10	Sophie	...	1

	p1	p1_conf	p1_dog		p2	p2_conf	p2_dog	\
360	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
361	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
362	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
363	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
364	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
365	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
366	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
367	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
368	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
4977	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4978	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4979	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4980	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4981	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4982	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4983	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4984	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4985	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
5481	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5482	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5483	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5484	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5485	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5486	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5487	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5488	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5489	clumber	0.946718	True	cocker_spaniel	0.015950	True		

	p3	p3_conf	p3_dog
360	kuvasz	0.005501	True
361	kuvasz	0.005501	True
362	kuvasz	0.005501	True

363	kuvasz	0.005501	True
364	kuvasz	0.005501	True
365	kuvasz	0.005501	True
366	kuvasz	0.005501	True
367	kuvasz	0.005501	True
368	kuvasz	0.005501	True
4977	chow	0.102654	True
4978	chow	0.102654	True
4979	chow	0.102654	True
4980	chow	0.102654	True
4981	chow	0.102654	True
4982	chow	0.102654	True
4983	chow	0.102654	True
4984	chow	0.102654	True
4985	chow	0.102654	True
5481	Lhasa	0.006519	True
5482	Lhasa	0.006519	True
5483	Lhasa	0.006519	True
5484	Lhasa	0.006519	True
5485	Lhasa	0.006519	True
5486	Lhasa	0.006519	True
5487	Lhasa	0.006519	True
5488	Lhasa	0.006519	True
5489	Lhasa	0.006519	True

[27 rows x 24 columns]

```
In [60]: # Change datatype of rating_numerator and denominator to float
twitter_archive_clean['rating_numerator'] = twitter_archive_clean['rating_numerator'].astype(float)
twitter_archive_clean['rating_denominator'] = twitter_archive_clean['rating_denominator'].astype(float)

In [61]: # Set correct numerators for specific tweets
twitter_archive_clean.loc[(twitter_archive_clean['tweet_id'] == 883482846933004288) & (twitter_archive_clean['reply_to_tweet_id'] == 883482846933004288)] = 1
twitter_archive_clean.loc[(twitter_archive_clean['tweet_id'] == 786709082849828864) & (twitter_archive_clean['reply_to_tweet_id'] == 786709082849828864)] = 1
twitter_archive_clean.loc[(twitter_archive_clean['tweet_id'] == 778027034220126208) & (twitter_archive_clean['reply_to_tweet_id'] == 778027034220126208)] = 1
twitter_archive_clean.loc[(twitter_archive_clean['tweet_id'] == 680494726643068929) & (twitter_archive_clean['reply_to_tweet_id'] == 680494726643068929)] = 1
```

29 Test

```
In [62]: twitter_archive_clean[twitter_archive_clean.text.str.contains(r"(\d+\.\d*\./\d+)")]
```

```
Out[62]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
360	883482846933004288	NaN	NaN	
361	883482846933004288	NaN	NaN	
362	883482846933004288	NaN	NaN	
363	883482846933004288	NaN	NaN	
364	883482846933004288	NaN	NaN	
365	883482846933004288	NaN	NaN	

366	883482846933004288	NaN	NaN
367	883482846933004288	NaN	NaN
368	883482846933004288	NaN	NaN
4977	786709082849828864	NaN	NaN
4978	786709082849828864	NaN	NaN
4979	786709082849828864	NaN	NaN
4980	786709082849828864	NaN	NaN
4981	786709082849828864	NaN	NaN
4982	786709082849828864	NaN	NaN
4983	786709082849828864	NaN	NaN
4984	786709082849828864	NaN	NaN
4985	786709082849828864	NaN	NaN
5481	778027034220126208	NaN	NaN
5482	778027034220126208	NaN	NaN
5483	778027034220126208	NaN	NaN
5484	778027034220126208	NaN	NaN
5485	778027034220126208	NaN	NaN
5486	778027034220126208	NaN	NaN
5487	778027034220126208	NaN	NaN
5488	778027034220126208	NaN	NaN
5489	778027034220126208	NaN	NaN

	timestamp \
360	2017-07-08 00:28:19 +0000
361	2017-07-08 00:28:19 +0000
362	2017-07-08 00:28:19 +0000
363	2017-07-08 00:28:19 +0000
364	2017-07-08 00:28:19 +0000
365	2017-07-08 00:28:19 +0000
366	2017-07-08 00:28:19 +0000
367	2017-07-08 00:28:19 +0000
368	2017-07-08 00:28:19 +0000
4977	2016-10-13 23:23:56 +0000
4978	2016-10-13 23:23:56 +0000
4979	2016-10-13 23:23:56 +0000
4980	2016-10-13 23:23:56 +0000
4981	2016-10-13 23:23:56 +0000
4982	2016-10-13 23:23:56 +0000
4983	2016-10-13 23:23:56 +0000
4984	2016-10-13 23:23:56 +0000
4985	2016-10-13 23:23:56 +0000
5481	2016-09-20 00:24:34 +0000
5482	2016-09-20 00:24:34 +0000
5483	2016-09-20 00:24:34 +0000
5484	2016-09-20 00:24:34 +0000
5485	2016-09-20 00:24:34 +0000
5486	2016-09-20 00:24:34 +0000
5487	2016-09-20 00:24:34 +0000

5488 2016-09-20 00:24:34 +0000
5489 2016-09-20 00:24:34 +0000

source
360 Twitter for iPhone
361 Twitter for iPhone
362 Twitter for iPhone
363 Twitter for iPhone
364 Twitter for iPhone
365 Twitter for iPhone
366 Twitter for iPhone
367 Twitter for iPhone
368 Twitter for iPhone
4977 Twitter for iPhone
4978 Twitter for iPhone
4979 Twitter for iPhone
4980 Twitter for iPhone
4981 Twitter for iPhone
4982 Twitter for iPhone
4983 Twitter for iPhone
4984 Twitter for iPhone
4985 Twitter for iPhone
5481 Twitter for iPhone
5482 Twitter for iPhone
5483 Twitter for iPhone
5484 Twitter for iPhone
5485 Twitter for iPhone
5486 Twitter for iPhone
5487 Twitter for iPhone
5488 Twitter for iPhone
5489 Twitter for iPhone

360 This is Bella. She hopes her smile made you smile. If not, she is also offering y
361 This is Bella. She hopes her smile made you smile. If not, she is also offering y
362 This is Bella. She hopes her smile made you smile. If not, she is also offering y
363 This is Bella. She hopes her smile made you smile. If not, she is also offering y
364 This is Bella. She hopes her smile made you smile. If not, she is also offering y
365 This is Bella. She hopes her smile made you smile. If not, she is also offering y
366 This is Bella. She hopes her smile made you smile. If not, she is also offering y
367 This is Bella. She hopes her smile made you smile. If not, she is also offering y
368 This is Bella. She hopes her smile made you smile. If not, she is also offering y
4977 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4978 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4979 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4980 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4981 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*
4982 This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*

365	13.50	10.0	Bella	...	1
366	13.50	10.0	Bella	...	1
367	13.50	10.0	Bella	...	1
368	13.50	10.0	Bella	...	1
4977	9.75	10.0	Logan	...	1
4978	9.75	10.0	Logan	...	1
4979	9.75	10.0	Logan	...	1
4980	9.75	10.0	Logan	...	1
4981	9.75	10.0	Logan	...	1
4982	9.75	10.0	Logan	...	1
4983	9.75	10.0	Logan	...	1
4984	9.75	10.0	Logan	...	1
4985	9.75	10.0	Logan	...	1
5481	11.27	10.0	Sophie	...	1
5482	11.27	10.0	Sophie	...	1
5483	11.27	10.0	Sophie	...	1
5484	11.27	10.0	Sophie	...	1
5485	11.27	10.0	Sophie	...	1
5486	11.27	10.0	Sophie	...	1
5487	11.27	10.0	Sophie	...	1
5488	11.27	10.0	Sophie	...	1
5489	11.27	10.0	Sophie	...	1

	p1	p1_conf	p1_dog		p2	p2_conf	p2_dog	\
360	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
361	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
362	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
363	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
364	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
365	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
366	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
367	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
368	golden_retriever	0.943082	True	Labrador_retriever	0.032409	True		
4977	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4978	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4979	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4980	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4981	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4982	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4983	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4984	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
4985	Pomeranian	0.467321	True	Persian_cat	0.122978	False		
5481	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5482	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5483	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5484	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5485	clumber	0.946718	True	cocker_spaniel	0.015950	True		
5486	clumber	0.946718	True	cocker_spaniel	0.015950	True		

5487	clumber	0.946718	True	cocker_spaniel	0.015950	True
5488	clumber	0.946718	True	cocker_spaniel	0.015950	True
5489	clumber	0.946718	True	cocker_spaniel	0.015950	True

	p3	p3_conf	p3_dog
360	kuvasz	0.005501	True
361	kuvasz	0.005501	True
362	kuvasz	0.005501	True
363	kuvasz	0.005501	True
364	kuvasz	0.005501	True
365	kuvasz	0.005501	True
366	kuvasz	0.005501	True
367	kuvasz	0.005501	True
368	kuvasz	0.005501	True
4977	chow	0.102654	True
4978	chow	0.102654	True
4979	chow	0.102654	True
4980	chow	0.102654	True
4981	chow	0.102654	True
4982	chow	0.102654	True
4983	chow	0.102654	True
4984	chow	0.102654	True
4985	chow	0.102654	True
5481	Lhasa	0.006519	True
5482	Lhasa	0.006519	True
5483	Lhasa	0.006519	True
5484	Lhasa	0.006519	True
5485	Lhasa	0.006519	True
5486	Lhasa	0.006519	True
5487	Lhasa	0.006519	True
5488	Lhasa	0.006519	True
5489	Lhasa	0.006519	True

[27 rows x 24 columns]

Define

Remove tweet without rating.

30 Code

```
In [63]: twitter_archive_clean = twitter_archive_clean[twitter_archive_clean.tweet_id != 810984652412424192]
```

31 Test

```
In [64]: twitter_archive_clean[twitter_archive_clean.tweet_id == 810984652412424192]
```

```
Out[64]: Empty DataFrame
```

```
Columns: [tweet_id, in_reply_to_status_id, in_reply_to_user_id, timestamp, source, text]
```

```
Index: []

[0 rows x 24 columns]
```

Define

Remove extra characters after '&' in twitter_archive_clean['text'].

32 Code

```
In [65]: twitter_archive_clean['text'] = twitter_archive_clean['text'].str.replace('&', '&')
```

33 Test

```
In [66]: twitter_archive_clean[twitter_archive_clean.text.str.contains('&')]

Out[66]: Empty DataFrame
Columns: [tweet_id, in_reply_to_status_id, in_reply_to_user_id, timestamp, source, text]
Index: []

[0 rows x 24 columns]
```

34 Define

Change sources to more readable categories.

35 Code

```
In [67]: # Remove url from sources
twitter_archive_clean['source'] = twitter_archive_clean['source'].str.replace('<a href=
twitter_archive_clean['source'] = twitter_archive_clean['source'].str.replace('<a href=
twitter_archive_clean['source'] = twitter_archive_clean['source'].str.replace('<a href=
twitter_archive_clean['source'] = twitter_archive_clean['source'].str.replace('<a href=

In [68]: # Change datatype to category
twitter_archive_clean['source'] = twitter_archive_clean['source'].astype('category')
```

36 Test

```
In [69]: twitter_archive_clean.source.value_counts()
```

```
Out[69]: Twitter for iPhone    6047
Twitter Web Client      70
Name: source, dtype: int64
```

```
In [70]: twitter_archive_clean.info()
```



```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 6117 entries, 0 to 6742
Data columns (total 24 columns):
tweet_id                6117 non-null int64
in_reply_to_status_id   36 non-null float64
in_reply_to_user_id     36 non-null float64
timestamp               6117 non-null object
source                 6117 non-null category
text                   6117 non-null object
expanded_urls          6117 non-null object
rating_numerator        6117 non-null float64
rating_denominator      6117 non-null float64
name                   4879 non-null object
dog_stage              1029 non-null object
retweet_count          6117 non-null int64
favorite_count         6117 non-null int64
jpg_url               6117 non-null object
img_num               6117 non-null int64
p1                    6117 non-null object
p1_conf              6117 non-null float64
p1_dog              6117 non-null bool
p2                    6117 non-null object
p2_conf              6117 non-null float64
p2_dog              6117 non-null bool
p3                    6117 non-null object
p3_conf              6117 non-null float64
p3_dog              6117 non-null bool
dtypes: bool(3), category(1), float64(7), int64(4), object(9)
memory usage: 1.0+ MB

```

37 Define

Change datatypes of timestamp to datetime, dog_stage to categorical, and tweet_id, in_reply_to_status_id, and in_reply_to_user_id to strings.

38 Code

```

In [71]: twitter_archive_clean['dog_stage'] = twitter_archive_clean['dog_stage'].astype('category')
twitter_archive_clean['timestamp'] = pd.to_datetime(twitter_archive_clean['timestamp'])
twitter_archive_clean['tweet_id'] = twitter_archive_clean['tweet_id'].astype('str')
twitter_archive_clean['in_reply_to_status_id'] = twitter_archive_clean['in_reply_to_status_id'].astype('str')
twitter_archive_clean['in_reply_to_user_id'] = twitter_archive_clean['in_reply_to_user_id'].astype('str')

```

39 Test

```
In [72]: twitter_archive_clean.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 6117 entries, 0 to 6742
Data columns (total 24 columns):
tweet_id          6117 non-null object
in_reply_to_status_id  6117 non-null object
in_reply_to_user_id  6117 non-null object
timestamp         6117 non-null datetime64[ns]
source            6117 non-null category
text              6117 non-null object
expanded_urls     6117 non-null object
rating_numerator   6117 non-null float64
rating_denominator 6117 non-null float64
name              4879 non-null object
dog_stage         1029 non-null category
retweet_count     6117 non-null int64
favorite_count    6117 non-null int64
jpg_url           6117 non-null object
img_num           6117 non-null int64
p1                6117 non-null object
p1_conf           6117 non-null float64
p1_dog            6117 non-null bool
p2                6117 non-null object
p2_conf           6117 non-null float64
p2_dog            6117 non-null bool
p3                6117 non-null object
p3_conf           6117 non-null float64
p3_dog            6117 non-null bool
dtypes: bool(3), category(2), datetime64[ns](1), float64(5), int64(3), object(10)
memory usage: 985.9+ KB
```

40 Store Data

```
In [73]: # Save clean DataFrame to csv file
         twitter_archive_clean.to_csv('twitter_archive_master.csv')
```

41 Explotary Data Analysis

```
In [74]: # Create copy of twitter_archive_clean to work off of using only my variables of interest
         time_df = twitter_archive_clean[['timestamp', 'retweet_count', 'favorite_count', 'rating_numerator']]

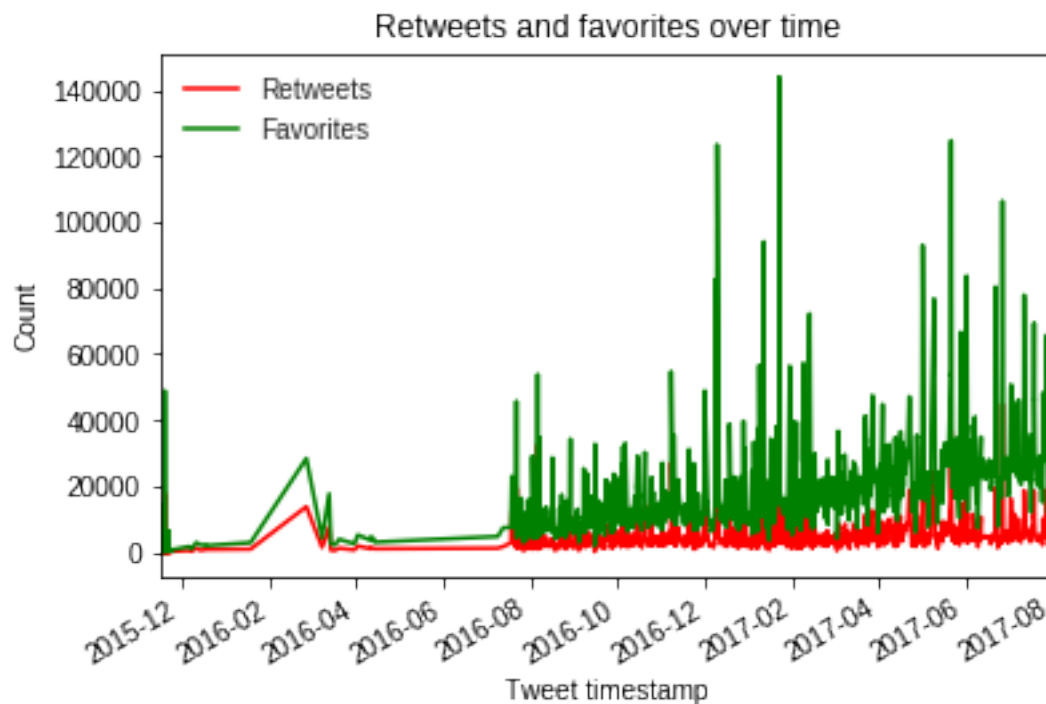
In [75]: # Set the index to be the timestamp so time is displayed properly in plots
         time_df.set_index('timestamp', inplace=True)
```

```

In [76]: # Create rating_ratio variable by dividing the rating numerator by the denominator to
# out of 10
time_df['rating_ratio'] = time_df['rating_numerator']/time_df['rating_denominator']

In [77]: time_df['retweet_count'].plot(color = 'red', label='Retweets')
time_df['favorite_count'].plot(color = 'green', label='Favorites')
plt.style.use('seaborn-darkgrid')
plt.legend(loc='upper left')
plt.xlabel('Tweet timestamp')
plt.ylabel('Count')
plt.title('Retweets and favorites over time')
plt.savefig('retweets_favorites.png')
plt.show()

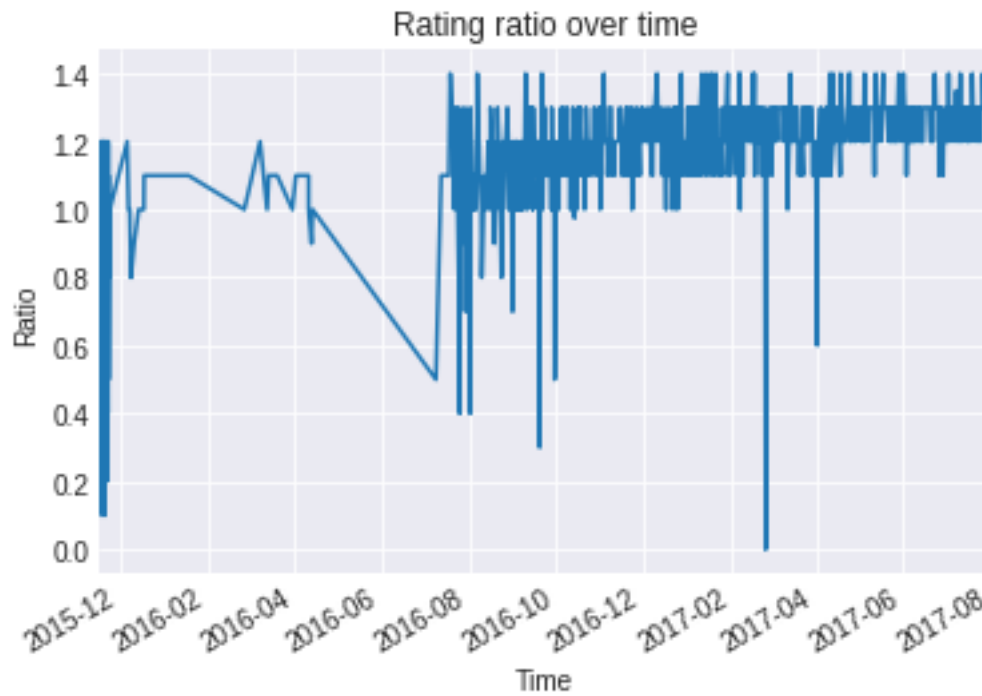
```



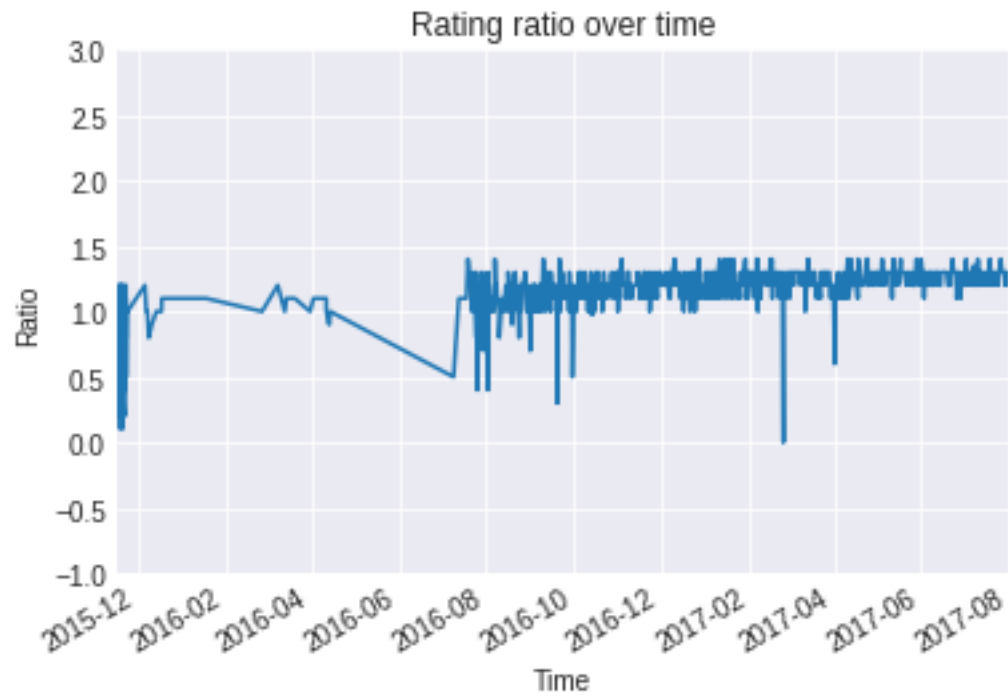
```

In [78]: time_df['rating_ratio'].plot()
plt.style.use('seaborn-darkgrid')
plt.xlabel('Time')
plt.ylabel('Ratio')
plt.title('Rating ratio over time')
plt.savefig('ratio.png')
plt.show()

```

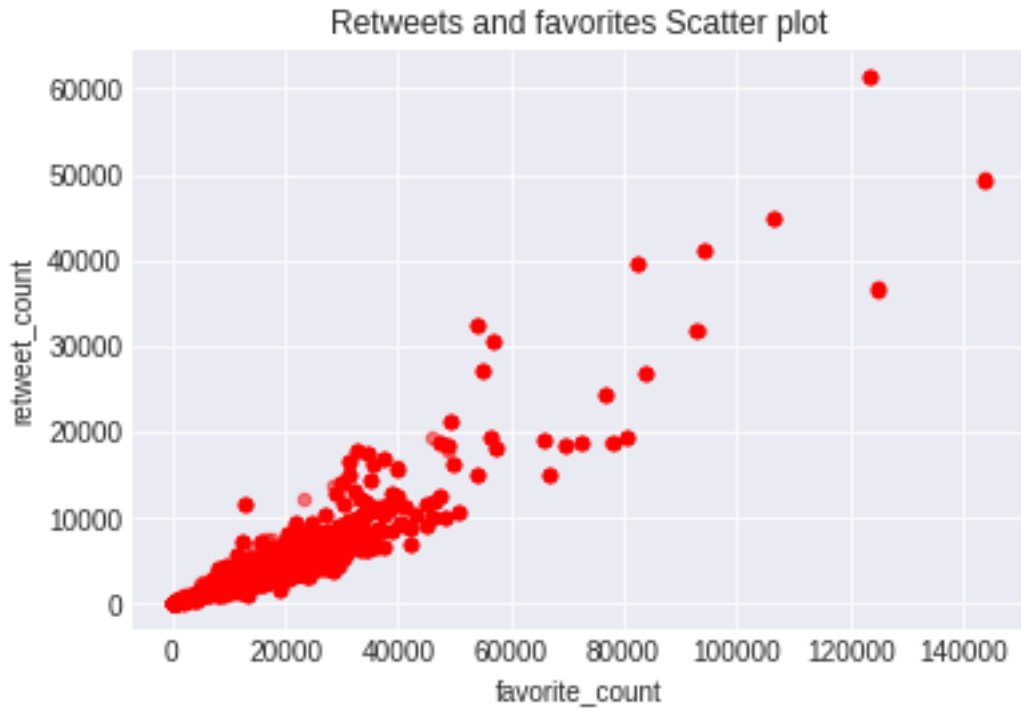


```
In [79]: # Limit y axis to zoom in on data and ignore outliers
time_df['rating_ratio'].plot()
plt.ylim(-1, 3)
plt.style.use('seaborn-darkgrid')
plt.xlabel('Time')
plt.ylabel('Ratio')
plt.title('Rating ratio over time')
plt.savefig('ratio_zoom.png')
plt.show()
```



```
In [80]: twitter_archive_clean.plot(kind = 'scatter', x = 'favorite_count', y = 'retweet_count',  
plt.xlabel('favorite_count')  
plt.ylabel('retweet_count')  
plt.title('Retweets and favorites Scatter plot')
```

```
Out[80]: Text(0.5,1,'Retweets and favorites Scatter plot')
```



```
In [81]: # Our range will be [0,16] taking of the two outliers (1776 and 420)
twitter_archive_clean.plot(x= 'timestamp', y = 'rating_numerator', ylim=[0,16], style =
plt.title('Rating plot over Time')
plt.xlabel('Date')
plt.ylabel('Rating')
```

```
Out[81]: Text(0,0.5, 'Rating')
```

