

wrangle_act

March 19, 2019

1 Wrangle and Analyze Data Project

1.1 Introduction

The purpose of this project is to wrangle data about WeRateDogs Twitter account. First, data is going to be obtained from multiple sources. Next, the data is going to be assessed in terms of both quality and tidiness (<https://ryanwingate.com/purpose/tidy-data/>). Finally, I will clean the data and perform simple analysis.

1.2 Gathering Data

```
In [1]: # import necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
% matplotlib inline
import requests
import os
```

```
In [2]: # load Twitter archive data and display 1 row
df_archive_original = pd.read_csv('twitter-archive-enhanced.csv')
df_archive_original.head(1)
```

```
Out[2]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
	timestamp	\		
0	2017-08-01 16:23:56 +0000			
	source	\		
0	<a href="http://twitter.com/download/iphone" r...			
	text	retweeted_status_id	\	
0	This is Phineas. He's a mystical boy. Only eve...	NaN		
	retweeted_status_user_id	retweeted_status_timestamp	\	
0	NaN	NaN		

```

                                expanded_urls  rating_numerator  \
0  https://twitter.com/dog_rates/status/892420643...          13

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

```

In [3]: # Download the tweet image predictions

```

url = 'https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image-predictio
local_filename = url.split('/')[-1]
r = requests.get(url)
f = open(local_filename, 'wb')
f.write(r.content)

```

Out[3]: 335079

In [4]: # open and look at the predictions

```

df_pred_original = pd.read_csv('image-predictions.tsv', sep='\t')
df_pred_original.head()

```

Out[4]:

	tweet_id	jpg_url
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.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 [5]: # Check for tweet_json.txt

```

# if tweet_json.txt doesn't exist, download programatically
url = 'https://s3.amazonaws.com/video.udacity-data.com/topher/2018/November/5be5fb7d_tweet_json.txt'
json_txt_file_path = 'tweet_json.txt'
if not os.path.isfile(json_txt_file_path):
    with open(json_txt_file_path, mode = 'wb') as file:
        file.write(requests.get(url).content)
# Load into DataFrame
df_json_original = pd.read_json(json_txt_file_path, lines=True)

```

```
df_json_original = df_json_original[['id', 'retweet_count', 'favorite_count']]
df_json_original.head()
```

```
Out [5]:
```

	id	retweet_count	favorite_count
0	892420643555336193	8853	39467
1	892177421306343426	6514	33819
2	891815181378084864	4328	25461
3	891689557279858688	8964	42908
4	891327558926688256	9774	41048

Make copies of the dataframes for the analysis.

```
In [6]: df_archive = df_archive_original.copy()
df_pred = df_pred_original.copy()
df_json = df_json_original.copy()
```

1.3 Assessing Data

```
In [7]: df_archive.head(1)
```

```
Out [7]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\		
0	892420643555336193	NaN	NaN			
	timestamp	\				
0	2017-08-01 16:23:56 +0000					
	source	\				
0	<a href="http://twitter.com/download/iphone" r...					
	text	retweeted_status_id	\			
0	This is Phineas. He's a mystical boy. Only eve...	NaN				
	retweeted_status_user_id	retweeted_status_timestamp	\			
0	NaN	NaN				
	expanded_urls	rating_numerator	\			
0	https://twitter.com/dog_rates/status/892420643...	13				
	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None

```
In [8]: df_pred.head(5)
```

```
Out [8]:
```

	tweet_id	jpg_url	\
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.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 [9]: df_json.head(1)
```

```
Out[9]:
```

	id	retweet_count	favorite_count
0	892420643555336193	8853	39467

```
In [10]: df_archive.info()
```

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

```
In [11]: df_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 [12]: df_json.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2354 entries, 0 to 2353
Data columns (total 3 columns):
id                2354 non-null int64
retweet_count     2354 non-null int64
favorite_count    2354 non-null int64
dtypes: int64(3)
memory usage: 55.2 KB

```

```
In [13]: df_archive.rating_denominator.value_counts()
```

```

Out[13]: 10      2333
         11        3
         50        3
         80        2
         20        2
          2        1
         16        1
         40        1
         70        1
         15        1
         90        1
        110        1
        120        1
        130        1
        150        1
        170        1
          7         1

```

```
0      1
Name: rating_denominator, dtype: int64
```

```
In [14]: df_archive.name.value_counts()
```

```
Out[14]: None      745
a      55
Charlie 12
Cooper 11
Oliver 11
Lucy    11
Tucker 10
Lola    10
Penny  10
Winston 9
Bo      9
the     8
Sadie   8
an      7
Daisy   7
Toby    7
Bailey  7
Buddy   7
Scout   6
Leo     6
Jax     6
Koda    6
Stanley 6
Milo    6
Dave    6
Bella   6
Oscar   6
Jack    6
Rusty   6
Sammy   5
...
Meera   1
Kirk    1
Sailor  1
Zoe     1
Gabby   1
Millie  1
Cleopatra 1
Kota    1
Mojo    1
Damon   1
Barclay 1
Hamrick 1
```

Stu	1
Chubbs	1
Ralphé	1
Frönq	1
Jeb	1
Arnold	1
Tycho	1
Edmund	1
Hector	1
Rhino	1
Ole	1
Duddles	1
Strider	1
my	1
Brockly	1
Mark	1
Beckham	1
Dook	1

Name: name, Length: 957, dtype: int64

1.3.1 Quality

1. Change the **id** column in `df_json` dataframe to **tweet_id** in order to match it with other dataframes. This will make is easier to merge dataframes at the last steps. In addition, I will change the data type from int to string because ids should not be numeric and they aren't intended to perform calculations.
2. Change **timestamp** data type in `df_archive`. This step will help if I was to analyze the data based on time or date.
3. Drop **retweets** rows in `df_archive`. As per project instructions, only original tweets must be included in the final master dataset.
4. Drop unnecessary columns in `df_archive`. It is not properly an issue, but it will make it easier to analyze the data. That is why I have 9 bullet points in this section.
5. Drop null values from **expanded_urls** in `df_archive`. The **expanded_urls** is the only column that has a few missing values. I could not determine why that was the case nor I could fill these values out; therefore, I will drop these values.
6. Drop tweets that are not dogs. This analysis should only incorporate tweets of dogs; therefore, the next step is to delete rows that contain non-dog tweets. In order to do that, I will merge the `df_pred` dataframe with `df_archive` and create a new dataframe `df_archive2`. The `df_pred` contains 3 predictions from the neural network whether a picture shows a dog. I assume that if any of the predictions suggest that it is a dog, I treated it as a dog. I checked a few tweets 'by hand' and it appeared to be true.
7. Clean **rating_denominator** in `df_archive`. Extract the values from text, delete tweets without ratings and tweets with multiple dogs. Change the data type to float.
8. Clean **rating_nominator** in `df_archive`. Extract the values from text, delete tweets without ratings and tweets with multiple dogs. Change the data type to float.
9. Clean names using regular expression in `df_archive`. A useful pattern to find these "fake names" is that they started with lowercase, instead, the "real names" always started with uppercase, moreover, the real ones usually are placed after words like "named", "is" etc...so

it's possible to clean this issue using a regex.

1.3.2 Tidiness

1. Melt 'doggo' 'floofer' 'pupper' 'puppo' in df_archive.
2. Join all data frames into twitter_archive_master.csv.

1.4 Cleaning Data

1.4.1 Quality

Define 1. Change the **id** column in df_json dataframe to **tweet_id** in order to match it with other dataframes. This will make it easier to merge dataframes at the last steps. In addition, I will change the data type from int to string because ids should not be numeric and they aren't intended to perform calculations.

Code

```
In [15]: # rename to `tweet_id`
df_json = df_json.rename(columns={'id': 'tweet_id'})

In [16]: # change `tweet_id` datatype to string in all 3 data frames
df_archive['tweet_id'] = df_archive['tweet_id'].astype(str)
df_pred['tweet_id'] = df_pred['tweet_id'].astype(str)
df_json['tweet_id'] = df_json['tweet_id'].astype(str)
```

Test

```
In [17]: df_json.head(1)

Out[17]:
```

	tweet_id	retweet_count	favorite_count
0	892420643555336193	8853	39467

```
In [18]: df_json.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2354 entries, 0 to 2353
Data columns (total 3 columns):
tweet_id      2354 non-null object
retweet_count  2354 non-null int64
favorite_count 2354 non-null int64
dtypes: int64(2), object(1)
memory usage: 55.2+ KB
```

Define 2. Change **timestamp** data type in df_archive. This step will help if I was to analyze the data based on time or date.

Code

```
In [19]: # change data type to datetime
df_archive['timestamp'] = pd.to_datetime(df_archive.timestamp)
```


Test

```
In [20]: df_archive.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2356 non-null object
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2356 non-null datetime64[ns]
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: datetime64[ns](1), float64(4), int64(2), object(10)
memory usage: 313.0+ KB
```

Define 3. Drop **retweets** rows in `df_archive`. As per project instructions, only original tweets must be included in the final master dataset.

Code

```
In [21]: # check how many retweets are in the dataframe
         df_archive.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2356 non-null object
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2356 non-null datetime64[ns]
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: datetime64[ns](1), float64(4), int64(2), object(10)
memory usage: 313.0+ KB

```

There are 181 retweets in the dataframe. Next, I will delete the retweets.

```

In [22]: # select retweets by checking the retweets columns (if they are empty it's an original
df_archive_notnull = df_archive[df_archive.retweeted_status_id.notnull()]
df_archive_notnull

```

```

Out[22]:
   tweet_id  in_reply_to_status_id  in_reply_to_user_id \
19   888202515573088257           NaN                NaN
32   886054160059072513           NaN                NaN
36   885311592912609280           NaN                NaN
68   879130579576475649           NaN                NaN
73   878404777348136964           NaN                NaN
74   878316110768087041           NaN                NaN
78   877611172832227328           NaN                NaN
91   874434818259525634           NaN                NaN
95   873697596434513921           NaN                NaN
97   873337748698140672           NaN                NaN
101  872668790621863937           NaN                NaN
109  871166179821445120           NaN                NaN
118  869988702071779329           NaN                NaN
124  868639477480148993           NaN                NaN
130  867072653475098625           NaN                NaN
132  866816280283807744           NaN                NaN
137  866094527597207552           NaN                NaN
146  863471782782697472           NaN                NaN
155  861769973181624320           NaN                NaN
159  860981674716409858           NaN                NaN
160  860924035999428608           NaN                NaN
165  860177593139703809           NaN                NaN
171  858860390427611136           NaN                NaN
180  857062103051644929           NaN                NaN
182  856602993587888130           NaN                NaN
185  856330835276025856           NaN                NaN
194  855245323840757760           NaN                NaN
195  855138241867124737           NaN                NaN
204  852936405516943360           NaN                NaN
211  851953902622658560           NaN                NaN
...      ...      ...      ...

```

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

	timestamp	source \
19	2017-07-21 01:02:36	<a href="http://twitter.com/download/iphone" r...
32	2017-07-15 02:45:48	<a href="http://twitter.com/download/iphone" r...
36	2017-07-13 01:35:06	<a href="http://twitter.com/download/iphone" r...
68	2017-06-26 00:13:58	<a href="http://twitter.com/download/iphone" r...
73	2017-06-24 00:09:53	<a href="http://twitter.com/download/iphone" r...
74	2017-06-23 18:17:33	<a href="http://twitter.com/download/iphone" r...
78	2017-06-21 19:36:23	<a href="http://twitter.com/download/iphone" r...
91	2017-06-13 01:14:41	<a href="http://twitter.com/download/iphone" r...
95	2017-06-11 00:25:14	<a href="http://twitter.com/download/iphone" r...
97	2017-06-10 00:35:19	<a href="http://twitter.com/download/iphone" r...
101	2017-06-08 04:17:07	<a href="http://twitter.com/download/iphone" r...
109	2017-06-04 00:46:17	<a href="http://twitter.com/download/iphone" r...
118	2017-05-31 18:47:24	<a href="http://twitter.com/download/iphone" r...
124	2017-05-28 01:26:04	<a href="http://twitter.com/download/iphone" r...
130	2017-05-23 17:40:04	<a href="http://twitter.com/download/iphone" r...
132	2017-05-23 00:41:20	<a href="http://twitter.com/download/iphone" r...

```

137 2017-05-21 00:53:21 <a href="http://twitter.com/download/iphone" r...
146 2017-05-13 19:11:30 <a href="http://twitter.com/download/iphone" r...
155 2017-05-09 02:29:07 <a href="http://twitter.com/download/iphone" r...
159 2017-05-06 22:16:42 <a href="http://twitter.com/download/iphone" r...
160 2017-05-06 18:27:40 <a href="http://twitter.com/download/iphone" r...
165 2017-05-04 17:01:34 <a href="http://twitter.com/download/iphone" r...
171 2017-05-01 01:47:28 <a href="http://twitter.com/download/iphone" r...
180 2017-04-26 02:41:43 <a href="http://twitter.com/download/iphone" r...
182 2017-04-24 20:17:23 <a href="http://twitter.com/download/iphone" r...
185 2017-04-24 02:15:55 <a href="http://twitter.com/download/iphone" r...
194 2017-04-21 02:22:29 <a href="http://twitter.com/download/iphone" r...
195 2017-04-20 19:16:59 <a href="http://twitter.com/download/iphone" r...
204 2017-04-14 17:27:40 <a href="http://twitter.com/download/iphone" r...
211 2017-04-12 00:23:33 <a href="http://twitter.com/download/iphone" r...
...
784 2016-09-11 22:20:06 <a href="http://twitter.com/download/iphone" r...
794 2016-09-07 01:47:12 <a href="http://twitter.com/download/iphone" r...
800 2016-09-05 02:00:22 <a href="http://twitter.com/download/iphone" r...
811 2016-09-01 02:21:21 <a href="http://twitter.com/download/iphone" r...
815 2016-08-31 15:19:06 <a href="http://twitter.com/download/iphone" r...
818 2016-08-30 22:04:05 <a href="http://twitter.com/download/iphone" r...
822 2016-08-29 03:00:36 <a href="http://twitter.com/download/iphone" r...
826 2016-08-27 00:47:53 <a href="http://twitter.com/download/iphone" r...
829 2016-08-25 20:35:48 <a href="http://twitter.com/download/iphone" r...
833 2016-08-24 21:02:45 <a href="http://twitter.com/download/iphone" r...
841 2016-08-20 05:08:29 <a href="http://twitter.com/download/iphone" r...
847 2016-08-18 01:03:45 <a href="http://twitter.com/download/iphone" r...
860 2016-08-10 00:16:21 <a href="http://twitter.com/download/iphone" r...
868 2016-08-06 02:27:27 <a href="http://twitter.com/download/iphone" r...
872 2016-08-05 01:19:35 <a href="http://twitter.com/download/iphone" r...
885 2016-08-01 16:43:19 <a href="http://twitter.com/download/iphone" r...
890 2016-07-31 01:50:18 <a href="http://twitter.com/download/iphone" r...
895 2016-07-29 22:53:27 <a href="http://twitter.com/download/iphone" r...
908 2016-07-26 00:08:05 <a href="http://twitter.com/download/iphone" r...
911 2016-07-25 15:26:30 <a href="http://twitter.com/download/iphone" r...
926 2016-07-18 03:06:01 <a href="http://twitter.com/download/iphone" r...
937 2016-07-13 18:42:44 <a href="http://twitter.com/download/iphone" r...
943 2016-07-12 03:11:42 <a href="http://twitter.com/download/iphone" r...
949 2016-07-11 01:11:51 <a href="http://twitter.com/download/iphone" r...
1012 2016-06-27 01:37:04 <a href="http://twitter.com/download/iphone" r...
1023 2016-06-25 01:52:36 <a href="http://twitter.com/download/iphone" r...
1043 2016-06-17 16:01:16 <a href="http://twitter.com/download/iphone" r...
1242 2016-03-21 19:31:59 <a href="http://twitter.com/download/iphone" r...
2259 2015-11-20 03:51:52 <a href="http://twitter.com" rel="nofollow">Tw...
2260 2015-11-20 03:51:47 <a href="http://twitter.com" rel="nofollow">Tw...

```

```

text retweeted_status_id \
19 RT @dog_rates: This is Canela. She attempted s... 8.874740e+17

```

32	RT @Athletics: 12/10 #BATP https://t.co/WxwJmv...	8.860537e+17
36	RT @dog_rates: This is Lilly. She just paralle...	8.305833e+17
68	RT @dog_rates: This is Emmy. She was adopted t...	8.780576e+17
73	RT @dog_rates: Meet Shadow. In an attempt to r...	8.782815e+17
74	RT @dog_rates: Meet Terrance. He's being yelle...	6.690004e+17
78	RT @rachel2195: @dog_rates the boyfriend and h...	8.768508e+17
91	RT @dog_rates: This is Coco. At first I though...	8.663350e+17
95	RT @dog_rates: This is Walter. He won't start ...	8.688804e+17
97	RT @dog_rates: This is Sierra. She's one preci...	8.732138e+17
101	RT @loganammosis: Penelope here is doing me qu...	8.726576e+17
109	RT @dog_rates: This is Dawn. She's just checki...	8.410770e+17
118	RT @dog_rates: We only rate dogs. This is quit...	8.591970e+17
124	RT @dog_rates: Say hello to Cooper. His expres...	8.685523e+17
130	RT @rachaeleasler: these @dog_rates hats are 1...	8.650134e+17
132	RT @dog_rates: This is Jamesy. He gives a kiss...	8.664507e+17
137	RT @dog_rates: Here's a pupper before and afte...	8.378202e+17
146	RT @dog_rates: Say hello to Quinn. She's quite...	8.630625e+17
155	RT @dog_rates: "Good afternoon class today we'...	8.066291e+17
159	RT @dog_rates: Meet Lorenzo. He's an avid nift...	8.605638e+17
160	RT @tallylott: h*ckin adorable promposal. 13/1...	8.609145e+17
165	RT @dog_rates: Ohboyohboyohboyohboyohboyohboy...	7.616730e+17
171	RT @dog_rates: Meet Winston. He knows he's a l...	8.395493e+17
180	RT @AaronChewning: First time wearing my @dog_...	8.570611e+17
182	RT @dog_rates: This is Luna. It's her first ti...	8.447048e+17
185	RT @Jenna_Marbles: @dog_rates Thanks for ratin...	8.563302e+17
194	RT @dog_rates: Meet George. He looks slightly ...	8.421635e+17
195	RT @frasercampbell_: oh my... what's that... b...	8.551225e+17
204	RT @dog_rates: I usually only share these on F...	8.316501e+17
211	RT @dog_rates: This is Astrid. She's a guide d...	8.293743e+17
...
784	RT @dog_rates: After so many requests, this is...	7.403732e+17
794	RT @dog_rates: Meet Fizz. She thinks love is a...	7.713808e+17
800	RT @dog_rates: This is Gromit. He's pupset bec...	7.652221e+17
811	RT @dog_rates: This is Frankie. He's wearing b...	6.733201e+17
815	RT @katieornah: @dog_rates learning a lot at c...	7.710021e+17
818	RT @dog_rates: Here's a doggo blowing bubbles...	7.392382e+17
822	RT @dog_rates: This is just downright precious...	7.410673e+17
826	RT @dog_rates: Ever seen a dog pet another dog...	7.069045e+17
829	RT @dog_rates: When it's Janet from accounting...	7.001438e+17
833	RT @dog_rates: This is Nollie. She's waving at...	7.399792e+17
841	RT @dog_rates: We only rate dogs... this is a ...	7.599238e+17
847	RT @dog_rates: This is Colby. He's currently r...	7.258423e+17
860	RT @dog_rates: Meet Eve. She's a raging alcoho...	6.732953e+17
868	RT @dog_rates: "Tristan do not speak to me wit...	6.853251e+17
872	RT @dog_rates: Oh. My. God. 13/10 magical af h...	7.116948e+17
885	RT @hownottodraw: The story/person behind @dog...	7.601538e+17
890	RT @dog_rates: This... is a Tyrannosaurus rex...	7.395441e+17
895	RT @dog_rates: AT DAWN...\nWE RIDE\n\n11/10 ht...	6.703191e+17

908	RT @dog_rates: This is Chompsky. He lives up t...	6.790626e+17
911	RT @jon_hill987: @dog_rates There is a cunning...	7.575971e+17
926	RT @dog_rates: This is Rubio. He has too much ...	6.791584e+17
937	RT @dog_rates: This is Carly. She's actually 2...	6.815232e+17
943	RT @dog_rates: HEY PUP WHAT'S THE PART OF THE ...	6.835159e+17
949	RT @dog_rates: Everyone needs to watch this. 1...	6.753544e+17
1012	RT @dog_rates: This pupper killed this great w...	7.047611e+17
1023	RT @dog_rates: This is Shaggy. He knows exactl...	6.678667e+17
1043	RT @dog_rates: Extremely intelligent dog here...	6.671383e+17
1242	RT @twitter: @dog_rates Awesome Tweet! 12/10. ...	7.119983e+17
2259	RT @dogratingrating: Exceptional talent. Origi...	6.675487e+17
2260	RT @dogratingrating: Unoriginal idea. Blatant ...	6.675484e+17

	retweeted_status_user_id	retweeted_status_timestamp	\
19	4.196984e+09	2017-07-19 00:47:34 +0000	
32	1.960740e+07	2017-07-15 02:44:07 +0000	
36	4.196984e+09	2017-02-12 01:04:29 +0000	
68	4.196984e+09	2017-06-23 01:10:23 +0000	
73	4.196984e+09	2017-06-23 16:00:04 +0000	
74	4.196984e+09	2015-11-24 03:51:38 +0000	
78	5.128045e+08	2017-06-19 17:14:49 +0000	
91	4.196984e+09	2017-05-21 16:48:45 +0000	
95	4.196984e+09	2017-05-28 17:23:24 +0000	
97	4.196984e+09	2017-06-09 16:22:42 +0000	
101	1.547674e+08	2017-06-08 03:32:35 +0000	
109	4.196984e+09	2017-03-13 00:02:39 +0000	
118	4.196984e+09	2017-05-02 00:04:57 +0000	
124	4.196984e+09	2017-05-27 19:39:34 +0000	
130	7.874618e+17	2017-05-18 01:17:25 +0000	
132	4.196984e+09	2017-05-22 00:28:40 +0000	
137	4.196984e+09	2017-03-04 00:21:08 +0000	
146	4.196984e+09	2017-05-12 16:05:02 +0000	
155	4.196984e+09	2016-12-07 22:38:52 +0000	
159	4.196984e+09	2017-05-05 18:36:06 +0000	
160	3.638908e+08	2017-05-06 17:49:42 +0000	
165	4.196984e+09	2016-08-05 21:19:27 +0000	
171	4.196984e+09	2017-03-08 18:52:12 +0000	
180	5.870972e+07	2017-04-26 02:37:47 +0000	
182	4.196984e+09	2017-03-23 00:18:10 +0000	
185	6.669901e+07	2017-04-24 02:13:14 +0000	
194	4.196984e+09	2017-03-16 00:00:07 +0000	
195	7.475543e+17	2017-04-20 18:14:33 +0000	
204	4.196984e+09	2017-02-14 23:43:18 +0000	
211	4.196984e+09	2017-02-08 17:00:26 +0000	
...	
784	4.196984e+09	2016-06-08 02:41:38 +0000	
794	4.196984e+09	2016-09-01 16:14:48 +0000	
800	4.196984e+09	2016-08-15 16:22:20 +0000	

811	4.196984e+09	2015-12-06 01:56:44 +0000
815	1.732729e+09	2016-08-31 15:10:07 +0000
818	4.196984e+09	2016-06-04 23:31:25 +0000
822	4.196984e+09	2016-06-10 00:39:48 +0000
826	4.196984e+09	2016-03-07 18:09:06 +0000
829	4.196984e+09	2016-02-18 02:24:13 +0000
833	4.196984e+09	2016-06-07 00:36:02 +0000
841	4.196984e+09	2016-08-01 01:28:46 +0000
847	4.196984e+09	2016-04-29 00:21:01 +0000
860	4.196984e+09	2015-12-06 00:17:55 +0000
868	4.196984e+09	2016-01-08 05:00:14 +0000
872	4.196984e+09	2016-03-20 23:23:54 +0000
885	1.950368e+08	2016-08-01 16:42:51 +0000
890	4.196984e+09	2016-06-05 19:47:03 +0000
895	4.196984e+09	2015-11-27 19:11:49 +0000
908	4.196984e+09	2015-12-21 22:15:18 +0000
911	2.804798e+08	2016-07-25 15:23:28 +0000
926	4.196984e+09	2015-12-22 04:35:49 +0000
937	4.196984e+09	2015-12-28 17:12:42 +0000
943	4.196984e+09	2016-01-03 05:11:12 +0000
949	4.196984e+09	2015-12-11 16:40:19 +0000
1012	4.196984e+09	2016-03-01 20:11:59 +0000
1023	4.196984e+09	2015-11-21 00:46:50 +0000
1043	4.196984e+09	2015-11-19 00:32:12 +0000
1242	7.832140e+05	2016-03-21 19:29:52 +0000
2259	4.296832e+09	2015-11-20 03:43:06 +0000
2260	4.296832e+09	2015-11-20 03:41:59 +0000

	expanded_urls	rating_numerator \
19	https://twitter.com/dog_rates/status/887473957...	13
32	https://twitter.com/dog_rates/status/886053434...	12
36	https://twitter.com/dog_rates/status/830583320...	13
68	https://twitter.com/dog_rates/status/878057613...	14
73	https://www.gofundme.com/3yd6y1c,https://twitt...	13
74	https://twitter.com/dog_rates/status/669000397...	11
78	https://twitter.com/rachel2195/status/87685077...	14
91	https://twitter.com/dog_rates/status/866334964...	12
95	https://twitter.com/dog_rates/status/868880397...	14
97	https://www.gofundme.com/help-my-baby-sierra-g...	12
101	https://twitter.com/loganamnosis/status/872657...	14
109	https://twitter.com/dog_rates/status/841077006...	12
118	https://twitter.com/dog_rates/status/859196978...	12
124	https://www.gofundme.com/3ti3nps,https://twitt...	12
130	https://twitter.com/rachaeleasler/status/86501...	13
132	https://twitter.com/dog_rates/status/866450705...	13
137	https://twitter.com/dog_rates/status/837820167...	12
146	https://www.gofundme.com/helpquinny,https://tw...	13
155	https://twitter.com/dog_rates/status/806629075...	13

159	https://www.gofundme.com/help-lorenzo-beat-can...	13
160	https://twitter.com/tallylott/status/860914485...	13
165	https://twitter.com/dog_rates/status/761672994...	10
171	https://twitter.com/dog_rates/status/839549326...	12
180	https://twitter.com/AaronChewning/status/85706...	13
182	https://twitter.com/dog_rates/status/844704788...	13
185	NaN	14
194	https://twitter.com/dog_rates/status/842163532...	12
195	https://twitter.com/frasercampbell_/status/855...	14
204	http://www.gofundme.com/bluethewhitehusky,http...	13
211	https://twitter.com/dog_rates/status/829374341...	13
...
784	https://twitter.com/dog_rates/status/740373189...	9
794	https://twitter.com/dog_rates/status/771380798...	11
800	https://twitter.com/dog_rates/status/765222098...	10
811	https://twitter.com/dog_rates/status/673320132...	11
815	https://twitter.com/katieornah/status/77100213...	12
818	https://twitter.com/dog_rates/status/739238157...	13
822	https://twitter.com/dog_rates/status/741067306...	12
826	https://vine.co/v/iXQAm5Lrgrh,https://vine.co/...	13
829	https://twitter.com/dog_rates/status/700143752...	10
833	https://twitter.com/dog_rates/status/739979191...	12
841	https://twitter.com/dog_rates/status/759923798...	10
847	https://twitter.com/dog_rates/status/725842289...	12
860	https://twitter.com/dog_rates/status/673295268...	8
868	https://twitter.com/dog_rates/status/685325112...	10
872	https://twitter.com/dog_rates/status/711694788...	13
885	https://weratedogs.com/pages/about-us,https://...	11
890	https://twitter.com/dog_rates/status/739544079...	10
895	https://twitter.com/dog_rates/status/670319130...	11
908	https://twitter.com/dog_rates/status/679062614...	11
911	https://twitter.com/jon_hill1987/status/7575971...	11
926	https://twitter.com/dog_rates/status/679158373...	11
937	https://twitter.com/dog_rates/status/681523177...	12
943	https://vine.co/v/ibvnzrauFuV,https://vine.co/...	11
949	https://twitter.com/dog_rates/status/675354435...	13
1012	https://twitter.com/dog_rates/status/704761120...	13
1023	https://twitter.com/dog_rates/status/667866724...	10
1043	https://twitter.com/dog_rates/status/667138269...	10
1242	https://twitter.com/twitter/status/71199827977...	12
2259	https://twitter.com/dogratingrating/status/667...	12
2260	https://twitter.com/dogratingrating/status/667...	5

	rating_denominator	name	doggo	floofer	pupper	puppo
19	10	Canela	None	None	None	None
32	10	None	None	None	None	None
36	10	Lilly	None	None	None	None
68	10	Emmy	None	None	None	None

73	10	Shadow	None	None	None	None
74	10	Terrance	None	None	None	None
78	10	None	None	None	pupper	None
91	10	Coco	None	None	None	None
95	10	Walter	None	None	None	None
97	10	Sierra	None	None	pupper	None
101	10	None	None	None	None	None
109	10	Dawn	None	None	None	None
118	10	quite	None	None	None	None
124	10	Cooper	None	None	None	None
130	10	None	None	None	None	None
132	10	Jamesy	None	None	pupper	None
137	10	None	None	None	pupper	None
146	10	Quinn	None	None	None	None
155	10	None	None	None	None	None
159	10	Lorenzo	None	None	None	None
160	10	None	None	None	None	None
165	10	None	None	None	None	None
171	10	Winston	None	None	None	None
180	10	None	None	None	None	None
182	10	Luna	None	None	None	None
185	10	None	None	None	None	None
194	10	George	None	None	None	None
195	10	None	None	None	None	None
204	10	None	None	None	None	None
211	10	Astrid	doggo	None	None	None
...
784	11	None	None	None	None	None
794	10	Fizz	None	None	None	None
800	10	Gromit	None	None	None	None
811	10	Frankie	None	None	None	None
815	10	None	None	None	pupper	None
818	10	None	doggo	None	None	None
822	10	just	doggo	None	pupper	None
826	10	None	None	None	None	None
829	10	None	None	None	pupper	None
833	10	Nollie	None	None	None	None
841	10	None	None	None	None	None
847	10	Colby	None	None	None	None
860	10	Eve	None	None	pupper	None
868	10	None	None	None	None	None
872	10	None	None	None	None	None
885	10	None	None	None	None	None
890	10	None	None	None	None	None
895	10	None	None	None	None	None
908	10	Chompsky	None	None	None	None
911	10	None	None	None	pupper	None
926	10	Rubio	None	None	None	None

937	10	Carly	None	None	None	None
943	10	None	None	None	None	None
949	10	None	None	None	None	None
1012	10	None	None	None	pupper	None
1023	10	Shaggy	None	None	None	None
1043	10	None	None	None	None	None
1242	10	None	None	None	None	None
2259	10	None	None	None	None	None
2260	10	None	None	None	None	None

[181 rows x 17 columns]

```
In [23]: # drop retweets (df_archive.retweeted_status rows that are not empty)
df_archive = df_archive.drop(df_archive.notnull.index, axis=0)
df_archive
```

```
Out[23]:
```

	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	
20	888078434458587136	NaN	NaN	
21	887705289381826560	NaN	NaN	
22	887517139158093824	NaN	NaN	
23	887473957103951883	NaN	NaN	
24	887343217045368832	NaN	NaN	
25	887101392804085760	NaN	NaN	
26	886983233522544640	NaN	NaN	
27	886736880519319552	NaN	NaN	
28	886680336477933568	NaN	NaN	
29	886366144734445568	NaN	NaN	
30	886267009285017600	8.862664e+17	2.281182e+09	
...	

2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2328	666396247373291520	NaN	NaN
2329	666373753744588802	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2332	666345417576210432	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	NaN	NaN
2336	666273097616637952	NaN	NaN
2337	666268910803644416	NaN	NaN
2338	666104133288665088	NaN	NaN
2339	666102155909144576	NaN	NaN
2340	666099513787052032	NaN	NaN
2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2343	666073100786774016	NaN	NaN
2344	666071193221509120	NaN	NaN
2345	666063827256086533	NaN	NaN
2346	666058600524156928	NaN	NaN
2347	666057090499244032	NaN	NaN
2348	666055525042405380	NaN	NaN
2349	666051853826850816	NaN	NaN
2350	666050758794694657	NaN	NaN
2351	666049248165822465	NaN	NaN
2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp	source \
0	2017-08-01 16:23:56	<a href="http://twitter.com/download/iphone" r...
1	2017-08-01 00:17:27	<a href="http://twitter.com/download/iphone" r...
2	2017-07-31 00:18:03	<a href="http://twitter.com/download/iphone" r...
3	2017-07-30 15:58:51	<a href="http://twitter.com/download/iphone" r...
4	2017-07-29 16:00:24	<a href="http://twitter.com/download/iphone" r...
5	2017-07-29 00:08:17	<a href="http://twitter.com/download/iphone" r...
6	2017-07-28 16:27:12	<a href="http://twitter.com/download/iphone" r...
7	2017-07-28 00:22:40	<a href="http://twitter.com/download/iphone" r...
8	2017-07-27 16:25:51	<a href="http://twitter.com/download/iphone" r...
9	2017-07-26 15:59:51	<a href="http://twitter.com/download/iphone" r...
10	2017-07-26 00:31:25	<a href="http://twitter.com/download/iphone" r...
11	2017-07-25 16:11:53	<a href="http://twitter.com/download/iphone" r...
12	2017-07-25 01:55:32	<a href="http://twitter.com/download/iphone" r...
13	2017-07-25 00:10:02	<a href="http://twitter.com/download/iphone" r...
14	2017-07-24 17:02:04	<a href="http://twitter.com/download/iphone" r...
15	2017-07-24 00:19:32	<a href="http://twitter.com/download/iphone" r...

```

16 2017-07-23 00:22:39 <a href="http://twitter.com/download/iphone" r...
17 2017-07-22 16:56:37 <a href="http://twitter.com/download/iphone" r...
18 2017-07-22 00:23:06 <a href="http://twitter.com/download/iphone" r...
20 2017-07-20 16:49:33 <a href="http://twitter.com/download/iphone" r...
21 2017-07-19 16:06:48 <a href="http://twitter.com/download/iphone" r...
22 2017-07-19 03:39:09 <a href="http://twitter.com/download/iphone" r...
23 2017-07-19 00:47:34 <a href="http://twitter.com/download/iphone" r...
24 2017-07-18 16:08:03 <a href="http://twitter.com/download/iphone" r...
25 2017-07-18 00:07:08 <a href="http://twitter.com/download/iphone" r...
26 2017-07-17 16:17:36 <a href="http://twitter.com/download/iphone" r...
27 2017-07-16 23:58:41 <a href="http://twitter.com/download/iphone" r...
28 2017-07-16 20:14:00 <a href="http://twitter.com/download/iphone" r...
29 2017-07-15 23:25:31 <a href="http://twitter.com/download/iphone" r...
30 2017-07-15 16:51:35 <a href="http://twitter.com/download/iphone" r...
...
2326 2015-11-17 00:24:19 <a href="http://twitter.com/download/iphone" r...
2327 2015-11-17 00:06:54 <a href="http://twitter.com/download/iphone" r...
2328 2015-11-16 23:23:41 <a href="http://twitter.com/download/iphone" r...
2329 2015-11-16 21:54:18 <a href="http://twitter.com/download/iphone" r...
2330 2015-11-16 21:10:36 <a href="http://twitter.com/download/iphone" r...
2331 2015-11-16 20:32:58 <a href="http://twitter.com/download/iphone" r...
2332 2015-11-16 20:01:42 <a href="http://twitter.com/download/iphone" r...
2333 2015-11-16 19:31:45 <a href="http://twitter.com/download/iphone" r...
2334 2015-11-16 16:37:02 <a href="http://twitter.com/download/iphone" r...
2335 2015-11-16 16:11:11 <a href="http://twitter.com/download/iphone" r...
2336 2015-11-16 15:14:19 <a href="http://twitter.com/download/iphone" r...
2337 2015-11-16 14:57:41 <a href="http://twitter.com/download/iphone" r...
2338 2015-11-16 04:02:55 <a href="http://twitter.com/download/iphone" r...
2339 2015-11-16 03:55:04 <a href="http://twitter.com/download/iphone" r...
2340 2015-11-16 03:44:34 <a href="http://twitter.com/download/iphone" r...
2341 2015-11-16 03:22:39 <a href="http://twitter.com/download/iphone" r...
2342 2015-11-16 02:38:37 <a href="http://twitter.com/download/iphone" r...
2343 2015-11-16 01:59:36 <a href="http://twitter.com/download/iphone" r...
2344 2015-11-16 01:52:02 <a href="http://twitter.com/download/iphone" r...
2345 2015-11-16 01:22:45 <a href="http://twitter.com/download/iphone" r...
2346 2015-11-16 01:01:59 <a href="http://twitter.com/download/iphone" r...
2347 2015-11-16 00:55:59 <a href="http://twitter.com/download/iphone" r...
2348 2015-11-16 00:49:46 <a href="http://twitter.com/download/iphone" r...
2349 2015-11-16 00:35:11 <a href="http://twitter.com/download/iphone" r...
2350 2015-11-16 00:30:50 <a href="http://twitter.com/download/iphone" r...
2351 2015-11-16 00:24:50 <a href="http://twitter.com/download/iphone" r...
2352 2015-11-16 00:04:52 <a href="http://twitter.com/download/iphone" r...
2353 2015-11-15 23:21:54 <a href="http://twitter.com/download/iphone" r...
2354 2015-11-15 23:05:30 <a href="http://twitter.com/download/iphone" r...
2355 2015-11-15 22:32:08 <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
20	This is Gerald. He was just told he didn't get...	NaN
21	This is Jeffrey. He has a monopoly on the pool...	NaN
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
23	This is Canela. She attempted some fancy porch...	NaN
24	You may not have known you needed to see this ...	NaN
25	This... is a Jubilant Antarctic House Bear. We...	NaN
26	This is Maya. She's very shy. Rarely leaves he...	NaN
27	This is Mingus. He's a wonderful father to his...	NaN
28	This is Derek. He's late for a dog meeting. 13...	NaN
29	This is Roscoe. Another pupper fallen victim t...	NaN
30	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
...
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN

2344	Here we have a northern speckled Rhododendron...	NaN
2345	This is the happiest dog you will ever see. Ve...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2348	Here is a Siberian heavily armored polar bear ...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2350	This is a truly beautiful English Wilson Staff...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	NaN	NaN
26	NaN	NaN
27	NaN	NaN
28	NaN	NaN
29	NaN	NaN
30	NaN	NaN
...
2326	NaN	NaN
2327	NaN	NaN
2328	NaN	NaN

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

	expanded_urls	rating_numerator \
0	https://twitter.com/dog_rates/status/892420643...	13
1	https://twitter.com/dog_rates/status/892177421...	13
2	https://twitter.com/dog_rates/status/891815181...	12
3	https://twitter.com/dog_rates/status/891689557...	13
4	https://twitter.com/dog_rates/status/891327558...	12
5	https://twitter.com/dog_rates/status/891087950...	13
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	13
7	https://twitter.com/dog_rates/status/890729181...	13
8	https://twitter.com/dog_rates/status/890609185...	13
9	https://twitter.com/dog_rates/status/890240255...	14
10	https://twitter.com/dog_rates/status/890006608...	13
11	https://twitter.com/dog_rates/status/889880896...	13
12	https://twitter.com/dog_rates/status/889665388...	13
13	https://twitter.com/dog_rates/status/889638837...	12
14	https://twitter.com/dog_rates/status/889531135...	13
15	https://twitter.com/dog_rates/status/889278841...	13
16	https://twitter.com/dog_rates/status/888917238...	12
17	https://twitter.com/dog_rates/status/888804989...	13
18	https://twitter.com/dog_rates/status/888554962...	13

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

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

4	10	Franklin	None	None	None	None
5	10	None	None	None	None	None
6	10	Jax	None	None	None	None
7	10	None	None	None	None	None
8	10	Zoey	None	None	None	None
9	10	Cassie	doggo	None	None	None
10	10	Koda	None	None	None	None
11	10	Bruno	None	None	None	None
12	10	None	None	None	None	puppo
13	10	Ted	None	None	None	None
14	10	Stuart	None	None	None	puppo
15	10	Oliver	None	None	None	None
16	10	Jim	None	None	None	None
17	10	Zeke	None	None	None	None
18	10	Ralphus	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None
28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
30	10	None	None	None	None	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2345	10	the	None	None	None	None
2346	10	the	None	None	None	None

```
[2175 rows x 17 columns]
```

```
In [24]: # confirm changes (retweeted_status_id should be 0)
df_archive.info()
```

Define 4. Drop unnecessary columns in `df_archive`. It is not properly an issue, but it will make it easier to analyze the data. That is why I have 9 bullet points in this section.

```
In [25]: # drop columns
df_archive.drop(['in_reply_to_status_id', 'in_reply_to_user_id', 'retweeted_status_id', 'in_reply_to_in_reply_to_status_id', 'in_reply_to_in_reply_to_user_id', 'retweeted_status_in_reply_to_status_id', 'retweeted_status_in_reply_to_user_id'], axis=1)
```

26

```
In [26]: # confirm changes
df_archive
```

```
Out[26]:
```

	tweet_id	timestamp	\
0	892420643555336193	2017-08-01 16:23:56	
1	892177421306343426	2017-08-01 00:17:27	
2	891815181378084864	2017-07-31 00:18:03	
3	891689557279858688	2017-07-30 15:58:51	
4	891327558926688256	2017-07-29 16:00:24	
5	891087950875897856	2017-07-29 00:08:17	
6	890971913173991426	2017-07-28 16:27:12	
7	890729181411237888	2017-07-28 00:22:40	
8	890609185150312448	2017-07-27 16:25:51	
9	890240255349198849	2017-07-26 15:59:51	
10	890006608113172480	2017-07-26 00:31:25	
11	889880896479866881	2017-07-25 16:11:53	
12	889665388333682689	2017-07-25 01:55:32	
13	889638837579907072	2017-07-25 00:10:02	
14	889531135344209921	2017-07-24 17:02:04	
15	889278841981685760	2017-07-24 00:19:32	
16	888917238123831296	2017-07-23 00:22:39	
17	888804989199671297	2017-07-22 16:56:37	
18	888554962724278272	2017-07-22 00:23:06	
20	888078434458587136	2017-07-20 16:49:33	
21	887705289381826560	2017-07-19 16:06:48	
22	887517139158093824	2017-07-19 03:39:09	
23	887473957103951883	2017-07-19 00:47:34	
24	887343217045368832	2017-07-18 16:08:03	
25	887101392804085760	2017-07-18 00:07:08	
26	886983233522544640	2017-07-17 16:17:36	
27	886736880519319552	2017-07-16 23:58:41	
28	886680336477933568	2017-07-16 20:14:00	
29	886366144734445568	2017-07-15 23:25:31	
30	886267009285017600	2017-07-15 16:51:35	
...	
2326	666411507551481857	2015-11-17 00:24:19	
2327	666407126856765440	2015-11-17 00:06:54	
2328	666396247373291520	2015-11-16 23:23:41	
2329	666373753744588802	2015-11-16 21:54:18	
2330	666362758909284353	2015-11-16 21:10:36	
2331	666353288456101888	2015-11-16 20:32:58	
2332	666345417576210432	2015-11-16 20:01:42	
2333	666337882303524864	2015-11-16 19:31:45	
2334	666293911632134144	2015-11-16 16:37:02	
2335	666287406224695296	2015-11-16 16:11:11	
2336	666273097616637952	2015-11-16 15:14:19	
2337	666268910803644416	2015-11-16 14:57:41	
2338	666104133288665088	2015-11-16 04:02:55	

2339 666102155909144576 2015-11-16 03:55:04
 2340 666099513787052032 2015-11-16 03:44:34
 2341 666094000022159362 2015-11-16 03:22:39
 2342 666082916733198337 2015-11-16 02:38:37
 2343 666073100786774016 2015-11-16 01:59:36
 2344 666071193221509120 2015-11-16 01:52:02
 2345 666063827256086533 2015-11-16 01:22:45
 2346 666058600524156928 2015-11-16 01:01:59
 2347 666057090499244032 2015-11-16 00:55:59
 2348 666055525042405380 2015-11-16 00:49:46
 2349 666051853826850816 2015-11-16 00:35:11
 2350 666050758794694657 2015-11-16 00:30:50
 2351 666049248165822465 2015-11-16 00:24:50
 2352 666044226329800704 2015-11-16 00:04:52
 2353 666033412701032449 2015-11-15 23:21:54
 2354 666029285002620928 2015-11-15 23:05:30
 2355 666020888022790149 2015-11-15 22:32:08

text \

0 This is Phineas. He's a mystical boy. Only eve...
 1 This is Tilly. She's just checking pup on you...
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 3 This is Darla. She commenced a snooze mid meal...
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 5 Here we have a majestic great white breaching ...
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 15 This is Oliver. You're witnessing one of his m...
 16 This is Jim. He found a fren. Taught him how t...
 17 This is Zeke. He has a new stick. Very proud o...
 18 This is Ralphus. He's powering up. Attempting ...
 20 This is Gerald. He was just told he didn't get...
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 22 I've yet to rate a Venezuelan Hover Wiener. Th...
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30 @NonWhiteHat @MayhewMayhem omg hello tanner yo...
 ...
 2326 This is quite the dog. Gets really excited whe...
 2327 This is a southern Vesuvius bumblegruff. Can d...
 2328 Oh goodness. A super rare northeast Qdoba kang...
 2329 Those are sunglasses and a jean jacket. 11/10 ...
 2330 Unique dog here. Very small. Lives in containe...
 2331 Here we have a mixed Asiago from the Galápagos...
 2332 Look at this jokester thinking seat belt laws ...
 2333 This is an extremely rare horned Parthenon. No...
 2334 This is a funny dog. Weird toes. Won't come do...
 2335 This is an Albanian 3 1/2 legged Episcopalian...
 2336 Can take selfies 11/10 https://t.co/ws2AMaWpW
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 2339 Oh my. Here you are seeing an Adobe Setter giv...
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 2341 This appears to be a Mongolian Presbyterian mi...
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 2346 Here is the Rand Paul of retrievers folks! He'...
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 2349 This is an odd dog. Hard on the outside but lo...
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 2351 Here we have a 1949 1st generation vulpix. Enj...
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 2353 Here is a very happy pup. Big fan of well-main...
 2354 This is a western brown Mitsubishi terrier. Up...
 2355 Here we have a Japanese Irish Setter. Lost eye...

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

	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None
1	10	Tilly	None	None	None	None
2	10	Archie	None	None	None	None
3	10	Darla	None	None	None	None
4	10	Franklin	None	None	None	None
5	10	None	None	None	None	None
6	10	Jax	None	None	None	None
7	10	None	None	None	None	None
8	10	Zoey	None	None	None	None
9	10	Cassie	doggo	None	None	None
10	10	Koda	None	None	None	None
11	10	Bruno	None	None	None	None
12	10	None	None	None	None	puppo
13	10	Ted	None	None	None	None
14	10	Stuart	None	None	None	puppo
15	10	Oliver	None	None	None	None
16	10	Jim	None	None	None	None
17	10	Zeke	None	None	None	None
18	10	Ralphus	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None
28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
30	10	None	None	None	None	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None

2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2345	10	the	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2348	10	a	None	None	None	None
2349	10	an	None	None	None	None
2350	10	a	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[2175 rows x 11 columns]

Define 5. Drop null values from **expanded_urls** in **df_archive**. The **expanded_urls** is the only column that has a few missing values. I could not determine why that was the case nor I could fill these values out; therefore, I will drop these values.

Code

```
In [27]: # drop null values
df_archive.dropna(inplace=True)
```

Test

```
In [28]: # confirm changes
df_archive.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2117 entries, 0 to 2355
Data columns (total 11 columns):
tweet_id          2117 non-null object
timestamp         2117 non-null datetime64[ns]
text              2117 non-null object
expanded_urls     2117 non-null object
rating_numerator  2117 non-null int64
rating_denominator 2117 non-null int64
name              2117 non-null object
doggo             2117 non-null object
floofer          2117 non-null object
pupper           2117 non-null object
puppo            2117 non-null object
dtypes: datetime64[ns](1), int64(2), object(8)
memory usage: 198.5+ KB
```

Define 6. Drop tweets that are not dogs. This analysis should only incorporate tweets of dogs; therefore, the next step is to delete rows that contain non-dog tweets. In order to do that, I

will merge the df_pred dataframe with df_archive and create a new dataframe df_archive2. The df_pred contains 3 predictions from the neural network whether a picture shows a dog. I assume that if any of the predictions suggest that it is a dog, I treated it as a dog. I checked a few tweets 'by hand' and it appeared to be true.

Code

```
In [29]: # merge predictions into the archive
        df_archive2 = df_archive.merge(df_pred, on=['tweet_id'], how='left')
```

```
In [30]: # check if the dataframes were merged correctly
        df_archive2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2117 entries, 0 to 2116
Data columns (total 22 columns):
tweet_id          2117 non-null object
timestamp         2117 non-null datetime64[ns]
text              2117 non-null object
expanded_urls     2117 non-null object
rating_numerator  2117 non-null int64
rating_denominator 2117 non-null int64
name              2117 non-null object
doggo             2117 non-null object
floofer           2117 non-null object
pupper           2117 non-null object
puppo            2117 non-null object
jpg_url          1994 non-null object
img_num          1994 non-null float64
p1               1994 non-null object
p1_conf          1994 non-null float64
p1_dog           1994 non-null object
p2               1994 non-null object
p2_conf          1994 non-null float64
p2_dog           1994 non-null object
p3               1994 non-null object
p3_conf          1994 non-null float64
p3_dog           1994 non-null object
dtypes: datetime64[ns](1), float64(4), int64(2), object(15)
memory usage: 380.4+ KB
```

```
In [31]: # select pictures if any of the predictions suggest that it is a dog
        df_archive2 = df_archive2.query('p1_dog == True | p2_dog == True | p3_dog == True')
```

Test

```
In [32]: # check how many dog tweets are left
        df_archive2.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1686 entries, 1 to 2116
Data columns (total 22 columns):
tweet_id          1686 non-null object
timestamp         1686 non-null datetime64[ns]
text              1686 non-null object
expanded_urls     1686 non-null object
rating_numerator  1686 non-null int64
rating_denominator 1686 non-null int64
name              1686 non-null object
doggo             1686 non-null object
floofer           1686 non-null object
pupper            1686 non-null object
puppo             1686 non-null object
jpg_url           1686 non-null object
img_num           1686 non-null float64
p1                1686 non-null object
p1_conf           1686 non-null float64
p1_dog            1686 non-null object
p2                1686 non-null object
p2_conf           1686 non-null float64
p2_dog            1686 non-null object
p3                1686 non-null object
p3_conf           1686 non-null float64
p3_dog            1686 non-null object
dtypes: datetime64[ns](1), float64(4), int64(2), object(15)
memory usage: 303.0+ KB

```

Define 7. Clean **rating_denominator** in **df_archive**. Extract the values from text, delete tweets without ratings and tweets with multiple dogs. Change the data type to float.

8. Clean **rating_denominator** in **df_archive**. Extract the values from text, delete tweets without ratings and tweets with multiple dogs. Change the data type to float.

Code

```

In [33]: # change column width to view full text
pd.options.display.max_colwidth = 200

In [34]: # import regular expression
import re

In [35]: # extract ratings from the text column into a new column 'ratings'.
# Some ratings need to be fixed manually because there are multiple xx/xx occurrences
# or there are multiple dogs in one picture
df_archive2['ratings'] = df_archive2['text'].str.extract(r'(\d+\.? \d*/\d+){1}', expand=1)

In [36]: # Extract rating numerators from the new column 'ratings' as float
df_archive2['rating_numerator'] = df_archive2['ratings'].str.split('/').str.get(0).astype(float)

```

```
In [37]: # Extract rating denominators from the new column 'ratings' as float
df_archive2['rating_denominator'] = df_archive2['ratings'].str.split('/').str.get(1).as

In [38]: # drop ratings column after the extraction
df_archive2.drop(['ratings'], axis=1, inplace=True)
```

In this section, I'm going to manually clean incorrectly extracted ratings. I will delete tweets with multiple dogs.

```
In [39]: df_archive2['rating_denominator'].value_counts()
```

```
Out[39]: 10.0      1669
         50.0        3
         80.0        2
         11.0        2
         130.0       1
         150.0       1
          2.0        1
         120.0       1
         110.0       1
         40.0        1
         90.0        1
         20.0        1
          7.0        1
         70.0        1
         Name: rating_denominator, dtype: int64
```

```
In [40]: df_archive2[df_archive2.rating_denominator == 70]
```

```
Out[40]:          tweet_id          timestamp \
341  820690176645140481  2017-01-15 17:52:40

341  The floofs have been released I repeat the floofs have been released. 84/70 https:

341  https://twitter.com/dog_rates/status/820690176645140481/photo/1,https://twitter.co

          rating_numerator  rating_denominator  name  doggo  floofer  pupper  ...  \
341              84.0              70.0  None  None  None  None  ...

          img_num          p1  p1_conf  p1_dog          p2  p2_conf  \
341          2.0  West_Highland_white_terrier  0.872064  True  kuvasz  0.059526

          p2_dog          p3  p3_conf  p3_dog
341  True  Samoyed  0.0374  True

[1 rows x 22 columns]
```

```

In [41]: # drop row 341 because it is multiple dogs
df_archive2.drop(341, inplace=True)

In [42]: df_archive2[df_archive2.rating_denominator == 7]

Out[42]:
      tweet_id      timestamp \
405  810984652412424192  2016-12-19 23:06:23

405  Meet Sam. She smiles 24/7 & secretly aspires to be a reindeer. \nKeep Sam smil

405  https://www.gofundme.com/sams-smile,https://twitter.com/dog_rates/status/810984652

      rating_numerator rating_denominator name doggo floofer pupper ... \
405                24.0                7.0 Sam None None None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf \
405      1.0 golden_retriever 0.871342 True Tibetan_mastiff 0.036708

      p2_dog      p3  p3_conf p3_dog
405  True Labrador_retriever 0.025823 True

[1 rows x 22 columns]

In [43]: # drop row 405 as there is no rating provided
df_archive2.drop(405, inplace=True)

In [44]: df_archive2[df_archive2.rating_denominator == 20]

Out[44]:
      tweet_id      timestamp \
953  722974582966214656  2016-04-21 02:25:47

      text \
953  Happy 4/20 from the squad! 13/10 for all https://t.co/eV1diwds8a

      expanded_urls \
953  https://twitter.com/dog_rates/status/722974582966214656/photo/1

      rating_numerator rating_denominator name doggo floofer pupper ... \
953                4.0                20.0 None None None None ...

      img_num      p1  p1_conf p1_dog      p2 \
953      1.0 Great_Dane 0.246762 True Greater_Swiss_Mountain_dog

      p2_conf p2_dog      p3  p3_conf p3_dog
953  0.126131 True Weimaraner 0.085297 True

[1 rows x 22 columns]

```

```

In [45]: df_archive2.loc[953, 'rating_denominator'] = 10
         df_archive2.loc[953, 'rating_numerator'] = 13

In [46]: df_archive2[df_archive2.rating_denominator == 40]

Out[46]:
          tweet_id          timestamp \
1218  697463031882764288  2016-02-10 16:51:59

1218  Happy Wednesday here's a bucket of pups. 44/40 would pet all at once https://t.co

          expanded_urls \
1218  https://twitter.com/dog_rates/status/697463031882764288/photo/1

          rating_numerator rating_denominator name doggo floofer pupper ... \
1218                44.0                40.0  None  None    None  None ...

          img_num          p1  p1_conf p1_dog          p2  p2_conf \
1218          1.0  Labrador_retriever  0.999885   True  golden_retriever  0.000098

          p2_dog  p3  p3_conf p3_dog
1218   True  pug  0.000008   True

[1 rows x 22 columns]

In [47]: # drop row 1218 because it is multiple dogs
         df_archive2.drop(1218, inplace=True)

In [48]: df_archive2[df_archive2.rating_denominator == 90]

Out[48]:
          tweet_id          timestamp \
1016  713900603437621249  2016-03-27 01:29:02

1016  Happy Saturday here's 9 puppies on a bench. 99/90 good work everybody https://t.co

          expanded_urls \
1016  https://twitter.com/dog_rates/status/713900603437621249/photo/1

          rating_numerator rating_denominator name doggo floofer pupper ... \
1016                99.0                90.0  None  None    None  None ...

          img_num          p1  p1_conf p1_dog          p2  p2_conf \
1016          1.0  golden_retriever  0.371816   True  cocker_spaniel  0.177413

          p2_dog          p3  p3_conf p3_dog
1016   True  Irish_setter  0.092725   True

[1 rows x 22 columns]

```

```

In [49]: # drop row 1016 because it is multiple dogs
df_archive2.drop(1016, inplace=True)

In [50]: df_archive2[df_archive2.rating_denominator == 110]

Out[50]:
      tweet_id      timestamp \
1411  684222868335505415  2016-01-05 04:00:18

      text \
1411  Someone help the girl is being mugged. Several are distracting her while two stea

      expanded_urls \
1411  https://twitter.com/dog_rates/status/684222868335505415/photo/1

      rating_numerator rating_denominator name doggo floofer pupper ... \
1411              121.0              110.0  None  None    None  None ...

      img_num      p1  p1_conf p1_dog      p2 \
1411      1.0  soft-coated_wheaten_terrier  0.791182  True  cocker_spaniel

      p2_conf p2_dog      p3  p3_conf p3_dog
1411  0.072444  True  teddy  0.071486  False

[1 rows x 22 columns]

In [51]: # drop row 1411 because it is multiple dogs
df_archive2.drop(1411, inplace=True)

In [52]: df_archive2[df_archive2.rating_denominator == 120]

Out[52]:
      tweet_id      timestamp \
1552  677716515794329600  2015-12-18 05:06:23

      text \
1552  IT'S PUPPERGEDDON. Total of 144/120 ...I think https://t.co/ZanVtAtvIq

      expanded_urls \
1552  https://twitter.com/dog_rates/status/677716515794329600/photo/1

      rating_numerator rating_denominator name doggo floofer pupper ... \
1552              144.0              120.0  None  None    None  None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf p2_dog      p3  p3_conf \
1552      1.0  teddy  0.662908  False  crib  0.031891  False  chow  0.025438

      p3_dog
1552  True

[1 rows x 22 columns]

```

```

In [53]: # drop row 1552 because it is multiple dogs
df_archive2.drop(1552, inplace=True)

In [54]: df_archive2[df_archive2.rating_denominator == 130]

Out[54]:
      tweet_id      timestamp \
1410  684225744407494656  2016-01-05 04:11:44

1410  Two sneaky puppies were not initially seen, moving the rating to 143/130. Please

1410  https://twitter.com/dog_rates/status/684225744407494656/photo/1,https://twitter.c

      rating_numerator rating_denominator name doggo floofer pupper ... \
1410                143.0                130.0 None None None None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf p2_dog \
1410      2.0  golden_retriever  0.203249  True  Samoyed  0.067958  True

      p3  p3_conf p3_dog
1410  Great_Pyrenees  0.065327  True

[1 rows x 22 columns]

In [55]: # drop row 1410 because it is multiple dogs
df_archive2.drop(1410, inplace=True)

In [56]: df_archive2[df_archive2.rating_denominator == 150]

Out[56]:
      tweet_id      timestamp \
702  758467244762497024  2016-07-28 01:00:57

      text \
702  Why does this never happen at my front door... 165/150 https://t.co/HmwrdfEfUE

      expanded_urls \
702  https://twitter.com/dog_rates/status/758467244762497024/video/1

      rating_numerator rating_denominator name doggo floofer pupper ... \
702                165.0                150.0 None None None None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf p2_dog \
702      1.0  Labrador_retriever  0.436377  True  Chihuahua  0.113956  True

      p3  p3_conf p3_dog
702  American_Staffordshire_terrier  0.099689  True

[1 rows x 22 columns]

```

```

In [57]: # drop row 1410 because it is multiple dogs
df_archive2.drop(702, inplace=True)

In [58]: df_archive2[df_archive2.rating_denominator == 2]

Out[58]:
      tweet_id      timestamp \
2096  666287406224695296  2015-11-16 16:11:11

2096  This is an Albanian 3 1/2 legged  Episcopalian. Loves well-polished hardwood floor

      expanded_urls \
2096  https://twitter.com/dog_rates/status/666287406224695296/photo/1

      rating_numerator rating_denominator name doggo floofer pupper ... \
2096                1.0                2.0   an  None   None   None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf p2_dog \
2096      1.0  Maltese_dog  0.857531  True  toy_poodle  0.063064  True

      p3  p3_conf p3_dog
2096  miniature_poodle  0.025581  True

[1 rows x 22 columns]

In [59]: # change row 2096
df_archive2.loc[2096, 'rating_denominator'] = 10
df_archive2.loc[2096, 'rating_numerator'] = 9

In [60]: df_archive2[df_archive2.rating_denominator == 11]

Out[60]:
      tweet_id      timestamp \
857  740373189193256964  2016-06-08 02:41:38
1438  682962037429899265  2016-01-01 16:30:13

857  After so many requests, this is Bretagne. She was the last surviving 9/11 search
1438  This is Darrel. He just robbed a 7/11 and is in a high speed police chase. Was j

857  https://twitter.com/dog_rates/status/740373189193256964/photo/1,https://twitter.o
1438

      rating_numerator rating_denominator   name doggo floofer pupper \
857                9.0                11.0   None  None   None  None
1438                7.0                11.0  Darrel  None   None  None

      ...  img_num      p1  p1_conf p1_dog      p2  p2_conf \
857  ...      3.0  golden_retriever  0.807644  True      kuvasz  0.101286

```



```

1438 ...          1.0                dingo 0.278600 False Chihuahua 0.155207

          p2_dog                p3    p3_conf p3_dog
857    True  Labrador_retriever 0.023785    True
1438    True                loupe 0.153598    False

[2 rows x 22 columns]

In [61]: df_archive2.loc[857, 'rating_denominator'] = 10
df_archive2.loc[857, 'rating_numerator'] = 14

In [62]: df_archive2.loc[1438, 'rating_denominator'] = 10
df_archive2.loc[1438, 'rating_numerator'] = 10

In [63]: df_archive2[df_archive2.rating_denominator == 80]

Out[63]:
          tweet_id          timestamp \
1041  710658690886586372 2016-03-18 02:46:49
1615  675853064436391936 2015-12-13 01:41:41

1041  Here's a brigade of puppies. All look very prepared for whatever happens next. 80
1615  Here we have an entire platoon of puppies. Total score: 88/80 would pet all at o

1041
1615  https://twitter.com/dog_rates/status/675853064436391936/photo/1,https://twitter.c

          rating_numerator  rating_denominator  name doggo floofer pupper ... \
1041                80.0                80.0  None  None    None  None ...
1615                88.0                80.0  None  None    None  None ...

          img_num                p1    p1_conf p1_dog                p2 \
1041          1.0  soft-coated_wheaten_terrier 0.948617    True    Dandie_Dinmont
1615          1.0    Labrador_retriever 0.868367    True    golden_retriever

          p2_conf p2_dog          p3    p3_conf p3_dog
1041  0.018664    True    cairn 0.015943    True
1615  0.043305    True    vizsla 0.028207    True

[2 rows x 22 columns]

In [64]: # drop rows 1041 and 1615 because it is multiple dogs
df_archive2.drop([1041,1615], inplace=True)

In [65]: df_archive2[df_archive2.rating_denominator == 50]

Out[65]:
          tweet_id          timestamp \
990    716439118184652801 2016-04-03 01:36:11

```

```
1061 709198395643068416 2016-03-14 02:04:08
1136 704054845121142784 2016-02-28 21:25:30
```

```
990                                     This is Bluebert. He just saw that both #FinalFur ma
1061 From left to right:\nCletus, Jerome, Alejandro, Burp, & Titson\nNone know whe
1136                                     Here is a whole f
```

```

                                     expanded_urls \
990 https://twitter.com/dog_rates/status/716439118184652801/photo/1
1061 https://twitter.com/dog_rates/status/709198395643068416/photo/1
1136 https://twitter.com/dog_rates/status/704054845121142784/photo/1
```

```

rating_numerator rating_denominator name doggo floofer pupper \
990              50.0              50.0 Bluebert None None None
1061              45.0              50.0 None None None None
1136              60.0              50.0 a None None None
```

```

... img_num p1 p1_conf p1_dog p2 \
990 ... 1.0 Siberian_husky 0.396495 True malamute
1061 ... 1.0 borzoi 0.490783 True wire-haired_fox_terrier
1136 ... 1.0 Great_Pyrenees 0.667939 True kuvasz
```

```

p2_conf p2_dog p3 p3_conf p3_dog
990 0.317053 True Eskimo_dog 0.273419 True
1061 0.083513 True English_setter 0.083184 True
1136 0.228764 True golden_retriever 0.043885 True
```

```
[3 rows x 22 columns]
```

```
In [66]: df_archive2.loc[990, 'rating_denominator'] = 10
df_archive2.loc[990, 'rating_numerator'] = 11
```

```
In [67]: # drop rows 1061 and 1136 because it is multiple dogs
df_archive2.drop([1061,1136], inplace=True)
```

Clean Rating Numerator

```
In [68]: df_archive2.rating_numerator.value_counts()
```

```
Out[68]: 12.00    423
          10.00    361
          11.00    357
          13.00    230
          9.00     133
          8.00     68
          7.00     31
          14.00     24
          6.00     16
```

```

5.00      13
4.00      6
3.00      5
2.00      2
13.50     1
11.27     1
0.00      1
11.26     1
9.75      1
Name: rating_numerator, dtype: int64

```

```
In [69]: df_archive2[df_archive2.rating_numerator == 0]
```

```

Out[69]:
      tweet_id      timestamp \
249  835152434251116546  2017-02-24 15:40:31

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

249  https://twitter.com/dog_rates/status/835152434251116546/photo/1,https://twitter.co

      rating_numerator  rating_denominator  name  doggo  floofer  pupper  ... \
249                0.0                10.0  None  None    None    None  ...

      img_num    p1    p1_conf  p1_dog                                p2    p2_conf \
249      3.0  swing  0.967066  False  American_Staffordshire_terrier  0.012731

      p2_dog                                p3    p3_conf  p3_dog
249    True  Staffordshire_bullterrier  0.007039    True

[1 rows x 22 columns]

```

```
In [70]: df_archive2[df_archive2.rating_numerator == 2]
```

```

Out[70]:
      tweet_id      timestamp \
1538  678424312106393600  2015-12-20 03:58:55
2010  667878741721415682  2015-11-21 01:34:35

1538  This is Crystal. She's a shitty fireman. No sense of urgency. People could be dyi
2010  This is Tedrick. He lives on the edge. Needs someone to hit the gas tho. Other t

      expanded_urls \
1538  https://twitter.com/dog_rates/status/678424312106393600/photo/1
2010  https://twitter.com/dog_rates/status/667878741721415682/photo/1

      rating_numerator  rating_denominator  name  doggo  floofer  pupper \
1538                2.0                10.0  Crystal  None    None    None

```

2010		2.0		10.0	Tedrick	None	None	None
------	--	-----	--	------	---------	------	------	------

	...	img_num		p1	p1_conf	p1_dog		p2 \
1538	...	1.0	Maltese_dog	0.759945	True		toy_poodle	
2010	...	1.0	seat_belt	0.200373	False	miniature_pinscher		

		p2_conf	p2_dog		p3	p3_conf	p3_dog
1538	0.101194	True	Shih-Tzu	0.056037	True		
2010	0.106003	True	schipperke	0.104733	True		

[2 rows x 22 columns]

```
In [71]: df_archive2[df_archive2.rating_numerator == 3]
```

```
Out[71]:
```

		tweet_id		timestamp	\
589	777885040357281792	2016-09-19	15:00:20		
977	718246886998687744	2016-04-08	01:19:36		
1036	711306686208872448	2016-03-19	21:41:44		
1706	673906403526995968	2015-12-07	16:46:21		
1948	668989615043424256	2015-11-24	03:08:48		

589	This is Wesley. He's clearly trespassing. Seems rather h*ckin violent too. Weapon
977	This is Alexanderson. He's got a weird ass birth mark. Dreadful at fetch.
1036	What hooligan sent in pictures w/out a dog in them? Churlish af. 3/10 just bc
1706	Guys I'm getting real tired of this. We only rate dogs. Please don't send in oth
1948	This is Bernie. He's taking his Halloween costume very seriously. Wants to be bak

589	https://twitter.com/dog_rates/status/777885040357281792/photo/1 , https://twitter.c
977	https://twitter.c
1036	https://twitter.com/dog_rates/status/711306686208872448/photo/1 , https://twitter.c
1706	https://twitter.c
1948	https://twitter.c

	rating_numerator	rating_denominator		name	doggo	floofer	pupper	\
589	3.0	10.0		Wesley	None	None	None	
977	3.0	10.0		Alexanderson	None	None	None	
1036	3.0	10.0		None	None	None	None	
1706	3.0	10.0		None	None	None	None	
1948	3.0	10.0		Bernie	None	None	None	

	...	img_num		p1	p1_conf	p1_dog	\
589	...	1.0	Afghan_hound	0.123529	True		
977	...	1.0	Chihuahua	0.354488	True		
1036	...	1.0	leatherback_turtle	0.280835	False		
1706	...	1.0	toilet_seat	0.683319	False		
1948	...	1.0	pug	0.917326	True		

		p2	p2_conf	p2_dog		p3	p3_conf	\
589		basset	0.119682	True	Siberian_husky	0.108709		
977		carton	0.159672	False	Siberian_husky	0.057498		
1036		loggerhead	0.123290	False	Dandie_Dinmont	0.086792		
1706	soft-coated_wheaten_terrier		0.048928	True	Siberian_husky	0.030386		
1948	waffle_iron		0.014918	False	Chihuahua	0.013524		

	p3_dog
589	True
977	True
1036	True
1706	True
1948	True

[5 rows x 22 columns]

```
In [72]: df_archive2[df_archive2.rating_numerator == 4]
```

```
Out[72]:
```

	tweet_id	timestamp	\
797	747816857231626240	2016-06-28 15:40:07	
1089	707420581654872064	2016-03-09 04:19:44	
1475	680940246314430465	2015-12-27 02:36:20	
1836	671122204919246848	2015-11-30 00:22:57	
2050	667176164155375616	2015-11-19 03:02:47	
2077	666649482315059201	2015-11-17 16:09:56	

797	Viewer discretion is advised. This is a terrible attack in progress. Not even i
1089	This is Keurig. He's a rare dog. Laughs like an idiot tho. Head is basically a we
1475	
1836	Two miniature golden retrievers here. Webbed paws. Don't walk very efficientl
2050	These are strange dogs. All have toupees. Long neck for dogs. In a shed of sorts?
2077	Cool dog. Enjoys couch. Low monotone bark. Very nice kicks. Pisses milk (must b

	expanded_urls	\
797	https://twitter.com/dog_rates/status/747816857231626240/photo/1	
1089	https://twitter.com/dog_rates/status/707420581654872064/photo/1	
1475	https://twitter.com/dog_rates/status/680940246314430465/photo/1	
1836	https://twitter.com/dog_rates/status/671122204919246848/photo/1	
2050	https://twitter.com/dog_rates/status/667176164155375616/photo/1	
2077	https://twitter.com/dog_rates/status/666649482315059201/photo/1	

	rating_numerator	rating_denominator	name	doggo	floofer	pupper	\
797	4.0	10.0	a	None	None	None	
1089	4.0	10.0	Keurig	None	None	None	
1475	4.0	10.0	Alice	None	None	None	
1836	4.0	10.0	None	None	None	None	

2050	4.0	10.0	None	None	None	None
2077	4.0	10.0	None	None	None	None

	...	img_num		p1	p1_conf	p1_dog	\
797	...	1.0		Pembroke	0.768923	True	
1089	...	1.0		ram	0.518215	False	
1475	...	1.0	soft-coated_wheaten_terrier		0.289598	True	
1836	...	1.0		goose	0.351957	False	
2050	...	1.0	soft-coated_wheaten_terrier		0.318981	True	
2077	...	1.0		Border_collie	0.447803	True	

		p2	p2_conf	p2_dog		p3	\
797		Chihuahua	0.029053	True	Shetland_sheepdog		
1089		kuvasz	0.149391	True	Great_Pyrenees		
1475	West_Highland_white_terrier		0.157195	True	toy_poodle		
1836		Chihuahua	0.101228	True	hen		
2050		Lakeland_terrier	0.215218	True	toy_poodle		
2077		English_springer	0.170497	True	collie		

	p3_conf	p3_dog
797	0.029035	True
1089	0.106003	True
1475	0.074435	True
1836	0.065818	False
2050	0.106014	True
2077	0.139206	True

[6 rows x 22 columns]

In [73]: df_archive2[df_archive2.rating_numerator == 5]

Out[73]:

	tweet_id	timestamp	\
561	781661882474196992	2016-09-30 01:08:10	
750	751583847268179968	2016-07-09 01:08:47	
1288	691483041324204033	2016-01-25 04:49:38	
1421	683849932751646720	2016-01-04 03:18:23	
1569	677301033169788928	2015-12-17 01:35:24	
1592	676588346097852417	2015-12-15 02:23:26	
1671	674646392044941312	2015-12-09 17:46:48	
1746	672980819271634944	2015-12-05 03:28:25	
1858	670782429121134593	2015-11-29 01:52:48	
1946	668994913074286592	2015-11-24 03:29:51	
1970	668631377374486528	2015-11-23 03:25:17	
2073	666776908487630848	2015-11-18 00:36:17	
2112	666049248165822465	2015-11-16 00:24:50	

561 Who keeps sending in pictures without dogs in them? This ne

750 Please stop sending it pictures that don't even have a doggo or pupper in
 1288 When bae says they can't go out but you see them with someone else that same nigh
 1421 This is Jiminy. He's not the brightest dog. Needs to
 1569 This is Juckson. He's totally on his
 1592 This is Bubbles. He kinda resembles a fish. Always makes eye contact with u n
 1671 Two gorgeous dogs here. Little waddling dog is a rebel. Refuses to look at cam
 1746 Extraordinary dog here. Looks large. Just a head. No body.
 1858 This dude slap
 1946 Two gorgeous pups here. Both have cute fake horns(adorable). Barn in the back
 1970 Meet Zeek. He is a grey Cumulonimbus. Zeek is hungry. Someone should fee
 2073 This is Josep. He is a Rye Manganese mix. Can drive w eyes closed. Very irre
 2112 Here we have a 1949 1st generation vulpix. Enjoys sweat t

561
 750
 1288 https://twitter.com/dog_rates/status/691483041324204033/photo/1,<https://twitter.c>
 1421
 1569
 1592
 1671
 1746
 1858
 1946
 1970
 2073
 2112

	rating_numerator	rating_denominator	name	doggo	floofer	pupper	\
561	5.0	10.0	None	None	None	None	
750	5.0	10.0	None	doggo	None	pupper	
1288	5.0	10.0	None	None	None	None	
1421	5.0	10.0	Jiminy	None	None	None	
1569	5.0	10.0	Juckson	None	None	None	
1592	5.0	10.0	Bubbles	None	None	None	
1671	5.0	10.0	None	None	None	None	
1746	5.0	10.0	None	None	None	None	
1858	5.0	10.0	None	None	None	None	
1946	5.0	10.0	None	None	None	None	
1970	5.0	10.0	Zeek	None	None	None	
2073	5.0	10.0	Josep	None	None	None	
2112	5.0	10.0	None	None	None	None	

	...	img_num	p1	p1_conf	p1_dog	\
561	...	1.0	Pembroke	0.438087	True	
750	...	1.0	dalmatian	0.868304	True	
1288	...	1.0	bloodhound	0.886232	True	
1421	...	1.0	hog	0.458855	False	

1569	...	1.0	Japanese_spaniel	0.661178	True
1592	...	1.0	Boston_bull	0.976577	True
1671	...	1.0	flat-coated_retriever	0.837448	True
1746	...	1.0	car_mirror	0.232754	False
1858	...	1.0	Chihuahua	0.952963	True
1946	...	1.0	hog	0.113789	False
1970	...	1.0	miniature_schnauzer	0.904549	True
2073	...	1.0	seat_belt	0.375057	False
2112	...	1.0	miniature_pinscher	0.560311	True

		p2	p2_conf	p2_dog		p3	p3_conf	\
561	golden_retriever	0.226954	True		collie	0.070652		
750	studio_couch	0.059623	False		snow_leopard	0.013876		
1288	black-and-tan_coonhound	0.077420	True		Gordon_setter	0.009826		
1421	Mexican_hairless	0.164906	True		wild_boar	0.111700		
1569	Pekinese	0.150119	True		Chihuahua	0.119720		
1592	French_bulldog	0.014324	True		Chihuahua	0.002302		
1671	groenendael	0.086166	True	Labrador_retriever		0.016052		
1746	basset	0.219461	True		beagle	0.112397		
1858	French_bulldog	0.036575	True		Boston_bull	0.001977		
1946	English_springer	0.089763	True		French_bulldog	0.082186		
1970	Australian_terrier	0.022529	True		silky_terrier	0.015243		
2073	miniature_pinscher	0.167175	True		Chihuahua	0.086951		
2112	Rottweiler	0.243682	True		Doberman	0.154629		

	p3_dog
561	True
750	False
1288	True
1421	False
1569	True
1592	True
1671	True
1746	True
1858	True
1946	True
1970	True
2073	True
2112	True

[13 rows x 22 columns]

```
In [74]: df_archive2[df_archive2.rating_numerator == 6]
```

```
Out[74]:
```

	tweet_id	timestamp	\
186	848212111729840128	2017-04-01 16:35:01	
786	748575535303884801	2016-06-30 17:54:50	
867	739485634323156992	2016-06-05 15:54:48	

1173 700462010979500032 2016-02-18 23:28:52
 1243 695051054296211456 2016-02-04 01:07:39
 1533 678740035362037760 2015-12-21 00:53:29
 1560 677565715327688705 2015-12-17 19:07:09
 1611 675891555769696257 2015-12-13 04:14:39
 1739 673240798075449344 2015-12-05 20:41:29
 1837 671115716440031232 2015-11-29 23:57:10
 1866 670704688707301377 2015-11-28 20:43:53
 1928 669375718304980992 2015-11-25 04:43:02
 2002 668113020489474048 2015-11-21 17:05:31
 2083 666430724426358785 2015-11-17 01:40:41
 2103 666082916733198337 2015-11-16 02:38:37
 2113 666044226329800704 2015-11-16 00:04:52

186 This is Jerry. He's doing a distinguished tongue slip. Slightly patronizing tbh.
 786 This is one of the most reckless puppers I've ever seen. Ho
 867 This is Kyle. He's a heavy drinker and an avid pot user.
 1173 This is Murphy. He's a mini golden retriever. Missing two
 1243 Meet Brian (pronounced "Kirk"). He's not amused by ur churl
 1533 Meet Tango. He's a large dog. Doesn't care much for pers
 1560 Contortionist pup here. Inside pentagram. Clearly worships
 1611 This is Donny. He's summoning the demon monster Babadook.
 1739 Magical floating dog here. Very calm. Always hangs by the p
 1837 Meet Phred. He isn't steering, looking at the road, or wear
 1866 Meet Danny. He's too good to look at the road when he'
 1928 This is Billl. He's trying to
 2002 This is Alfie. He's that one hypocritical gym teach
 2083 Oh boy what a pup! Sunglasses take this one to t
 2103 Here we have a well-established sunblockersp
 2113 This is a purebred Piers Morgan. Loves to Netflix and ch

186 <https://twitter.c>
 786 <https://twitter.c>
 867 https://twitter.com/dog_rates/status/739485634323156992/photo/1,<https://twitter.c>
 1173 <https://twitter.c>
 1243 https://twitter.com/dog_rates/status/695051054296211456/photo/1,<https://twitter.c>
 1533 <https://twitter.c>
 1560 <https://twitter.c>
 1611 <https://twitter.c>
 1739 <https://twitter.c>
 1837 <https://twitter.c>
 1866 <https://twitter.c>
 1928 <https://twitter.c>
 2002 <https://twitter.c>
 2083 <https://twitter.c>
 2103 <https://twitter.c>

	rating_numerator	rating_denominator	name	doggo	floofer	pupper	\
186	6.0	10.0	Jerry	None	None	None	
786	6.0	10.0	one	None	None	None	
867	6.0	10.0	Kyle	None	None	None	
1173	6.0	10.0	Murphy	None	None	None	
1243	6.0	10.0	Brian	None	None	None	
1533	6.0	10.0	Tango	None	None	None	
1560	6.0	10.0	None	None	None	None	
1611	6.0	10.0	Donny	None	None	None	
1739	6.0	10.0	None	None	None	None	
1837	6.0	10.0	Phred	None	None	None	
1866	6.0	10.0	Danny	None	None	None	
1928	6.0	10.0	Billl	None	None	None	
2002	6.0	10.0	Alfie	None	None	None	
2083	6.0	10.0	None	None	None	None	
2103	6.0	10.0	None	None	None	None	
2113	6.0	10.0	a	None	None	None	

	...	img_num	p1	p1_conf	p1_dog	\
186	...	1.0	Bedlington_terrier	0.333486	True	
786	...	1.0	muzzle	0.176172	False	
867	...	2.0	Walker_hound	0.640256	True	
1173	...	1.0	hamster	0.678651	False	
1243	...	1.0	Boston_bull	0.761454	True	
1533	...	1.0	seat_belt	0.787164	False	
1560	...	1.0	basset	0.397295	True	
1611	...	1.0	Italian_greyhound	0.305637	True	
1739	...	1.0	Airedale	0.443004	True	
1837	...	1.0	malinois	0.406341	True	
1866	...	1.0	Norwich_terrier	0.419838	True	
1928	...	1.0	Airedale	0.168762	True	
2002	...	1.0	Pembroke	0.548896	True	
2083	...	1.0	llama	0.505184	False	
2103	...	1.0	pug	0.489814	True	
2113	...	1.0	Rhodesian_ridgeback	0.408143	True	

		p2	p2_conf	p2_dog		p3	\
186	Ibizan_hound	0.245797	True			wallaby	
786	seat_belt	0.160953	False		soft-coated-wheaten_terrier		
867	English_foxhound	0.229799	True			beagle	
1173	Pomeranian	0.110268	True			Angora	
1243	pug	0.075395	True			Chihuahua	
1533	sunglasses	0.045739	False			beagle	
1560	Welsh_springer_spaniel	0.199554	True			purse	
1611	whippet	0.232057	True			Great_Dane	
1739	brown_bear	0.114162	False		Chesapeake_Bay_retriever		

1837	kelpie	0.143366	True	dingo
1866	cairn	0.351876	True	Norfolk_terrier
1928	Norfolk_terrier	0.107479	True	Lakeland_terrier
2002	Cardigan	0.191101	True	collie
2083	Irish_terrier	0.104109	True	dingo
2103	bull_mastiff	0.404722	True	French_bulldog
2113	redbone	0.360687	True	miniature_pinscher

	p3_conf	p3_dog
186	0.131647	False
786	0.086499	True
867	0.037754	True
1173	0.104139	False
1243	0.041598	True
1533	0.022525	True
1560	0.105641	False
1611	0.117806	True
1739	0.094639	True
1837	0.129802	False
1866	0.051094	True
1928	0.097846	True
2002	0.059814	True
2083	0.062071	False
2103	0.048960	True
2113	0.222752	True

[16 rows x 22 columns]

```
In [75]: df_archive2[df_archive2.rating_numerator == 7]
```

```
Out[75]:
```

	tweet_id	timestamp	\
696	759099523532779520	2016-07-29 18:53:24	
819	745712589599014916	2016-06-22 20:18:30	
1111	706166467411222528	2016-03-05 17:16:20	
1200	698703483621523456	2016-02-14 03:01:06	
1231	696405997980676096	2016-02-07 18:51:43	
1352	687494652870668288	2016-01-14 04:41:12	
1358	687312378585812992	2016-01-13 16:36:55	
1385	685641971164143616	2016-01-09 01:59:19	
1504	679854723806179328	2015-12-24 02:42:51	
1511	679530280114372609	2015-12-23 05:13:38	
1513	679511351870550016	2015-12-23 03:58:25	
1528	678798276842360832	2015-12-21 04:44:55	
1643	675146535592706048	2015-12-11 02:54:12	
1723	673612854080196609	2015-12-06 21:19:54	
1732	673345638550134785	2015-12-06 03:38:05	
1765	672538107540070400	2015-12-03 22:09:14	
1779	672239279297454080	2015-12-03 02:21:48	

1789 672068090318987265 2015-12-02 15:01:33
 1843 670833812859932673 2015-11-29 05:16:59
 1854 670792680469889025 2015-11-29 02:33:32
 1859 670780561024270336 2015-11-29 01:45:22
 1962 668815180734689280 2015-11-23 15:35:39
 1999 668171859951755264 2015-11-21 20:59:20
 2034 667491009379606528 2015-11-19 23:53:52
 2036 667455448082227200 2015-11-19 21:32:34
 2041 667405339315146752 2015-11-19 18:13:27
 2058 667090893657276420 2015-11-18 21:23:57
 2081 666437273139982337 2015-11-17 02:06:42
 2084 666428276349472768 2015-11-17 01:30:57
 2088 666407126856765440 2015-11-17 00:06:54
 2115 666029285002620928 2015-11-15 23:05:30

696 Meet Toby. He has a drinking problem. Inflatable marijuana plant in the back is a
 819 This is Percy. He fell asleep at the wheel. Irresponsible af.
 1111 This is Dexter. He's a shy pup. Doesn't bark much. Dreadful fetcher. Has r
 1200 This is Rusty. He has no respect for POULTRY products. Un
 1231 This is Berb. He just found out that they have made 31 Kidz Bop CD's. Downright
 1352 This is Marq
 1358 "You got any games on your phone" 7/10
 1385 This is Otis. He just passed a cop while going 61
 1504 This is Bruce. He's a rare pup. Covered in Frosted Flakes. Nifty gold teeth. Over
 1511 Guys this really needs to stop. We've been over this way too many times. This is
 1513 Say hello to William. He makes fun of others because he's terrified of his
 1528 This is L
 1643 This is Coops. He's yelling at the carp
 1723 This is Jeffri. He's a speckled ice pupper. Very lazy. Enjoys the occasional
 1732 This is Gerald. He's a fluffy lil yellow pup. Always looks like his favorite
 1765 Meet Darby. He's a Fiscal Tutankhamen
 1779 This is Ridley.
 1789 Say hello to Gizmo. He's upset because he's not sure if he's really big or the
 1843 This is Jett
 1854 This is Antony. He's a Sheraton Tetrahedron. Skips awkwardly. Doesn't look when h
 1859 This is Traviss. He has no ears. Two rare dogs in background. I bet they all
 1962 This is a wild Toblerone from Papua New Guinea. Mouth always open. Addicted to
 1999 This is a Trans Siberian Kellogg named Alfonso. Huge ass eyeballs. Act
 2034 Two dogs in this one. Both are rare Jujitsu Pythagoreans. One slightly whiter t
 2036 This is Reese and Twips. Reese protects Twips. Both think they're too good for
 2041 This i
 2058 This is Clybe. He is an Anemone Valdez. One ear works. Can look in 2 different di
 2081 Here we see a lone northeastern Cumberbatch. Half ladybug. Only builds with bri
 2084 Here we have an Austrian Pulitzer. Collectors edition. Le
 2088 This is a southern Vesuvius bumblegruff. Can drive a truck (wow). Made friends w
 2115 This is a western brown Mitsubishi terrier. Upset about leaf. Actually 2 dogs he

	expanded_urls \
696	https://twitter.com/dog_rates/status/759099523532779520/photo/1
819	https://twitter.com/dog_rates/status/745712589599014916/photo/1
1111	https://twitter.com/dog_rates/status/706166467411222528/photo/1
1200	https://twitter.com/dog_rates/status/698703483621523456/photo/1
1231	https://twitter.com/dog_rates/status/696405997980676096/photo/1
1352	https://twitter.com/dog_rates/status/687494652870668288/photo/1
1358	https://twitter.com/dog_rates/status/687312378585812992/photo/1
1385	https://twitter.com/dog_rates/status/685641971164143616/photo/1
1504	https://twitter.com/dog_rates/status/679854723806179328/photo/1
1511	https://twitter.com/dog_rates/status/679530280114372609/photo/1
1513	https://twitter.com/dog_rates/status/679511351870550016/photo/1
1528	https://twitter.com/dog_rates/status/678798276842360832/photo/1
1643	https://twitter.com/dog_rates/status/675146535592706048/photo/1
1723	https://twitter.com/dog_rates/status/673612854080196609/photo/1
1732	https://twitter.com/dog_rates/status/673345638550134785/photo/1
1765	https://twitter.com/dog_rates/status/672538107540070400/photo/1
1779	https://twitter.com/dog_rates/status/672239279297454080/photo/1
1789	https://twitter.com/dog_rates/status/672068090318987265/photo/1
1843	https://twitter.com/dog_rates/status/670833812859932673/photo/1
1854	https://twitter.com/dog_rates/status/670792680469889025/photo/1
1859	https://twitter.com/dog_rates/status/670780561024270336/photo/1
1962	https://twitter.com/dog_rates/status/668815180734689280/photo/1
1999	https://twitter.com/dog_rates/status/668171859951755264/photo/1
2034	https://twitter.com/dog_rates/status/667491009379606528/photo/1
2036	https://twitter.com/dog_rates/status/667455448082227200/photo/1
2041	https://twitter.com/dog_rates/status/667405339315146752/photo/1
2058	https://twitter.com/dog_rates/status/667090893657276420/photo/1
2081	https://twitter.com/dog_rates/status/666437273139982337/photo/1
2084	https://twitter.com/dog_rates/status/666428276349472768/photo/1
2088	https://twitter.com/dog_rates/status/666407126856765440/photo/1
2115	https://twitter.com/dog_rates/status/666029285002620928/photo/1

	rating_numerator	rating_denominator	name	doggo	floofer	pupper \
696	7.0	10.0	Toby	None	None	None
819	7.0	10.0	Percy	None	None	None
1111	7.0	10.0	Dexter	None	None	None
1200	7.0	10.0	Rusty	None	None	None
1231	7.0	10.0	Berb	None	None	None
1352	7.0	10.0	Marq	None	None	None
1358	7.0	10.0	None	None	None	pupper
1385	7.0	10.0	Otis	None	None	pupper
1504	7.0	10.0	Bruce	None	None	None
1511	7.0	10.0	a	None	None	None
1513	7.0	10.0	William	None	None	None
1528	7.0	10.0	Linda	None	None	None
1643	7.0	10.0	Coops	None	None	None
1723	7.0	10.0	Jeffri	None	None	pupper

1732	7.0	10.0	Gerald	None	None	None
1765	7.0	10.0	Darby	None	None	None
1779	7.0	10.0	Ridley	None	None	None
1789	7.0	10.0	Gizmo	None	None	None
1843	7.0	10.0	Jett	None	None	None
1854	7.0	10.0	Antony	None	None	None
1859	7.0	10.0	Traviss	None	None	None
1962	7.0	10.0	a	None	None	None
1999	7.0	10.0	a	None	None	None
2034	7.0	10.0	None	None	None	None
2036	7.0	10.0	Reese	None	None	None
2041	7.0	10.0	Biden	None	None	None
2058	7.0	10.0	Clybe	None	None	None
2081	7.0	10.0	None	None	None	None
2084	7.0	10.0	None	None	None	None
2088	7.0	10.0	a	None	None	None
2115	7.0	10.0	a	None	None	None

	img_num	p1	p1_conf	p1_dog	\
696	1.0	Shetland_sheepdog	0.129034	True	
819	1.0	seat_belt	0.379055	False	
1111	1.0	Samoyed	0.430418	True	
1200	1.0	Brittany_spaniel	0.931963	True	
1231	1.0	borzoi	0.132845	True	
1352	1.0	Rottweiler	0.391471	True	
1358	1.0	seat_belt	0.703561	False	
1385	1.0	Lakeland_terrier	0.253839	True	
1504	1.0	llama	0.887963	False	
1511	1.0	dalmatian	0.750256	True	
1513	1.0	Chihuahua	0.761972	True	
1528	1.0	Airedale	0.583122	True	
1643	1.0	dingo	0.288447	False	
1723	1.0	Newfoundland	0.223101	True	
1732	1.0	hamster	0.761025	False	
1765	1.0	Siamese_cat	0.383937	False	
1779	1.0	pug	0.332536	True	
1789	1.0	pug	0.863385	True	
1843	1.0	Pekinese	0.609853	True	
1854	1.0	brown_bear	0.882426	False	
1859	1.0	Labrador_retriever	0.244889	True	
1962	1.0	redbone	0.461172	True	
1999	1.0	Chihuahua	0.664834	True	
2034	1.0	borzoi	0.852088	True	
2036	1.0	Tibetan_terrier	0.676376	True	
2041	1.0	Saint_Bernard	0.381377	True	
2058	1.0	Chihuahua	0.959514	True	
2081	1.0	Chihuahua	0.671853	True	
2084	1.0	Pembroke	0.371361	True	

2088	...	1.0	black-and-tan_coonhound	0.529139	True
2115	...	1.0	redbone	0.506826	True

		p2	p2_conf	p2_dog	p3 \
696		kelpie	0.117508	True	Siberian_husky
819		chow	0.062754	True	minibus
1111		kuvasz	0.279600	True	Great_Pyrenees
1200	Welsh_springer_spaniel		0.030695	True	beagle
1231	Walker_hound		0.086005	True	Great_Pyrenees
1352	miniature_pinscher		0.273595	True	Tibetan_mastiff
1358	Great_Dane		0.139909	True	Weimaraner
1385	Airedale		0.213349	True	three-toed_sloth
1504	ram		0.095021	False	chow
1511	jaguar		0.169007	False	zebra
1513	black-footed_ferret		0.150605	False	squirrel_monkey
1528	silky_terrier		0.129567	True	Lakeland_terrier
1643	Cardigan		0.229944	True	Pembroke
1723	Leonberg		0.111106	True	shovel
1732	weasel		0.069362	False	Pomeranian
1765	Chihuahua		0.160274	True	giant_panda
1779	French_bulldog		0.258124	True	bull_mastiff
1789	shopping_cart		0.125746	False	Border_terrier
1843	Persian_cat		0.265442	False	Japanese_spaniel
1854	toy_poodle		0.031355	True	miniature_poodle
1859	American_black_bear		0.056994	False	brown_bear
1962	Italian_greyhound		0.270733	True	miniature_pinscher
1999	cowboy_boot		0.060343	False	giant_panda
2034	ice_bear		0.132264	False	weasel
2036	Irish_terrier		0.054933	True	Yorkshire_terrier
2041	Leonberg		0.127998	True	golden_retriever
2058	Italian_greyhound		0.005370	True	Pomeranian
2081	beagle		0.124680	True	Saluki
2084	chow		0.249394	True	Pomeranian
2088	bloodhound		0.244220	True	flat-coated_retriever
2115	miniature_pinscher		0.074192	True	Rhodesian_ridgeback

	p3_conf	p3_dog
696	0.106708	True
819	0.052423	False
1111	0.117480	True
1200	0.012896	True
1231	0.065582	True
1352	0.041692	True
1358	0.021112	True
1385	0.083834	False
1504	0.002307	True
1511	0.006481	False
1513	0.028148	False

```

1528  0.094727  True
1643  0.190407  True
1723  0.085626  False
1732  0.064627  True
1765  0.147745  False
1779  0.120873  True
1789  0.002972  True
1843  0.027460  True
1854  0.025743  True
1859  0.053993  False
1962  0.109752  True
1999  0.059838  False
2034  0.005730  False
2036  0.040576  True
2041  0.069357  True
2058  0.002641  True
2081  0.044094  True
2084  0.241878  True
2088  0.173810  True
2115  0.072010  True

```

```
[31 rows x 22 columns]
```

```
In [76]: df_archive2[df_archive2.rating_numerator == 13.5]
```

```

Out[76]:
      tweet_id      timestamp \
41  883482846933004288  2017-07-08 00:28:19

41  This is Bella. She hopes her smile made you smile. If not, she is also offering you

41  https://twitter.com/dog_rates/status/883482846933004288/photo/1,https://twitter.com

      rating_numerator  rating_denominator  name doggo floofer pupper ... \
41                13.5                10.0  Bella  None    None  None ...

      img_num      p1  p1_conf p1_dog      p2  p2_conf \
41      1.0  golden_retriever  0.943082  True  Labrador_retriever  0.032409

      p2_dog      p3  p3_conf p3_dog
41  True  kuvasz  0.005501  True

[1 rows x 22 columns]

```

Test

```

In [77]: # confirm that all denominators are 10
df_archive2['rating_denominator'].value_counts()

```



```
Out[77]: 10.0      1674
         Name: rating_denominator, dtype: int64
```

```
In [78]: # check for changed data type (float for ratings)
         df_archive2.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1674 entries, 1 to 2116
Data columns (total 22 columns):
tweet_id      1674 non-null object
timestamp     1674 non-null datetime64[ns]
text          1674 non-null object
expanded_urls 1674 non-null object
rating_numerator 1674 non-null float64
rating_denominator 1674 non-null float64
name          1674 non-null object
doggo         1674 non-null object
floofer       1674 non-null object
pupper       1674 non-null object
puppo        1674 non-null object
jpg_url       1674 non-null object
img_num       1674 non-null float64
p1            1674 non-null object
p1_conf       1674 non-null float64
p1_dog        1674 non-null object
p2            1674 non-null object
p2_conf       1674 non-null float64
p2_dog        1674 non-null object
p3            1674 non-null object
p3_conf       1674 non-null float64
p3_dog        1674 non-null object
dtypes: datetime64[ns](1), float64(6), object(15)
memory usage: 300.8+ KB
```

```
In [79]: # confirm changes
         df_archive2.rating_numerator.value_counts()
```

```
Out[79]: 12.00      423
         10.00      361
         11.00      357
         13.00      230
         9.00       133
         8.00        68
         7.00        31
         14.00        24
         6.00        16
         5.00        13
         4.00         6
```

```

3.00      5
2.00      2
13.50     1
11.27     1
0.00      1
11.26     1
9.75      1
Name: rating_numerator, dtype: int64

```

All rating numerators look clean.

Define 8. Clean names using regular expression in `df_archive`. A useful pattern to find these "fake names" is that they started with lowercase, instead, the "real names" always started with uppercase, moreover, the real ones usually are placed after words like "named", "is" etc...so it's possible to clean this issue using a regex

Code

```

In [80]: # extract correct names from text
df_archive2['name'] = df_archive2['text'].str.extract(r'([is|named|to|meet]+\s[A-Z] [\W])')

In [81]: # Change 'I' to NaN
df_archive2.loc[df_archive2['name'] == "I", 'name'] = np.nan

```

Test

```

In [82]: df_archive2['name'].value_counts()

```

```

Out[82]: Cooper      10
Charlie      10
Lucy         10
Tucker       9
Oliver       9
Penny        8
Daisy        7
Sadie         7
Winston       7
Toby          7
Lola           6
Koda           6
Jax            6
Bo             6
Bella          5
Oscar          5
Zoey           5
Rusty          5
Stanley        5
Reggie         5
Leo            5
Louis          4
Jack           4

```

Gus	4
Maggie	4
Oakley	4
Sophie	4
Milo	4
Finn	4
Boomer	4
..	
Gingivitis	1
Kona	1
Rand	1
Yoshi	1
Pumpkin	1
Rueben	1
Jim	1
Bobble	1
Napolean	1
Lili	1
Murphy	1
Diogi	1
Franq	1
Kuyu	1
Stephanus	1
Molly	1
Winifred	1
Timison	1
Mollie	1
Bobbay	1
Blanket	1
Pawnd	1
Lilli	1
Bradley	1
Crimson	1
Kevon	1
Orion	1
Kayla	1
Eriq	1
Kenny	1

Name: name, Length: 942, dtype: int64

In [83]: df_archive2.head()

```
Out[83]:
```

	tweet_id	timestamp	\
1	892177421306343426	2017-08-01 00:17:27	
2	891815181378084864	2017-07-31 00:18:03	
3	891689557279858688	2017-07-30 15:58:51	
4	891327558926688256	2017-07-29 16:00:24	
5	891087950875897856	2017-07-29 00:08:17	

```

1 This is Tilly. She's just checking pup on you. Hopes you're doing ok. If not, she's
2           This is Archie. He is a rare Norwegian Pouncing Corgo. Lives in the
3           This is Darla. She commen
4 This is Franklin. He would like you to stop calling him "cute." He is a very fierce
5 Here we have a majestic great white breaching off South Africa's coast. Absolutely h

```

```

1 https://twitter.com/
2 https://twitter.com/
3 https://twitter.com/
4 https://twitter.com/dog_rates/status/891327558926688256/photo/1,https://twitter.com/
5 https://twitter.com/

```

	rating_numerator	rating_denominator	name	doggo	floofer	pupper	...	\
1	13.0	10.0	Tilly	None	None	None	...	
2	12.0	10.0	Archie	None	None	None	...	
3	13.0	10.0	Darla	None	None	None	...	
4	12.0	10.0	Franklin	None	None	None	...	
5	13.0	10.0	NaN	None	None	None	...	

	img_num	p1	p1_conf	p1_dog	p2	\
1	1.0	Chihuahua	0.323581	True	Pekinese	
2	1.0	Chihuahua	0.716012	True	malamute	
3	1.0	paper_towel	0.170278	False	Labrador_retriever	
4	2.0	basset	0.555712	True	English_springer	
5	1.0	Chesapeake_Bay_retriever	0.425595	True	Irish_terrier	

	p2_conf	p2_dog	p3	p3_conf	p3_dog
1	0.090647	True	papillon	0.068957	True
2	0.078253	True	kelpie	0.031379	True
3	0.168086	True	spatula	0.040836	False
4	0.225770	True	German_short-haired_pointer	0.175219	True
5	0.116317	True	Indian_elephant	0.076902	False

[5 rows x 22 columns]

In [84]: df_archive2.info()

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1674 entries, 1 to 2116
Data columns (total 22 columns):
tweet_id      1674 non-null object
timestamp     1674 non-null datetime64[ns]
text          1674 non-null object
expanded_urls  1674 non-null object
rating_numerator  1674 non-null float64

```

```

rating_denominator    1674 non-null float64
name                  1321 non-null object
doggo                 1674 non-null object
floofer              1674 non-null object
pupper              1674 non-null object
puppo               1674 non-null object
jpg_url             1674 non-null object
img_num            1674 non-null float64
p1                 1674 non-null object
p1_conf            1674 non-null float64
p1_dog            1674 non-null object
p2                 1674 non-null object
p2_conf            1674 non-null float64
p2_dog            1674 non-null object
p3                 1674 non-null object
p3_conf            1674 non-null float64
p3_dog            1674 non-null object
dtypes: datetime64[ns](1), float64(6), object(15)
memory usage: 300.8+ KB

```

1.4.2 Tidiness

1. Melt 'doggo' 'floofer' 'pupper' 'puppo' in df_archive.

```
In [85]: df_archive2.head(1)
```

```

Out[85]:
   tweet_id      timestamp \
1  892177421306343426  2017-08-01 00:17:27

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

   expanded_urls \
1  https://twitter.com/dog_rates/status/892177421306343426/photo/1

   rating_numerator  rating_denominator  name doggo floofer pupper ... \
1              13.0              10.0  Tilly  None    None    None ...

   img_num      p1    p1_conf  p1_dog      p2    p2_conf  p2_dog      p3 \
1      1.0  Chihuahua  0.323581    True  Pekinese  0.090647    True  papillon

   p3_conf  p3_dog
1  0.068957    True

[1 rows x 22 columns]

```

```
In [86]: df_archive2['stage'] = df_archive2.doggo + df_archive2.floofer + df_archive2.pupper + d
df_archive2.head(1)
```

```

Out[86]:          tweet_id          timestamp \
1  892177421306343426 2017-08-01 00:17:27

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

          expanded_urls \
1  https://twitter.com/dog_rates/status/892177421306343426/photo/1

    rating_numerator rating_denominator  name doggo floofer pupper \
1                13.0                10.0  Tilly  None    None    None

    ...          p1  p1_conf  p1_dog          p2  p2_conf  p2_dog \
1    ...          Chihuahua 0.323581    True  Pekinese 0.090647    True

          p3  p3_conf  p3_dog          stage
1  papillon 0.068957    True  NoneNoneNoneNone

[1 rows x 23 columns]

```

```

In [87]: df_archive2.stage.value_counts()

```

```

Out[87]: NoneNoneNoneNone      1414
NoneNonepupperNone           168
doggoNoneNoneNone             54
NoneNoneNonepuppo             21
doggoNonepupperNone            8
NoneflooferNoneNone            7
doggoflooferNoneNone           1
doggoNoneNonepuppo             1
Name: stage, dtype: int64

```

```

In [88]: df_archive2.loc[df_archive2.stage == 'NoneNoneNoneNone', 'stage'] = 'None'
df_archive2.loc[df_archive2.stage == 'NoneNonepupperNone', 'stage'] = 'pupper'
df_archive2.loc[df_archive2.stage == 'doggoNoneNoneNone', 'stage'] = 'doggo'
df_archive2.loc[df_archive2.stage == 'NoneNoneNonepuppo', 'stage'] = 'puppo'
df_archive2.loc[df_archive2.stage == 'doggoNonepupperNone', 'stage'] = 'multiple'
df_archive2.loc[df_archive2.stage == 'NoneflooferNoneNone', 'stage'] = 'floofer'
df_archive2.loc[df_archive2.stage == 'doggoflooferNoneNone', 'stage'] = 'multiple'
df_archive2.loc[df_archive2.stage == 'doggoNoneNonepuppo', 'stage'] = 'multiple'

```

```

In [89]: df_archive2.stage.value_counts()

```

```

Out[89]: None      1414
pupper      168
doggo       54
puppo       21
multiple    10
floofer      7
Name: stage, dtype: int64

```

```
In [90]: # drop individual stages columns
df_archive2.drop(['floofer','doggo','pupper', 'puppo'], axis=1, inplace = True)
df_archive2.head(1)
```

```
Out[90]:
```

	tweet_id	timestamp	expanded_urls	rating_numerator	rating_denominator	name	jpg_url	img_num	p1	p1_conf	p1_dog	p2	p2_conf	p2_dog	p3	p3_conf	p3_dog	stage
1	892177421306343426	2017-08-01 00:17:27	https://twitter.com/dog_rates/status/892177421306343426/photo/1	13.0	10.0	Tilly	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg	1.0	Chihuahua	0.323581	True	Pekinese	0.090647	True	papillon	0.068957	True	None

```
In [91]: df_pred.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2075 entries, 0 to 2074
Data columns (total 12 columns):
tweet_id    2075 non-null object
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(1), object(5)
memory usage: 152.1+ KB
```

2. Join all data frames into twitter_archive_master.csv.

```
In [92]: twitter_archive_master = df_archive2.merge(df_json, on=['tweet_id'], how='left')

In [93]: twitter_archive_master.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1674 entries, 0 to 1673
Data columns (total 21 columns):
tweet_id          1674 non-null object
timestamp         1674 non-null datetime64[ns]
text              1674 non-null object
expanded_urls     1674 non-null object
rating_numerator  1674 non-null float64
rating_denominator 1674 non-null float64
name              1321 non-null object
jpg_url           1674 non-null object
img_num           1674 non-null float64
p1                1674 non-null object
p1_conf           1674 non-null float64
p1_dog            1674 non-null object
p2                1674 non-null object
p2_conf           1674 non-null float64
p2_dog            1674 non-null object
p3                1674 non-null object
p3_conf           1674 non-null float64
p3_dog            1674 non-null object
stage             1674 non-null object
retweet_count     1674 non-null int64
favorite_count    1674 non-null int64
dtypes: datetime64[ns](1), float64(6), int64(2), object(12)
memory usage: 287.7+ KB

```

```
In [94]: twitter_archive_master.head(1)
```

```

Out[94]:
   tweet_id      timestamp \
0  892177421306343426  2017-08-01 00:17:27

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

   expanded_urls \
0  https://twitter.com/dog_rates/status/892177421306343426/photo/1

   rating_numerator  rating_denominator  name \
0                13.0                10.0  Tilly

   jpg_url  img_num  p1 \
0  https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg  1.0  Chihuahua

   ...  p1_dog  p2  p2_conf  p2_dog  p3  p3_conf \
0  ...      True  Pekinese  0.090647  True  papillon  0.068957

```



```

p3_dog stage retweet_count favorite_count
0    True  None           6514           33819

```

```
[1 rows x 21 columns]
```

```

In [95]: # save to csv
twitter_archive_master.to_csv('twitter_archive_master.csv', index=False)

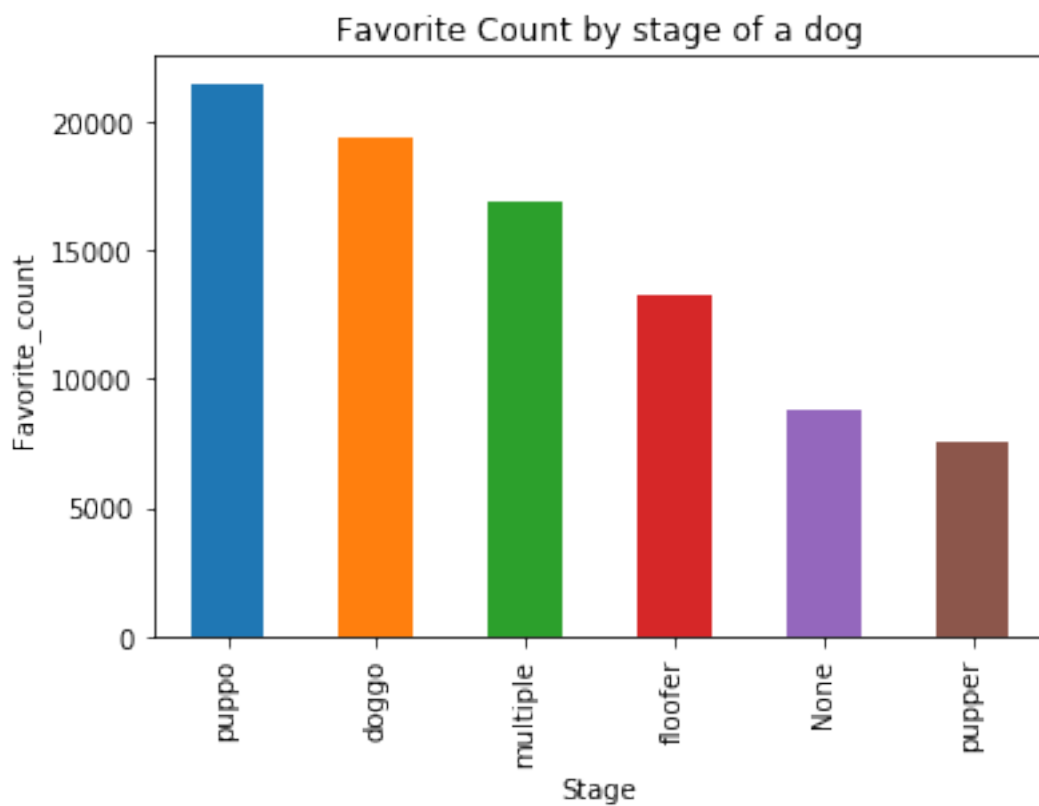
```

1.5 Analysis

```

In [96]: ax = twitter_archive_master.groupby('stage').favorite_count.mean().sort_values(ascending=True)
ax.set_xlabel("Stage");
ax.set_ylabel("Favorite_count");

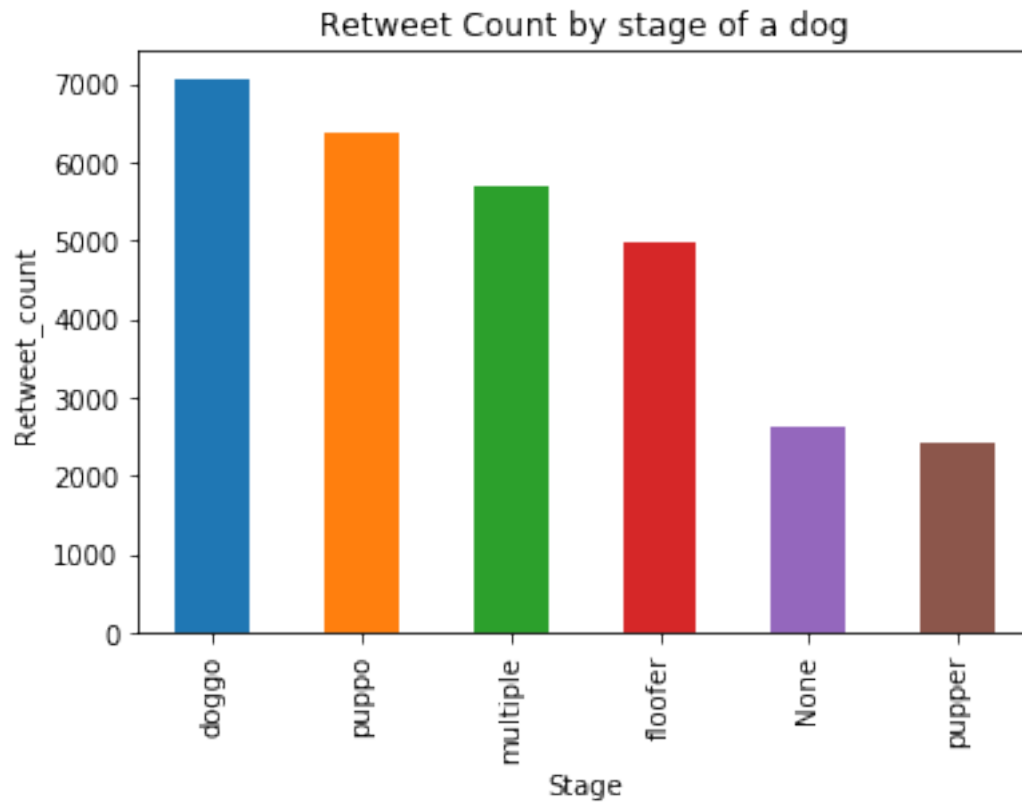
```



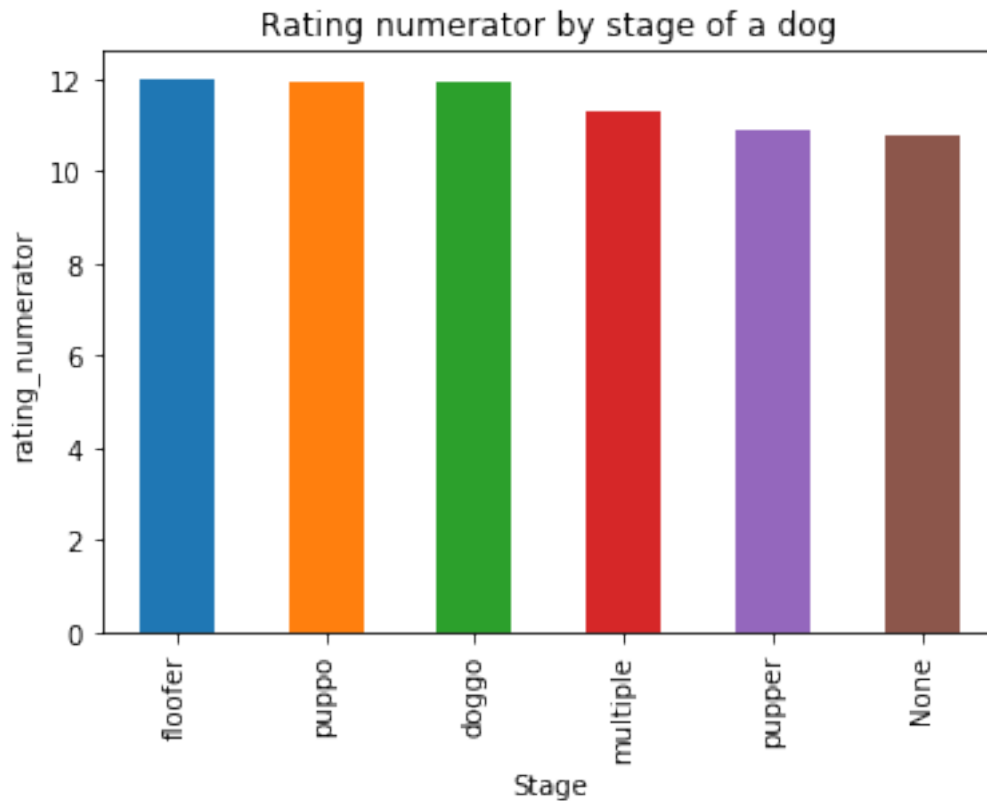
```

In [97]: ax = twitter_archive_master.groupby('stage').retweet_count.mean().sort_values(ascending=True)
ax.set_xlabel("Stage");
ax.set_ylabel("Retweet_count");

```



```
In [98]: ax = twitter_archive_master.groupby('stage').rating_numerator.mean().sort_values(ascending=True)
ax.set_xlabel("Stage");
ax.set_ylabel("rating_numerator");
```



The chart shows that tweets that contain dogs' stages like 'puppo' or 'doggo' tend to have higher count favorite count. On the other hand, 'puppers' have the lowest favorite count, even lower than 'None'. The same pattern is present in the 'retweet_count'. In addition, dogs that have 'stage' added to the tweets, tend to have higher rating numerators. 'Floofers' tend to have the greatest rating numerators.

```
In [99]: twitter_archive_master.rating_numerator.mean()
```

```
Out[99]: 10.840370370370369
```

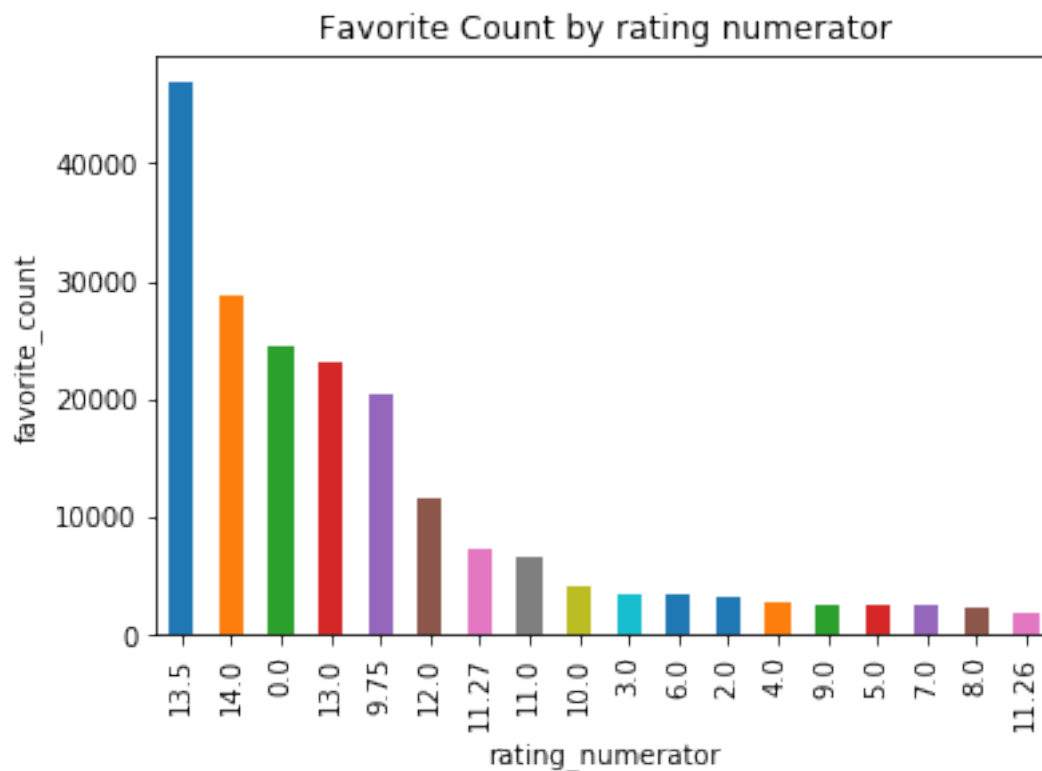
Mean rating_numerator for the tweets is 10.84.

```
In [100]: twitter_archive_master[twitter_archive_master.stage == 'floofer'].rating_numerator.mean()
```

```
Out[100]: 12.0
```

Mean rating_numerator for floofers is 12.

```
In [101]: ax = twitter_archive_master.groupby('rating_numerator').favorite_count.mean().sort_val
ax.set_xlabel("rating_numerator");
ax.set_ylabel("favorite_count");
```



```
In [102]: twitter_archive_master.rating_numerator.value_counts()
```

```
Out[102]: 12.00    423
          10.00    361
          11.00    357
          13.00    230
          9.00     133
          8.00     68
          7.00     31
          14.00     24
          6.00     16
          5.00     13
          4.00      6
          3.00      5
          2.00      2
          13.50      1
          11.27      1
          0.00       1
          11.26      1
          9.75       1
          Name: rating_numerator, dtype: int64
```

The chart shows that dogs with higher ratings also tend to get higher favorite count. An exception is rating numerator of 0; however, only one dog received that rating; therefore, we can ignore that results as the sample size is not big enough.

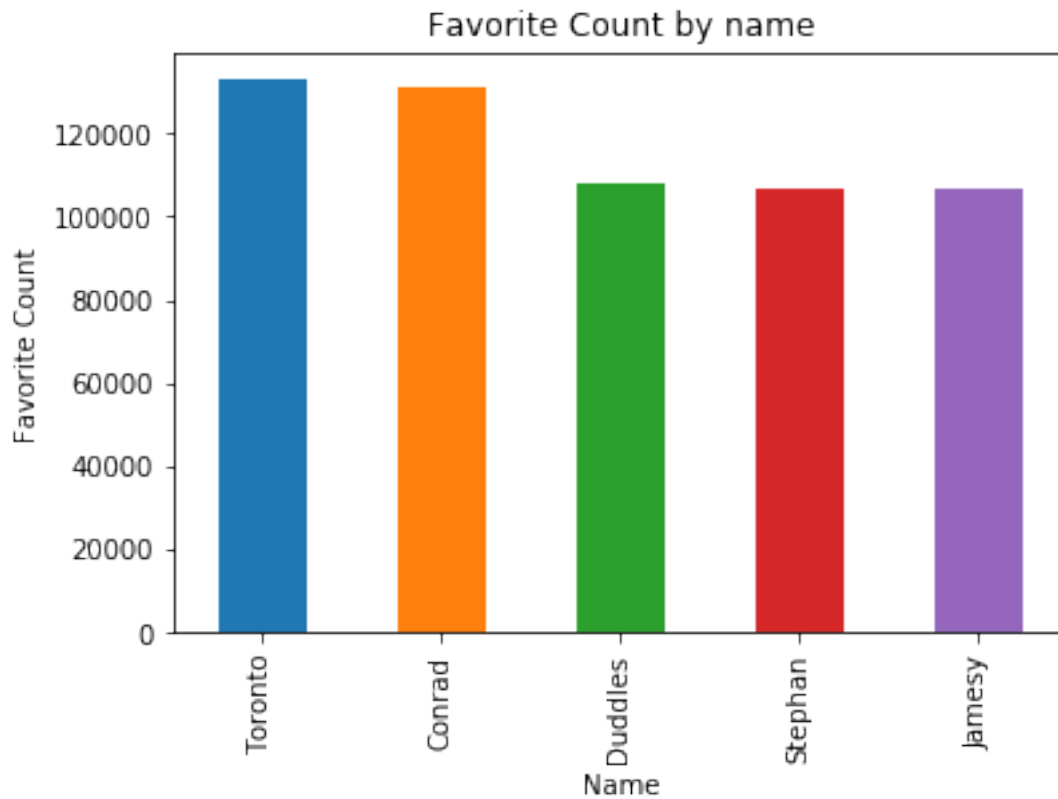
```
In [103]: twitter_archive_master.favorite_count.mean()
```

```
Out[103]: 9268.3655913978491
```

```
In [104]: twitter_archive_master.groupby('rating_numerator').favorite_count.mean().sort_values(ascending=True)
```

```
Out[104]: rating_numerator
13.50      46860.000000
14.00      28772.416667
0.00       24574.000000
13.00      23103.182609
9.75       20296.000000
12.00      11473.617021
11.27       7320.000000
11.00       6661.815126
10.00       4124.229917
3.00        3384.200000
6.00        3382.312500
2.00        3162.500000
4.00        2780.166667
9.00        2524.849624
5.00        2518.923077
7.00        2422.290323
8.00        2233.882353
11.26       1879.000000
Name: favorite_count, dtype: float64
```

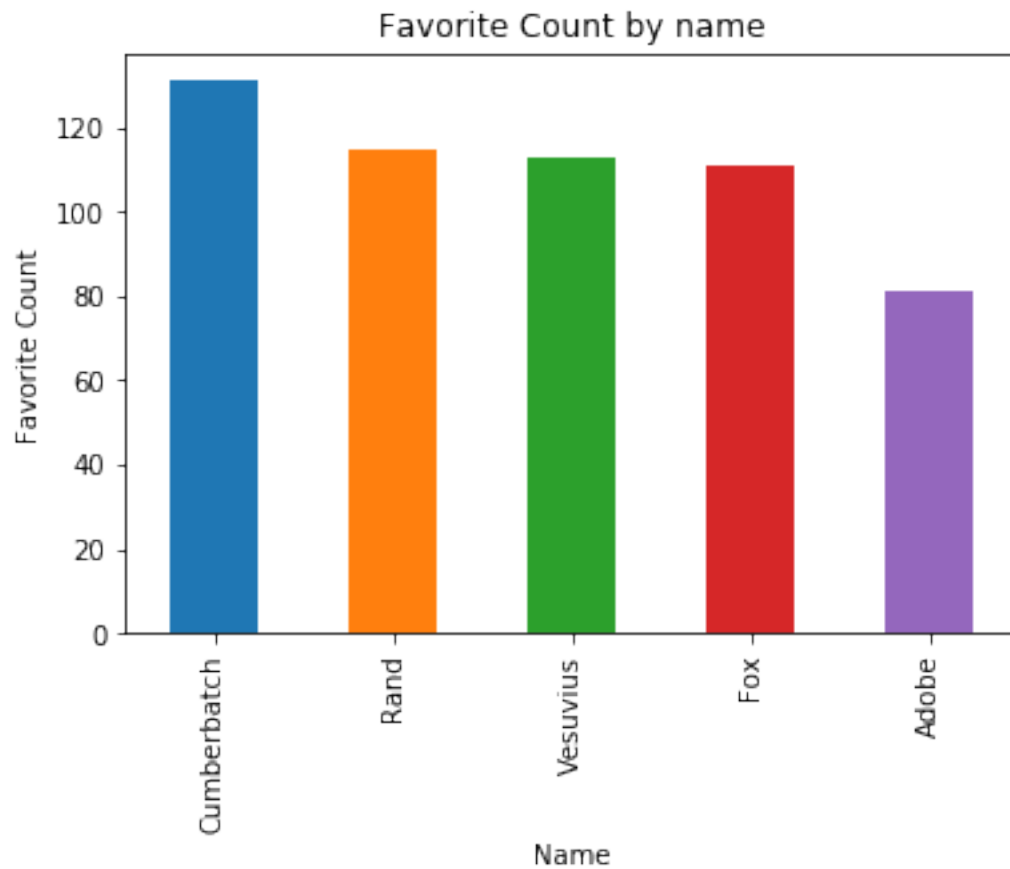
```
In [105]: ax = twitter_archive_master.groupby('name').favorite_count.mean().sort_values(ascending=True)
ax.set_xlabel("Name");
ax.set_ylabel("Favorite Count");
```



```
In [106]: twitter_archive_master.groupby('name').favorite_count.mean().sort_values(ascending=False)
```

```
Out[106]: name
Toronto    132810.0
Conrad     131075.0
Duddles    107956.0
Stephan    107015.0
Jamesy     106827.0
Name: favorite_count, dtype: float64
```

```
In [107]: ax = twitter_archive_master.groupby('name').favorite_count.mean().sort_values(ascending=False)
ax.set_xlabel("Name");
ax.set_ylabel("Favorite Count");
```



```
In [108]: twitter_archive_master.groupby('name').favorite_count.mean().sort_values(ascending=False)
```

```
Out[108]: name
Cumberbatch    131.0
Rand           115.0
Vesuvius       113.0
Fox            111.0
Adobe          81.0
Name: favorite_count, dtype: float64
```