

In [1]:

```
#mentioning all the libraries
import numpy as np
import pandas as pd
import requests
import tweepy
import json
import matplotlib.pyplot as plt
import warnings
```

GATHER

In [2]:

```
#reading and loading a csv file
twitter=pd.read_csv('twitter_archive_enhanced.csv')
```

In [3]:

```
#loading a tsv file
url="https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image_predictions/image-predictions.tsv"
response = requests.get(url)

with open('image_predictions.tsv', 'wb') as file:
    file.write(response.content)

image = pd.read_csv('image_predictions.tsv', sep='\t')
```

In [4]:

```
#creating API object. I have removed original values of consumer_key, consumer_secret, OAUTH_TOKEN, OAUTH_TOKEN_SECRET
consumer_key = 'my consumer key'
consumer_secret = 'my consumer secret'
OAUTH_TOKEN = 'my oauth token'
OAUTH_TOKEN_SECRET = 'my token secret'
auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
auth.set_access_token(OAUTH_TOKEN, OAUTH_TOKEN_SECRET)
api = tweepy.API(auth, wait_on_rate_limit = True, wait_on_rate_limit_notify = True)
```

In [5]:

```
#writing json data to tweet_json.txt file
with open('tweet_json.txt', 'a', encoding='utf8') as file:
    for tweet_id in twitter['tweet_id']:
        try:
            tweet = api.get_status(tweet_id, tweet_mode='extended')
            json.dump(tweet._json, file)
            file.write('\n')
        except:
            continue
```

Rate limit reached. Sleeping for: 722
Rate limit reached. Sleeping for: 724

In [6]:

```
#append each data to tweet_list line by line
tweet_list = []

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

for line in file:
    try:
        data = json.loads(line)
        tweet_list.append(data)
    except:
        continue

file.close()
```

In [7]:

```
#creating dataframe for tweet_data
tweet_data = pd.DataFrame()
tweet_data['id'] = list(map(lambda tweet: tweet['id'], tweet_list))
tweet_data['retweet_count'] = list(map(lambda tweet: tweet['retweet_count'],
tweet_list))
tweet_data['favorite_count'] = list(map(lambda tweet:
tweet['favorite_count'], tweet_list))
```

We have three dataframes named as :twitter , tweet_data , image

ASSESS

In [8]:

```
# first ten entries of twitter dataframe
twitter.head(10)
```

Out[8]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08-01 16:23:56 +0000	<a href="http://twit r...
1	892177421306343426	NaN	NaN	2017-08-01 00:17:27 +0000	<a href="http://twit r...
2				2017-07-	

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	<a href="http://twit
	891815181378084864	NaN	NaN	00:18:03 +0000	r...
3	891689557279858688	NaN	NaN	2017-07-30 15:58:51 +0000	<a href="http://twit r...
4	891327558926688256	NaN	NaN	2017-07-29 16:00:24 +0000	<a href="http://twit r...
5	891087950875897856	NaN	NaN	2017-07-29 00:08:17 +0000	<a href="http://twit r...
6	890971913173991426	NaN	NaN	2017-07-28 16:27:12 +0000	<a href="http://twit r...
7	890729181411237888	NaN	NaN	2017-07-28 00:22:40 +0000	<a href="http://twit r...
8	890609185150312448	NaN	NaN	2017-07-27 16:25:51 +0000	<a href="http://twit r...
9	890240255349198849	NaN	NaN	2017-07-26 15:59:51 +0000	<a href="http://twit r...



In [9]:

```
# first 10 entries of image dataframe
image.head(10)
```

Out [9]:

	tweet_id	jpg_url	img_num	
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	1	Welsh_
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1	redbon
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1	German
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-IEu.jpg	1	Rhodes
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1	miniatur
5	666050758794694657	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg	1	Bernes
6	666051853826850816	https://pbs.twimg.com/media/CT5KoJ1WoAAJash.jpg	1	box_tur
7	666055525042405380	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg	1	chow
8	666057090499244032	https://pbs.twimg.com/media/CT5PY90WoAAQGLo.jpg	1	shoppir
9	666058600524156928	https://pbs.twimg.com/media/CT5Qw94XAAA_2dP.jpg	1	miniatur

In [10]:

```
# first 10 entries of tweet_data dataframe
tweet_data.head(10)
```

Out[10]:

	id	retweet_count	favorite_count
0	892420643555336193	8580	38778
1	892177421306343426	6304	33207
2	891815181378084864	4186	25009
3	891689557279858688	8706	42146
4	891327558926688256	9468	40312
5	891087950875897856	3136	20207
6	890971913173991426	2087	11847
7	890729181411237888	19021	65476
8	890609185150312448	4292	27769
9	890240255349198849	7464	31923

In [11]:

```
twitter.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 [12]:

```
image.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 [13]:

```
tweet_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 23752 entries, 0 to 23751
Data columns (total 3 columns):
id            23752 non-null int64
retweet_count 23752 non-null int64
favorite_count 23752 non-null int64
dtypes: int64(3)
memory usage: 556.8 KB
```

In [14]:

```
# checking for null values in twitter dataframe for each column
twitter.isnull().sum()
```

Out[14]:

```
tweet_id      0
in_reply_to_status_id  2278
in_reply_to_user_id  2278
timestamp      0
source         0
text          0
```

```
retweeted_status_id      2175
retweeted_status_user_id  2175
retweeted_status_timestamp 2175
expanded_urls            59
rating_numerator          0
rating_denominator        0
name                     0
doggo                    0
floofer                  0
pupper                   0
puppo                    0
dtype: int64
```

In [15]:

```
# checking for null values in image dataframe for each column
image.isnull().sum()
```

Out[15]:

```
tweet_id      0
jpg_url       0
img_num       0
p1            0
p1_conf       0
p1_dog        0
p2           0
p2_conf       0
p2_dog        0
p3           0
p3_conf       0
p3_dog        0
dtype: int64
```

In [16]:

```
#checking for null values in tweet_data dataframe for each column
tweet_data.isnull().sum()
```

Out[16]:

```
id          0
retweet_count  0
favorite_count  0
dtype: int64
```

In [17]:

```
# inspecting the name column of twitter dataframe
twitter['name'].value_counts()
```

Out[17]:

```
None      745
a          55
Charlie    12
Lucy       11
Cooper     11
Oliver     11
Tucker     10
Penny      10
Lola       10
```

Bo	9
Winston	9
the	8
Sadie	8
Buddy	7
Daisy	7
Bailey	7
Toby	7
an	7
Koda	6
Oscar	6
Dave	6
Milo	6
Scout	6
Stanley	6
Leo	6
Jax	6
Bella	6
Rusty	6
Jack	6
Alfie	5

...

Moofasa	1
Glacier	1
Socks	1
Biden	1
Snickers	1
Chaz	1
Pumpkin	1
Eevee	1
Thor	1
Crumpet	1
Kane	1
Eazy	1
Julius	1
Charleson	1
Brownie	1
Rambo	1
Batdog	1
Tedrick	1
Rueben	1
Kulet	1
Nigel	1
Hanz	1
Bones	1
Diogi	1
Blakely	1
Sephie	1
Tug	1
Willow	1
Maks	1
Bubba	1

Name: name, Length: 957, dtype: int64

In [18]:

```
#checking those texts which contain any decimal rating as rating_numerator
values can be wrong for those particular rows
twitter[twitter['text'].str.contains(r"(\d+\.\d*\//\d+)")]
```

/opt/conda/lib/python3.6/site-packages/ipykernel_launcher.py:2: UserWarning

```
/cp/conda/lib/python3.7/site-packages/ipykernel_launcher.py? : This pattern has match groups. To actually get the groups, use str.extract.
```

Out [18]:

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

In [19]:

```
#Checking if any dog has 0 as its rating_denominator value as that value is wrong
twitter[twitter['rating_denominator']==0]
```

Out [19]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
313	835246439529840640	8.352460e+17	26259576.0	2017-02-24 21:54:03 +0000	<a href="http://t r...

In [20]:

```
#checking for those rows in which rating denominator is not having value 10
twitter[twitter['rating_denominator']!=10]
```

Out[20]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
313	835246439529840640	8.352460e+17	2.625958e+07	2017-02-24 21:54:03 +0000	<a href="http:// r...
342	832088576586297345	8.320875e+17	3.058208e+07	2017-02-16 04:45:50 +0000	<a href="http:// r...
433	820690176645140481	NaN	NaN	2017-01-15 17:52:40 +0000	<a href="http:// r...
516	810984652412424192	NaN	NaN	2016-12-19 23:06:23 +0000	<a href="http:// r...
784	775096608509886464	NaN	NaN	2016-09-11 22:20:06 +0000	<a href="http:// r...
902	758467244762497024	NaN	NaN	2016-07-28 01:00:57 +0000	<a href="http:// r...
1068	740373189193256964	NaN	NaN	2016-06-08 02:41:38 +0000	<a href="http:// r...
1120	731156023742988288	NaN	NaN	2016-05-13 16:15:54 +0000	<a href="http:// r...

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
1165				2016-04-	
	722974582966214656	NaN	NaN	21 02:25:47 +0000	<a href="http://r...
1202				2016-04-03 01:36:11 +0000	<a href="http://r...
1228				2016-03-27 01:29:02 +0000	<a href="http://r...
1254				2016-03-18 02:46:49 +0000	<a href="http://r...
1274				2016-03-14 02:04:08 +0000	<a href="http://r...
1351				2016-02-28 21:25:30 +0000	<a href="http://r...
1433				2016-02-10 16:51:59 +0000	<a href="http://r...
1598				2016-01-10 04:04:10 +0000	<a href="http://r...
1634				2016-01-05 04:11:44 +0000	<a href="http://r...
1635				2016-01-05 04:00:18 +0000	<a href="http://r...
1662				2016-01-01 16:30:13 +0000	<a href="http://r...
1663				2016-01-	<a

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	href="http://
	682808988178739200	6827884417	11969844100	01 06:22:03	r...
1779	677716515794329600	NaN	NaN	2015-12-18 05:06:23 +0000	<a href="http://r...
1843	675853064436391936	NaN	NaN	2015-12-13 01:41:41 +0000	<a href="http://r...
2335	666287406224695296	NaN	NaN	2015-11-16 16:11:11 +0000	<a href="http://r...

In [21]:

```
image.head(3)
```

Out [21]:

	tweet_id	jpg_url	img_num	
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	1	Welsh s
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1	redbone
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1	German

In [22]:

```
#checking for duplicate values in tweet_data dataframe
tweet_data.duplicated().sum()
```

Out [22]:

14003

In [23]:

```
#checking for duplicate values in twitter dataframe
twitter.duplicated().sum()
```

Out [23]:

0

In [24]:

```
#checking for duplicate values in image dataframe
image.duplicated().sum()
```

Out [24]:

0

QUALITY ISSUES WITH DATAFRAMES

-> data type conflicts (time_stamp, in_reply_to_user_id, in_reply_to_status_id, tweet_id , etc...)

-> duplicacy of records(In tweet_data dataframe)

-> dataframe consists of retweets

-> Incorrect dog names in twitter dataframe

-> tweets with no images are present

-> record containing wrong value of rating_denominator as 0

-> some columns such as retweeted_status_id and retweeted_status_user_id are not required

-> the rating_numerator value of records having decimal rating in their text is not written properly

TIDINESS ISSUES

-> The distinct dog type columns (can be resolved by combining dog type columns into one column)

-> merging all the three dataframes together to get a new dataframe

CLEAN

In [25]:

```
# creating copies of dataframes for cleaning purpose
twitter_clean=twitter.copy()
image_clean=image.copy()
tweet_data_clean=tweet_data.copy()
```

DEFINE

Removing retweets from twitter_clean dataframe

CODE

In [26]:

```
# if retweet status id is null then there is no retweet
twitter_clean=twitter_clean[twitter_clean['retweeted_status_id'].isnull()]
```

In [27]:

```
twitter_clean
```

Out[27]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08-01 16:23:56 +0000	<a href="http://r...
1	892177421306343426	NaN	NaN	2017-08-01 00:17:27 +0000	<a href="http://r...
2	891815181378084864	NaN	NaN	2017-07-31 00:18:03 +0000	<a href="http://r...
3	891689557279858688	NaN	NaN	2017-07-30 15:58:51 +0000	<a href="http://r...
4	891327558926688256	NaN	NaN	2017-07-29 16:00:24 +0000	<a href="http://r...
5	891087950875897856	NaN	NaN	2017-07-29 00:08:17 +0000	<a href="http://r...
6	890971913173991426	NaN	NaN	2017-07-28 16:27:12 +0000	<a href="http://r...
7	890729181411237888	NaN	NaN	2017-07-28 00:22:40 +0000	<a href="http://r...
8	890609185150312448	NaN	NaN	2017-07-27 16:25:51 +0000	<a href="http://r...
9	890240255349198849	NaN	NaN	2017-07-26 15:59:51 +0000	<a href="http://r...
10	890006608113172480	NaN	NaN	2017-07-26 00:31:25	<a href="http://_

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	r...
11	889880896479866881	NaN	NaN	2017-07-25 16:11:53 +0000	<a href="http://r...
12	889665388333682689	NaN	NaN	2017-07-25 01:55:32 +0000	<a href="http://r...
13	889638837579907072	NaN	NaN	2017-07-25 00:10:02 +0000	<a href="http://r...
14	889531135344209921	NaN	NaN	2017-07-24 17:02:04 +0000	<a href="http://r...
15	889278841981685760	NaN	NaN	2017-07-24 00:19:32 +0000	<a href="http://r...
16	888917238123831296	NaN	NaN	2017-07-23 00:22:39 +0000	<a href="http://r...
17	888804989199671297	NaN	NaN	2017-07-22 16:56:37 +0000	<a href="http://r...
18	888554962724278272	NaN	NaN	2017-07-22 00:23:06 +0000	<a href="http://r...
20	888078434458587136	NaN	NaN	2017-07-20 16:49:33 +0000	<a href="http://r...
21	887705289381826560	NaN	NaN	2017-07-19 16:06:48 +0000	<a href="http://r...
22	887517139158093824	NaN	NaN	2017-07-19 03:39:09 +0000	<a href="http://r...
23				2017-07-	

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	<a href="http://r...
	887473957103951883	NaN	NaN	00:47:34+0000	
24	887343217045368832	NaN	NaN	2017-07-18 16:08:03+0000	<a href="http://r...
25	887101392804085760	NaN	NaN	2017-07-18 00:07:08+0000	<a href="http://r...
26	886983233522544640	NaN	NaN	2017-07-17 16:17:36+0000	<a href="http://r...
27	886736880519319552	NaN	NaN	2017-07-16 23:58:41+0000	<a href="http://r...
28	886680336477933568	NaN	NaN	2017-07-16 20:14:00+0000	<a href="http://r...
29	886366144734445568	NaN	NaN	2017-07-15 23:25:31+0000	<a href="http://r...
30	886267009285017600	8.862664e+17	2.281182e+09	2017-07-15 16:51:35+0000	<a href="http://r...
...
2326	666411507551481857	NaN	NaN	2015-11-17 00:24:19+0000	<a href="http://r...
2327	666407126856765440	NaN	NaN	2015-11-17 00:06:54+0000	<a href="http://r...
2328	666396247373291520	NaN	NaN	2015-11-16 23:23:41+0000	<a href="http://r...
2329				2015-11-16 23:23:41+0000	<a href="http://r...

	666373753744588802	NaN	NaN	16 21:54:18	href="http://
				+0000	r...
2330	666362758909284353	NaN	NaN	2015-11-16 21:10:36 +0000	<a href="http://r...
2331	666353288456101888	NaN	NaN	2015-11-16 20:32:58 +0000	<a href="http://r...
2332	666345417576210432	NaN	NaN	2015-11-16 20:01:42 +0000	<a href="http://r...
2333	666337882303524864	NaN	NaN	2015-11-16 19:31:45 +0000	<a href="http://r...
2334	666293911632134144	NaN	NaN	2015-11-16 16:37:02 +0000	<a href="http://r...
2335	666287406224695296	NaN	NaN	2015-11-16 16:11:11 +0000	<a href="http://r...
2336	666273097616637952	NaN	NaN	2015-11-16 15:14:19 +0000	<a href="http://r...
2337	666268910803644416	NaN	NaN	2015-11-16 14:57:41 +0000	<a href="http://r...
2338	666104133288665088	NaN	NaN	2015-11-16 04:02:55 +0000	<a href="http://r...
2339	666102155909144576	NaN	NaN	2015-11-16 03:55:04 +0000	<a href="http://r...
2340	666099513787052032	NaN	NaN	2015-11-16 03:44:34 +0000	<a href="http://r...

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
2341	666094000022159362	NaN	NaN	2015-11-16 03:22:39 +0000	<a href="http://r...
2342	666082916733198337	NaN	NaN	2015-11-16 02:38:37 +0000	<a href="http://r...
2343	666073100786774016	NaN	NaN	2015-11-16 01:59:36 +0000	<a href="http://r...
2344	666071193221509120	NaN	NaN	2015-11-16 01:52:02 +0000	<a href="http://r...
2345	666063827256086533	NaN	NaN	2015-11-16 01:22:45 +0000	<a href="http://r...
2346	666058600524156928	NaN	NaN	2015-11-16 01:01:59 +0000	<a href="http://r...
2347	666057090499244032	NaN	NaN	2015-11-16 00:55:59 +0000	<a href="http://r...
2348	666055525042405380	NaN	NaN	2015-11-16 00:49:46 +0000	<a href="http://r...
2349	666051853826850816	NaN	NaN	2015-11-16 00:35:11 +0000	<a href="http://r...
2350	666050758794694657	NaN	NaN	2015-11-16 00:30:50 +0000	<a href="http://r...
2351	666049248165822465	NaN	NaN	2015-11-16 00:24:50 +0000	<a href="http://r...
2352				2015-11-	<a

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	href="http://
	66604422632999704	NaN	NaN	2015-11-15 00:04:52 +0000	r...
2353	666033412701032449	NaN	NaN	2015-11-15 23:21:54 +0000	<a href="http://r...
2354	666029285002620928	NaN	NaN	2015-11-15 23:05:30 +0000	<a href="http://r...
2355	666020888022790149	NaN	NaN	2015-11-15 22:32:08 +0000	<a href="http://r...

2175 rows × 17 columns



TEST

In [28]:

```
# Through info we can observe that there is no non null value present for r
retweeted_status_id
twitter_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2175 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2175 non-null int64
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2175 non-null object
source                  2175 non-null object
text                    2175 non-null object
retweeted_status_id     0 non-null float64
retweeted_status_user_id 0 non-null float64
retweeted_status_timestamp 0 non-null object
expanded_urls           2117 non-null object
rating_numerator         2175 non-null int64
rating_denominator       2175 non-null int64
name                    2175 non-null object
doggo                   2175 non-null object
floofer                 2175 non-null object
pupper                  2175 non-null object
puppo                   2175 non-null object
dtypes: float64(4), int64(3), object(10)
memory usage: 305.9+ KB
```

DEFINE

removing columns related to retweets from twitter_clean dataframe

CODE

In [29]:

```
columns=['retweeted_status_id','retweeted_status_user_id','retweeted_status_
timestamp']
twitter_clean.drop(columns,inplace=True,axis=1)
```

TEST

In [30]:

```
# to check if the columns related to retweets are deleted
twitter_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2175 entries, 0 to 2355
Data columns (total 14 columns):
tweet_id                2175 non-null int64
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2175 non-null object
source                  2175 non-null object
text                    2175 non-null object
expanded_urls           2117 non-null object
rating_numerator         2175 non-null int64
rating_denominator       2175 non-null int64
name                    2175 non-null object
doggo                   2175 non-null object
floofer                 2175 non-null object
pupper                  2175 non-null object
puppo                   2175 non-null object
dtypes: float64(2), int64(3), object(9)
memory usage: 254.9+ KB
```

DEFINE

Removing the invalid names of dogs (We observed while assessing that the dog names having all lowercase letters are wrong names)

CODE

In [31]:

```
# replacing invalid names with None
for name in twitter_clean['name']:
    if name.islower() == True:
        twitter_clean.replace(name, 'None', inplace=True)
```

TEST

In [32]:

```
# to check whether invalid dog names are removed and replaced with None
twitter_clean['name'].value_counts()
```

Out[32]:

None	784
Charlie	11
Lucy	11
Oliver	10
Cooper	10
Penny	9
Tucker	9
Winston	8
Sadie	8
Lola	8
Daisy	7
Toby	7
Bo	6
Jax	6
Bailey	6
Bella	6
Stanley	6
Oscar	6
Koda	6
Louis	5
Rusty	5
Scout	5
Buddy	5
Bentley	5
Dave	5
Milo	5
Leo	5
Chester	5
Alfie	4
Brody	4
...	
Opie	1
Traviss	1
Kota	1
Rupert	1
Shooter	1
Bronte	1
Eve	1
Blanket	1
Hercules	1
William	1
Apollo	1
Chompsky	1
Terrance	1
Ziva	1
Bobby	1
Rufio	1
Sandra	1
Cleopatra	1
Hero	1
Eugene	1
Alejandro	1
Ozzie	1
Shikha	1
Jessiga	1
Fido	1
Cecil	1
Brat	1

```
Bell          1
Danny         1
Chadrick      1
Name: name, Length: 931, dtype: int64
```

DEFINE

to replace invalid ratings with the decimal rating present in the text (if present)

CODE

In [33]:

```
#column width is adjusted to make the full text visible
pd.set_option('display.max_colwidth', -1)
twitter_clean[twitter_clean['text'].str.contains(r"(\d+\.\d*\//\d+) ")]
```

```
/opt/conda/lib/python3.6/site-packages/ipykernel_launcher.py:3: UserWarning
: This pattern has match groups. To actually get the groups, use str.extrac
t.
```

```
This is separate from the ipykernel package so we can avoid doing imports
until
```

Out[33]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
45	883482846933004288	NaN	NaN	2017-07-08 00:28:19 +0000	<a href="http:// rel="nofollow
695	786709082849828864	NaN	NaN	2016-10-13 23:23:56 +0000	<a href="http:// rel="nofollow
763	778027034220126208	NaN	NaN	2016-09-20 00:24:34 +0000	<a href="http:// rel="nofollow
1689	681340665377193984	6.813394e+17	4.196984e+09	2015-12-28 05:07:27 +0000	<a href="http:// rel="nofollow

1712	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
	680494726643068929	NaN	NaN	2015-12-25 21:06:00 +0000	<a href="http://rel="nofollow"



In [34]:

```
# replacing the invalid ratings with the correct decimal ratings
twitter_clean.loc[twitter_clean['tweet_id']==883482846933004288,"rating_numerator"]=13.5
twitter_clean.loc[twitter_clean['tweet_id']==786709082849828864,"rating_numerator"]=9.75
twitter_clean.loc[twitter_clean['tweet_id']==778027034220126208,"rating_numerator"]=11.27
twitter_clean.loc[twitter_clean['tweet_id']==681340665377193984,"rating_numerator"]=9.5
twitter_clean.loc[twitter_clean['tweet_id']==680494726643068929,"rating_numerator"]=11.26
```

TEST

In [35]:

```
# checking whether updation of rating_denominator and rating_numerator is done
twitter_clean[twitter_clean['text'].str.contains(r"(\d+\.\d*/\d+) ")]
```

/opt/conda/lib/python3.6/site-packages/ipykernel_launcher.py:2: UserWarning: This pattern has match groups. To actually get the groups, use str.extract.

Out [35]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
45	883482846933004288	NaN	NaN	2017-07-08 00:28:19 +0000	<a href="http://rel="nofollow"
695	786709082849828864	NaN	NaN	2016-10-13 23:23:56 +0000	<a href="http://rel="nofollow"
763	778027034220126208	NaN	NaN	2016-09-20 00:04:24	<a href="http://

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	rel="nofollow"
1689	681340665377193984	6.813394e+17	4.196984e+09	2015-12-28 05:07:27 +0000	<a href="http:// rel="nofollow"
1712	680494726643068929	NaN	NaN	2015-12-25 21:06:00 +0000	<a href="http:// rel="nofollow"

DEFINE

correcting the rating_numerator and rating_denominator values where the value of rating_denominator is 0

CODE

In [36]:

```
#checking for the rows having rating_denominator = 0
twitter_clean[twitter_clean['rating_denominator']==0]
```

Out[36]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
313	835246439529840640	8.352460e+17	26259576.0	2017-02-24 21:54:03 +0000	<a href="http://t rel="nofollow"

In [37]:

```
#inspecting the text to check whether rating is available or not
twitter_clean.loc[twitter_clean['rating_denominator']==0].text
```

Out[37]:

```
313      @jonnysun @Lin_Manuel ok jomny I know you're excited but 960/00 isn'
t a valid rating, 13/10 is tho
Name: text, dtype: object
```

In [38]:

```
# changing the invalid rating values to valid values
twitter_clean.loc[twitter_clean['tweet_id']==835246439529840640,"rating_nu
erator"]=13
twitter_clean.loc[twitter_clean['tweet_id']==835246439529840640,"rating_den
ominator"]=10
```

TEST

In [39]:

```
# checking whether changes are performed for the particular tweet_id
twitter_clean.loc[twitter_clean['tweet_id']==835246439529840640]
```

Out[39]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
313	835246439529840640	8.352460e+17	26259576.0	2017-02-24 21:54:03 +0000	<a href="http://t rel="nofollow



DEFINE

Merging the columns of dog type into one column

CODE

In [40]:

```
#defining a new column dog_type and removing the columns which are not req
uired
twitter_clean['dog_type']=twitter_clean.text.str.extract('(puppo|pupper|flo
fer|doggo)',expand=True)
columns=['doggo','floofer','pupper','puppo']
twitter_clean=twitter_clean.drop(columns,axis=1)
```



TEST

In [41]:

```
#to check whether the columns are merged into one column dog_type
twitter_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2175 entries, 0 to 2355
Data columns (total 11 columns):
tweet id                2175 non-null int64
```



```

in_reply_to_status_id    78 non-null float64
in_reply_to_user_id      78 non-null float64
timestamp                2175 non-null object
source                   2175 non-null object
text                     2175 non-null object
expanded_urls            2117 non-null object
rating_numerator          2175 non-null float64
rating_denominator       2175 non-null int64
name                     2175 non-null object
dog_type                 364 non-null object
dtypes: float64(3), int64(2), object(6)
memory usage: 203.9+ KB

```

DEFINE

changing data types of columns of twitter_clean dataframe

CODE

In [42]:

```

#changing data type of rating_numerator and rating_denominator to float
twitter_clean['rating_numerator']=twitter_clean.rating_numerator.astype('float')
twitter_clean['rating_denominator']=twitter_clean.rating_denominator.astype('float')

```

In [43]:

```

#changing data type of dog_type to category data type
#changing data type of timestamp to datetime data type
twitter_clean['dog_type']=twitter_clean['dog_type'].astype('category')
twitter_clean['timestamp']=pd.to_datetime(twitter_clean['timestamp'])

```

In [44]:

```

#changing datatype of mentioned columns to string
twitter_clean['in_reply_to_status_id'] =
twitter_clean['in_reply_to_status_id'].astype('str')
twitter_clean['in_reply_to_user_id'] = twitter_clean['in_reply_to_user_id'].astype('str')
twitter_clean['tweet_id'] = twitter_clean['tweet_id'].astype('str')

```

TEST

In [45]:

```
twitter_clean.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2175 entries, 0 to 2355
Data columns (total 11 columns):
tweet_id                2175 non-null object
in_reply_to_status_id   2175 non-null object
in_reply_to_user_id     2175 non-null object
timestamp                2175 non-null datetime64[ns]
source                  2175 non-null object

```

```

text                2175 non-null object
expanded_urls       2117 non-null object
rating_numerator    2175 non-null float64
rating_denominator  2175 non-null float64
name                2175 non-null object
dog_type            364 non-null category
dtypes: category(1), datetime64[ns](1), float64(2), object(7)
memory usage: 189.2+ KB

```

DEFINE

changing the rating_numerator and rating_denominator where rating denominator is not equal to 10

CODE

In [46]:

```

# checking for those records where rating_denominator is not equal to 10
twitter_clean[twitter_clean['rating_denominator']!=10]

```

Out [46]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
342	832088576586297345	8.3208754756e+17	30582082.0	2017-02-16 04:45:50	<a href="http:// rel="nofollow
433	820690176645140481	nan	nan	2017-01-15 17:52:40	<a href="http:// rel="nofollow
516	810984652412424192	nan	nan	2016-12-19 23:06:23	<a href="http:// rel="nofollow
902	758467244762497024	nan	nan	2016-07-28 01:00:57	<a href="http:// rel="nofollow
1068	740373189193256964	nan	nan	2016-06-08 02:41:38	<a href="http:// rel="nofollow
1120	731156023742988288	nan	nan	2016-05-13	<a href="http://

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	rel="nofollow"
1165	722974582966214656	nan	nan	2016-04-21 02:25:47	<a href="http:// rel="nofollow"
1202	716439118184652801	nan	nan	2016-04-03 01:36:11	<a href="http:// rel="nofollow"
1228	713900603437621249	nan	nan	2016-03-27 01:29:02	<a href="http:// rel="nofollow"
1254	710658690886586372	nan	nan	2016-03-18 02:46:49	<a href="http:// rel="nofollow"
1274	709198395643068416	nan	nan	2016-03-14 02:04:08	<a href="http:// rel="nofollow"
1351	704054845121142784	nan	nan	2016-02-28 21:25:30	<a href="http:// rel="nofollow"
1433	697463031882764288	nan	nan	2016-02-10 16:51:59	<a href="http:// rel="nofollow"
1598	686035780142297088	6.86034024801e+17	4196983835.0	2016-01-10 04:04:10	<a href="http:// rel="nofollow"
1634	684225744407494656	6.84222868336e+17	4196983835.0	2016-01-05 04:11:44	<a href="http:// rel="nofollow"
1635	684222868335505415	nan	nan	2016-01-05	<a href="http://

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	rel="nofollow"
1662	682962037429899265	nan	nan	2016-01-01 16:30:13	<a href="http:// rel="nofollow"
1663	682808988178739200	6.82788441538e+17	4196983835.0	2016-01-01 06:22:03	<a href="http:// rel="nofollow"
1779	677716515794329600	nan	nan	2015-12-18 05:06:23	<a href="http:// rel="nofollow"
1843	675853064436391936	nan	nan	2015-12-13 01:41:41	<a href="http:// rel="nofollow"
2335	666287406224695296	nan	nan	2015-11-16 16:11:11	<a href="http:// rel="nofollow"

In [47]:

```
#changing incorrect values by inspecting the text in which the correct rating is available
twitter_clean.loc[twitter_clean['tweet_id']=='722974582966214656',"rating_numerator"]=13
twitter_clean.loc[twitter_clean['tweet_id']=='722974582966214656',"rating_denominator"]=10
```

In [48]:

```
#one more record found whose rating was available in text
twitter_clean.loc[twitter_clean['tweet_id']=='716439118184652801',"rating_numerator"]=11
twitter_clean.loc[twitter_clean['tweet_id']=='716439118184652801',"rating_denominator"]=10
```

we observed that when ratings are done for group of dogs then rating_denominator is of other value than 10. We can ignore those records

TEST

In [49]:

```
#checking if values of rating_numerator and rating_denominator is changed
twitter_clean.loc[twitter_clean['tweet_id']=='716439118184652801']
```

Out [49]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
1202	716439118184652801	nan	nan	2016-04-03 01:36:11	<a href="http://rel="nofollow"

In [50]:

```
#checking that value of rating_numerator and rating_denominator is changed
for another tweet_id also
twitter_clean.loc[twitter_clean['tweet_id']=='722974582966214656']
```

Out [50]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
1165	722974582966214656	nan	nan	2016-04-21 02:25:47	<a href="http://rel="nofollow"

DEFINE

removing rows that does not contain expanded_urls for images

code

In [51]:

```
#removing rows not containing images
twitter_clean = twitter_clean.dropna(subset=['expanded_urls'])
```

TEST

In [52]:

```
#to check whether records not having image urls are removed from the dataframe
twitter_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2117 entries, 0 to 2355
Data columns (total 11 columns):
```

```
Data columns (total 11 columns):
tweet_id                2117 non-null object
in_reply_to_status_id   2117 non-null object
in_reply_to_user_id     2117 non-null object
timestamp               2117 non-null datetime64[ns]
source                  2117 non-null object
text                    2117 non-null object
expanded_urls           2117 non-null object
rating_numerator         2117 non-null float64
rating_denominator       2117 non-null float64
name                    2117 non-null object
dog_type                 356 non-null category
dtypes: category(1), datetime64[ns](1), float64(2), object(7)
memory usage: 184.2+ KB
```

DEFINE

removing duplicated records from tweet_data_clean dataframe

CODE

In [53]:

```
#checking duplicated records
tweet_data_clean.duplicated().sum()
```

Out[53]:

14003

In [54]:

```
#removing duplicated records
tweet_data_clean=tweet_data_clean.drop_duplicates()
```

TEST

In [55]:

```
#checking whether duplicated rows are deleted
tweet_data_clean.duplicated().sum()
```

Out[55]:

0

DEFINE

changing column name and data type of column named 'id' of tweet_data_clean dataframe

CODE

In [56]:

```
#renaming the id column of tweet_data_clean as tweet_id
tweet_data_clean=tweet_data_clean.rename(columns={'id':'tweet_id'})
```

In [57]:

```
# changing the data type
tweet_data_clean['tweet_id'] = tweet_data_clean['tweet_id'].astype('str')
```

TEST

In [58]:

```
#to check the change in data type and name of column
tweet_data_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9749 entries, 0 to 23751
Data columns (total 3 columns):
tweet_id      9749 non-null object
retweet_count  9749 non-null int64
favorite_count 9749 non-null int64
dtypes: int64(2), object(1)
memory usage: 304.7+ KB
```

DEFINE

changing data type of columns of image_clean dataframe

CODE

In [59]:

```
#information about data types
image_clean.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 [60]:

```
#changing data type of tweet_id column to string
image_clean['tweet_id'] = image_clean['tweet_id'].astype('str')
```

TEST

In [61]:

```
image_clean.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
```

DEFINE

merging dataframes tweet_data_clean, image_clean, twitter_clean

CODE

In [62]:

```
#merging dataframes on the basis of tweet_id
master_df = pd.merge(twitter_clean, tweet_data_clean, on='tweet_id', how='inner')
master_df = pd.merge(master_df, image_clean, on='tweet_id', how='inner')
```

TEST

In [63]:

```
#final dataframe is master_df
master_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 8445 entries, 0 to 8444
Data columns (total 24 columns):
tweet_id      8445 non-null object
in_reply_to_status_id  8445 non-null object
in_reply_to_user_id    8445 non-null object
timestamp      8445 non-null datetime64[ns]
source         8445 non-null object
text           8445 non-null object
expanded_urls   8445 non-null object
rating_numerator 8445 non-null float64
rating_denominator 8445 non-null float64
name           8445 non-null object
dog_type       1504 non-null category
retweet_count   8445 non-null int64
favorite_count  8445 non-null int64
```



```

jpg_url      8445 non-null object
img_num      8445 non-null int64
p1           8445 non-null object
p1_conf      8445 non-null float64
p1_dog       8445 non-null bool
p2           8445 non-null object
p2_conf      8445 non-null float64
p2_dog       8445 non-null bool
p3           8445 non-null object
p3_conf      8445 non-null float64
p3_dog       8445 non-null bool
dtypes: bool(3), category(1), datetime64[ns](1), float64(5), int64(3), object(11)
memory usage: 1.4+ MB

```

In [64]:

```
master_df.head()
```

Out[64]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	nan	nan	2017-08-01 16:23:56	
1	892420643555336193	nan	nan	2017-08-01 16:23:56	
2	892420643555336193	nan	nan	2017-08-01 16:23:56	
3	892420643555336193	nan	nan	2017-08-01 16:23:56	
4	892420643555336193	nan	nan	2017-08-01 16:23:56	

5 rows × 24 columns



STORING DATA

In [84]:

```
#writing data to twitter_archive_master.csv
master_df.to_csv('twitter_archive_master.csv',index=False)
```

ANALYZING

In [85]:

```
data=pd.read_csv('twitter_archive_master.csv')
```

In [86]:

```
data.head(3)
```

Out [86]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08-01 16:23:56	
1	892420643555336193	NaN	NaN	2017-08-01 16:23:56	
2	892420643555336193	NaN	NaN	2017-08-01 16:23:56	

3 rows × 25 columns



In [87]:

```
#making a new column named rating ratio for further analysis
data['rating_ratio']=data['rating_numerator']/data['rating_denominator']
```

In [77]:

```
#again displaying the dataframe
data.head()
```

Out [77]:

	Unnamed: 0	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	0	892420643555336193	NaN	NaN	2017-08-01 16:23:56	<a href=
1	1	892420643555336193	NaN	NaN	2017-08-01 16:23:56	<a href=
2	2	892420643555336193	NaN	NaN	2017-08-01 16:23:56	<a href=
3	3	892420643555336193	NaN	NaN	2017-08-01 16:23:56	<a href=
4	4	892420643555336193	NaN	NaN	2017-08-01 16:23:56	<a href=

5 rows × 26 columns



In [88]:

```
#plotting rating ratio
%matplotlib inline
data['rating ratio'].value_counts()
```

Out[88]:

```
1.200000    2252
1.100000    1665
1.300000    1587
1.000000    1434
0.900000     431
0.800000     287
1.400000     233
0.700000     149
0.500000      95
0.600000      91
0.300000      57
0.400000      50
0.200000      32
0.100000      18
```

```

0.000000    10
1.350000     9
0.636364     8
42.000000     8
0.975000     8
0.818182     7
3.428571     4
1.127000     4
177.600000     3
1.126000     3
Name: rating_ratio, dtype: int64

```

In [89]:

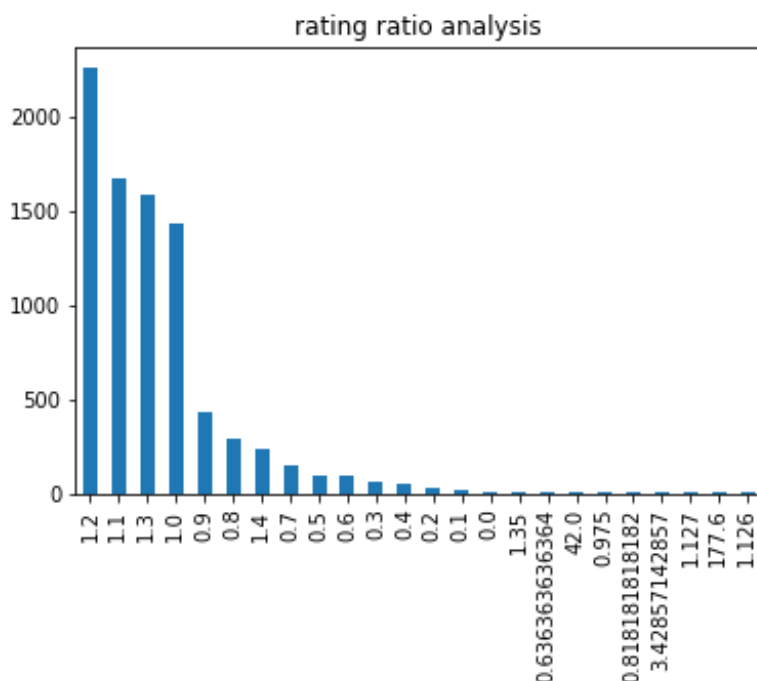
```

data['rating_ratio'].value_counts().plot(kind='bar')
plt.title('rating_ratio analysis')

```

Out[89]:

```
Text(0.5,1,'rating_ratio analysis')
```



This graph gives the insight of how rating ratio differs and which rating ratio is the highest

In [90]:

```
data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8445 entries, 0 to 8444
Data columns (total 25 columns):
tweet_id                8445 non-null int64
in_reply_to_status_id   86 non-null float64
in_reply_to_user_id     86 non-null float64
timestamp               8445 non-null object
source                  8445 non-null object
text                    8445 non-null object
expanded_urls           8445 non-null object
rating_numerator         8445 non-null float64
rating_denominator       8445 non-null float64
name                    8445 non-null object

```

```

dog_type          1504 non-null object
retweet_count     8445 non-null int64
favorite_count    8445 non-null int64
jpg_url           8445 non-null object
img_num           8445 non-null int64
p1                8445 non-null object
p1_conf           8445 non-null float64
p1_dog            8445 non-null bool
p2                8445 non-null object
p2_conf           8445 non-null float64
p2_dog            8445 non-null bool
p3                8445 non-null object
p3_conf           8445 non-null float64
p3_dog            8445 non-null bool
rating ratio      8445 non-null float64
dtypes: bool(3), float64(8), int64(4), object(10)
memory usage: 1.4+ MB

```

In [91]:

```
data.dog_type.value_counts()
```

Out[91]:

```

pupper      900
doggo       406
puppo       176
floofer      22
Name: dog_type, dtype: int64

```

In [92]:

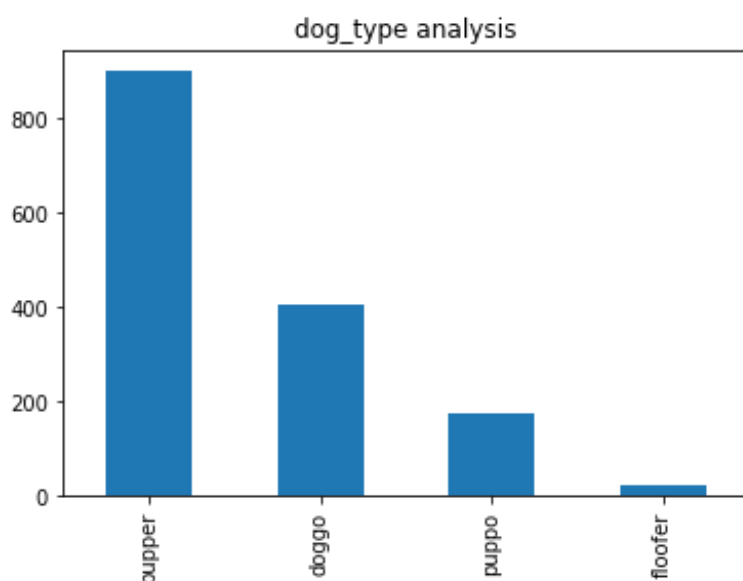
```

data['dog_type'].value_counts().plot(kind='bar')
plt.title('dog_type analysis')

```

Out[92]:

```
Text(0.5,1,'dog_type analysis')
```



We observe through this graph pupper is more in number that is pupper is more common dog type

In [93]:

In [99]:

```
#extracting month
data['timestamp']=pd.to_datetime(data['timestamp'])
data['month'] = data['timestamp'].dt.month
```

In [101]:

```
data.head(3)
```

Out[101]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08-01 16:23:56	
1	892420643555336193	NaN	NaN	2017-08-01 16:23:56	
2	892420643555336193	NaN	NaN	2017-08-01 16:23:56	

3 rows × 26 columns



In [102]:

```
#extracting year
data['year'] = data['timestamp'].dt.year
```

In [103]:

```
data.head(3)
```

Out[103]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08-01 16:23:56	
1				2017-08-	<a

	892420643555336193 tweet_id	NaN in_reply_to_status_id	NaN in_reply_to_user_id	01 timestamp 16:23:56	href="http://twit rel="nofollow">
2	892420643555336193	NaN	NaN	2017-08- 01 16:23:56	

3 rows × 27 columns



In [114]:

```
plotting_detail = pd.DataFrame(data.groupby('month')['retweet_count'].count
())
```

In [115]:

```
plotting_detail
```

Out[115]:

	retweet_count
month	
1	944
2	803
3	763
4	481
5	561
6	695
7	725
8	318
9	334
10	379
11	927
12	1515

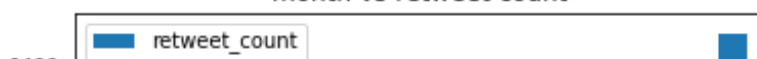
In [119]:

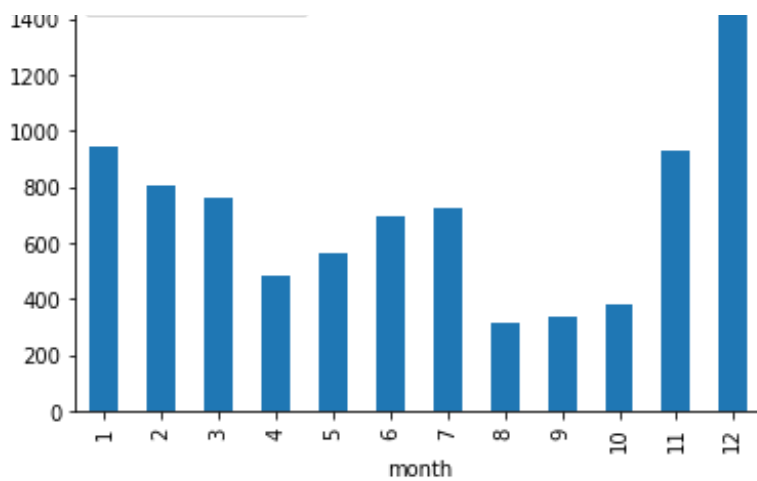
```
plotting_detail.plot(kind='bar',title='month vs retweet count')
```

Out[119]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f9837256748>

month vs retweet count





we observed that in 12th month retweet count is maximum

In [138]:

```
plotting_detail1 = pd.DataFrame(data.groupby('year')['retweet_count'].count())
```

In [139]:

```
plotting_detail1
```

Out[139]:

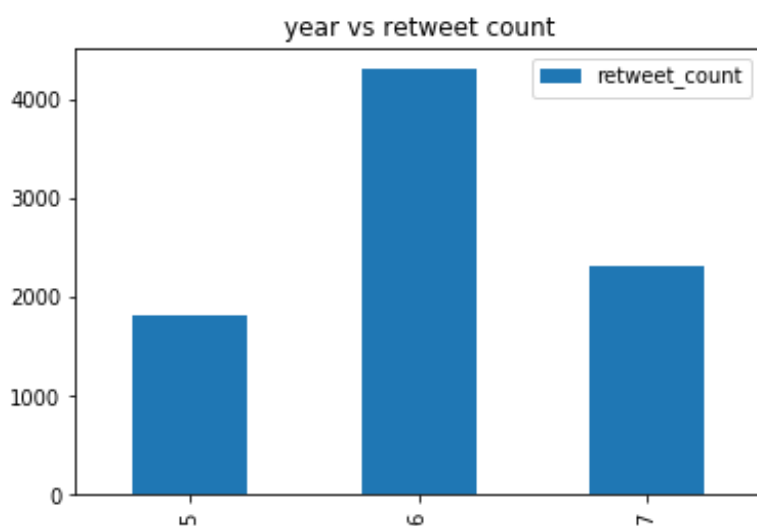
	retweet_count
year	
2015	1821
2016	4303
2017	2321

In [127]:

```
plotting_detail1.plot(kind='bar',title='year vs retweet count')
```

Out[127]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f983d38da58>



201

201
year

201

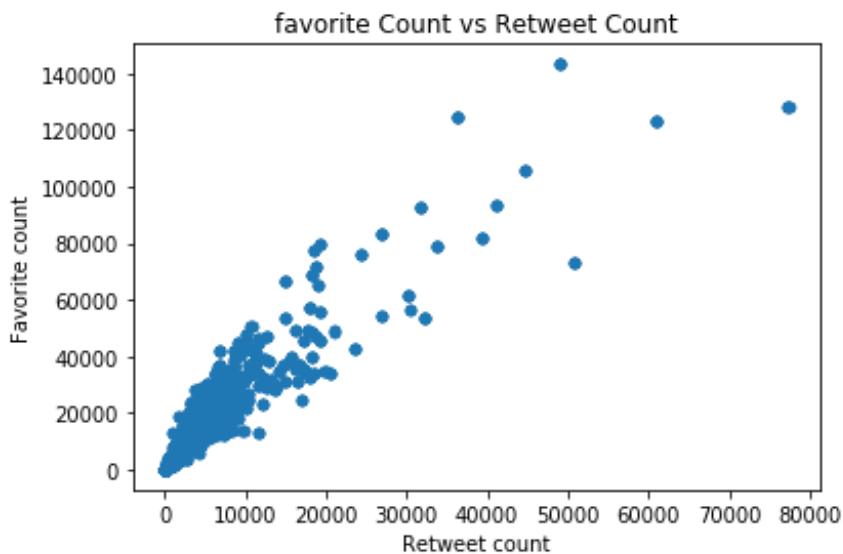
From above bar graph we noticed that 2016 year witnessed maximum retweet count

In [133]:

```
data.plot(x="retweet_count",y="favorite_count",kind="scatter")
plt.xlabel("Retweet count")
plt.ylabel("Favorite count")
plt.title("favorite Count vs Retweet Count")
plt.figure(figsize=(10,10))
```

Out[133]:

<matplotlib.figure.Figure at 0x7f983da63160>



<matplotlib.figure.Figure at 0x7f983da63160>

above graph shows strong correlation between favorite tweets and retweets

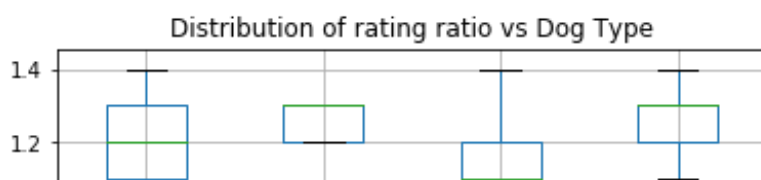
In [145]:

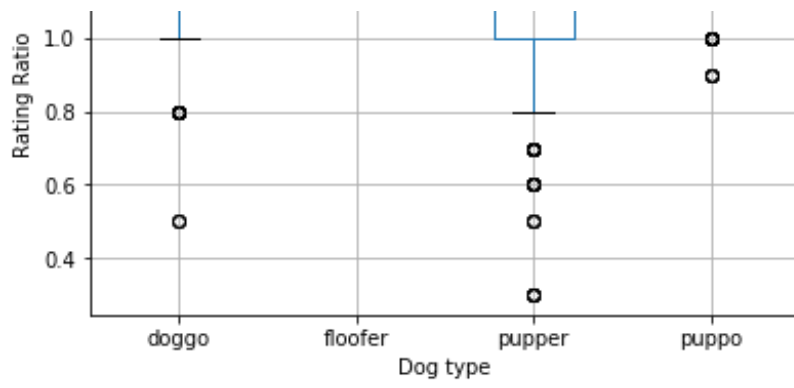
```
data.dog_type.value_counts()
data.boxplot(column='rating_ratio', by='dog_type')
plt.xlabel("Dog type")
plt.ylabel("Rating Ratio")
plt.suptitle("")
plt.title("Distribution of rating ratio vs Dog Type")
```

/opt/conda/lib/python3.6/site-packages/numpy/core/fromnumeric.py:57: Future Warning: reshape is deprecated and will raise in a subsequent release. Please use .values.reshape(...) instead
return getattr(obj, method)(*args, **kwds)

Out[145]:

Text(0.5,1,'Distribution of rating ratio vs Dog Type')





puppo dog type has highest median rating ratio and pupper has the lowest median rating ratio