

wrangle_act

August 7, 2021

1 Wrangling and Analyze Data

1.1 Gathering

```
In [1]: # Importing the required modules
import pandas as pd
import numpy as np
import tweepy
import json
import requests
import seaborn as sns
from matplotlib import pyplot as plt
%matplotlib inline

In [2]: # Reading the required files(csv and tsv)
df_twitter = pd.read_csv('twitter-archive-enhanced.csv')
df_img_pred = pd.read_csv('image-predictions.tsv', sep = '\t')

In [3]: import tweepy
from tweepy import OAuthHandler
import json
from timeit import default_timer as timer

consumer_secret = 'HIDDEN'
access_token = 'HIDDEN'
access_secret = 'HIDDEN'

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

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

tweet_ids = df_1.tweet_id.values
len(tweet_ids)

count = 0
fails_dict = {}
```

```

start = timer()

with open('tweet_json.txt', 'w') as outfile:
    for tweet_id in tweet_ids:
        count += 1
        print(str(count) + ": " + str(tweet_id))
        try:
            tweet = api.get_status(tweet_id, tweet_mode='extended')
            print("Success")
            json.dump(tweet._json, outfile)
            outfile.write('\n')
        except tweepy.TweepError as e:
            print("Fail")
            fails_dict[tweet_id] = e
        pass
end = timer()
print(end - start)
print(fails_dict)

```

Out[3]: 'import tweepy\nfrom tweepy import OAuthHandler\nimport json\nfrom timeit import default

In [4]: *# Reading json file*
df_tweets = pd.read_json("tweet-json.txt", lines = True)

1.2 Assess

In [5]: df_twitter.shape

Out[5]: (2356, 17)

In [6]: df_twitter

```

Out[6]:
   tweet_id  in_reply_to_status_id  in_reply_to_user_id \
0    892420643555336193           NaN                NaN
1    892177421306343426           NaN                NaN
2    891815181378084864           NaN                NaN
3    891689557279858688           NaN                NaN
4    891327558926688256           NaN                NaN
5    891087950875897856           NaN                NaN
6    890971913173991426           NaN                NaN
7    890729181411237888           NaN                NaN
8    890609185150312448           NaN                NaN
9    890240255349198849           NaN                NaN
10   890006608113172480           NaN                NaN
11   889880896479866881           NaN                NaN

```

12	889665388333682689	NaN	NaN
13	889638837579907072	NaN	NaN
14	889531135344209921	NaN	NaN
15	889278841981685760	NaN	NaN
16	888917238123831296	NaN	NaN
17	888804989199671297	NaN	NaN
18	888554962724278272	NaN	NaN
19	888202515573088257	NaN	NaN
20	888078434458587136	NaN	NaN
21	887705289381826560	NaN	NaN
22	887517139158093824	NaN	NaN
23	887473957103951883	NaN	NaN
24	887343217045368832	NaN	NaN
25	887101392804085760	NaN	NaN
26	886983233522544640	NaN	NaN
27	886736880519319552	NaN	NaN
28	886680336477933568	NaN	NaN
29	886366144734445568	NaN	NaN
...
2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2328	666396247373291520	NaN	NaN
2329	666373753744588802	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2332	666345417576210432	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	NaN	NaN
2336	666273097616637952	NaN	NaN
2337	666268910803644416	NaN	NaN
2338	666104133288665088	NaN	NaN
2339	666102155909144576	NaN	NaN
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 \
0	2017-08-01 16:23:56 +0000
1	2017-08-01 00:17:27 +0000
2	2017-07-31 00:18:03 +0000
3	2017-07-30 15:58:51 +0000
4	2017-07-29 16:00:24 +0000
5	2017-07-29 00:08:17 +0000
6	2017-07-28 16:27:12 +0000
7	2017-07-28 00:22:40 +0000
8	2017-07-27 16:25:51 +0000
9	2017-07-26 15:59:51 +0000
10	2017-07-26 00:31:25 +0000
11	2017-07-25 16:11:53 +0000
12	2017-07-25 01:55:32 +0000
13	2017-07-25 00:10:02 +0000
14	2017-07-24 17:02:04 +0000
15	2017-07-24 00:19:32 +0000
16	2017-07-23 00:22:39 +0000
17	2017-07-22 16:56:37 +0000
18	2017-07-22 00:23:06 +0000
19	2017-07-21 01:02:36 +0000
20	2017-07-20 16:49:33 +0000
21	2017-07-19 16:06:48 +0000
22	2017-07-19 03:39:09 +0000
23	2017-07-19 00:47:34 +0000
24	2017-07-18 16:08:03 +0000
25	2017-07-18 00:07:08 +0000
26	2017-07-17 16:17:36 +0000
27	2017-07-16 23:58:41 +0000
28	2017-07-16 20:14:00 +0000
29	2017-07-15 23:25:31 +0000
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2326	2015-11-17 00:24:19 +0000
2327	2015-11-17 00:06:54 +0000
2328	2015-11-16 23:23:41 +0000
2329	2015-11-16 21:54:18 +0000
2330	2015-11-16 21:10:36 +0000
2331	2015-11-16 20:32:58 +0000
2332	2015-11-16 20:01:42 +0000
2333	2015-11-16 19:31:45 +0000
2334	2015-11-16 16:37:02 +0000
2335	2015-11-16 16:11:11 +0000
2336	2015-11-16 15:14:19 +0000
2337	2015-11-16 14:57:41 +0000
2338	2015-11-16 04:02:55 +0000
2339	2015-11-16 03:55:04 +0000

2340 2015-11-16 03:44:34 +0000
 2341 2015-11-16 03:22:39 +0000
 2342 2015-11-16 02:38:37 +0000
 2343 2015-11-16 01:59:36 +0000
 2344 2015-11-16 01:52:02 +0000
 2345 2015-11-16 01:22:45 +0000
 2346 2015-11-16 01:01:59 +0000
 2347 2015-11-16 00:55:59 +0000
 2348 2015-11-16 00:49:46 +0000
 2349 2015-11-16 00:35:11 +0000
 2350 2015-11-16 00:30:50 +0000
 2351 2015-11-16 00:24:50 +0000
 2352 2015-11-16 00:04:52 +0000
 2353 2015-11-15 23:21:54 +0000
 2354 2015-11-15 23:05:30 +0000
 2355 2015-11-15 22:32:08 +0000

source \

0 <a href="http://twitter.com/download/iphone" r...
 1 <a href="http://twitter.com/download/iphone" r...
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2355 <a href="http://twitter.com/download/iphone" r...

```

	text	retweeted_status_id \
0	This is Phineas. He's a mystical boy. Only eve...	NaN
1	This is Tilly. She's just checking pup on you...	NaN
2	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	This is Darla. She commenced a snooze mid meal...	NaN
4	This is Franklin. He would like you to stop ca...	NaN
5	Here we have a majestic great white breaching ...	NaN
6	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	When you watch your owner call another dog a g...	NaN
8	This is Zoey. She doesn't want to be one of th...	NaN
9	This is Cassie. She is a college pup. Studying...	NaN
10	This is Koda. He is a South Australian decksha...	NaN
11	This is Bruno. He is a service shark. Only get...	NaN
12	Here's a puppo that seems to be on the fence a...	NaN
13	This is Ted. He does his best. Sometimes that'...	NaN
14	This is Stuart. He's sporting his favorite fan...	NaN

15	This is Oliver. You're witnessing one of his m...	NaN
16	This is Jim. He found a fren. Taught him how t...	NaN
17	This is Zeke. He has a new stick. Very proud o...	NaN
18	This is Ralphus. He's powering up. Attempting ...	NaN
19	RT @dog_rates: This is Canela. She attempted s...	8.874740e+17
20	This is Gerald. He was just told he didn't get...	NaN
21	This is Jeffrey. He has a monopoly on the pool...	NaN
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
23	This is Canela. She attempted some fancy porch...	NaN
24	You may not have known you needed to see this ...	NaN
25	This... is a Jubilant Antarctic House Bear. We...	NaN
26	This is Maya. She's very shy. Rarely leaves he...	NaN
27	This is Mingus. He's a wonderful father to his...	NaN
28	This is Derek. He's late for a dog meeting. 13...	NaN
29	This is Roscoe. Another pupper fallen victim t...	NaN
...
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaNwPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2345	This is the happiest dog you will ever see. Ve...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2348	Here is a Siberian heavily armored polar bear ...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2350	This is a truly beautiful English Wilson Staff...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

retweeted_status_user_id retweeted_status_timestamp \

0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	4.196984e+09	2017-07-19 00:47:34 +0000
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	NaN	NaN
26	NaN	NaN
27	NaN	NaN
28	NaN	NaN
29	NaN	NaN
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2326	NaN	NaN
2327	NaN	NaN
2328	NaN	NaN
2329	NaN	NaN
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2331	NaN	NaN
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2333	NaN	NaN
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2349	NaN	NaN
2350	NaN	NaN
2351	NaN	NaN
2352	NaN	NaN
2353	NaN	NaN
2354	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator	\
0	https://twitter.com/dog_rates/status/892420643...	13	
1	https://twitter.com/dog_rates/status/892177421...	13	
2	https://twitter.com/dog_rates/status/891815181...	12	
3	https://twitter.com/dog_rates/status/891689557...	13	
4	https://twitter.com/dog_rates/status/891327558...	12	
5	https://twitter.com/dog_rates/status/891087950...	13	
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	13	
7	https://twitter.com/dog_rates/status/890729181...	13	
8	https://twitter.com/dog_rates/status/890609185...	13	
9	https://twitter.com/dog_rates/status/890240255...	14	
10	https://twitter.com/dog_rates/status/890006608...	13	
11	https://twitter.com/dog_rates/status/889880896...	13	
12	https://twitter.com/dog_rates/status/889665388...	13	
13	https://twitter.com/dog_rates/status/889638837...	12	
14	https://twitter.com/dog_rates/status/889531135...	13	
15	https://twitter.com/dog_rates/status/889278841...	13	
16	https://twitter.com/dog_rates/status/888917238...	12	
17	https://twitter.com/dog_rates/status/888804989...	13	
18	https://twitter.com/dog_rates/status/888554962...	13	
19	https://twitter.com/dog_rates/status/887473957...	13	
20	https://twitter.com/dog_rates/status/888078434...	12	
21	https://twitter.com/dog_rates/status/887705289...	13	
22	https://twitter.com/dog_rates/status/887517139...	14	
23	https://twitter.com/dog_rates/status/887473957...	13	
24	https://twitter.com/dog_rates/status/887343217...	13	
25	https://twitter.com/dog_rates/status/887101392...	12	
26	https://twitter.com/dog_rates/status/886983233...	13	
27	https://www.gofundme.com/mingusneedsus,https:/...	13	
28	https://twitter.com/dog_rates/status/886680336...	13	
29	https://twitter.com/dog_rates/status/886366144...	12	
...	
2326	https://twitter.com/dog_rates/status/666411507...	2	
2327	https://twitter.com/dog_rates/status/666407126...	7	

2328	https://twitter.com/dog_rates/status/666396247...	9
2329	https://twitter.com/dog_rates/status/666373753...	11
2330	https://twitter.com/dog_rates/status/666362758...	6
2331	https://twitter.com/dog_rates/status/666353288...	8
2332	https://twitter.com/dog_rates/status/666345417...	10
2333	https://twitter.com/dog_rates/status/666337882...	9
2334	https://twitter.com/dog_rates/status/666293911...	3
2335	https://twitter.com/dog_rates/status/666287406...	1
2336	https://twitter.com/dog_rates/status/666273097...	11
2337	https://twitter.com/dog_rates/status/666268910...	10
2338	https://twitter.com/dog_rates/status/666104133...	1
2339	https://twitter.com/dog_rates/status/666102155...	11
2340	https://twitter.com/dog_rates/status/666099513...	8
2341	https://twitter.com/dog_rates/status/666094000...	9
2342	https://twitter.com/dog_rates/status/666082916...	6
2343	https://twitter.com/dog_rates/status/666073100...	10
2344	https://twitter.com/dog_rates/status/666071193...	9
2345	https://twitter.com/dog_rates/status/666063827...	10
2346	https://twitter.com/dog_rates/status/666058600...	8
2347	https://twitter.com/dog_rates/status/666057090...	9
2348	https://twitter.com/dog_rates/status/666055525...	10
2349	https://twitter.com/dog_rates/status/666051853...	2
2350	https://twitter.com/dog_rates/status/666050758...	10
2351	https://twitter.com/dog_rates/status/666049248...	5
2352	https://twitter.com/dog_rates/status/666044226...	6
2353	https://twitter.com/dog_rates/status/666033412...	9
2354	https://twitter.com/dog_rates/status/666029285...	7
2355	https://twitter.com/dog_rates/status/666020888...	8

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

18	10	Ralphus	None	None	None	None
19	10	Canela	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None
28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2345	10	the	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2348	10	a	None	None	None	None
2349	10	an	None	None	None	None
2350	10	a	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[2356 rows x 17 columns]

```
In [7]: # A concise summary of a DataFrame
df_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 [8]: # Viewing data to determine appearance
df_twitter.retweeted_status_timestamp.value_counts().head(3)

```

```

Out[8]: 2016-01-25 00:26:41 +0000    1
        2016-06-08 02:41:38 +0000    1
        2016-07-27 00:40:12 +0000    1
        Name: retweeted_status_timestamp, dtype: int64

```

```

In [9]: # Viewing data to determine appearance
df_twitter.retweeted_status_id.value_counts().head(3)

```

```

Out[9]: 7.757333e+17    1
        7.507196e+17    1
        6.742918e+17    1
        Name: retweeted_status_id, dtype: int64

```

```

In [10]: # Descriptive statistics of a DataFrame
df_twitter.describe()

```

```

Out[10]:

```

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

50%	7.196279e+17	7.038708e+17	4.196984e+09
75%	7.993373e+17	8.257804e+17	4.196984e+09
max	8.924206e+17	8.862664e+17	8.405479e+17

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

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

In [11]: df_twitter.source.value_counts()

Out[11]: Twitter for iPhone
 Vine - Make a Scene
 Twitter Web Client
 TweetDeck
 Name: source, dtype: int64

In [12]: df_twitter.rating_denominator.value_counts()

Out[12]:

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 [13]: df_twitter.rating_numerator.value_counts()
```

```
Out[13]: 12      558
11      464
10      461
13      351
9       158
8       102
7        55
14       54
5        37
6        32
3        19
4        17
1         9
2         9
420       2
0         2
15        2
75        2
80        1
20        1
24        1
26        1
44        1
50        1
60        1
165       1
84        1
88        1
144       1
182       1
143       1
666       1
960       1
1776      1
17        1
27        1
45        1
99        1
121       1
204       1
Name: rating_numerator, dtype: int64
```

```
In [14]: df_img_pred.shape
```

```
Out[14]: (2075, 12)
```

```
In [15]: df_img_pred
```

```
Out[15]:
```

	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
5	666050758794694657	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg
6	666051853826850816	https://pbs.twimg.com/media/CT5KoJ1WoAAJash.jpg
7	666055525042405380	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg
8	666057090499244032	https://pbs.twimg.com/media/CT5PY90WoAAQGLo.jpg
9	666058600524156928	https://pbs.twimg.com/media/CT5Qw94XAAA-2dP.jpg
10	666063827256086533	https://pbs.twimg.com/media/CT5Vg_wXIAAXfnj.jpg
11	666071193221509120	https://pbs.twimg.com/media/CT5cN_3WEAA10oZ.jpg
12	666073100786774016	https://pbs.twimg.com/media/CT5d9DZXAAALcwe.jpg
13	666082916733198337	https://pbs.twimg.com/media/CT5m4VGWEAAAtKc8.jpg
14	666094000022159362	https://pbs.twimg.com/media/CT5w9gUW4AAsBNN.jpg
15	666099513787052032	https://pbs.twimg.com/media/CT51-JJUEAA6hV8.jpg
16	666102155909144576	https://pbs.twimg.com/media/CT54YGiWUAEZnoK.jpg
17	666104133288665088	https://pbs.twimg.com/media/CT56LSZWAAALJj2.jpg
18	666268910803644416	https://pbs.twimg.com/media/CT8QCd1WEAADXws.jpg
19	666273097616637952	https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg
20	666287406224695296	https://pbs.twimg.com/media/CT8g3BpUEAAUfjg.jpg
21	666293911632134144	https://pbs.twimg.com/media/CT8mx7KW4AEQu8N.jpg
22	666337882303524864	https://pbs.twimg.com/media/CT90wFIWEAMuRje.jpg
23	666345417576210432	https://pbs.twimg.com/media/CT9Vn7PWAA_ZCM.jpg
24	666353288456101888	https://pbs.twimg.com/media/CT9cx0tUEAAhNN_.jpg
25	666362758909284353	https://pbs.twimg.com/media/CT9lXGsUcAAyUft.jpg
26	666373753744588802	https://pbs.twimg.com/media/CT9vZEYWUAA1Z05.jpg
27	666396247373291520	https://pbs.twimg.com/media/CT-D2ZHWIAA3gK1.jpg
28	666407126856765440	https://pbs.twimg.com/media/CT-NvwmW4AAAugGZ.jpg
29	666411507551481857	https://pbs.twimg.com/media/CT-RugiWIAELEaq.jpg
...
2045	886366144734445568	https://pbs.twimg.com/media/DE0BTnQUwAApKEH.jpg
2046	886680336477933568	https://pbs.twimg.com/media/DE4fEDzWAAAYHMM.jpg
2047	886736880519319552	https://pbs.twimg.com/media/DE5Se8FXcAAJFx4.jpg
2048	886983233522544640	https://pbs.twimg.com/media/DE8yicJW0AAAABJ.jpg
2049	887101392804085760	https://pbs.twimg.com/media/DE-eAq6UwAA-jaE.jpg
2050	887343217045368832	https://pbs.twimg.com/ext_tw_video_thumb/88734...
2051	887473957103951883	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2052	887517139158093824	https://pbs.twimg.com/ext_tw_video_thumb/88751...
2053	887705289381826560	https://pbs.twimg.com/media/DFHDQBbXgAEqY7t.jpg

2054	888078434458587136	https://pbs.twimg.com/media/DFMWn56WsAAkA7B.jpg
2055	888202515573088257	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2056	888554962724278272	https://pbs.twimg.com/media/DFTH_0-UQAACu20.jpg
2057	888804989199671297	https://pbs.twimg.com/media/DFWra-3VYAA2piG.jpg
2058	888917238123831296	https://pbs.twimg.com/media/DFYRgsOUQAARgh0.jpg
2059	889278841981685760	https://pbs.twimg.com/ext_tw_video_thumb/88927...
2060	889531135344209921	https://pbs.twimg.com/media/DFg_2PVW0AEHN3p.jpg
2061	889638837579907072	https://pbs.twimg.com/media/DFihzFfXsAYGDPR.jpg
2062	889665388333682689	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg
2063	889880896479866881	https://pbs.twimg.com/media/DFl99B1WsAITKsg.jpg
2064	890006608113172480	https://pbs.twimg.com/media/DFnwsY4WAAAMliS.jpg
2065	890240255349198849	https://pbs.twimg.com/media/DFrEyVuW0AA03t9.jpg
2066	890609185150312448	https://pbs.twimg.com/media/DFwUU__XcAEpyXI.jpg
2067	890729181411237888	https://pbs.twimg.com/media/DFyBahAVwAAhUTd.jpg
2068	890971913173991426	https://pbs.twimg.com/media/DF1eOmZXUAAALUcq.jpg
2069	891087950875897856	https://pbs.twimg.com/media/DF3HwyEWsAABqE6.jpg
2070	891327558926688256	https://pbs.twimg.com/media/DF6hr6BUMAAZzgT.jpg
2071	891689557279858688	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg
2072	891815181378084864	https://pbs.twimg.com/media/DGBdLU1WsAAANxJ9.jpg
2073	892177421306343426	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg
2074	892420643555336193	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg

	img_num		p1	p1_conf	p1_dog	\
0	1	Welsh_springer_spaniel	0.465074		True	
1	1	redbone	0.506826		True	
2	1	German_shepherd	0.596461		True	
3	1	Rhodesian_ridgeback	0.408143		True	
4	1	miniature_pinscher	0.560311		True	
5	1	Bernese_mountain_dog	0.651137		True	
6	1	box_turtle	0.933012		False	
7	1	chow	0.692517		True	
8	1	shopping_cart	0.962465		False	
9	1	miniature_poodle	0.201493		True	
10	1	golden_retriever	0.775930		True	
11	1	Gordon_setter	0.503672		True	
12	1	Walker_hound	0.260857		True	
13	1	pug	0.489814		True	
14	1	bloodhound	0.195217		True	
15	1	Lhasa	0.582330		True	
16	1	English_setter	0.298617		True	
17	1	hen	0.965932		False	
18	1	desktop_computer	0.086502		False	
19	1	Italian_greyhound	0.176053		True	
20	1	Maltese_dog	0.857531		True	
21	1	three-toed_sloth	0.914671		False	
22	1	ox	0.416669		False	
23	1	golden_retriever	0.858744		True	
24	1	malamute	0.336874		True	

25	1	guinea_pig	0.996496	False
26	1	soft-coated_wheaten_terrier	0.326467	True
27	1	Chihuahua	0.978108	True
28	1	black-and-tan_coonhound	0.529139	True
29	1	coho	0.404640	False
...
2045	1	French_bulldog	0.999201	True
2046	1	convertible	0.738995	False
2047	1	kuvasz	0.309706	True
2048	2	Chihuahua	0.793469	True
2049	1	Samoyed	0.733942	True
2050	1	Mexican_hairless	0.330741	True
2051	2	Pembroke	0.809197	True
2052	1	limousine	0.130432	False
2053	1	basset	0.821664	True
2054	1	French_bulldog	0.995026	True
2055	2	Pembroke	0.809197	True
2056	3	Siberian_husky	0.700377	True
2057	1	golden_retriever	0.469760	True
2058	1	golden_retriever	0.714719	True
2059	1	whippet	0.626152	True
2060	1	golden_retriever	0.953442	True
2061	1	French_bulldog	0.991650	True
2062	1	Pembroke	0.966327	True
2063	1	French_bulldog	0.377417	True
2064	1	Samoyed	0.957979	True
2065	1	Pembroke	0.511319	True
2066	1	Irish_terrier	0.487574	True
2067	2	Pomeranian	0.566142	True
2068	1	Appenzeller	0.341703	True
2069	1	Chesapeake_Bay_retriever	0.425595	True
2070	2	basset	0.555712	True
2071	1	paper_towel	0.170278	False
2072	1	Chihuahua	0.716012	True
2073	1	Chihuahua	0.323581	True
2074	1	orange	0.097049	False

	p2	p2_conf	p2_dog	p3 \
0	collie	0.156665	True	Shetland_sheepdog
1	miniature_pinscher	0.074192	True	Rhodesian_ridgeback
2	malinois	0.138584	True	bloodhound
3	redbone	0.360687	True	miniature_pinscher
4	Rottweiler	0.243682	True	Doberman
5	English_springer	0.263788	True	Greater_Swiss_Mountain_dog
6	mud_turtle	0.045885	False	terrapin
7	Tibetan_mastiff	0.058279	True	fur_coat
8	shopping_basket	0.014594	False	golden_retriever
9	komondor	0.192305	True	soft-coated_wheaten_terrier

10	Tibetan_mastiff	0.093718	True	Labrador_retriever
11	Yorkshire_terrier	0.174201	True	Pekinese
12	English_foxhound	0.175382	True	Ibizan_hound
13	bull_mastiff	0.404722	True	French_bulldog
14	German_shepherd	0.078260	True	malinois
15	Shih-Tzu	0.166192	True	Dandie_Dinmont
16	Newfoundland	0.149842	True	borzoi
17	cock	0.033919	False	partridge
18	desk	0.085547	False	bookcase
19	toy_terrier	0.111884	True	basenji
20	toy_poodle	0.063064	True	miniature_poodle
21	otter	0.015250	False	great_grey_owl
22	Newfoundland	0.278407	True	groenendael
23	Chesapeake_Bay_retriever	0.054787	True	Labrador_retriever
24	Siberian_husky	0.147655	True	Eskimo_dog
25	skunk	0.002402	False	hamster
26	Afghan_hound	0.259551	True	briard
27	toy_terrier	0.009397	True	papillon
28	bloodhound	0.244220	True	flat-coated_retriever
29	barracouta	0.271485	False	gar
...
2045	Chihuahua	0.000361	True	Boston_bull
2046	sports_car	0.139952	False	car_wheel
2047	Great_Pyrenees	0.186136	True	Dandie_Dinmont
2048	toy_terrier	0.143528	True	can_opener
2049	Eskimo_dog	0.035029	True	Staffordshire_bullterrier
2050	sea_lion	0.275645	False	Weimaraner
2051	Rhodesian_ridgeback	0.054950	True	beagle
2052	tow_truck	0.029175	False	shopping_cart
2053	redbone	0.087582	True	Weimaraner
2054	pug	0.000932	True	bull_mastiff
2055	Rhodesian_ridgeback	0.054950	True	beagle
2056	Eskimo_dog	0.166511	True	malamute
2057	Labrador_retriever	0.184172	True	English_setter
2058	Tibetan_mastiff	0.120184	True	Labrador_retriever
2059	borzoi	0.194742	True	Saluki
2060	Labrador_retriever	0.013834	True	redbone
2061	boxer	0.002129	True	Staffordshire_bullterrier
2062	Cardigan	0.027356	True	basenji
2063	Labrador_retriever	0.151317	True	muzzle
2064	Pomeranian	0.013884	True	chow
2065	Cardigan	0.451038	True	Chihuahua
2066	Irish_setter	0.193054	True	Chesapeake_Bay_retriever
2067	Eskimo_dog	0.178406	True	Pembroke
2068	Border_collie	0.199287	True	ice_lolly
2069	Irish_terrier	0.116317	True	Indian_elephant
2070	English_springer	0.225770	True	German_short-haired_pointer
2071	Labrador_retriever	0.168086	True	spatula

2072	malamute	0.078253	True	kelpie
2073	Pekinese	0.090647	True	papillon
2074	bagel	0.085851	False	banana

	p3_conf	p3_dog
0	0.061428	True
1	0.072010	True
2	0.116197	True
3	0.222752	True
4	0.154629	True
5	0.016199	True
6	0.017885	False
7	0.054449	False
8	0.007959	True
9	0.082086	True
10	0.072427	True
11	0.109454	True
12	0.097471	True
13	0.048960	True
14	0.075628	True
15	0.089688	True
16	0.133649	True
17	0.000052	False
18	0.079480	False
19	0.111152	True
20	0.025581	True
21	0.013207	False
22	0.102643	True
23	0.014241	True
24	0.093412	True
25	0.000461	False
26	0.206803	True
27	0.004577	True
28	0.173810	True
29	0.189945	False
...
2045	0.000076	True
2046	0.044173	False
2047	0.086346	True
2048	0.032253	False
2049	0.029705	True
2050	0.134203	True
2051	0.038915	True
2052	0.026321	False
2053	0.026236	True
2054	0.000903	True
2055	0.038915	True
2056	0.111411	True

2057	0.073482	True
2058	0.105506	True
2059	0.027351	True
2060	0.007958	True
2061	0.001498	True
2062	0.004633	True
2063	0.082981	False
2064	0.008167	True
2065	0.029248	True
2066	0.118184	True
2067	0.076507	True
2068	0.193548	False
2069	0.076902	False
2070	0.175219	True
2071	0.040836	False
2072	0.031379	True
2073	0.068957	True
2074	0.076110	False

[2075 rows x 12 columns]

```
In [16]: # A concise summary of a DataFrame
df_img_pred.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2075 entries, 0 to 2074
Data columns (total 12 columns):
tweet_id    2075 non-null int64
jpg_url     2075 non-null object
img_num     2075 non-null int64
p1          2075 non-null object
p1_conf     2075 non-null float64
p1_dog      2075 non-null bool
p2          2075 non-null object
p2_conf     2075 non-null float64
p2_dog      2075 non-null bool
p3          2075 non-null object
p3_conf     2075 non-null float64
p3_dog      2075 non-null bool
dtypes: bool(3), float64(3), int64(2), object(4)
memory usage: 152.1+ KB
```

```
In [17]: # Descriptive statistics of a DataFrame
df_img_pred.describe()
```

```
Out[17]:
```

	tweet_id	img_num	p1_conf	p2_conf	p3_conf
count	2.075000e+03	2075.000000	2075.000000	2.075000e+03	2.075000e+03
mean	7.384514e+17	1.203855	0.594548	1.345886e-01	6.032417e-02

std	6.785203e+16	0.561875	0.271174	1.006657e-01	5.090593e-02
min	6.660209e+17	1.000000	0.044333	1.011300e-08	1.740170e-10
25%	6.764835e+17	1.000000	0.364412	5.388625e-02	1.622240e-02
50%	7.119988e+17	1.000000	0.588230	1.181810e-01	4.944380e-02
75%	7.932034e+17	1.000000	0.843855	1.955655e-01	9.180755e-02
max	8.924206e+17	4.000000	1.000000	4.880140e-01	2.734190e-01

In [18]: df_tweets.shape

Out[18]: (2354, 31)

In [19]: df_tweets

Out[19]:

	contributors	coordinates	created_at	display_text_range	\
0	NaN	NaN	2017-08-01 16:23:56	[0, 85]	
1	NaN	NaN	2017-08-01 00:17:27	[0, 138]	
2	NaN	NaN	2017-07-31 00:18:03	[0, 121]	
3	NaN	NaN	2017-07-30 15:58:51	[0, 79]	
4	NaN	NaN	2017-07-29 16:00:24	[0, 138]	
5	NaN	NaN	2017-07-29 00:08:17	[0, 138]	
6	NaN	NaN	2017-07-28 16:27:12	[0, 140]	
7	NaN	NaN	2017-07-28 00:22:40	[0, 118]	
8	NaN	NaN	2017-07-27 16:25:51	[0, 122]	
9	NaN	NaN	2017-07-26 15:59:51	[0, 133]	
10	NaN	NaN	2017-07-26 00:31:25	[0, 130]	
11	NaN	NaN	2017-07-25 16:11:53	[0, 107]	
12	NaN	NaN	2017-07-25 01:55:32	[0, 106]	
13	NaN	NaN	2017-07-25 00:10:02	[0, 91]	
14	NaN	NaN	2017-07-24 17:02:04	[0, 118]	
15	NaN	NaN	2017-07-24 00:19:32	[0, 138]	
16	NaN	NaN	2017-07-23 00:22:39	[0, 86]	
17	NaN	NaN	2017-07-22 16:56:37	[0, 128]	
18	NaN	NaN	2017-07-22 00:23:06	[0, 87]	
19	NaN	NaN	2017-07-20 16:49:33	[0, 127]	
20	NaN	NaN	2017-07-19 16:06:48	[0, 127]	
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...
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2335	NaN	NaN	2015-11-16 14:57:41	[0, 82]
2336	NaN	NaN	2015-11-16 04:02:55	[0, 134]
2337	NaN	NaN	2015-11-16 03:55:04	[0, 128]
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4	{'hashtags': [{'text': 'BarkWeek', 'indices': ...
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3	{'media': [{'id': 891689552724799489, 'id_str'...	42908
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2353	{'media': [{'id': 666020881337073664, 'id_str'...	2535

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1	False	This is Tilly. She's just checking pup on you...	NaN
2	False	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	False	This is Darla. She commenced a snooze mid meal...	NaN
4	False	This is Franklin. He would like you to stop ca...	NaN
5	False	Here we have a majestic great white breaching ...	NaN
6	False	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	False	When you watch your owner call another dog a g...	NaN
8	False	This is Zoey. She doesn't want to be one of th...	NaN
9	False	This is Cassie. She is a college pup. Studying...	NaN
10	False	This is Koda. He is a South Australian decksha...	NaN
11	False	This is Bruno. He is a service shark. Only get...	NaN
12	False	Here's a puppo that seems to be on the fence a...	NaN
13	False	This is Ted. He does his best. Sometimes that'...	NaN
14	False	This is Stuart. He's sporting his favorite fan...	NaN
15	False	This is Oliver. You're witnessing one of his m...	NaN
16	False	This is Jim. He found a fren. Taught him how t...	NaN
17	False	This is Zeke. He has a new stick. Very proud o...	NaN
18	False	This is Ralphus. He's powering up. Attempting ...	NaN
19	False	This is Gerald. He was just told he didn't get...	NaN
20	False	This is Jeffrey. He has a monopoly on the pool...	NaN
21	True	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
22	False	This is Canela. She attempted some fancy porch...	NaN
23	False	You may not have known you needed to see this ...	NaN
24	False	This... is a Jubilant Antarctic House Bear. We...	NaN
25	False	This is Maya. She's very shy. Rarely leaves he...	NaN
26	False	This is Mingus. He's a wonderful father to his...	NaN
27	False	This is Derek. He's late for a dog meeting. 13...	NaN
28	False	This is Roscoe. Another pupper fallen victim t...	NaN
29	False	@NonWhiteHat @MayhewMayhem omg hello tanner yo...	NaN
...
2324	False	This is quite the dog. Gets really excited whe...	NaN
2325	False	This is a southern Vesuvius bumblegruff. Can d...	NaN
2326	False	Oh goodness. A super rare northeast Qdoba kang...	NaN
2327	False	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2328	False	Unique dog here. Very small. Lives in containe...	NaN

2329	False	Here we have a mixed Asiago from the Galápagos...	NaN
2330	False	Look at this jokester thinking seat belt laws ...	NaN
2331	False	This is an extremely rare horned Parthenon. No...	NaN
2332	False	This is a funny dog. Weird toes. Won't come do...	NaN
2333	False	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2334	False	Can take selfies 11/10 https://t.co/ws2AMaWpW	NaN
2335	False	Very concerned about fellow dog trapped in com...	NaN
2336	False	Not familiar with this breed. No tail (weird)...	NaN
2337	False	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2338	False	Can stand on stump for what seems like a while...	NaN
2339	False	This appears to be a Mongolian Presbyterian mi...	NaN
2340	False	Here we have a well-established sunblockerspan...	NaN
2341	False	Let's hope this flight isn't Malaysian (lol). ...	NaN
2342	False	Here we have a northern speckled Rhododendron...	NaN
2343	False	This is the happiest dog you will ever see. Ve...	NaN
2344	False	Here is the Rand Paul of retrievers folks! He'...	NaN
2345	False	My oh my. This is a rare blond Canadian terrie...	NaN
2346	False	Here is a Siberian heavily armored polar bear ...	NaN
2347	False	This is an odd dog. Hard on the outside but lo...	NaN
2348	False	This is a truly beautiful English Wilson Staff...	NaN
2349	False	Here we have a 1949 1st generation vulpix. Enj...	NaN
2350	False	This is a purebred Piers Morgan. Loves to Netf...	NaN
2351	False	Here is a very happy pup. Big fan of well-main...	NaN
2352	False	This is a western brown Mitsubishi terrier. Up...	NaN
2353	False	Here we have a Japanese Irish Setter. Lost eye...	NaN

	...	\
0	...	
1	...	
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2353	NaN	532	False	NaN

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[illegible]


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[2354 rows x 31 columns]
```

```

In [20]: # A concise summary of a DataFrame
df_tweets.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2354 entries, 0 to 2353
Data columns (total 31 columns):
contributors          0 non-null float64
coordinates           0 non-null float64
created_at            2354 non-null datetime64[ns]
display_text_range    2354 non-null object
entities              2354 non-null object
extended_entities     2073 non-null object
favorite_count        2354 non-null int64
favorited             2354 non-null bool
full_text             2354 non-null object
geo                   0 non-null float64
id                    2354 non-null int64
id_str                2354 non-null int64
in_reply_to_screen_name 78 non-null object
in_reply_to_status_id  78 non-null float64
in_reply_to_status_id_str 78 non-null float64
in_reply_to_user_id    78 non-null float64
in_reply_to_user_id_str 78 non-null float64
is_quote_status       2354 non-null bool
lang                  2354 non-null object
place                 1 non-null object
possibly_sensitive     2211 non-null float64
possibly_sensitive_appealable 2211 non-null float64
quoted_status         28 non-null object
quoted_status_id      29 non-null float64
quoted_status_id_str   29 non-null float64
retweet_count         2354 non-null int64
retweeted             2354 non-null bool
retweeted_status       179 non-null object
source                2354 non-null object
truncated             2354 non-null bool
user                  2354 non-null object
dtypes: bool(4), datetime64[ns](1), float64(11), int64(4), object(11)
memory usage: 505.8+ KB

```

```

In [21]: # Descriptive statistics of a DataFrame

```

```
df_tweets.describe()
```

```
Out[21]:
```

	contributors	coordinates	favorite_count	geo	id \
count	0.0	0.0	2354.000000	0.0	2.354000e+03
mean	NaN	NaN	8080.968564	NaN	7.426978e+17
std	NaN	NaN	11814.771334	NaN	6.852812e+16
min	NaN	NaN	0.000000	NaN	6.660209e+17
25%	NaN	NaN	1415.000000	NaN	6.783975e+17
50%	NaN	NaN	3603.500000	NaN	7.194596e+17
75%	NaN	NaN	10122.250000	NaN	7.993058e+17
max	NaN	NaN	132810.000000	NaN	8.924206e+17

	id_str	in_reply_to_status_id	in_reply_to_status_id_str \
count	2.354000e+03	7.800000e+01	7.800000e+01
mean	7.426978e+17	7.455079e+17	7.455079e+17
std	6.852812e+16	7.582492e+16	7.582492e+16
min	6.660209e+17	6.658147e+17	6.658147e+17
25%	6.783975e+17	6.757419e+17	6.757419e+17
50%	7.194596e+17	7.038708e+17	7.038708e+17
75%	7.993058e+17	8.257804e+17	8.257804e+17
max	8.924206e+17	8.862664e+17	8.862664e+17

	in_reply_to_user_id	in_reply_to_user_id_str	possibly_sensitive \
count	7.800000e+01	7.800000e+01	2211.0
mean	2.014171e+16	2.014171e+16	0.0
std	1.252797e+17	1.252797e+17	0.0
min	1.185634e+07	1.185634e+07	0.0
25%	3.086374e+08	3.086374e+08	0.0
50%	4.196984e+09	4.196984e+09	0.0
75%	4.196984e+09	4.196984e+09	0.0
max	8.405479e+17	8.405479e+17	0.0

	possibly_sensitive_appealable	quoted_status_id	quoted_status_id_str \
count	2211.0	2.900000e+01	2.900000e+01
mean	0.0	8.162686e+17	8.162686e+17
std	0.0	6.164161e+16	6.164161e+16
min	0.0	6.721083e+17	6.721083e+17
25%	0.0	7.888183e+17	7.888183e+17
50%	0.0	8.340867e+17	8.340867e+17
75%	0.0	8.664587e+17	8.664587e+17
max	0.0	8.860534e+17	8.860534e+17

	retweet_count
count	2354.000000
mean	3164.797366
std	5284.770364
min	0.000000
25%	624.500000

50%	1473.500000
75%	3652.000000
max	79515.000000

1.2.1 Quality

Twitter archive enhanced table:

- Invalid values in rating_denominator and rating_numerator columns.
- Inappropriate data types of timestamp, retweeted_status_timestamp, retweeted_status_id, retweeted_status_user_id columns.
- HTML tегs in source column.
- None instead of Null in some columns.
- Invalid names in name column.
- Unnecessary columns.

Image Predictions table: * Missing records(2075 rows instead of 2356). * Unnecessary columns.

Tweets table: * Unnecessary columns(contributors, coordinates, geo, place, ...).

1.2.2 Tidiness

- Four columns instead of one(doggo, floofer, pupper, puppo).
 - Three columns for prediction, prediction confidence and prediction dog.
 - No uniform DataFrames(for convinience).
-
-

1.3 Clean

```
In [22]: # Making copies of our initial DataFrames
df_twitter_clean = df_twitter.copy()
df_img_pred_clean = df_img_pred.copy()
df_tweets_clean = df_tweets.copy()
```

1.3.1 Quality I

Tweets: Unnecessary columns(contributors, coordinates, geo, place, ...)

Define Drop unnecessary columns for analysis

Code

```
In [23]: df_tweets_clean = df_tweets_clean[['id', 'favorite_count', 'retweet_count']]
df_tweets_clean.rename(columns = {'id': 'tweet_id'}, inplace = True)
```

Test

```
In [24]: df_tweets_clean.head()
```

```
Out[24]:
```

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

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

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

1.3.2 Tidiness I

No uniform DataFrames to convinience

Define Merging three DataFrames into one

Code

```
In [26]: general = pd.merge(df_img_pred_clean, df_tweets_clean, on = 'tweet_id')
         general_clean = pd.merge(df_twitter_clean, general, on = 'tweet_id')
```

Test

```
In [27]: general_clean.head(3)
```

```
Out[27]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	

	timestamp	\
0	2017-08-01 16:23:56 +0000	

```

1 2017-08-01 00:17:27 +0000
2 2017-07-31 00:18:03 +0000

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

                                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

    retweeted_status_user_id retweeted_status_timestamp \
0                               NaN                        NaN
1                               NaN                        NaN
2                               NaN                        NaN

                                expanded_urls      ...      p1_conf \
0 https://twitter.com/dog_rates/status/892420643...      ...      0.097049
1 https://twitter.com/dog_rates/status/892177421...      ...      0.323581
2 https://twitter.com/dog_rates/status/891815181...      ...      0.716012

    p1_dog      p2      p2_conf p2_dog      p3      p3_conf p3_dog \
0 False      bagel  0.085851 False      banana  0.076110 False
1 True      Pekinese  0.090647 True      papillon  0.068957 True
2 True      malamute  0.078253 True      kelpie  0.031379 True

    favorite_count retweet_count
0          39467          8853
1          33819          6514
2          25461          4328

[3 rows x 30 columns]

```

```
In [28]: general_clean.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2073 entries, 0 to 2072
Data columns (total 30 columns):
tweet_id      2073 non-null int64
in_reply_to_status_id      23 non-null float64
in_reply_to_user_id      23 non-null float64
timestamp      2073 non-null object
source      2073 non-null object
text      2073 non-null object
retweeted_status_id      79 non-null float64
retweeted_status_user_id      79 non-null float64

```

```

retweeted_status_timestamp    79 non-null object
expanded_urls                 2073 non-null object
rating_numerator              2073 non-null int64
rating_denominator            2073 non-null int64
name                          2073 non-null object
doggo                         2073 non-null object
floofer                       2073 non-null object
pupper                        2073 non-null object
puppo                         2073 non-null object
jpg_url                       2073 non-null object
img_num                       2073 non-null int64
p1                             2073 non-null object
p1_conf                       2073 non-null float64
p1_dog                        2073 non-null bool
p2                             2073 non-null object
p2_conf                       2073 non-null float64
p2_dog                        2073 non-null bool
p3                             2073 non-null object
p3_conf                       2073 non-null float64
p3_dog                        2073 non-null bool
favorite_count                2073 non-null int64
retweet_count                 2073 non-null int64
dtypes: bool(3), float64(7), int64(6), object(14)
memory usage: 459.5+ KB

```

Four columns instead of one(doggo, floofer, pupper, puppo)

Define Transform doggo, floofer, pupper, puppo columns into one(stages_of_dog) using melt method

Code

```
In [29]: general_clean.columns
```

```

Out[29]: Index(['tweet_id', 'in_reply_to_status_id', 'in_reply_to_user_id', 'timestamp',
               'source', 'text', 'retweeted_status_id', 'retweeted_status_user_id',
               'retweeted_status_timestamp', 'expanded_urls', 'rating_numerator',
               'rating_denominator', 'name', 'doggo', 'floofer', 'pupper', 'puppo',
               'jpg_url', 'img_num', 'p1', 'p1_conf', 'p1_dog', 'p2', 'p2_conf',
               'p2_dog', 'p3', 'p3_conf', 'p3_dog', 'favorite_count', 'retweet_count'],
              dtype='object')

```

```

In [30]: general_clean = pd.melt(general_clean, id_vars=['tweet_id', 'in_reply_to_status_id', 'i
               'source', 'text', 'retweeted_status_id', 'retweeted_status_user_id',
               'retweeted_status_timestamp', 'expanded_urls', 'rating_numerator',
               'rating_denominator', 'name', 'jpg_url', 'img_num', 'p1', 'p1_conf', 'p1_dog', '
               'p2_dog', 'p3', 'p3_conf', 'p3_dog', 'favorite_count', 'retweet_count'], var_nam

```

Test

In [31]: general_clean

```
Out[31]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	
5	891087950875897856	NaN	NaN	
6	890971913173991426	NaN	NaN	
7	890729181411237888	NaN	NaN	
8	890609185150312448	NaN	NaN	
9	890240255349198849	NaN	NaN	
10	890006608113172480	NaN	NaN	
11	889880896479866881	NaN	NaN	
12	889665388333682689	NaN	NaN	
13	889638837579907072	NaN	NaN	
14	889531135344209921	NaN	NaN	
15	889278841981685760	NaN	NaN	
16	888917238123831296	NaN	NaN	
17	888804989199671297	NaN	NaN	
18	888554962724278272	NaN	NaN	
19	888078434458587136	NaN	NaN	
20	887705289381826560	NaN	NaN	
21	887517139158093824	NaN	NaN	
22	887473957103951883	NaN	NaN	
23	887343217045368832	NaN	NaN	
24	887101392804085760	NaN	NaN	
25	886983233522544640	NaN	NaN	
26	886736880519319552	NaN	NaN	
27	886680336477933568	NaN	NaN	
28	886366144734445568	NaN	NaN	
29	886258384151887873	NaN	NaN	
...	
8262	666411507551481857	NaN	NaN	
8263	666407126856765440	NaN	NaN	
8264	666396247373291520	NaN	NaN	
8265	666373753744588802	NaN	NaN	
8266	666362758909284353	NaN	NaN	
8267	666353288456101888	NaN	NaN	
8268	666345417576210432	NaN	NaN	
8269	666337882303524864	NaN	NaN	
8270	666293911632134144	NaN	NaN	
8271	666287406224695296	NaN	NaN	
8272	666273097616637952	NaN	NaN	
8273	666268910803644416	NaN	NaN	
8274	666104133288665088	NaN	NaN	

8275	666102155909144576	NaN	NaN
8276	666099513787052032	NaN	NaN
8277	666094000022159362	NaN	NaN
8278	666082916733198337	NaN	NaN
8279	666073100786774016	NaN	NaN
8280	666071193221509120	NaN	NaN
8281	666063827256086533	NaN	NaN
8282	666058600524156928	NaN	NaN
8283	666057090499244032	NaN	NaN
8284	666055525042405380	NaN	NaN
8285	666051853826850816	NaN	NaN
8286	666050758794694657	NaN	NaN
8287	666049248165822465	NaN	NaN
8288	666044226329800704	NaN	NaN
8289	666033412701032449	NaN	NaN
8290	666029285002620928	NaN	NaN
8291	666020888022790149	NaN	NaN

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


```

29      2017-07-15 16:17:19 +0000
...
8262    2015-11-17 00:24:19 +0000
8263    2015-11-17 00:06:54 +0000
8264    2015-11-16 23:23:41 +0000
8265    2015-11-16 21:54:18 +0000
8266    2015-11-16 21:10:36 +0000
8267    2015-11-16 20:32:58 +0000
8268    2015-11-16 20:01:42 +0000
8269    2015-11-16 19:31:45 +0000
8270    2015-11-16 16:37:02 +0000
8271    2015-11-16 16:11:11 +0000
8272    2015-11-16 15:14:19 +0000
8273    2015-11-16 14:57:41 +0000
8274    2015-11-16 04:02:55 +0000
8275    2015-11-16 03:55:04 +0000
8276    2015-11-16 03:44:34 +0000
8277    2015-11-16 03:22:39 +0000
8278    2015-11-16 02:38:37 +0000
8279    2015-11-16 01:59:36 +0000
8280    2015-11-16 01:52:02 +0000
8281    2015-11-16 01:22:45 +0000
8282    2015-11-16 01:01:59 +0000
8283    2015-11-16 00:55:59 +0000
8284    2015-11-16 00:49:46 +0000
8285    2015-11-16 00:35:11 +0000
8286    2015-11-16 00:30:50 +0000
8287    2015-11-16 00:24:50 +0000
8288    2015-11-16 00:04:52 +0000
8289    2015-11-15 23:21:54 +0000
8290    2015-11-15 23:05:30 +0000
8291    2015-11-15 22:32:08 +0000

```

```

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

```

[illegible]

	text	retweeted_status_id \
0	This is Phineas. He's a mystical boy. Only eve...	NaN
1	This is Tilly. She's just checking pup on you...	NaN
2	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	This is Darla. She commenced a snooze mid meal...	NaN
4	This is Franklin. He would like you to stop ca...	NaN
5	Here we have a majestic great white breaching ...	NaN
6	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	When you watch your owner call another dog a g...	NaN
8	This is Zoey. She doesn't want to be one of th...	NaN
9	This is Cassie. She is a college pup. Studying...	NaN
10	This is Koda. He is a South Australian decksha...	NaN
11	This is Bruno. He is a service shark. Only get...	NaN
12	Here's a puppo that seems to be on the fence a...	NaN
13	This is Ted. He does his best. Sometimes that'...	NaN
14	This is Stuart. He's sporting his favorite fan...	NaN
15	This is Oliver. You're witnessing one of his m...	NaN
16	This is Jim. He found a fren. Taught him how t...	NaN
17	This is Zeke. He has a new stick. Very proud o...	NaN
18	This is Ralphus. He's powering up. Attempting ...	NaN
19	This is Gerald. He was just told he didn't get...	NaN
20	This is Jeffrey. He has a monopoly on the pool...	NaN
21	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
22	This is Canela. She attempted some fancy porch...	NaN
23	You may not have known you needed to see this ...	NaN
24	This... is a Jubilant Antarctic House Bear. We...	NaN
25	This is Maya. She's very shy. Rarely leaves he...	NaN
26	This is Mingus. He's a wonderful father to his...	NaN
27	This is Derek. He's late for a dog meeting. 13...	NaN
28	This is Roscoe. Another pupper fallen victim t...	NaN
29	This is Waffles. His doggles are pupside down...	NaN
...
8262	This is quite the dog. Gets really excited whe...	NaN
8263	This is a southern Vesuvius bumblegruff. Can d...	NaN
8264	Oh goodness. A super rare northeast Qdoba kang...	NaN
8265	Those are sunglasses and a jean jacket. 11/10 ...	NaN
8266	Unique dog here. Very small. Lives in containe...	NaN
8267	Here we have a mixed Asiago from the Galápagos...	NaN
8268	Look at this jokester thinking seat belt laws ...	NaN
8269	This is an extremely rare horned Parthenon. No...	NaN
8270	This is a funny dog. Weird toes. Won't come do...	NaN
8271	This is an Albanian 3 1/2 legged Episcopalian...	NaN
8272	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
8273	Very concerned about fellow dog trapped in com...	NaN
8274	Not familiar with this breed. No tail (weird)...	NaN
8275	Oh my. Here you are seeing an Adobe Setter giv...	NaN
8276	Can stand on stump for what seems like a while...	NaN
8277	This appears to be a Mongolian Presbyterian mi...	NaN

8278	Here we have a well-established sunblockerspan...	NaN
8279	Let's hope this flight isn't Malaysian (lol). ...	NaN
8280	Here we have a northern speckled Rhododendron...	NaN
8281	This is the happiest dog you will ever see. Ve...	NaN
8282	Here is the Rand Paul of retrievers folks! He'...	NaN
8283	My oh my. This is a rare blond Canadian terrie...	NaN
8284	Here is a Siberian heavily armored polar bear ...	NaN
8285	This is an odd dog. Hard on the outside but lo...	NaN
8286	This is a truly beautiful English Wilson Staff...	NaN
8287	Here we have a 1949 1st generation vulpix. Enj...	NaN
8288	This is a purebred Piers Morgan. Loves to Netf...	NaN
8289	Here is a very happy pup. Big fan of well-main...	NaN
8290	This is a western brown Mitsubishi terrier. Up...	NaN
8291	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN
10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	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
...
8262	NaN	NaN

8263	NaN	NaN
8264	NaN	NaN
8265	NaN	NaN
8266	NaN	NaN
8267	NaN	NaN
8268	NaN	NaN
8269	NaN	NaN
8270	NaN	NaN
8271	NaN	NaN
8272	NaN	NaN
8273	NaN	NaN
8274	NaN	NaN
8275	NaN	NaN
8276	NaN	NaN
8277	NaN	NaN
8278	NaN	NaN
8279	NaN	NaN
8280	NaN	NaN
8281	NaN	NaN
8282	NaN	NaN
8283	NaN	NaN
8284	NaN	NaN
8285	NaN	NaN
8286	NaN	NaN
8287	NaN	NaN
8288	NaN	NaN
8289	NaN	NaN
8290	NaN	NaN
8291	NaN	NaN

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

17 https://twitter.com/dog_rates/status/888804989...
 18 https://twitter.com/dog_rates/status/888554962...
 19 https://twitter.com/dog_rates/status/888078434...
 20 https://twitter.com/dog_rates/status/887705289...
 21 https://twitter.com/dog_rates/status/887517139...
 22 https://twitter.com/dog_rates/status/887473957...
 23 https://twitter.com/dog_rates/status/887343217...
 24 https://twitter.com/dog_rates/status/887101392...
 25 https://twitter.com/dog_rates/status/886983233...
 26 https://www.gofundme.com/mingusneedsus,https://...
 27 https://twitter.com/dog_rates/status/886680336...
 28 https://twitter.com/dog_rates/status/886366144...
 29 https://twitter.com/dog_rates/status/886258384...
 ...
 8262 https://twitter.com/dog_rates/status/666411507...
 8263 https://twitter.com/dog_rates/status/666407126...
 8264 https://twitter.com/dog_rates/status/666396247...
 8265 https://twitter.com/dog_rates/status/666373753...
 8266 https://twitter.com/dog_rates/status/666362758...
 8267 https://twitter.com/dog_rates/status/666353288...
 8268 https://twitter.com/dog_rates/status/666345417...
 8269 https://twitter.com/dog_rates/status/666337882...
 8270 https://twitter.com/dog_rates/status/666293911...
 8271 https://twitter.com/dog_rates/status/666287406...
 8272 https://twitter.com/dog_rates/status/666273097...
 8273 https://twitter.com/dog_rates/status/666268910...
 8274 https://twitter.com/dog_rates/status/666104133...
 8275 https://twitter.com/dog_rates/status/666102155...
 8276 https://twitter.com/dog_rates/status/666099513...
 8277 https://twitter.com/dog_rates/status/666094000...
 8278 https://twitter.com/dog_rates/status/666082916...
 8279 https://twitter.com/dog_rates/status/666073100...
 8280 https://twitter.com/dog_rates/status/666071193...
 8281 https://twitter.com/dog_rates/status/666063827...
 8282 https://twitter.com/dog_rates/status/666058600...
 8283 https://twitter.com/dog_rates/status/666057090...
 8284 https://twitter.com/dog_rates/status/666055525...
 8285 https://twitter.com/dog_rates/status/666051853...
 8286 https://twitter.com/dog_rates/status/666050758...
 8287 https://twitter.com/dog_rates/status/666049248...
 8288 https://twitter.com/dog_rates/status/666044226...
 8289 https://twitter.com/dog_rates/status/666033412...
 8290 https://twitter.com/dog_rates/status/666029285...
 8291 https://twitter.com/dog_rates/status/666020888...

	p2	p2_conf	p2_dog	p3 \
0	bagel	0.085851	False	banana
1	Pekinese	0.090647	True	papillon

2	malamute	0.078253	True	kelpie
3	Labrador_retriever	0.168086	True	spatula
4	English_springer	0.225770	True	German_short-haired_pointer
5	Irish_terrier	0.116317	True	Indian_elephant
6	Border_collie	0.199287	True	ice_lolly
7	Eskimo_dog	0.178406	True	Pembroke
8	Irish_setter	0.193054	True	Chesapeake_Bay_retriever
9	Cardigan	0.451038	True	Chihuahua
10	Pomeranian	0.013884	True	chow
11	Labrador_retriever	0.151317	True	muzzle
12	Cardigan	0.027356	True	basenji
13	boxer	0.002129	True	Staffordshire_bullterrier
14	Labrador_retriever	0.013834	True	redbone
15	borzoi	0.194742	True	Saluki
16	Tibetan_mastiff	0.120184	True	Labrador_retriever
17	Labrador_retriever	0.184172	True	English_setter
18	Eskimo_dog	0.166511	True	malamute
19	pug	0.000932	True	bull_mastiff
20	redbone	0.087582	True	Weimaraner
21	tow_truck	0.029175	False	shopping_cart
22	Rhodesian_ridgeback	0.054950	True	beagle
23	sea_lion	0.275645	False	Weimaraner
24	Eskimo_dog	0.035029	True	Staffordshire_bullterrier
25	toy_terrier	0.143528	True	can_opener
26	Great_Pyrenees	0.186136	True	Dandie_Dinmont
27	sports_car	0.139952	False	car_wheel
28	Chihuahua	0.000361	True	Boston_bull
29	shower_cap	0.025286	False	Siamese_cat
...
8262	barracouta	0.271485	False	gar
8263	bloodhound	0.244220	True	flat-coated_retriever
8264	toy_terrier	0.009397	True	papillon
8265	Afghan_hound	0.259551	True	briard
8266	skunk	0.002402	False	hamster
8267	Siberian_husky	0.147655	True	Eskimo_dog
8268	Chesapeake_Bay_retriever	0.054787	True	Labrador_retriever
8269	Newfoundland	0.278407	True	groenendael
8270	otter	0.015250	False	great_grey_owl
8271	toy_poodle	0.063064	True	miniature_poodle
8272	toy_terrier	0.111884	True	basenji
8273	desk	0.085547	False	bookcase
8274	cock	0.033919	False	partridge
8275	Newfoundland	0.149842	True	borzoi
8276	Shih-Tzu	0.166192	True	Dandie_Dinmont
8277	German_shepherd	0.078260	True	malinois
8278	bull_mastiff	0.404722	True	French_bulldog
8279	English_foxhound	0.175382	True	Ibizan_hound
8280	Yorkshire_terrier	0.174201	True	Pekinese

8281	Tibetan_mastiff	0.093718	True	Labrador_retriever
8282	komondor	0.192305	True	soft-coated_wheaten_terrier
8283	shopping_basket	0.014594	False	golden_retriever
8284	Tibetan_mastiff	0.058279	True	fur_coat
8285	mud_turtle	0.045885	False	terrapin
8286	English_springer	0.263788	True	Greater_Swiss_Mountain_dog
8287	Rottweiler	0.243682	True	Doberman
8288	redbone	0.360687	True	miniature_pinscher
8289	malinois	0.138584	True	bloodhound
8290	miniature_pinscher	0.074192	True	Rhodesian_ridgeback
8291	collie	0.156665	True	Shetland_sheepdog

	p3_conf	p3_dog	favorite_count	retweet_count	stage_of_dog	value
0	0.076110	False	39467	8853	doggo	None
1	0.068957	True	33819	6514	doggo	None
2	0.031379	True	25461	4328	doggo	None
3	0.040836	False	42908	8964	doggo	None
4	0.175219	True	41048	9774	doggo	None
5	0.076902	False	20562	3261	doggo	None
6	0.193548	False	12041	2158	doggo	None
7	0.076507	True	56848	16716	doggo	None
8	0.118184	True	28226	4429	doggo	None
9	0.029248	True	32467	7711	doggo	doggo
10	0.008167	True	31166	7624	doggo	None
11	0.082981	False	28268	5156	doggo	None
12	0.004633	True	38818	8538	doggo	None
13	0.001498	True	27672	4735	doggo	None
14	0.007958	True	15359	2321	doggo	None
15	0.027351	True	25652	5637	doggo	None
16	0.105506	True	29611	4709	doggo	None
17	0.073482	True	26080	4559	doggo	None
18	0.111411	True	20290	3732	doggo	None
19	0.000903	True	22201	3653	doggo	None
20	0.026236	True	30779	5609	doggo	None
21	0.026321	False	46959	12082	doggo	None
22	0.038915	True	69871	18781	doggo	None
23	0.134203	True	34222	10737	doggo	None
24	0.029705	True	31061	6167	doggo	None
25	0.032253	False	35859	8084	doggo	None
26	0.086346	True	12306	3443	doggo	None
27	0.044173	False	22798	4610	doggo	None
28	0.000076	True	21524	3316	doggo	None
29	0.002849	False	28469	6523	doggo	None
...
8262	0.189945	False	459	339	puppo	None
8263	0.173810	True	113	44	puppo	None
8264	0.004577	True	172	92	puppo	None
8265	0.206803	True	194	100	puppo	None

8266	0.000461	False	804	595	puppo	None
8267	0.093412	True	229	77	puppo	None
8268	0.014241	True	307	146	puppo	None
8269	0.102643	True	204	96	puppo	None
8270	0.013207	False	522	368	puppo	None
8271	0.025581	True	152	71	puppo	None
8272	0.111152	True	184	82	puppo	None
8273	0.079480	False	108	37	puppo	None
8274	0.000052	False	14765	6871	puppo	None
8275	0.133649	True	81	16	puppo	None
8276	0.089688	True	164	73	puppo	None
8277	0.075628	True	169	79	puppo	None
8278	0.048960	True	121	47	puppo	None
8279	0.097471	True	335	174	puppo	None
8280	0.109454	True	154	67	puppo	None
8281	0.072427	True	496	232	puppo	None
8282	0.082086	True	115	61	puppo	None
8283	0.007959	True	304	146	puppo	None
8284	0.054449	False	448	261	puppo	None
8285	0.017885	False	1253	879	puppo	None
8286	0.016199	True	136	60	puppo	None
8287	0.154629	True	111	41	puppo	None
8288	0.222752	True	311	147	puppo	None
8289	0.116197	True	128	47	puppo	None
8290	0.072010	True	132	48	puppo	None
8291	0.061428	True	2535	532	puppo	None

[8292 rows x 28 columns]

In [32]: `general_clean.value.value_counts()`

```
Out[32]: None      7959
pupper      221
doggo       80
puppo       24
floofer      8
Name: value, dtype: int64
```

1.3.3 Quality II

Unnecessary columns

Define Drop unnecessary columns

Code

```
In [33]: general_clean.drop(['in_reply_to_status_id', 'in_reply_to_user_id', 'retweeted_status_id',  
                             'retweeted_status_user_id', 'retweeted_status_timestamp'], axis=1,
```

Test

```
In [34]: general_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 8292 entries, 0 to 8291  
Data columns (total 23 columns):  
tweet_id          8292 non-null int64  
timestamp         8292 non-null object  
source            8292 non-null object  
text              8292 non-null object  
expanded_urls     8292 non-null object  
rating_numerator  8292 non-null int64  
rating_denominator 8292 non-null int64  
name              8292 non-null object  
jpg_url           8292 non-null object  
img_num           8292 non-null int64  
p1                8292 non-null object  
p1_conf           8292 non-null float64  
p1_dog            8292 non-null bool  
p2                8292 non-null object  
p2_conf           8292 non-null float64  
p2_dog            8292 non-null bool  
p3                8292 non-null object  
p3_conf           8292 non-null float64  
p3_dog            8292 non-null bool  
favorite_count    8292 non-null int64  
retweet_count     8292 non-null int64  
stage_of_dog      8292 non-null object  
value             8292 non-null object  
dtypes: bool(3), float64(3), int64(6), object(11)  
memory usage: 1.3+ MB
```

Data type of timestamp column

Define Change the type of data using `to_datetime()` function

Code

```
In [35]: general_clean.timestamp = pd.to_datetime(general_clean.timestamp)
```

Test

```
In [36]: general_clean.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8292 entries, 0 to 8291
Data columns (total 23 columns):
tweet_id          8292 non-null int64
timestamp         8292 non-null datetime64[ns]
source            8292 non-null object
text              8292 non-null object
expanded_urls     8292 non-null object
rating_numerator  8292 non-null int64
rating_denominator 8292 non-null int64
name              8292 non-null object
jpg_url           8292 non-null object
img_num           8292 non-null int64
p1                8292 non-null object
p1_conf           8292 non-null float64
p1_dog            8292 non-null bool
p2                8292 non-null object
p2_conf           8292 non-null float64
p2_dog            8292 non-null bool
p3                8292 non-null object
p3_conf           8292 non-null float64
p3_dog            8292 non-null bool
favorite_count    8292 non-null int64
retweet_count     8292 non-null int64
stage_of_dog      8292 non-null object
value             8292 non-null object
dtypes: bool(3), datetime64[ns](1), float64(3), int64(6), object(10)
memory usage: 1.3+ MB
```

Invalid values in rating_denominator and rating_numerator columns

Define Find the quotient of rating_denominator and rating_numerator columns to find the rating (It seems that the invalid values in the rating_denominator and rating_numerator columns are not random, because when a large value appears in some column, approximately the same value appears in another column (f.e. 140 and 134, 76 and 70. Not like 90 and 10 or 12 and 50). Because of this, I will find the quotient of these columns so that there is not much difference between the values)

Code

```
In [37]: general_clean['rating'] = general_clean.rating_numerator / general_clean.rating_denominator

In [38]: general_clean.drop(['rating_numerator', 'rating_denominator'], axis = 1, inplace = True)
```

Test

In [39]: general_clean

```
Out[39]:
```

	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	
19	888078434458587136	2017-07-20 16:49:33	
20	887705289381826560	2017-07-19 16:06:48	
21	887517139158093824	2017-07-19 03:39:09	
22	887473957103951883	2017-07-19 00:47:34	
23	887343217045368832	2017-07-18 16:08:03	
24	887101392804085760	2017-07-18 00:07:08	
25	886983233522544640	2017-07-17 16:17:36	
26	886736880519319552	2017-07-16 23:58:41	
27	886680336477933568	2017-07-16 20:14:00	
28	886366144734445568	2017-07-15 23:25:31	
29	886258384151887873	2017-07-15 16:17:19	
...	
8262	666411507551481857	2015-11-17 00:24:19	
8263	666407126856765440	2015-11-17 00:06:54	
8264	666396247373291520	2015-11-16 23:23:41	
8265	666373753744588802	2015-11-16 21:54:18	
8266	666362758909284353	2015-11-16 21:10:36	
8267	666353288456101888	2015-11-16 20:32:58	
8268	666345417576210432	2015-11-16 20:01:42	
8269	666337882303524864	2015-11-16 19:31:45	
8270	666293911632134144	2015-11-16 16:37:02	
8271	666287406224695296	2015-11-16 16:11:11	
8272	666273097616637952	2015-11-16 15:14:19	
8273	666268910803644416	2015-11-16 14:57:41	
8274	666104133288665088	2015-11-16 04:02:55	

8275 666102155909144576 2015-11-16 03:55:04
 8276 666099513787052032 2015-11-16 03:44:34
 8277 666094000022159362 2015-11-16 03:22:39
 8278 666082916733198337 2015-11-16 02:38:37
 8279 666073100786774016 2015-11-16 01:59:36
 8280 666071193221509120 2015-11-16 01:52:02
 8281 666063827256086533 2015-11-16 01:22:45
 8282 666058600524156928 2015-11-16 01:01:59
 8283 666057090499244032 2015-11-16 00:55:59
 8284 666055525042405380 2015-11-16 00:49:46
 8285 666051853826850816 2015-11-16 00:35:11
 8286 666050758794694657 2015-11-16 00:30:50
 8287 666049248165822465 2015-11-16 00:24:50
 8288 666044226329800704 2015-11-16 00:04:52
 8289 666033412701032449 2015-11-15 23:21:54
 8290 666029285002620928 2015-11-15 23:05:30
 8291 666020888022790149 2015-11-15 22:32:08

```

                                source \
0  <a href="http://twitter.com/download/iphone" r...
1  <a href="http://twitter.com/download/iphone" r...
2  <a href="http://twitter.com/download/iphone" r...
3  <a href="http://twitter.com/download/iphone" r...
4  <a href="http://twitter.com/download/iphone" r...
5  <a href="http://twitter.com/download/iphone" r...
6  <a href="http://twitter.com/download/iphone" r...
7  <a href="http://twitter.com/download/iphone" r...
8  <a href="http://twitter.com/download/iphone" r...
9  <a href="http://twitter.com/download/iphone" r...
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14 <a href="http://twitter.com/download/iphone" r...
15 <a href="http://twitter.com/download/iphone" r...
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26 <a href="http://twitter.com/download/iphone" r...
27 <a href="http://twitter.com/download/iphone" r...
28 <a href="http://twitter.com/download/iphone" r...
  
```

```

29      <a href="http://twitter.com/download/iphone" r...
...
8262    <a href="http://twitter.com/download/iphone" r...
8263    <a href="http://twitter.com/download/iphone" r...
8264    <a href="http://twitter.com/download/iphone" r...
8265    <a href="http://twitter.com/download/iphone" r...
8266    <a href="http://twitter.com/download/iphone" r...
8267    <a href="http://twitter.com/download/iphone" r...
8268    <a href="http://twitter.com/download/iphone" r...
8269    <a href="http://twitter.com/download/iphone" r...
8270    <a href="http://twitter.com/download/iphone" r...
8271    <a href="http://twitter.com/download/iphone" r...
8272    <a href="http://twitter.com/download/iphone" r...
8273    <a href="http://twitter.com/download/iphone" r...
8274    <a href="http://twitter.com/download/iphone" r...
8275    <a href="http://twitter.com/download/iphone" r...
8276    <a href="http://twitter.com/download/iphone" r...
8277    <a href="http://twitter.com/download/iphone" r...
8278    <a href="http://twitter.com/download/iphone" r...
8279    <a href="http://twitter.com/download/iphone" r...
8280    <a href="http://twitter.com/download/iphone" r...
8281    <a href="http://twitter.com/download/iphone" r...
8282    <a href="http://twitter.com/download/iphone" r...
8283    <a href="http://twitter.com/download/iphone" r...
8284    <a href="http://twitter.com/download/iphone" r...
8285    <a href="http://twitter.com/download/iphone" r...
8286    <a href="http://twitter.com/download/iphone" r...
8287    <a href="http://twitter.com/download/iphone" r...
8288    <a href="http://twitter.com/download/iphone" r...
8289    <a href="http://twitter.com/download/iphone" r...
8290    <a href="http://twitter.com/download/iphone" r...
8291    <a href="http://twitter.com/download/iphone" r...

```

text \

```

0      This is Phineas. He's a mystical boy. Only eve...
1      This is Tilly. She's just checking pup on you...
2      This is Archie. He is a rare Norwegian Pouncin...
3      This is Darla. She commenced a snooze mid meal...
4      This is Franklin. He would like you to stop ca...
5      Here we have a majestic great white breaching ...
6      Meet Jax. He enjoys ice cream so much he gets ...
7      When you watch your owner call another dog a g...
8      This is Zoey. She doesn't want to be one of th...
9      This is Cassie. She is a college pup. Studying...
10     This is Koda. He is a South Australian decksha...
11     This is Bruno. He is a service shark. Only get...
12     Here's a puppo that seems to be on the fence a...
13     This is Ted. He does his best. Sometimes that'...

```

14 This is Stuart. He's sporting his favorite fan...
 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 ...
 19 This is Gerald. He was just told he didn't get...
 20 This is Jeffrey. He has a monopoly on the pool...
 21 I've yet to rate a Venezuelan Hover Wiener. Th...
 22 This is Canela. She attempted some fancy porch...
 23 You may not have known you needed to see this ...
 24 This... is a Jubilant Antarctic House Bear. We...
 25 This is Maya. She's very shy. Rarely leaves he...
 26 This is Mingus. He's a wonderful father to his...
 27 This is Derek. He's late for a dog meeting. 13...
 28 This is Roscoe. Another pupper fallen victim t...
 29 This is Waffles. His doggles are pupside down...
 ...
 8262 This is quite the dog. Gets really excited whe...
 8263 This is a southern Vesuvius bumblegruff. Can d...
 8264 Oh goodness. A super rare northeast Qdoba kang...
 8265 Those are sunglasses and a jean jacket. 11/10 ...
 8266 Unique dog here. Very small. Lives in containe...
 8267 Here we have a mixed Asiago from the Galápagos...
 8268 Look at this jokester thinking seat belt laws ...
 8269 This is an extremely rare horned Parthenon. No...
 8270 This is a funny dog. Weird toes. Won't come do...
 8271 This is an Albanian 3 1/2 legged Episcopalian...
 8272 Can take selfies 11/10 <https://t.co/ws2AMaNwPW>
 8273 Very concerned about fellow dog trapped in com...
 8274 Not familiar with this breed. No tail (weird)...
 8275 Oh my. Here you are seeing an Adobe Setter giv...
 8276 Can stand on stump for what seems like a while...
 8277 This appears to be a Mongolian Presbyterian mi...
 8278 Here we have a well-established sunblockerspan...
 8279 Let's hope this flight isn't Malaysian (lol). ...
 8280 Here we have a northern speckled Rhododendron...
 8281 This is the happiest dog you will ever see. Ve...
 8282 Here is the Rand Paul of retrievers folks! He'...
 8283 My oh my. This is a rare blond Canadian terrie...
 8284 Here is a Siberian heavily armored polar bear ...
 8285 This is an odd dog. Hard on the outside but lo...
 8286 This is a truly beautiful English Wilson Staff...
 8287 Here we have a 1949 1st generation vulpix. Enj...
 8288 This is a purebred Piers Morgan. Loves to Netf...
 8289 Here is a very happy pup. Big fan of well-main...
 8290 This is a western brown Mitsubishi terrier. Up...
 8291 Here we have a Japanese Irish Setter. Lost eye...

	expanded_urls	name \
0	https://twitter.com/dog_rates/status/892420643...	Phineas
1	https://twitter.com/dog_rates/status/892177421...	Tilly
2	https://twitter.com/dog_rates/status/891815181...	Archie
3	https://twitter.com/dog_rates/status/891689557...	Darla
4	https://twitter.com/dog_rates/status/891327558...	Franklin
5	https://twitter.com/dog_rates/status/891087950...	None
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	Jax
7	https://twitter.com/dog_rates/status/890729181...	None
8	https://twitter.com/dog_rates/status/890609185...	Zoey
9	https://twitter.com/dog_rates/status/890240255...	Cassie
10	https://twitter.com/dog_rates/status/890006608...	Koda
11	https://twitter.com/dog_rates/status/889880896...	Bruno
12	https://twitter.com/dog_rates/status/889665388...	None
13	https://twitter.com/dog_rates/status/889638837...	Ted
14	https://twitter.com/dog_rates/status/889531135...	Stuart
15	https://twitter.com/dog_rates/status/889278841...	Oliver
16	https://twitter.com/dog_rates/status/888917238...	Jim
17	https://twitter.com/dog_rates/status/888804989...	Zeke
18	https://twitter.com/dog_rates/status/888554962...	Ralphus
19	https://twitter.com/dog_rates/status/888078434...	Gerald
20	https://twitter.com/dog_rates/status/887705289...	Jeffrey
21	https://twitter.com/dog_rates/status/887517139...	such
22	https://twitter.com/dog_rates/status/887473957...	Canela
23	https://twitter.com/dog_rates/status/887343217...	None
24	https://twitter.com/dog_rates/status/887101392...	None
25	https://twitter.com/dog_rates/status/886983233...	Maya
26	https://www.gofundme.com/mingusneedsus,https:/...	Mingus
27	https://twitter.com/dog_rates/status/886680336...	Derek
28	https://twitter.com/dog_rates/status/886366144...	Roscoe
29	https://twitter.com/dog_rates/status/886258384...	Waffles
...
8262	https://twitter.com/dog_rates/status/666411507...	quite
8263	https://twitter.com/dog_rates/status/666407126...	a
8264	https://twitter.com/dog_rates/status/666396247...	None
8265	https://twitter.com/dog_rates/status/666373753...	None
8266	https://twitter.com/dog_rates/status/666362758...	None
8267	https://twitter.com/dog_rates/status/666353288...	None
8268	https://twitter.com/dog_rates/status/666345417...	None
8269	https://twitter.com/dog_rates/status/666337882...	an
8270	https://twitter.com/dog_rates/status/666293911...	a
8271	https://twitter.com/dog_rates/status/666287406...	an
8272	https://twitter.com/dog_rates/status/666273097...	None
8273	https://twitter.com/dog_rates/status/666268910...	None
8274	https://twitter.com/dog_rates/status/666104133...	None
8275	https://twitter.com/dog_rates/status/666102155...	None
8276	https://twitter.com/dog_rates/status/666099513...	None
8277	https://twitter.com/dog_rates/status/666094000...	None

8278	https://twitter.com/dog_rates/status/666082916...	None
8279	https://twitter.com/dog_rates/status/666073100...	None
8280	https://twitter.com/dog_rates/status/666071193...	None
8281	https://twitter.com/dog_rates/status/666063827...	the
8282	https://twitter.com/dog_rates/status/666058600...	the
8283	https://twitter.com/dog_rates/status/666057090...	a
8284	https://twitter.com/dog_rates/status/666055525...	a
8285	https://twitter.com/dog_rates/status/666051853...	an
8286	https://twitter.com/dog_rates/status/666050758...	a
8287	https://twitter.com/dog_rates/status/666049248...	None
8288	https://twitter.com/dog_rates/status/666044226...	a
8289	https://twitter.com/dog_rates/status/666033412...	a
8290	https://twitter.com/dog_rates/status/666029285...	a
8291	https://twitter.com/dog_rates/status/666020888...	None

	jpg_url	img_num \
0	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg	1
1	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg	1
2	https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg	1
3	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg	1
4	https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.jpg	2
5	https://pbs.twimg.com/media/DF3HwyEWsAABqE6.jpg	1
6	https://pbs.twimg.com/media/DF1eOmZXUAAALUc.jpg	1
7	https://pbs.twimg.com/media/DFyBahAVwAAhUTd.jpg	2
8	https://pbs.twimg.com/media/DFwUU_XcAEpyXI.jpg	1
9	https://pbs.twimg.com/media/DFrEyVuW0AAO3t9.jpg	1
10	https://pbs.twimg.com/media/DFnwSY4WAAAMliS.jpg	1
11	https://pbs.twimg.com/media/DF199B1WsAITKsg.jpg	1
12	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg	1
13	https://pbs.twimg.com/media/DFihzFfXsAYGDPR.jpg	1
14	https://pbs.twimg.com/media/DFg_2PVW0AEHN3p.jpg	1
15	https://pbs.twimg.com/ext_tw_video_thumb/88927...	1
16	https://pbs.twimg.com/media/DFYRgsOUQAARGh0.jpg	1
17	https://pbs.twimg.com/media/DFWra-3VYAA2piG.jpg	1
18	https://pbs.twimg.com/media/DFTH_0-UQAACu20.jpg	3
19	https://pbs.twimg.com/media/DFMwn56WsAAka7B.jpg	1
20	https://pbs.twimg.com/media/DFHDQBbXgAEqY7t.jpg	1
21	https://pbs.twimg.com/ext_tw_video_thumb/88751...	1
22	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg	2
23	https://pbs.twimg.com/ext_tw_video_thumb/88734...	1
24	https://pbs.twimg.com/media/DE-eAq6UwAA-jaE.jpg	1
25	https://pbs.twimg.com/media/DE8yicJW0AAAavBJ.jpg	2
26	https://pbs.twimg.com/media/DE5Se8FXcAAJFx4.jpg	1
27	https://pbs.twimg.com/media/DE4fEDzWAAayHMM.jpg	1
28	https://pbs.twimg.com/media/DE0BTnQUwAApKEH.jpg	1
29	https://pbs.twimg.com/media/DEyfTG4UMAE4aE9.jpg	1
...
8262	https://pbs.twimg.com/media/CT-RugiWIAELEaq.jpg	1

8263	https://pbs.twimg.com/media/CT-NvwmW4AAugGZ.jpg	1
8264	https://pbs.twimg.com/media/CT-D2ZHWIAA3gK1.jpg	1
8265	https://pbs.twimg.com/media/CT9vZEYWUAA1Z05.jpg	1
8266	https://pbs.twimg.com/media/CT9lXGsUcAAyUFt.jpg	1
8267	https://pbs.twimg.com/media/CT9cx0tUEAAhNN_.jpg	1
8268	https://pbs.twimg.com/media/CT9Vn7PW0AA_ZCM.jpg	1
8269	https://pbs.twimg.com/media/CT90wFIWEAMuRje.jpg	1
8270	https://pbs.twimg.com/media/CT8mx7KW4AEQu8N.jpg	1
8271	https://pbs.twimg.com/media/CT8g3BpUEAAuFjg.jpg	1
8272	https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg	1
8273	https://pbs.twimg.com/media/CT8QCd1WEAADXws.jpg	1
8274	https://pbs.twimg.com/media/CT56LSZW0AALJj2.jpg	1
8275	https://pbs.twimg.com/media/CT54YGiWUAEZnoK.jpg	1
8276	https://pbs.twimg.com/media/CT51-JJUEAA6hV8.jpg	1
8277	https://pbs.twimg.com/media/CT5w9gUW4AAsBNN.jpg	1
8278	https://pbs.twimg.com/media/CT5m4VGWEAAAtKc8.jpg	1
8279	https://pbs.twimg.com/media/CT5d9DZXAAALcwe.jpg	1
8280	https://pbs.twimg.com/media/CT5cN_3WEAA1OoZ.jpg	1
8281	https://pbs.twimg.com/media/CT5Vg_wXIAAXfnj.jpg	1
8282	https://pbs.twimg.com/media/CT5Qw94XAAA_2dP.jpg	1
8283	https://pbs.twimg.com/media/CT5PY90W0AAQGLo.jpg	1
8284	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg	1
8285	https://pbs.twimg.com/media/CT5KoJ1WoAAJash.jpg	1
8286	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg	1
8287	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1
8288	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg	1
8289	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1
8290	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1
8291	https://pbs.twimg.com/media/CT4udnOWwAA0aMy.jpg	1

	p1	p1_conf	...	p2_conf	p2_dog	\
0	orange	0.097049	...	0.085851	False	
1	Chihuahua	0.323581	...	0.090647	True	
2	Chihuahua	0.716012	...	0.078253	True	
3	paper_towel	0.170278	...	0.168086	True	
4	basset	0.555712	...	0.225770	True	
5	Chesapeake_Bay_retriever	0.425595	...	0.116317	True	
6	Appenzeller	0.341703	...	0.199287	True	
7	Pomeranian	0.566142	...	0.178406	True	
8	Irish_terrier	0.487574	...	0.193054	True	
9	Pembroke	0.511319	...	0.451038	True	
10	Samoyed	0.957979	...	0.013884	True	
11	French_bulldog	0.377417	...	0.151317	True	
12	Pembroke	0.966327	...	0.027356	True	
13	French_bulldog	0.991650	...	0.002129	True	
14	golden_retriever	0.953442	...	0.013834	True	
15	whippet	0.626152	...	0.194742	True	
16	golden_retriever	0.714719	...	0.120184	True	

17	golden_retriever	0.469760	...	0.184172	True
18	Siberian_husky	0.700377	...	0.166511	True
19	French_bulldog	0.995026	...	0.000932	True
20	basset	0.821664	...	0.087582	True
21	limousine	0.130432	...	0.029175	False
22	Pembroke	0.809197	...	0.054950	True
23	Mexican_hairless	0.330741	...	0.275645	False
24	Samoyed	0.733942	...	0.035029	True
25	Chihuahua	0.793469	...	0.143528	True
26	kuvasz	0.309706	...	0.186136	True
27	convertible	0.738995	...	0.139952	False
28	French_bulldog	0.999201	...	0.000361	True
29	pug	0.943575	...	0.025286	False
...
8262	coho	0.404640	...	0.271485	False
8263	black-and-tan_coonhound	0.529139	...	0.244220	True
8264	Chihuahua	0.978108	...	0.009397	True
8265	soft-coated_wheaten_terrier	0.326467	...	0.259551	True
8266	guinea_pig	0.996496	...	0.002402	False
8267	malamute	0.336874	...	0.147655	True
8268	golden_retriever	0.858744	...	0.054787	True
8269	ox	0.416669	...	0.278407	True
8270	three-toed_sloth	0.914671	...	0.015250	False
8271	Maltese_dog	0.857531	...	0.063064	True
8272	Italian_greyhound	0.176053	...	0.111884	True
8273	desktop_computer	0.086502	...	0.085547	False
8274	hen	0.965932	...	0.033919	False
8275	English_setter	0.298617	...	0.149842	True
8276	Lhasa	0.582330	...	0.166192	True
8277	bloodhound	0.195217	...	0.078260	True
8278	pug	0.489814	...	0.404722	True
8279	Walker_hound	0.260857	...	0.175382	True
8280	Gordon_setter	0.503672	...	0.174201	True
8281	golden_retriever	0.775930	...	0.093718	True
8282	miniature_poodle	0.201493	...	0.192305	True
8283	shopping_cart	0.962465	...	0.014594	False
8284	chow	0.692517	...	0.058279	True
8285	box_turtle	0.933012	...	0.045885	False
8286	Bernese_mountain_dog	0.651137	...	0.263788	True
8287	miniature_pinscher	0.560311	...	0.243682	True
8288	Rhodesian_ridgeback	0.408143	...	0.360687	True
8289	German_shepherd	0.596461	...	0.138584	True
8290	redbone	0.506826	...	0.074192	True
8291	Welsh_springer_spaniel	0.465074	...	0.156665	True
	p3	p3_conf	p3_dog	favorite_count	\
0	banana	0.076110	False	39467	
1	papillon	0.068957	True	33819	

2		kelpie	0.031379	True	25461
3		spatula	0.040836	False	42908
4	German_short-haired_pointer		0.175219	True	41048
5		Indian_elephant	0.076902	False	20562
6		ice_lolly	0.193548	False	12041
7		Pembroke	0.076507	True	56848
8	Chesapeake_Bay_retriever		0.118184	True	28226
9		Chihuahua	0.029248	True	32467
10		chow	0.008167	True	31166
11		muzzle	0.082981	False	28268
12		basenji	0.004633	True	38818
13	Staffordshire_bullterrier		0.001498	True	27672
14		redbone	0.007958	True	15359
15		Saluki	0.027351	True	25652
16	Labrador_retriever		0.105506	True	29611
17		English_setter	0.073482	True	26080
18		malamute	0.111411	True	20290
19		bull_mastiff	0.000903	True	22201
20		Weimaraner	0.026236	True	30779
21		shopping_cart	0.026321	False	46959
22		beagle	0.038915	True	69871
23		Weimaraner	0.134203	True	34222
24	Staffordshire_bullterrier		0.029705	True	31061
25		can_opener	0.032253	False	35859
26		Dandie_Dinmont	0.086346	True	12306
27		car_wheel	0.044173	False	22798
28		Boston_bull	0.000076	True	21524
29		Siamese_cat	0.002849	False	28469
...	
8262		gar	0.189945	False	459
8263	flat-coated_retriever		0.173810	True	113
8264		papillon	0.004577	True	172
8265		briard	0.206803	True	194
8266		hamster	0.000461	False	804
8267		Eskimo_dog	0.093412	True	229
8268	Labrador_retriever		0.014241	True	307
8269		groenendael	0.102643	True	204
8270		great_grey_owl	0.013207	False	522
8271	miniature_poodle		0.025581	True	152
8272		basenji	0.111152	True	184
8273		bookcase	0.079480	False	108
8274		partridge	0.000052	False	14765
8275		borzoi	0.133649	True	81
8276	Dandie_Dinmont		0.089688	True	164
8277		malinois	0.075628	True	169
8278	French_bulldog		0.048960	True	121
8279	Ibizan_hound		0.097471	True	335
8280		Pekinese	0.109454	True	154

8281	Labrador_retriever	0.072427	True	496
8282	soft-coated_wheaten_terrier	0.082086	True	115
8283	golden_retriever	0.007959	True	304
8284	fur_coat	0.054449	False	448
8285	terrapin	0.017885	False	1253
8286	Greater_Swiss_Mountain_dog	0.016199	True	136
8287	Doberman	0.154629	True	111
8288	miniature_pinscher	0.222752	True	311
8289	bloodhound	0.116197	True	128
8290	Rhodesian_ridgeback	0.072010	True	132
8291	Shetland_sheepdog	0.061428	True	2535

	retweet_count	stage_of_dog	value	rating
0	8853	doggo	None	1.3
1	6514	doggo	None	1.3
2	4328	doggo	None	1.2
3	8964	doggo	None	1.3
4	9774	doggo	None	1.2
5	3261	doggo	None	1.3
6	2158	doggo	None	1.3
7	16716	doggo	None	1.3
8	4429	doggo	None	1.3
9	7711	doggo	doggo	1.4
10	7624	doggo	None	1.3
11	5156	doggo	None	1.3
12	8538	doggo	None	1.3
13	4735	doggo	None	1.2
14	2321	doggo	None	1.3
15	5637	doggo	None	1.3
16	4709	doggo	None	1.2
17	4559	doggo	None	1.3
18	3732	doggo	None	1.3
19	3653	doggo	None	1.2
20	5609	doggo	None	1.3
21	12082	doggo	None	1.4
22	18781	doggo	None	1.3
23	10737	doggo	None	1.3
24	6167	doggo	None	1.2
25	8084	doggo	None	1.3
26	3443	doggo	None	1.3
27	4610	doggo	None	1.3
28	3316	doggo	None	1.2
29	6523	doggo	None	1.3
...
8262	339	puppo	None	0.2
8263	44	puppo	None	0.7
8264	92	puppo	None	0.9
8265	100	puppo	None	1.1

8266	595	puppo	None	0.6
8267	77	puppo	None	0.8
8268	146	puppo	None	1.0
8269	96	puppo	None	0.9
8270	368	puppo	None	0.3
8271	71	puppo	None	0.5
8272	82	puppo	None	1.1
8273	37	puppo	None	1.0
8274	6871	puppo	None	0.1
8275	16	puppo	None	1.1
8276	73	puppo	None	0.8
8277	79	puppo	None	0.9
8278	47	puppo	None	0.6
8279	174	puppo	None	1.0
8280	67	puppo	None	0.9
8281	232	puppo	None	1.0
8282	61	puppo	None	0.8
8283	146	puppo	None	0.9
8284	261	puppo	None	1.0
8285	879	puppo	None	0.2
8286	60	puppo	None	1.0
8287	41	puppo	None	0.5
8288	147	puppo	None	0.6
8289	47	puppo	None	0.9
8290	48	puppo	None	0.7
8291	532	puppo	None	0.8

[8292 rows x 22 columns]

HTML tags in source column

Define Remove HTML tags from the begin and the end of the data using strip method

Code

```
In [40]: general_clean.source = general_clean.source.str.strip('< ="/.abcdefghijklmnopqrstuvwxyz')
```

```
In [41]: general_clean.source =general_clean.source.str.strip('</a>')
```

Test

```
In [42]: general_clean.head()
```

```
Out[42]:
```

	tweet_id	timestamp	source \
0	892420643555336193	2017-08-01 16:23:56	Twitter for iPhone
1	892177421306343426	2017-08-01 00:17:27	Twitter for iPhone

```

2 891815181378084864 2017-07-31 00:18:03 Twitter for iPhone
3 891689557279858688 2017-07-30 15:58:51 Twitter for iPhone
4 891327558926688256 2017-07-29 16:00:24 Twitter for iPhone

```

```

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

```

```

                                expanded_urls      name \
0 https://twitter.com/dog_rates/status/892420643... Phineas
1 https://twitter.com/dog_rates/status/892177421... Tilly
2 https://twitter.com/dog_rates/status/891815181... Archie
3 https://twitter.com/dog_rates/status/891689557... Darla
4 https://twitter.com/dog_rates/status/891327558... Franklin

```

```

                                jpg_url  img_num      p1 \
0 https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg      1      orange
1 https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg      1      Chihuahua
2 https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg      1      Chihuahua
3 https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg      1  paper_towel
4 https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.jpg      2      basset

```

```

      p1_conf  ...      p2_conf p2_dog      p3      p3_conf \
0 0.097049 ...      0.085851 False      banana 0.076110
1 0.323581 ...      0.090647 True      papillon 0.068957
2 0.716012 ...      0.078253 True      kelpie 0.031379
3 0.170278 ...      0.168086 True      spatula 0.040836
4 0.555712 ...      0.225770 True German_short-haired_pointer 0.175219

```

```

      p3_dog  favorite_count  retweet_count  stage_of_dog  value rating
0 False      39467      8853      doggo      None      1.3
1 True      33819      6514      doggo      None      1.3
2 True      25461      4328      doggo      None      1.2
3 False      42908      8964      doggo      None      1.3
4 True      41048      9774      doggo      None      1.2

```

```
[5 rows x 22 columns]
```

```
In [43]: general_clean.source.value_counts()
```

```

Out[43]: Twitter for iPhone      8128
Twitter Web Client      120
TweetDeck      44
Name: source, dtype: int64

```

Invalid names in name column

Define Change the invalid names to the None

Code

```
In [44]: def name(n):  
         return general_clean.name == n  
  
In [45]: general_clean.loc[(general_clean['name'].str.len() < 2) |  
                           (name('an')) |  
                           (name('by')) |  
                           (name('my')) |  
                           (name('Al')) |  
                           (name('JD')) |  
                           (name('Mo')) |  
                           (name('the')), 'name'] = "None"
```

Test

```
In [46]: general_clean.name.value_counts()
```

```
Out[46]: None          2608  
Charlie           44  
Tucker            40  
Oliver            40  
Penny             40  
Lucy              40  
Cooper            40  
Lola               32  
Bo                32  
Winston           32  
Sadie             32  
Toby              28  
Daisy             28  
Koda              24  
Dave              24  
Bella             24  
Milo              24  
Rusty             24  
Stanley           24  
Scout             24  
Jax               24  
Bailey            24  
Louis             20  
Oscar             20  
Buddy             20  
Chester           20
```


Alfie	20
Leo	20
Larry	20
Sunny	16
...	
Darby	4
Rupert	4
Edmund	4
Trip	4
Josep	4
Maks	4
Tupawc	4
Rey	4
Richie	4
Severus	4
Bungalo	4
Evy	4
Mauve	4
Jaycob	4
Boots	4
Maya	4
Buckley	4
Jackie	4
Eugene	4
Shikha	4
Cedrick	4
Ridley	4
Keet	4
Rinna	4
Julius	4
Spencer	4
Mary	4
Edd	4
Pawnd	4
Jarod	4

Name: name, Length: 927, dtype: int64

None instead of Null in some columns

Define Change None to the NaN

Code

```
In [47]: general_clean.loc[(general_clean.value == 'None'), 'value'] = 'NaN'
```

Test

In [48]: general_clean.value

```
Out[48]: 0      NaN
          1      NaN
          2      NaN
          3      NaN
          4      NaN
          5      NaN
          6      NaN
          7      NaN
          8      NaN
          9      doggo
         10      NaN
         11      NaN
         12      NaN
         13      NaN
         14      NaN
         15      NaN
         16      NaN
         17      NaN
         18      NaN
         19      NaN
         20      NaN
         21      NaN
         22      NaN
         23      NaN
         24      NaN
         25      NaN
         26      NaN
         27      NaN
         28      NaN
         29      NaN
          ...
        8262      NaN
        8263      NaN
        8264      NaN
        8265      NaN
        8266      NaN
        8267      NaN
        8268      NaN
        8269      NaN
        8270      NaN
        8271      NaN
        8272      NaN
        8273      NaN
        8274      NaN
        8275      NaN
```

```

8276      NaN
8277      NaN
8278      NaN
8279      NaN
8280      NaN
8281      NaN
8282      NaN
8283      NaN
8284      NaN
8285      NaN
8286      NaN
8287      NaN
8288      NaN
8289      NaN
8290      NaN
8291      NaN
Name: value, Length: 8292, dtype: object

```

1.3.4 Tidiness II

Three columns of prediction, prediction confidence and prediction dog

Define Create new lists: prediction and confidence. If p1_dog, p2_dog or p3_dog(or all) is True then append data from p1, p1_conf, p2, p2_conf, p3, p3_conf columns to the lists that we created. After add this lists to our DataFrame.

Code

```
In [49]: general_clean.iloc[:, 8:17].head(4)
```

```

Out[49]:
      p1  p1_conf  p1_dog      p2  p2_conf  p2_dog  \
0  orange  0.097049  False    bagel  0.085851  False
1  Chihuahua  0.323581   True    Pekinese  0.090647   True
2  Chihuahua  0.716012   True    malamute  0.078253   True
3  paper_towel  0.170278  False  Labrador_retriever  0.168086   True

      p3  p3_conf  p3_dog
0  banana  0.076110  False
1  papillon  0.068957   True
2  kelpie  0.031379   True
3  spatula  0.040836  False

```

```
In [50]: prediction = []
         confidence = []
```

```

for i in range(len(general_clean.p1_dog)):
    if general_clean.p1_dog[i] == True:
        prediction.append(general_clean.p1[i])
        confidence.append(general_clean.p1_conf[i])
    elif general_clean.p2_dog[i] == True:
        prediction.append(general_clean.p2[i])
        confidence.append(general_clean.p2_conf[i])
    elif general_clean.p3_dog[i] == True:
        prediction.append(general_clean.p3[i])
        confidence.append(general_clean.p3_conf[i])
    else:
        prediction.append('unknown')
        confidence.append(0.0)

```

```

general_clean['prediction'] = prediction
general_clean['confidence'] = confidence

```

```

In [51]: general_clean.drop(['p1', 'p1_conf', 'p1_dog', 'p2', 'p2_conf', 'p2_dog', 'p3', 'p3_conf',
                             'p3_dog'], axis=1, inplace=True)

```

Test

```

In [52]: general_clean

```

```

Out [52]:
      tweet_id      timestamp      source \
0  892420643555336193  2017-08-01 16:23:56  Twitter for iPhone
1  892177421306343426  2017-08-01 00:17:27  Twitter for iPhone
2  891815181378084864  2017-07-31 00:18:03  Twitter for iPhone
3  891689557279858688  2017-07-30 15:58:51  Twitter for iPhone
4  891327558926688256  2017-07-29 16:00:24  Twitter for iPhone
5  891087950875897856  2017-07-29 00:08:17  Twitter for iPhone
6  890971913173991426  2017-07-28 16:27:12  Twitter for iPhone
7  890729181411237888  2017-07-28 00:22:40  Twitter for iPhone
8  890609185150312448  2017-07-27 16:25:51  Twitter for iPhone
9  890240255349198849  2017-07-26 15:59:51  Twitter for iPhone
10 890006608113172480  2017-07-26 00:31:25  Twitter for iPhone
11 889880896479866881  2017-07-25 16:11:53  Twitter for iPhone
12 889665388333682689  2017-07-25 01:55:32  Twitter for iPhone
13 889638837579907072  2017-07-25 00:10:02  Twitter for iPhone
14 889531135344209921  2017-07-24 17:02:04  Twitter for iPhone
15 889278841981685760  2017-07-24 00:19:32  Twitter for iPhone
16 888917238123831296  2017-07-23 00:22:39  Twitter for iPhone
17 888804989199671297  2017-07-22 16:56:37  Twitter for iPhone
18 888554962724278272  2017-07-22 00:23:06  Twitter for iPhone
19 888078434458587136  2017-07-20 16:49:33  Twitter for iPhone
20 887705289381826560  2017-07-19 16:06:48  Twitter for iPhone
21 887517139158093824  2017-07-19 03:39:09  Twitter for iPhone
22 887473957103951883  2017-07-19 00:47:34  Twitter for iPhone

```

23	887343217045368832	2017-07-18	16:08:03	Twitter for iPhone
24	887101392804085760	2017-07-18	00:07:08	Twitter for iPhone
25	886983233522544640	2017-07-17	16:17:36	Twitter for iPhone
26	886736880519319552	2017-07-16	23:58:41	Twitter for iPhone
27	886680336477933568	2017-07-16	20:14:00	Twitter for iPhone
28	886366144734445568	2017-07-15	23:25:31	Twitter for iPhone
29	886258384151887873	2017-07-15	16:17:19	Twitter for iPhone
...
8262	666411507551481857	2015-11-17	00:24:19	Twitter for iPhone
8263	666407126856765440	2015-11-17	00:06:54	Twitter for iPhone
8264	666396247373291520	2015-11-16	23:23:41	Twitter for iPhone
8265	666373753744588802	2015-11-16	21:54:18	Twitter for iPhone
8266	666362758909284353	2015-11-16	21:10:36	Twitter for iPhone
8267	666353288456101888	2015-11-16	20:32:58	Twitter for iPhone
8268	666345417576210432	2015-11-16	20:01:42	Twitter for iPhone
8269	666337882303524864	2015-11-16	19:31:45	Twitter for iPhone
8270	666293911632134144	2015-11-16	16:37:02	Twitter for iPhone
8271	666287406224695296	2015-11-16	16:11:11	Twitter for iPhone
8272	666273097616637952	2015-11-16	15:14:19	Twitter for iPhone
8273	666268910803644416	2015-11-16	14:57:41	Twitter for iPhone
8274	666104133288665088	2015-11-16	04:02:55	Twitter for iPhone
8275	666102155909144576	2015-11-16	03:55:04	Twitter for iPhone
8276	666099513787052032	2015-11-16	03:44:34	Twitter for iPhone
8277	666094000022159362	2015-11-16	03:22:39	Twitter for iPhone
8278	666082916733198337	2015-11-16	02:38:37	Twitter for iPhone
8279	666073100786774016	2015-11-16	01:59:36	Twitter for iPhone
8280	666071193221509120	2015-11-16	01:52:02	Twitter for iPhone
8281	666063827256086533	2015-11-16	01:22:45	Twitter for iPhone
8282	666058600524156928	2015-11-16	01:01:59	Twitter for iPhone
8283	666057090499244032	2015-11-16	00:55:59	Twitter for iPhone
8284	666055525042405380	2015-11-16	00:49:46	Twitter for iPhone
8285	666051853826850816	2015-11-16	00:35:11	Twitter for iPhone
8286	666050758794694657	2015-11-16	00:30:50	Twitter for iPhone
8287	666049248165822465	2015-11-16	00:24:50	Twitter for iPhone
8288	666044226329800704	2015-11-16	00:04:52	Twitter for iPhone
8289	666033412701032449	2015-11-15	23:21:54	Twitter for iPhone
8290	666029285002620928	2015-11-15	23:05:30	Twitter for iPhone
8291	666020888022790149	2015-11-15	22:32:08	Twitter for iPhone

text \

0	This is Phineas. He's a mystical boy. Only eve...
1	This is Tilly. She's just checking pup on you...
2	This is Archie. He is a rare Norwegian Pouncin...
3	This is Darla. She commenced a snooze mid meal...
4	This is Franklin. He would like you to stop ca...
5	Here we have a majestic great white breaching ...
6	Meet Jax. He enjoys ice cream so much he gets ...
7	When you watch your owner call another dog a g...

8 This is Zoey. She doesn't want to be one of th...
 9 This is Cassie. She is a college pup. Studying...
 10 This is Koda. He is a South Australian decksha...
 11 This is Bruno. He is a service shark. Only get...
 12 Here's a puppo that seems to be on the fence a...
 13 This is Ted. He does his best. Sometimes that'...
 14 This is Stuart. He's sporting his favorite fan...
 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 ...
 19 This is Gerald. He was just told he didn't get...
 20 This is Jeffrey. He has a monopoly on the pool...
 21 I've yet to rate a Venezuelan Hover Wiener. Th...
 22 This is Canela. She attempted some fancy porch...
 23 You may not have known you needed to see this ...
 24 This... is a Jubilant Antarctic House Bear. We...
 25 This is Maya. She's very shy. Rarely leaves he...
 26 This is Mingus. He's a wonderful father to his...
 27 This is Derek. He's late for a dog meeting. 13...
 28 This is Roscoe. Another pupper fallen victim t...
 29 This is Waffles. His doggles are pupside down...
 ...
 8262 This is quite the dog. Gets really excited whe...
 8263 This is a southern Vesuvius bumblegruff. Can d...
 8264 Oh goodness. A super rare northeast Qdoba kang...
 8265 Those are sunglasses and a jean jacket. 11/10 ...
 8266 Unique dog here. Very small. Lives in containe...
 8267 Here we have a mixed Asiago from the Galápagos...
 8268 Look at this jokester thinking seat belt laws ...
 8269 This is an extremely rare horned Parthenon. No...
 8270 This is a funny dog. Weird toes. Won't come do...
 8271 This is an Albanian 3 1/2 legged Episcopalian...
 8272 Can take selfies 11/10 <https://t.co/ws2AMaWpW>
 8273 Very concerned about fellow dog trapped in com...
 8274 Not familiar with this breed. No tail (weird)...
 8275 Oh my. Here you are seeing an Adobe Setter giv...
 8276 Can stand on stump for what seems like a while...
 8277 This appears to be a Mongolian Presbyterian mi...
 8278 Here we have a well-established sunblockerspan...
 8279 Let's hope this flight isn't Malaysian (lol). ...
 8280 Here we have a northern speckled Rhododendron...
 8281 This is the happiest dog you will ever see. Ve...
 8282 Here is the Rand Paul of retrievers folks! He'...
 8283 My oh my. This is a rare blond Canadian terrie...
 8284 Here is a Siberian heavily armored polar bear ...
 8285 This is an odd dog. Hard on the outside but lo...
 8286 This is a truly beautiful English Wilson Staff...

8287 Here we have a 1949 1st generation vulpix. Enj...
 8288 This is a purebred Piers Morgan. Loves to Netf...
 8289 Here is a very happy pup. Big fan of well-main...
 8290 This is a western brown Mitsubishi terrier. Up...
 8291 Here we have a Japanese Irish Setter. Lost eye...

	expanded_urls	name \
0	https://twitter.com/dog_rates/status/892420643...	Phineas
1	https://twitter.com/dog_rates/status/892177421...	Tilly
2	https://twitter.com/dog_rates/status/891815181...	Archie
3	https://twitter.com/dog_rates/status/891689557...	Darla
4	https://twitter.com/dog_rates/status/891327558...	Franklin
5	https://twitter.com/dog_rates/status/891087950...	None
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	Jax
7	https://twitter.com/dog_rates/status/890729181...	None
8	https://twitter.com/dog_rates/status/890609185...	Zoey
9	https://twitter.com/dog_rates/status/890240255...	Cassie
10	https://twitter.com/dog_rates/status/890006608...	Koda
11	https://twitter.com/dog_rates/status/889880896...	Bruno
12	https://twitter.com/dog_rates/status/889665388...	None
13	https://twitter.com/dog_rates/status/889638837...	Ted
14	https://twitter.com/dog_rates/status/889531135...	Stuart
15	https://twitter.com/dog_rates/status/889278841...	Oliver
16	https://twitter.com/dog_rates/status/888917238...	Jim
17	https://twitter.com/dog_rates/status/888804989...	Zeke
18	https://twitter.com/dog_rates/status/888554962...	Ralphus
19	https://twitter.com/dog_rates/status/888078434...	Gerald
20	https://twitter.com/dog_rates/status/887705289...	Jeffrey
21	https://twitter.com/dog_rates/status/887517139...	such
22	https://twitter.com/dog_rates/status/887473957...	Canela
23	https://twitter.com/dog_rates/status/887343217...	None
24	https://twitter.com/dog_rates/status/887101392...	None
25	https://twitter.com/dog_rates/status/886983233...	Maya
26	https://www.gofundme.com/mingusneedsus,https://...	Mingus
27	https://twitter.com/dog_rates/status/886680336...	Derek
28	https://twitter.com/dog_rates/status/886366144...	Roscoe
29	https://twitter.com/dog_rates/status/886258384...	Waffles
...
8262	https://twitter.com/dog_rates/status/666411507...	quite
8263	https://twitter.com/dog_rates/status/666407126...	None
8264	https://twitter.com/dog_rates/status/666396247...	None
8265	https://twitter.com/dog_rates/status/666373753...	None
8266	https://twitter.com/dog_rates/status/666362758...	None
8267	https://twitter.com/dog_rates/status/666353288...	None
8268	https://twitter.com/dog_rates/status/666345417...	None
8269	https://twitter.com/dog_rates/status/666337882...	None
8270	https://twitter.com/dog_rates/status/666293911...	None
8271	https://twitter.com/dog_rates/status/666287406...	None

8272	https://twitter.com/dog_rates/status/666273097...	None
8273	https://twitter.com/dog_rates/status/666268910...	None
8274	https://twitter.com/dog_rates/status/666104133...	None
8275	https://twitter.com/dog_rates/status/666102155...	None
8276	https://twitter.com/dog_rates/status/666099513...	None
8277	https://twitter.com/dog_rates/status/666094000...	None
8278	https://twitter.com/dog_rates/status/666082916...	None
8279	https://twitter.com/dog_rates/status/666073100...	None
8280	https://twitter.com/dog_rates/status/666071193...	None
8281	https://twitter.com/dog_rates/status/666063827...	None
8282	https://twitter.com/dog_rates/status/666058600...	None
8283	https://twitter.com/dog_rates/status/666057090...	None
8284	https://twitter.com/dog_rates/status/666055525...	None
8285	https://twitter.com/dog_rates/status/666051853...	None
8286	https://twitter.com/dog_rates/status/666050758...	None
8287	https://twitter.com/dog_rates/status/666049248...	None
8288	https://twitter.com/dog_rates/status/666044226...	None
8289	https://twitter.com/dog_rates/status/666033412...	None
8290	https://twitter.com/dog_rates/status/666029285...	None
8291	https://twitter.com/dog_rates/status/666020888...	None

	jpg_url	img_num \
0	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg	1
1	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg	1
2	https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg	1
3	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg	1
4	https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.jpg	2
5	https://pbs.twimg.com/media/DF3HwyEWsAABqE6.jpg	1
6	https://pbs.twimg.com/media/DF1eOmZXUAAALUcq.jpg	1
7	https://pbs.twimg.com/media/DFyBahAVwAAhUTd.jpg	2
8	https://pbs.twimg.com/media/DFwUU_XcAEpyXI.jpg	1
9	https://pbs.twimg.com/media/DFrEyVuW0AAO3t9.jpg	1
10	https://pbs.twimg.com/media/DFnwSY4WAAAMliS.jpg	1
11	https://pbs.twimg.com/media/DF199B1WsAITKsg.jpg	1
12	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg	1
13	https://pbs.twimg.com/media/DFihzFfXsAYGDPR.jpg	1
14	https://pbs.twimg.com/media/DFg_2PVW0AEHN3p.jpg	1
15	https://pbs.twimg.com/ext_tw_video_thumb/88927...	1
16	https://pbs.twimg.com/media/DFYRgsOUQAARGh0.jpg	1
17	https://pbs.twimg.com/media/DFWra-3VYAA2piG.jpg	1
18	https://pbs.twimg.com/media/DFTH_0-UQAACu20.jpg	3
19	https://pbs.twimg.com/media/DFMwn56WsAAkA7B.jpg	1
20	https://pbs.twimg.com/media/DFHDQBbXgAEqY7t.jpg	1
21	https://pbs.twimg.com/ext_tw_video_thumb/88751...	1
22	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg	2
23	https://pbs.twimg.com/ext_tw_video_thumb/88734...	1
24	https://pbs.twimg.com/media/DE-eAq6UwAA-jaE.jpg	1
25	https://pbs.twimg.com/media/DE8yicJW0AAAavBJ.jpg	2

26	https://pbs.twimg.com/media/DE5Se8FXcAAJFx4.jpg	1
27	https://pbs.twimg.com/media/DE4fEDzWAAyHMM.jpg	1
28	https://pbs.twimg.com/media/DE0BTnQUwAApKEH.jpg	1
29	https://pbs.twimg.com/media/DEyftG4UMAE4aE9.jpg	1
...
8262	https://pbs.twimg.com/media/CT-RugiWIAELEaq.jpg	1
8263	https://pbs.twimg.com/media/CT-NvwmW4AAugGZ.jpg	1
8264	https://pbs.twimg.com/media/CT-D2ZHWIAA3gK1.jpg	1
8265	https://pbs.twimg.com/media/CT9vZEYUWAA1Z05.jpg	1
8266	https://pbs.twimg.com/media/CT9lXGsUcAAyUft.jpg	1
8267	https://pbs.twimg.com/media/CT9cx0tUEAAhNN_.jpg	1
8268	https://pbs.twimg.com/media/CT9Vn7PW0AA_ZCM.jpg	1
8269	https://pbs.twimg.com/media/CT90wFIWEAMuRje.jpg	1
8270	https://pbs.twimg.com/media/CT8mx7KW4AEQu8N.jpg	1
8271	https://pbs.twimg.com/media/CT8g3BpUEAAuFjg.jpg	1
8272	https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg	1
8273	https://pbs.twimg.com/media/CT8QCd1WEAADXws.jpg	1
8274	https://pbs.twimg.com/media/CT56LSZW0AALJj2.jpg	1
8275	https://pbs.twimg.com/media/CT54YGiWUAEZnoK.jpg	1
8276	https://pbs.twimg.com/media/CT51-JJUEAA6hV8.jpg	1
8277	https://pbs.twimg.com/media/CT5w9gUW4AAsBNN.jpg	1
8278	https://pbs.twimg.com/media/CT5m4VGWEAAAtKc8.jpg	1
8279	https://pbs.twimg.com/media/CT5d9DZXAAALcwe.jpg	1
8280	https://pbs.twimg.com/media/CT5cN_3WEAA10oZ.jpg	1
8281	https://pbs.twimg.com/media/CT5Vg_wXIAAXfnj.jpg	1
8282	https://pbs.twimg.com/media/CT5Qw94XAAA_2dP.jpg	1
8283	https://pbs.twimg.com/media/CT5PY90W0AAQGLo.jpg	1
8284	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg	1
8285	https://pbs.twimg.com/media/CT5K0J1W0AAJash.jpg	1
8286	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg	1
8287	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1
8288	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg	1
8289	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1
8290	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1
8291	https://pbs.twimg.com/media/CT4udn0WwAAOaMy.jpg	1

	favorite_count	retweet_count	stage_of_dog	value	rating	\
0	39467	8853	doggo	NaN	1.3	
1	33819	6514	doggo	NaN	1.3	
2	25461	4328	doggo	NaN	1.2	
3	42908	8964	doggo	NaN	1.3	
4	41048	9774	doggo	NaN	1.2	
5	20562	3261	doggo	NaN	1.3	
6	12041	2158	doggo	NaN	1.3	
7	56848	16716	doggo	NaN	1.3	
8	28226	4429	doggo	NaN	1.3	
9	32467	7711	doggo	doggo	1.4	
10	31166	7624	doggo	NaN	1.3	

11	28268	5156	doggo	NaN	1.3
12	38818	8538	doggo	NaN	1.3
13	27672	4735	doggo	NaN	1.2
14	15359	2321	doggo	NaN	1.3
15	25652	5637	doggo	NaN	1.3
16	29611	4709	doggo	NaN	1.2
17	26080	4559	doggo	NaN	1.3
18	20290	3732	doggo	NaN	1.3
19	22201	3653	doggo	NaN	1.2
20	30779	5609	doggo	NaN	1.3
21	46959	12082	doggo	NaN	1.4
22	69871	18781	doggo	NaN	1.3
23	34222	10737	doggo	NaN	1.3
24	31061	6167	doggo	NaN	1.2
25	35859	8084	doggo	NaN	1.3
26	12306	3443	doggo	NaN	1.3
27	22798	4610	doggo	NaN	1.3
28	21524	3316	doggo	NaN	1.2
29	28469	6523	doggo	NaN	1.3
...
8262	459	339	puppo	NaN	0.2
8263	113	44	puppo	NaN	0.7
8264	172	92	puppo	NaN	0.9
8265	194	100	puppo	NaN	1.1
8266	804	595	puppo	NaN	0.6
8267	229	77	puppo	NaN	0.8
8268	307	146	puppo	NaN	1.0
8269	204	96	puppo	NaN	0.9
8270	522	368	puppo	NaN	0.3
8271	152	71	puppo	NaN	0.5
8272	184	82	puppo	NaN	1.1
8273	108	37	puppo	NaN	1.0
8274	14765	6871	puppo	NaN	0.1
8275	81	16	puppo	NaN	1.1
8276	164	73	puppo	NaN	0.8
8277	169	79	puppo	NaN	0.9
8278	121	47	puppo	NaN	0.6
8279	335	174	puppo	NaN	1.0
8280	154	67	puppo	NaN	0.9
8281	496	232	puppo	NaN	1.0
8282	115	61	puppo	NaN	0.8
8283	304	146	puppo	NaN	0.9
8284	448	261	puppo	NaN	1.0
8285	1253	879	puppo	NaN	0.2
8286	136	60	puppo	NaN	1.0
8287	111	41	puppo	NaN	0.5
8288	311	147	puppo	NaN	0.6
8289	128	47	puppo	NaN	0.9

8290	132	48	puppo	NaN	0.7
8291	2535	532	puppo	NaN	0.8

		prediction	confidence
0		unknown	0.000000
1		Chihuahua	0.323581
2		Chihuahua	0.716012
3		Labrador_retriever	0.168086
4		basset	0.555712
5		Chesapeake_Bay_retriever	0.425595
6		Appenzeller	0.341703
7		Pomeranian	0.566142
8		Irish_terrier	0.487574
9		Pembroke	0.511319
10		Samoyed	0.957979
11		French_bulldog	0.377417
12		Pembroke	0.966327
13		French_bulldog	0.991650
14		golden_retriever	0.953442
15		whippet	0.626152
16		golden_retriever	0.714719
17		golden_retriever	0.469760
18		Siberian_husky	0.700377
19		French_bulldog	0.995026
20		basset	0.821664
21		unknown	0.000000
22		Pembroke	0.809197
23		Mexican_hairless	0.330741
24		Samoyed	0.733942
25		Chihuahua	0.793469
26		kuvasz	0.309706
27		unknown	0.000000
28		French_bulldog	0.999201
29		pug	0.943575
...	
8262		unknown	0.000000
8263		black-and-tan_coonhound	0.529139
8264		Chihuahua	0.978108
8265		soft-coated_wheaten_terrier	0.326467
8266		unknown	0.000000
8267		malamute	0.336874
8268		golden_retriever	0.858744
8269		Newfoundland	0.278407
8270		unknown	0.000000
8271		Maltese_dog	0.857531
8272		Italian_greyhound	0.176053
8273		unknown	0.000000
8274		unknown	0.000000

8275	English_setter	0.298617
8276	Lhasa	0.582330
8277	bloodhound	0.195217
8278	pug	0.489814
8279	Walker_hound	0.260857
8280	Gordon_setter	0.503672
8281	golden_retriever	0.775930
8282	miniature_poodle	0.201493
8283	golden_retriever	0.007959
8284	chow	0.692517
8285	unknown	0.000000
8286	Bernese_mountain_dog	0.651137
8287	miniature_pinscher	0.560311
8288	Rhodesian_ridgeback	0.408143
8289	German_shepherd	0.596461
8290	redbone	0.506826
8291	Welsh_springer_spaniel	0.465074

[8292 rows x 15 columns]

1.3.5 Quality III

Some breeds with a lowercase and with _ sign

Define Replace _ sign with space and change the capital letters to uppercase letter

Code

```
In [53]: general_clean.prediction = general_clean.prediction.str.replace('_', ' ').str.capitalize
```

Test

```
In [54]: general_clean.prediction
```

```
Out[54]: 0          Unknown
          1      Chihuahua
          2      Chihuahua
          3  Labrador retriever
          4          Basset
          5  Chesapeake bay retriever
          6      Appenzeller
          7      Pomeranian
          8      Irish terrier
          9      Pembroke
```

10		Samoyed
11		French bulldog
12		Pembroke
13		French bulldog
14		Golden retriever
15		Whippet
16		Golden retriever
17		Golden retriever
18		Siberian husky
19		French bulldog
20		Basset
21		Unknown
22		Pembroke
23		Mexican hairless
24		Samoyed
25		Chihuahua
26		Kuvasz
27		Unknown
28		French bulldog
29		Pug
	...	
8262		Unknown
8263	Black-and-tan	coonhound
8264		Chihuahua
8265	Soft-coated wheaten	terrier
8266		Unknown
8267		Malamute
8268		Golden retriever
8269		Newfoundland
8270		Unknown
8271		Maltese dog
8272	Italian	greyhound
8273		Unknown
8274		Unknown
8275	English	setter
8276		Lhasa
8277		Bloodhound
8278		Pug
8279		Walker hound
8280		Gordon setter
8281		Golden retriever
8282		Miniature poodle
8283		Golden retriever
8284		Chow
8285		Unknown
8286	Bernese	mountain dog
8287	Miniature	pinscher
8288	Rhodesian	ridgeback

```

8289                German shepherd
8290                Redbone
8291        Welsh springer spaniel
Name: prediction, Length: 8292, dtype: object

```

1.4 Storing a clean DataFrame

```
In [55]: twitter_archive_master = general_clean.copy()
```

```
In [56]: twitter_archive_master.head(2)
```

```

Out [56]:
      tweet_id      timestamp      source \
0  892420643555336193  2017-08-01 16:23:56  Twitter for iPhone
1  892177421306343426  2017-08-01 00:17:27  Twitter for iPhone

      text \
0  This is Phineas. He's a mystical boy. Only eve...
1  This is Tilly. She's just checking pup on you...

      expanded_urls      name \
0  https://twitter.com/dog_rates/status/892420643...  Phineas
1  https://twitter.com/dog_rates/status/892177421...  Tilly

      jpg_url  img_num  favorite_count \
0  https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg      1      39467
1  https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg      1      33819

      retweet_count  stage_of_dog  value  rating  prediction  confidence
0           8853      doggo      NaN    1.3    Unknown    0.000000
1           6514      doggo      NaN    1.3    Chihuahua    0.323581

```

```
In [57]: twitter_archive_master.to_csv('twitter_archive_master.csv', index=False)
```

2 Analyze and Visualization Data.

- *What are the most common names?*
- *Which breed gets the highest rating on average??*
- *What is the most common stage of the dog in a DataFrame?*

To get the appropriate DataFrames to answer the questions above, i used SQLite program. My SQL queries for all questions are given in each section of the relevant question below.

2.0.1 What are the most common names?

```
SELECT count(name) / 4 as num_of_names, name
FROM twitter_archive_master
GROUP BY name
HAVING name NOT LIKE "None"
ORDER BY num_of_names DESC
LIMIT 20
```

```
In [58]: the_most_common_names = pd.read_csv('the_most_common_names.csv')
```

```
In [59]: the_most_common_names
```

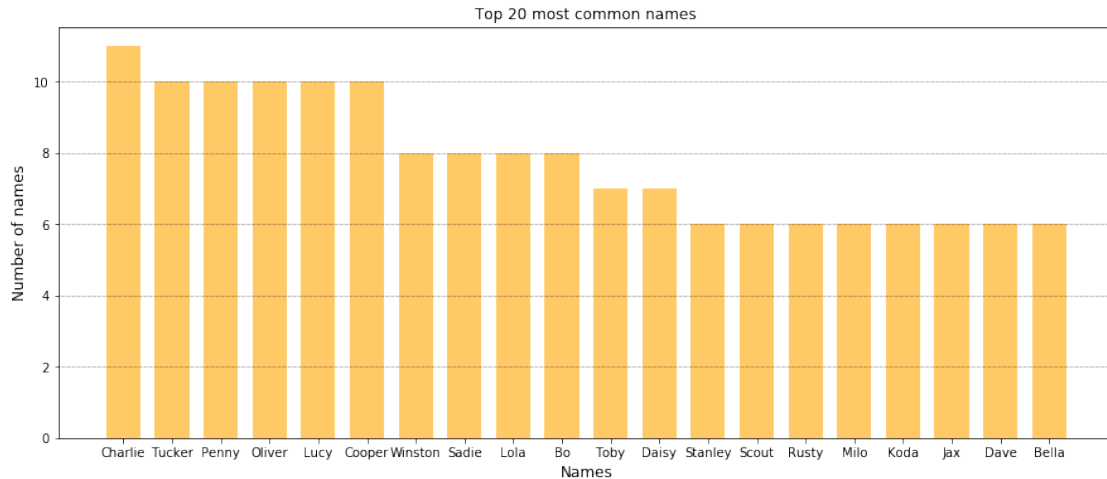
```
Out[59]:
```

	num_of_names	name
0	11	Charlie
1	10	Tucker
2	10	Penny
3	10	Oliver
4	10	Lucy
5	10	Cooper
6	8	Winston
7	8	Sadie
8	8	Lola
9	8	Bo
10	7	Toby
11	7	Daisy
12	6	Stanley
13	6	Scout
14	6	Rusty
15	6	Milo
16	6	Koda
17	6	Jax
18	6	Dave
19	6	Bella

```
In [60]: index = range(len(the_most_common_names.name))
         index
```

```
Out[60]: range(0, 20)
```

```
In [61]: plt.figure(figsize = (15,6))
         plt.bar(the_most_common_names.name.sort_values(), the_most_common_names.num_of_names, w
         plt.xticks(index, the_most_common_names.name)
         plt.xlabel('Names', fontsize = 12)
         plt.ylabel('Number of names', fontsize = 12)
         plt.title('Top 20 most common names')
         plt.grid(axis = 'y', linestyle=':', linewidth=0.5, color='black');
```



2.0.2 Which breed gets the highest rating??

```
SELECT sum(rating) / count(prediction) as avg_rating, prediction, count(prediction) / 4 as number_of_appearances
FROM twitter_archive_master
GROUP BY prediction
HAVING avg_rating NOT LIKE 1.58699690402476 AND number_of_appearances > 1
ORDER BY avg_rating DESC
LIMIT 15
```

I have pulled data in which breeds are found and evaluated more than once.

```
In [62]: avg_rating = pd.read_csv('avg_rating_prediction.csv')
```

```
In [63]: avg_rating
```

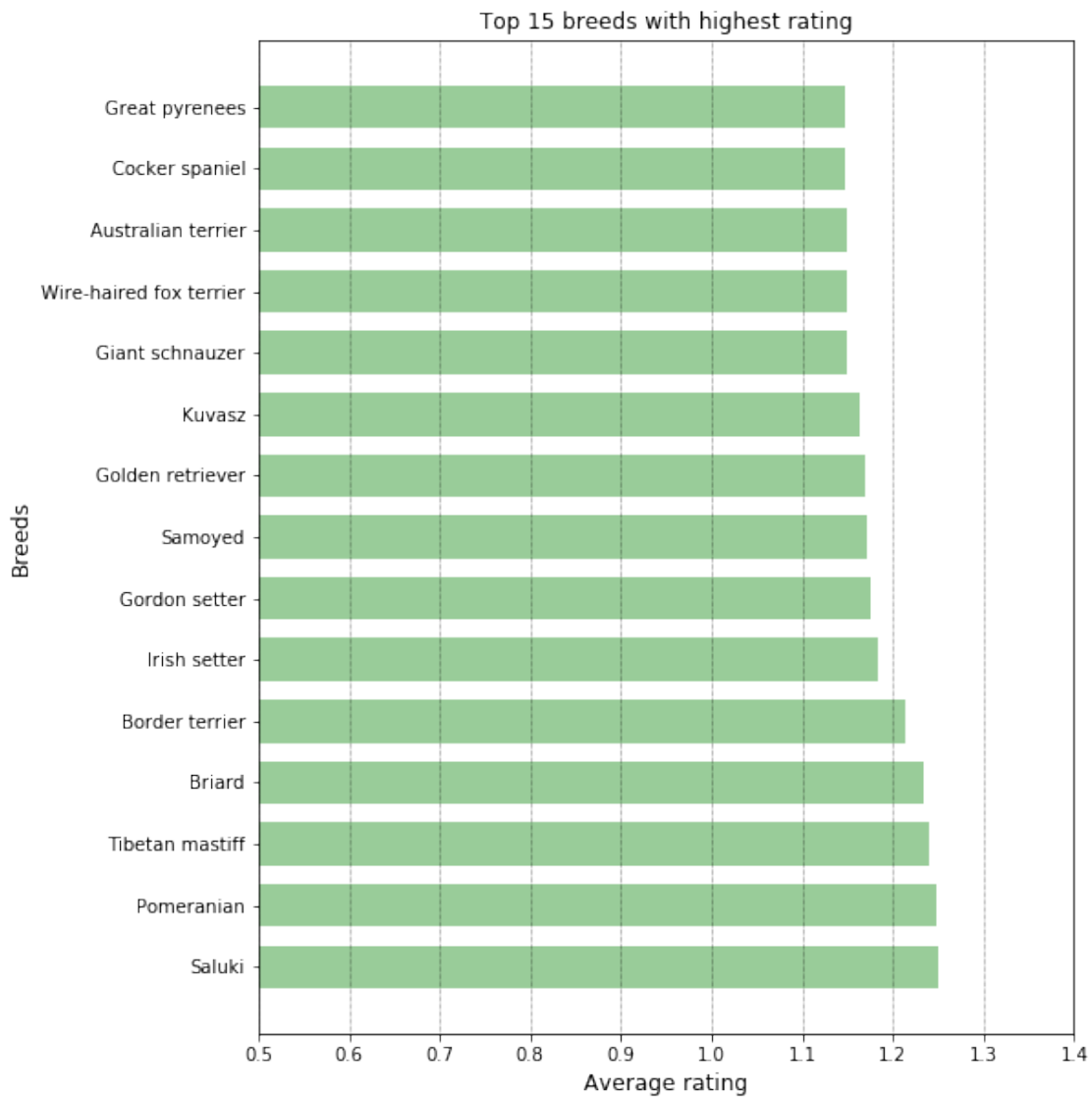
```
Out[63]:
```

	avg_rating	prediction	number_of_appearances
0	1.250000	Saluki	4
1	1.247619	Pomeranian	42
2	1.240000	Tibetan mastiff	5
3	1.233333	Briard	3
4	1.214286	Border terrier	7
5	1.183333	Irish setter	6
6	1.175000	Gordon setter	4
7	1.171739	Samoyed	46
8	1.168478	Golden retriever	173
9	1.163158	Kuvasz	19
10	1.150000	Giant schnauzer	4
11	1.150000	Wire-haired fox terrier	2
12	1.150000	Australian terrier	2
13	1.147059	Cocker spaniel	34
14	1.146667	Great pyrenees	15


```
In [64]: index2 = range(len(avg_rating.prediction))
         index2
```

```
Out[64]: range(0, 15)
```

```
In [65]: plt.figure(figsize = (8, 10))
         plt.barh(avg_rating.prediction.sort_values(), avg_rating.avg_rating, height = 0.7, color='green')
         plt.yticks(index2, avg_rating.prediction)
         plt.xlabel('Average rating', fontsize = 12)
         plt.ylabel('Breeds', fontsize = 12)
         plt.title('Top 15 breeds with highest rating')
         plt.xlim(0.5, 1.4)
         plt.grid(axis = 'x', linestyle=':', linewidth=0.5, color='black');
```



2.0.3 What is the most common stage of the dog in a DataFrame?

```
SELECT count(value) as num_of_appearances, value as stage_of_dogs
FROM twitter_archive_master
GROUP BY stage_of_dogs
HAVING stage_of_dogs NOT LIKE "NaN"
```

```
In [66]: num_app_dog_stages = pd.read_csv('share_of_stages_of_dogs.csv')
```

```
In [67]: num_app_dog_stages
```

```
Out[67]:
```

	num_of_appearances	stage_of_dogs
0	80	doggo
1	8	floofer
2	221	pupper
3	24	puppo

```
In [68]: prop_of_dog_stages = num_app_dog_stages.copy()
```

```
In [69]: prop_of_dog_stages.num_of_appearances = prop_of_dog_stages.num_of_appearances / 333
```

```
In [70]: prop_of_dog_stages.rename(columns = {'num_of_appearances': 'propotion'}, inplace = True)
```

```
In [71]: prop_of_dog_stages
```

```
Out[71]:
```

	propotion	stage_of_dogs
0	0.240240	doggo
1	0.024024	floofer
2	0.663664	pupper
3	0.072072	puppo

```
In [72]: propotions = prop_of_dog_stages.propotion
labels=prop_of_dog_stages.stage_of_dogs
explode = [0.08, 0.08, 0.1, 0.08]
radius = 1.1
colors = ['#ABDED7', '#FFC98B', '#8BBA48', '#DE5B6D']

plt.pie(propotions, labels = labels, explode = explode,
        startangle = 290, autopct = '%1.2f%%', radius = radius,
        colors = colors, shadow = True,
        wedgeprops = {'edgecolor' : 'black'});

plt.title('Percentage of dog stages among all dogs in DataFrame');
plt.tight_layout();
```

Percentage of dog stages among all dogs in DataFrame

