# Project: Wrangle and Analyze Data by Z. McLaughlin

### Gather

```
In [1]: # Note:
        # The api access info has been removed
        # There is a variable switch = "NO" so that that the api query portion
         of the code will not run
        # To run the twitter api query set switch = "YES" and enter the access
         keys.
In [2]: # import statements for all of the packages needed
        # Required libraries
        import pandas as pd
        import numpy as np
        import tweepy
        import json
        import requests
        # Other libraries that will be needed
        import os
        from bs4 import BeautifulSoup
        import matplotlib.pyplot as plt
        import seaborn as sns
        from datetime import datetime
        % matplotlib inline
```

In [3]: # Reading in the csv file provided: twitter-archive-enhanced.csv

df\_tae = pd.read\_csv('./data/twitter-archive-enhanced.csv')
 df\_tae.head()

Out[3]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08- 01 16:23:56 +0000	<a href="http:/ r</a 
1	892177421306343426	NaN	NaN	2017-08- 01 00:17:27 +0000	<a href="http:/ r</a 
2	891815181378084864	NaN	NaN	2017-07- 31 00:18:03 +0000	<a href="http:/ r</a 
3	891689557279858688	NaN	NaN	2017-07- 30 15:58:51 +0000	<a href="http:/ r</a 
4	891327558926688256	NaN	NaN	2017-07- 29 16:00:24 +0000	<a href="http:/ r</a 

```
In [4]: # Downloading the file image_predictions.tsv

url = 'https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad
    _image-predictions/image-predictions.tsv'
    r=requests.get(url)
    open('./data/image-predictions.tsv', 'wb').write(r.content)

# Reading in the tsv file for cleaning

df_tip = pd.read_csv('./data/image-predictions.tsv',sep='\t')
    df_tip.head()
```

#### Out[4]:

	tweet_id	jpg_url	img_num	
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	1	W
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1	re
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1	G
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-IEu.jpg	1	R
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1	m

```
In [6]: # Setup authorization for tweepy - authorization keys will need to be ad
        ded to run the notebook
        if switch == 'YES':
            consumer_key = ''
            consumer_secret = ''
            access_token = ''
            access_secret = ''
            auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
            auth.set_access_token(access_token, access_secret)
            api = tweepy.API(auth)
In [7]: # Downloading json via twitter api to text file: tweet_json.txt
        if switch == "YES":
            tweet_errors = []
            tweets = df_tae.tweet_id
            for tweet in tweets:
                try:
                    # Getting the tweepy json
                    tweepy info = api.get status(tweet, wait on rate limit = True
        , wait on rate limit notify = True, tweet mode='extended')
                    # Saving tweepy json off to tweet_json.txt file
                    tweepy file = open("./data/tweet json.txt", "a")
                    json.dump(tweepy_info._json, tweepy_file)
                    tweepy file.write("\n")
                    tweepy file.close()
                # Saving errors to a file
                except Exception as e:
                    tweet errors.append(str(tweet))
                    print("Error")
                    print("This tweet had a problem: " + str(tweet))
                    error file = open("./data/errors.txt","a")
                    error_file.write(str(tweet) + ': ' + str(e) + '\n')
                    error_file.close()
```

print(e)

```
In [8]: # Reads the contents of tweet_json.txt and extracts the retweet and like
         counts.
        # Creates a new dataframe wiht the info and saves it off to: tweet_extra
        _info.csv
        tweet_id = []
        retweet_count = [] # retweet_count
        like_count = [] # favourites_count
        # Opens the files with json info and reads line by line and extracts inf
        ο.
        with open('./data/tweet_json.txt') as f:
            lines = f.readlines()
        for line in lines:
            line=line.rstrip()
            data=json.loads(line)
            tid= data['id']
            rc = data['retweet_count']
            lc = data['favorite count']
            tweet_id.append(tid)
            retweet_count.append(rc)
            like_count.append(lc)
        # Creating new dataframe
        columns = ['tweet id','retweet count','like count']
        df ti = pd.DataFrame(columns=columns)
        df ti['tweet id'] = tweet id
        df_ti['retweet_count'] = retweet_count
        df_ti['like_count'] = like_count
        # Writing dataframe to csv file
        df ti.to csv('./data/tweet extra info.csv',index=False)
        print('The following file has been saved to disk: tweet_extra_info.csv'
```

The following file has been saved to disk: tweet extra info.csv

```
In [9]: # Reading in the info file for further use

df_info = pd.read_csv('./data/tweet_extra_info.csv')
    df_info.head()

df_info.describe()
```

Out[9]:

	tweet_id	retweet_count	like_count
count	2.344000e+03	2344.000000	2344.000000
mean	7.422890e+17	3010.996587	8037.963311
std	6.835057e+16	5009.803282	12097.967741
min	6.660209e+17	0.000000	0.000000
25%	6.783704e+17	603.750000	1398.500000
50%	7.187854e+17	1403.000000	3524.500000
75%	7.986989e+17	3504.500000	9937.750000
max	8.924206e+17	77042.000000	142895.000000

## **Assess**

The following areas for cleaning were identified:

Project requirement: Assess and clean at least 8 quality issues and at least 2 tidiness issues in this dataset.

#### Quality: Issues with content/dirty data:

- 1. twitter-archive-enhanced.csv: Not all tweets are still valid While querying the api several of the tweet ids came back with an error. Cleaned
- 2. twitter-archive-enhanced.csv: Instead of using null for no content there is "None" under doggo, floofer, pupper,puppo. Cleaned
- 3. twitter-archive-enhanced.csv: Contains retweets and replies when we only want original tweets Cleaned
- 4. twitter-archive-enhanced.csv: Only want tweets that have images (Some have no images or have videos) Cleaned
- 5. twitter-archive-enhanced.csv: Rating denominator min is 0 and max is 170 when should be 10. Cleaned
- 6. twitter-archive-enhanced.csv: Source contains whole HTML string instead of just simple source Cleaned
- 7. twitter-archive-enhanced.csv: timestamp & retweeted\_status\_timestamp are not dates Cleaned (timestamp only)
- 8. twitter-archive-enhanced.csv: Dog names are sometimes showing as the, a, an Cleaned

#### Tidiness: Issues with structure that prevent easy analysis. Messy data.

- · Each variable forms a column.
- · Each observation forms a row.
- Each type of observational unit forms a table.
- 1. twitter-archive-enhanced.csv: Text field includes short links already expanded in another column. Cleaned
- 2. Three tables that could all be in one table.

## Used a combination of exploring the data manually and programmatically.

- Used google docs to view the tables in csv to get general view of the tables.
- While cleaning each item further investigated other areas leading to the list above.
- Assess order and the cleaning steps were different.

In [10]: df\_tae.head()

Out[10]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08- 01 16:23:56 +0000	<a href="http:/ r</a 
1	892177421306343426	NaN	NaN	2017-08- 01 00:17:27 +0000	<a href="http:/ r</a 
2	891815181378084864	NaN	NaN	2017-07- 31 00:18:03 +0000	<a href="http:/ r</a 
3	891689557279858688	NaN	NaN	2017-07- 30 15:58:51 +0000	<a href="http:/ r</a 
4	891327558926688256	NaN	NaN	2017-07- 29 16:00:24 +0000	<a href="http:/ r</a 

## In [11]: df\_tae.info() df\_tae.describe()

2356 non-null object 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 2356 non-null int64 rating numerator rating\_denominator 2356 non-null int64 name 2356 non-null object doggo 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

floofer

#### Out[11]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	retweeted_status_id	retv
count	2.356000e+03	7.800000e+01	7.800000e+01	1.810000e+02	1.81
mean	7.427716e+17	7.455079e+17	2.014171e+16	7.720400e+17	1.24
std	6.856705e+16	7.582492e+16	1.252797e+17	6.236928e+16	9.59
min	6.660209e+17	6.658147e+17	1.185634e+07	6.661041e+17	7.83
25%	6.783989e+17	6.757419e+17	3.086374e+08	7.186315e+17	4.19
50%	7.196279e+17	7.038708e+17	4.196984e+09	7.804657e+17	4.19
75%	7.993373e+17	8.257804e+17	4.196984e+09	8.203146e+17	4.19
max	8.924206e+17	8.862664e+17	8.405479e+17	8.874740e+17	7.87

2356 non-null object

```
In [12]: df_tip.head()
```

#### Out[12]:

	tweet_id	jpg_url	img_num	
C	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	1	W
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1	re
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1	G
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-IEu.jpg	1	R
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1	m

```
In [13]: df_tip.info()
    df_tip.describe()
```

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

#### Out[13]:

	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 [14]: df_info.head()
```

#### Out[14]:

	tweet_id	retweet_count	like_count
0	892420643555336193	8553	38673
1	892177421306343426	6287	33127
2	891815181378084864	4166	24943
3	891689557279858688	8675	42044
4	891327558926688256	9441	40193

```
In [15]: df_info.info()
df_info.describe()
```

#### Out[15]: \_

	tweet_id	retweet_count	like_count
count	2.344000e+03	2344.000000	2344.000000
mean	7.422890e+17	3010.996587	8037.963311
std	6.835057e+16	5009.803282	12097.967741
min	6.660209e+17	0.000000	0.000000
25%	6.783704e+17	603.750000	1398.500000
50%	7.187854e+17	1403.000000	3524.500000
75%	7.986989e+17	3504.500000	9937.750000
max	8.924206e+17	77042.000000	142895.000000

## Clean

```
In [16]: # Make a copy of the data

df_tae_clean = df_tae.copy()
df_tae_clean.head()
```

Out[16]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	timestamp	
0	892420643555336193	NaN	NaN	2017-08- 01 16:23:56 +0000	<a href="http:/ r</a 
1	892177421306343426	NaN	NaN	2017-08- 01 00:17:27 +0000	<a href="http:/ r</a 
2	891815181378084864	NaN	NaN	2017-07- 31 00:18:03 +0000	<a href="http:/ r</a 
3	891689557279858688	NaN	NaN	2017-07- 30 15:58:51 +0000	<a href="http:/ r</a 
4	891327558926688256	NaN	NaN	2017-07- 29 16:00:24 +0000	<a href="http:/ r</a 

**Quality Improvement - Remove replies and retweets** 

```
In [17]: # twitter-archive-enhanced.csv: Contains retweets and replies when we on
ly want original tweets
# Remove rows that have content in 'in_reply_to_status_id'

# Idenfity reply rows and review

reply_info = df_tae_clean[(df_tae_clean['in_reply_to_status_id'].notnull
())]
reply_info.info()

# Cut the rows reply rows that are not necessary

df_tae_clean = df_tae_clean[(df_tae_clean['in_reply_to_status_id'].isnull
l())]
df_tae_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 78 entries, 30 to 2298
Data columns (total 17 columns):
tweet id
                              78 non-null int64
in_reply_to_status_id
                              78 non-null float64
in reply to user id
                              78 non-null float64
timestamp
                              78 non-null object
                              78 non-null object
source
                              78 non-null object
text
retweeted status id
                              0 non-null float64
retweeted status user id
                              0 non-null float64
retweeted status timestamp
                              0 non-null object
expanded urls
                              23 non-null object
                              78 non-null int64
rating numerator
rating denominator
                              78 non-null int64
name
                              78 non-null object
doggo
                              78 non-null object
                              78 non-null object
floofer
                              78 non-null object
pupper
                              78 non-null object
puppo
dtypes: float64(4), int64(3), object(10)
memory usage: 11.0+ KB
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2278 entries, 0 to 2355
Data columns (total 17 columns):
tweet id
                              2278 non-null int64
in reply to status id
                              0 non-null float64
in reply to user id
                              0 non-null float64
                              2278 non-null object
timestamp
source
                              2278 non-null object
text
                              2278 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
                              2274 non-null object
rating numerator
                              2278 non-null int64
rating denominator
                              2278 non-null int64
                              2278 non-null object
name
                              2278 non-null object
doggo
                              2278 non-null object
floofer
                              2278 non-null object
pupper
                              2278 non-null object
puppo
dtypes: float64(4), int64(3), object(10)
memory usage: 320.3+ KB
```

```
In [18]: # Remove rows that have content in 'retweeted_status_id'
    # Idenfity retweet rows and review

retweet_info = df_tae_clean[(df_tae_clean['retweeted_status_id'].notnull
    ())]
    retweet_info.info()

# Cut the rows reply rows that are not necessary

df_tae_clean = df_tae_clean[(df_tae_clean['retweeted_status_id'].isnull
    ())]
    df_tae_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 181 entries, 19 to 2260
Data columns (total 17 columns):
tweet id
                              181 non-null int64
in_reply_to_status_id
                              0 non-null float64
in reply to user id
                              0 non-null float64
timestamp
                              181 non-null object
                              181 non-null object
source
                              181 non-null object
text
retweeted status id
                              181 non-null float64
retweeted status user id
                              181 non-null float64
retweeted status timestamp
                              181 non-null object
expanded urls
                              180 non-null object
                              181 non-null int64
rating numerator
rating denominator
                              181 non-null int64
name
                              181 non-null object
doggo
                              181 non-null object
floofer
                              181 non-null object
                              181 non-null object
pupper
                              181 non-null object
puppo
dtypes: float64(4), int64(3), object(10)
memory usage: 25.5+ KB
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2097 entries, 0 to 2355
Data columns (total 17 columns):
tweet id
                              2097 non-null int64
in reply to status id
                              0 non-null float64
in reply to user id
                              0 non-null float64
                              2097 non-null object
timestamp
source
                              2097 non-null object
text
                              2097 non-null object
retweeted status id
                              0 non-null float64
retweeted status user id
                              0 non-null float64
retweeted status timestamp
                              0 non-null object
                              2094 non-null object
expanded urls
rating numerator
                              2097 non-null int64
rating denominator
                              2097 non-null int64
                              2097 non-null object
name
                              2097 non-null object
doggo
                              2097 non-null object
floofer
                              2097 non-null object
pupper
                              2097 non-null object
puppo
dtypes: float64(4), int64(3), object(10)
memory usage: 294.9+ KB
```

Tidyness Improvement - Remove columns related to replies and retweets

```
In [19]: # Now that we've eliminated the replys and retweets we can delete the co
lumns
# that save that information.

# Columns to be deleted - in_reply_to_status_id, in_reply_to_user_id, re
tweeted_status_id,retweeted_status_user_id, retweeted_status_timestamp

columns = ['in_reply_to_status_id', 'in_reply_to_user_id','retweeted_status_id','retweeted_status_timestamp']

df_tae_clean.drop(columns, inplace=True, axis=1)

df_tae_clean.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2097 entries. 0 to 2355
```

```
Int64Index: 2097 entries, 0 to 2355
Data columns (total 12 columns):
                        2097 non-null int64
tweet id
timestamp
                        2097 non-null object
source
                        2097 non-null object
text
                        2097 non-null object
expanded_urls
                        2094 non-null object
rating_numerator 2097 non-null int64 rating_denominator 2097 non-null int64
name
                        2097 non-null object
                        2097 non-null object
doggo
                        2097 non-null object
floofer
                        2097 non-null object
pupper
                        2097 non-null object
puppo
dtypes: int64(3), object(9)
memory usage: 213.0+ KB
```

Quality Improvement - Removing tweets that were not found using the api

```
In [20]: # Need to remove the tweets that couldn't be found using the api
         # use the error file created during the api step using tweepy
         tweet_errors = pd.read_csv('./data/errors.txt', sep=":", header=None)
         tweet errors.head()
         tweets_to_remove = tweet_errors[0].values
         # de-duplicate
         tweets_to_remove_u = []
         for tweet in tweets_to_remove:
             if tweet not in tweets_to_remove_u:
                 tweets_to_remove_u.append(tweet)
         # Remove any rows from clean data that couldn't be found using the api
         for tweet in tweets_to_remove_u:
              # A lot of the tweets that errored out were replies or retweets
              if df_tae_clean[df_tae_clean['tweet_id']== tweet].empty == False:
                 df_tae_clean = df_tae_clean[df_tae_clean['tweet_id']!= tweet]
         # Test that the tweet in question got removed (turned out there was only
          one.)
         df_tae_clean[df_tae_clean['tweet_id'] == 754011816964026368]
         df_tae_clean.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 2096 entries, 0 to 2355
         Data columns (total 12 columns):
                    2096 non-null int64
2096 non-null object
         tweet id
         timestamp
         source
                             2096 non-null object
                              2096 non-null object
         text
```

**Quality - Change timestamp to date format** 

```
In [21]: # timestamp to datetime
    df_tae_clean.timestamp = pd.to_datetime(df_tae_clean.timestamp)

# test
    df_tae_clean.info()

df_tae_clean.head()

# See another possible problem area - The number of exampnded_urls does
    n't match the number of tweets
    # Investigate

df_tae_clean[(df_tae_clean['expanded_urls'].isnull())]
    # Result: These tweets don't have images so they will be removed via ot
    her cleaning
```

<class 'pandas.core.frame.DataFrame'> Int64Index: 2096 entries, 0 to 2355 Data columns (total 12 columns): tweet\_id 2096 non-null int64 timestamp 2096 non-null datetime64[ns] 2096 non-null object source 2096 non-null object text 2093 non-null object expanded urls 2096 non-null int64 rating\_numerator rating\_denominator 2096 non-null int64 name 2096 non-null object doggo 2096 non-null object 2096 non-null object floofer 2096 non-null object pupper 2096 non-null object puppo dtypes: datetime64[ns](1), int64(3), object(8) memory usage: 212.9+ KB

#### Out[21]:

	tweet_id	timestamp	source	text
375	828361771580813312	2017-02- 05 21:56:51	<a <br="" href="http://twitter.com">rel="nofollow"&gt;Tw</a>	Beebop and Doobert should start a band 12/10 w
707	785515384317313025	2016-10- 10 16:20:36	<a href="http://twitter.com/download/iphone" r</a 	Today, 10/10, should be National Dog Rates Day
1445	696518437233913856	2016-02- 08 02:18:30	<a href="http://twitter.com/download/iphone" r</a 	Oh my god 10/10 for every little hot dog pupper

```
['None' 'doggo']
['None' 'floofer']
['None' 'pupper']
['None' 'puppo']
```

	tweet_id	timestamp	source	te
0	892420643555336193	2017-08- 01 16:23:56	<a href="http://twitter.com/download/iphone" r</a 	This is Phineas. He's a mystical bo Only eve
1	892177421306343426	2017-08- 01 00:17:27	<a href="http://twitter.com/download/iphone" r</a 	This is Tilly. She's just checking pup on you
2	891815181378084864	5181378084864 2017-07- <a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Archie. He i a rare Norwegian Pouncin</td></a>		This is Archie. He i a rare Norwegian Pouncin
3	891689557279858688	2017-07- 30 15:58:51	<a href="http://twitter.com/download/iphone" r</a 	This is Darla She commence a snooze mid meal
4	891327558926688256	2017-07- 29 16:00:24	<a href="http://twitter.com/download/iphone" r</a 	This is Franklin. He would like you to stop ca
5	891087950875897856	2017-07- 29 00:08:17	<a href="http://twitter.com/download/iphone" r</a 	Here we have a majestic great white breaching
6   2017-07-		href="http://twitter.com/download/iphone"	Meet Jax. He enjoys ice cream s much he gets	
7	890729181411237888	2017-07- 28 00:22:40	<a href="http://twitter.com/download/iphone" r</a 	When you watch your owner call another dog a g

	tweet_id	timestamp	source	te
8	890609185150312448	2017-07- 27 16:25:51	<a href="http://twitter.com/download/iphone" r</a 	This is Zoey She doesn's want to be one of th
9	890240255349198849	2017-07- 26 15:59:51	<a href="http://twitter.com/download/iphone" r</a 	This is Cassie. She is a college pup. Studying
10	890006608113172480	2017-07- 26 00:31:25	<a href="http://twitter.com/download/iphone" r</a 	This is Koda He is a South Australian decksha
11	889880896479866881	2017-07- 25 16:11:53	<a href="http://twitter.com/download/iphone" r</a 	This is Bruno. He is a service shark. Only get
12	889665388333682689	2017-07- 25 01:55:32	<a href="http://twitter.com/download/iphone" r</a 	Here's a puppo that seems to be on the fencea
13	889638837579907072	2017-07- 25 00:10:02	<a href="http://twitter.com/download/iphone" r</a 	This is Ted. He does his best. Sometimes that'
14	889531135344209921	2017-07- 24 17:02:04	<a href="http://twitter.com/download/iphone" r</a 	This is Stuart. He's sporting his favorite fan.
15	889278841981685760	2017-07- 24 00:19:32	<a href="http://twitter.com/download/iphone" r</a 	This is Oliver. You'n witnessing one of his m
16	888917238123831296	2017-07- 23 00:22:39	<a href="http://twitter.com/download/iphone" r</a 	This is Jim. He found a fren. Taught him how t

	tweet_id	timestamp	source	te
17	888804989199671297	2017-07- 22 16:56:37	<a href="http://twitter.com/download/iphone" r</a 	This is Zeke He has a new stick. Very proud 0
18	888554962724278272	2017-07- 22 00:23:06	<a href="http://twitter.com/download/iphone" r</a 	This is Ralphus. He's powering up Attempting
20	888078434458587136	2017-07- 20 16:49:33	<a href="http://twitter.com/download/iphone" r</a 	This is Gerald. He was just tol- he didn't get
21	887705289381826560	2017-07- 19 16:06:48	<a href="http://twitter.com/download/iphone" r</a 	This is Jeffrey. He has a monopoly c the pool
22	887517139158093824	2017-07- 19 03:39:09	<a href="http://twitter.com/download/iphone" r</a 	I've yet to rate a Venezuelan Hover Wiener. Th
23	887473957103951883	2017-07- 19 00:47:34	<a href="http://twitter.com/download/iphone" r</a 	This is Canela. She attempted some fancy porch
24	887343217045368832	2017-07- 18 16:08:03	<a href="http://twitter.com/download/iphone" r</a 	You may no have knowr you needed to see this.
25	887101392804085760	2017-07- 18 00:07:08	<a href="http://twitter.com/download/iphone" r</a 	This is a Jubilant Antarctic House Bear We

	tweet_id	timestamp	source	te
26	886983233522544640	2017-07- 17 16:17:36	<a href="http://twitter.com/download/iphone" r</a 	This is May: She's very shy. Rarely leaves he
27	886736880519319552	2017-07- 16 23:58:41	<a href="http://twitter.com/download/iphone" r</a 	This is Mingus. He's a wonderful father to his
28	886680336477933568	2017-07- 16 20:14:00	<a href="http://twitter.com/download/iphone" r</a 	This is Derek. He's late for a dog meeting. 13
29	886366144734445568	2017-07- 15 23:25:31	<a href="http://twitter.com/download/iphone" r</a 	This is Roscoe. Another pupper falle victim t
31	886258384151887873	2017-07- 15 16:17:19	<a href="http://twitter.com/download/iphone" r</a 	This is Waffles. His doggles are pupside down
33	885984800019947520	2017-07- 14 22:10:11	<a href="http://twitter.com/download/iphone" r</a 	Viewer discretion advised. This is Jimbo. He w
34	885528943205470208	2017-07- 13 15:58:47	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Maisey. She fell asleep mid- excavation.</th></a>	This is Maisey. She fell asleep mid- excavation.
35	885518971528720385	2017-07- 13 15:19:09	<a href="http://twitter.com/download/iphone" r</a 	I have a new hero and his name is Howard. 14/1

	tweet_id	timestamp	source	te
37	885167619883638784	2017-07- 12 16:03:00	<a href="http://twitter.com/download/iphone" r</a 	Here we have a corg undercover as a malamute
38	884925521741709313	2017-07- 12 00:01:00	<a href="http://twitter.com/download/iphone" r</a 	This is Earl. He found a hat. Nervou about wh
39	884876753390489601	2017-07- 11 20:47:12	<a href="http://twitter.com/download/iphone" r</a 	This is Lola. It's her first time outside Mus
40	884562892145688576	2017-07- 11 00:00:02	<a href="http://twitter.com/download/iphone" r</a 	This is Kevin. He's just so happy. 13/1 what
41	884441805382717440	2017-07- 10 15:58:53	<a href="http://twitter.com/download/iphone" r</a 	I present to you, Pup in Hat. Pup in Hat is gr
42	884247878851493888	2017-07- 10 03:08:17	<a href="http://twitter.com/download/iphone" r</a 	OMG HE DIDN'T MEAN TO HE WAS JUST TRYING A LIT
43	884162670584377345	2017-07- 09 21:29:42	<a href="http://twitter.com/download/iphone" r</a 	Meet Yogi. He doesn't have any important dog m
44	883838122936631299	2017-07- 09 00:00:04	<a href="http://twitter.com/download/iphone" r</a 	This is Noal He can't believe someone made th

	tweet_id	timestamp	source	te
45	883482846933004288	2017-07- 08 00:28:19	<a href="http://twitter.com/download/iphone" r</a 	This is Bella She hopes her smile made you sm
46	883360690899218434	2017-07- 07 16:22:55	<a href="http://twitter.com/download/iphone" r<="" th=""><th>Meet Grizzwald. He may be the floofiest floofe</th></a>	Meet Grizzwald. He may be the floofiest floofe
47	883117836046086144	2017-07- 07 00:17:54	<a href="http://twitter.com/download/iphone" r</a 	Please only send dogs. We don't rate mechanics.
48	882992080364220416	2017-07- 06 15:58:11	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Rusty. He wasn't read for the first p</th></a>	This is Rusty. He wasn't read for the first p
49	882762694511734784	2017-07- 06 00:46:41	<a href="http://twitter.com/download/iphone" r</a 	This is Gus. He's quite the cheeky pupper. Alr.
50	882627270321602560	2017-07- 05 15:48:34	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Stanley. He has his first swim lessor </th></a>	This is Stanley. He has his first swim lessor 
51	882268110199369728	2017-07- 04 16:01:23	<a href="http://twitter.com/download/iphone" r</a 	This is Alfy. You're witnessing his first wate
52	882045870035918850	2017-07- 04 01:18:17	<a href="http://twitter.com/download/iphone" r</a 	This is Koko Her owner, inspired by Barney, r
53	881906580714921986	2017-07- 03 16:04:48	<a href="http://twitter.com/download/iphone" r</a 	This is Rey. He's a Benebop Cumberfloo 12/10

Quality Improvement - Change obvious non-name items (a, an, the) to Null for name

```
In [23]: # Changing non-names in name column to Null (testing if first chacter is
    lower case)

df_tae_clean['name'] = df_tae_clean['name'].astype(str)
    names = df_tae_clean['name'].unique()
    names_to_change_to_null = []
    for nam in names:
        if nam[0].islower():
            names_to_change_to_null.append(nam)
    print(names_to_change_to_null)

for ncn in names_to_change_to_null:
        df_tae_clean['name'].replace(ncn, np.nan,inplace = True)

print(df_tae_clean['name'].unique())

df_tae_clean.head(100)
```

```
['nan', 'such', 'a', 'quite', 'not', 'one', 'incredibly', 'very', 'my',
'his', 'an', 'actually', 'just', 'getting', 'mad', 'this', 'unacceptabl
e', 'all', 'old', 'infuriating', 'the', 'by', 'officially', 'life', 'li
ght', 'space']
['Phineas' 'Tilly' 'Archie' 'Darla' 'Franklin' nan 'Jax' 'Zoey' 'Cassi
 'Koda' 'Bruno' 'Ted' 'Stuart' 'Oliver' 'Jim' 'Zeke' 'Ralphus' 'Gerald'
 'Jeffrey' 'Canela' 'Maya' 'Mingus' 'Derek' 'Roscoe' 'Waffles' 'Jimbo'
 'Maisey' 'Earl' 'Lola' 'Kevin' 'Yogi' 'Noah' 'Bella' 'Grizzwald' 'Rust
 'Gus' 'Stanley' 'Alfy' 'Koko' 'Rey' 'Gary' 'Elliot' 'Louis' 'Jesse'
 'Romeo' 'Bailey' 'Duddles' 'Jack' 'Steven' 'Beau' 'Snoopy' 'Shadow' 'E
 'Aja' 'Penny' 'Dante' 'Nelly' 'Ginger' 'Benedict' 'Venti' 'Goose' 'Nug
 'Cash' 'Jed' 'Sebastian' 'Sierra' 'Monkey' 'Harry' 'Kody' 'Lassie' 'Ro
 'Napolean' 'Boomer' 'Cody' 'Rumble' 'Clifford' 'Dewey' 'Scout' 'Gizmo'
 'Walter' 'Cooper' 'Harold' 'Shikha' 'Lili' 'Jamesy' 'Coco' 'Sammy'
 'Meatball' 'Paisley' 'Albus' 'Neptune' 'Belle' 'Quinn' 'Zooey' 'Dave'
 'Jersey' 'Hobbes' 'Burt' 'Lorenzo' 'Carl' 'Jordy' 'Milky' 'Trooper'
 'Sophie' 'Wyatt' 'Rosie' 'Thor' 'Oscar' 'Callie' 'Cermet' 'Marlee' 'Ar
ya'
 'Einstein' 'Alice' 'Rumpole' 'Benny' 'Aspen' 'Jarod' 'Wiggles' 'Genera
 'Sailor' 'Iggy' 'Snoop' 'Kyle' 'Leo' 'Riley' 'Noosh' 'Odin' 'Jerry'
 'Georgie' 'Rontu' 'Cannon' 'Furzey' 'Daisy' 'Tuck' 'Barney' 'Vixen'
 'Jarvis' 'Mimosa' 'Pickles' 'Brady' 'Luna' 'Charlie' 'Margo' 'Sadie'
 'Hank' 'Tycho' 'Indie' 'Winnie' 'George' 'Bentley' 'Max' 'Dawn' 'Maddi
 'Monty' 'Sojourner' 'Winston' 'Odie' 'Arlo' 'Vincent' 'Lucy' 'Clark'
 'Mookie' 'Meera' 'Ava' 'Eli' 'Ash' 'Tucker' 'Tobi' 'Chester' 'Wilson'
 'Sunshine' 'Lipton' 'Bronte' 'Poppy' 'Gidget' 'Rhino' 'Willow' 'Orion'
 'Eevee' 'Smiley' 'Miguel' 'Emanuel' 'Kuyu' 'Dutch' 'Pete' 'Scooter'
 'Reggie' 'Lilly' 'Samson' 'Mia' 'Astrid' 'Malcolm' 'Dexter' 'Alfie'
 'Fiona' 'Mutt' 'Bear' 'Doobert' 'Beebop' 'Alexander' 'Sailer' 'Brutus'
 'Kona' 'Boots' 'Ralphie' 'Loki' 'Cupid' 'Pawnd' 'Pilot' 'Ike' 'Mo' 'To
by'
 'Sweet' 'Pablo' 'Nala' 'Crawford' 'Gabe' 'Jimison' 'Duchess' 'Harlso'
 'Sundance' 'Luca' 'Flash' 'Sunny' 'Howie' 'Jazzy' 'Anna' 'Finn' 'Bo'
 'Wafer' 'Tom' 'Florence' 'Autumn' 'Buddy' 'Dido' 'Eugene' 'Ken' 'Strud
 'Tebow' 'Chloe' 'Timber' 'Binky' 'Moose' 'Dudley' 'Comet' 'Akumi' 'Tit
 'Olivia' 'Alf' 'Oshie' 'Chubbs' 'Sky' 'Atlas' 'Eleanor' 'Layla' 'Rock
 'Baron' 'Tyr' 'Bauer' 'Swagger' 'Brandi' 'Mary' 'Moe' 'Halo' 'Augie'
 'Craig' 'Sam' 'Hunter' 'Pavlov' 'Phil' 'Kyro' 'Wallace' 'Ito' 'Seamus'
 'Ollie' 'Stephan' 'Lennon' 'Major' 'Duke' 'Sansa' 'Shooter' 'Django'
 'Diogi' 'Sonny' 'Marley' 'Severus' 'Ronnie' 'Milo' 'Bones' 'Mauve' 'Ch
ef'
 'Doc' 'Peaches' 'Sobe' 'Longfellow' 'Mister' 'Iroh' 'Pancake' 'Snicku'
 'Ruby' 'Brody' 'Mack' 'Nimbus' 'Laika' 'Maximus' 'Dobby' 'Moreton' 'Ju
 'Maude' 'Lily' 'Newt' 'Benji' 'Nida' 'Robin' 'Monster' 'BeBe' 'Remus'
 'Levi' 'Mabel' 'Misty' 'Betty' 'Mosby' 'Maggie' 'Bruce' 'Happy' 'Ralph
у'
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'Brownie' 'Rizzy' 'Stella' 'Butter' 'Frank' 'Tonks' 'Lincoln' 'Rory'
 'Logan' 'Dale' 'Rizzo' 'Arnie' 'Mattie' 'Pinot' 'Dallas' 'Hero' 'Frank
 'Stormy' 'Reginald' 'Balto' 'Mairi' 'Loomis' 'Godi' 'Cali' 'Deacon'
 'Timmy' 'Sampson' 'Chipson' 'Combo' 'Oakley' 'Dash' 'Hercules' 'Jay'
 'Strider' 'Wesley' 'Solomon' 'Huck' 'O' 'Blue' 'Anakin' 'Finley'
 'Sprinkles' 'Heinrich' 'Shakespeare' 'Chelsea' 'Bungalo' 'Chip' 'Grey'
 'Roosevelt' 'Willem' 'Davey' 'Dakota' 'Fizz' 'Dixie' 'Al' 'Jackson'
 'Carbon' 'Klein' 'DonDon' 'Kirby' 'Lou' 'Chevy' 'Tito' 'Philbert' 'Lou
 'Rupert' 'Rufus' 'Brudge' 'Shadoe' 'Angel' 'Brat' 'Tove' 'Gromit' 'Aub
 'Kota' 'Leela' 'Glenn' 'Shelby' 'Sephie' 'Bonaparte' 'Albert' 'Wishes'
 'Rose' 'Theo' 'Rocco' 'Fido' 'Emma' 'Spencer' 'Lilli' 'Boston'
 'Brandonald' 'Corey' 'Leonard' 'Beckham' 'Devón' 'Gert' 'Watson' 'Keit
h'
 'Dex' 'Ace' 'Tayzie' 'Grizzie' 'Fred' 'Gilbert' 'Meyer' 'Zoe' 'Stewie'
 'Calvin' 'Lilah' 'Spanky' 'Jameson' 'Piper' 'Atticus' 'Blu' 'Dietrich'
 'Divine' 'Tripp' 'Cora' 'Huxley' 'Keurig' 'Bookstore' 'Linus' 'Abby'
 'Shiloh' 'Gustav' 'Arlen' 'Percy' 'Lenox' 'Sugar' 'Harvey' 'Blanket'
 'Geno' 'Stark' 'Beya' 'Kilo' 'Kayla' 'Maxaroni' 'Bell' 'Doug' 'Edmund'
 'Aqua' 'Theodore' 'Baloo' 'Chase' 'Nollie' 'Rorie' 'Simba' 'Charles'
 'Bayley' 'Axel' 'Storkson' 'Remy' 'Chadrick' 'Kellogg' 'Buckley' 'Livv
 'Terry' 'Hermione' 'Ralpher' 'Aldrick' 'Larry' 'Rooney' 'Crystal' 'Ziv
 'Stefan' 'Pupcasso' 'Puff' 'Flurpson' 'Coleman' 'Enchilada' 'Raymond'
 'Rueben' 'Cilantro' 'Karll' 'Sprout' 'Blitz' 'Bloop' 'Colby' 'Lillie'
 'Ashleigh' 'Kreggory' 'Sarge' 'Luther' 'Ivar' 'Jangle' 'Schnitzel' 'Pa
 'Berkeley' 'Ralphé' 'Charleson' 'Clyde' 'Harnold' 'Sid' 'Pippa' 'Otis'
 'Carper' 'Bowie' 'Alexanderson' 'Suki' 'Barclay' 'Skittle' 'Ebby' 'Flá
 'Smokey' 'Link' 'Jennifur' 'Ozzy' 'Bluebert' 'Stephanus' 'Bubbles' 'Ze
 'Bertson' 'Nico' 'Michelangelope' 'Siba' 'Calbert' 'Curtis' 'Travis'
 'Thumas' 'Kanu' 'Lance' 'Opie' 'Stubert' 'Kane' 'Olive' 'Chuckles'
 'Staniel' 'Sora' 'Beemo' 'Gunner' 'Lacy' 'Tater' 'Olaf' 'Cecil' 'Vinc
e'
 'Karma' 'Billy' 'Walker' 'Rodney' 'Klevin' 'Malikai' 'Bobble' 'River'
 'Jebberson' 'Remington' 'Farfle' 'Jiminus' 'Harper' 'Clarkus' 'Finnegu
 'Cupcake' 'Kathmandu' 'Ellie' 'Katie' 'Kara' 'Adele' 'Zara' 'Ambrose'
 'Jimothy' 'Bode' 'Terrenth' 'Reese' 'Chesterson' 'Lucia' 'Bisquick'
 'Ralphson' 'Socks' 'Rambo' 'Rudy' 'Fiji' 'Rilo' 'Bilbo' 'Coopson' 'Yod
 'Millie' 'Chet' 'Crouton' 'Daniel' 'Kaia' 'Murphy' 'Dotsy' 'Eazy' 'Coo
 'Fillup' 'Miley' 'Charl' 'Reagan' 'Yukon' 'CeCe' 'Cuddles' 'Claude'
 'Jessiga' 'Carter' 'Ole' 'Pherb' 'Blipson' 'Reptar' 'Trevith' 'Berb'
 'Colin' 'Brian' 'Oliviér' 'Grady' 'Kobe' 'Freddery' 'Bodie' 'Dunkin'
 'Wally' 'Tupawc' 'Amber' 'Herschel' 'Edgar' 'Teddy' 'Kingsley' 'Brockl
 'Richie' 'Molly' 'Vinscent' 'Cedrick' 'Hazel' 'Lolo' 'Eriq' 'Phred'
 'Oddie' 'Maxwell' 'Geoff' 'Covach' 'Durg' 'Fynn' 'Ricky' 'Herald' 'Luc
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ky'
 'Ferg' 'Trip' 'Clarence' 'Hamrick' 'Brad' 'Pubert' 'Frönq' 'Derby'
 'Lizzie' 'Ember' 'Blakely' 'Opal' 'Marq' 'Kramer' 'Barry' 'Tyrone'
 'Gordon' 'Baxter' 'Mona' 'Horace' 'Crimson' 'Birf' 'Hammond' 'Lorelei'
 'Marty' 'Brooks' 'Petrick' 'Hubertson' 'Gerbald' 'Oreo' 'Bruiser' 'Per
 'Bobby' 'Jeph' 'Obi' 'Tino' 'Kulet' 'Sweets' 'Lupe' 'Tiger' 'Jiminy'
 'Griffin' 'Banjo' 'Brandy' 'Lulu' 'Darrel' 'Taco' 'Joey' 'Patrick' 'Kr
eg'
 'Todo' 'Tess' 'Ulysses' 'Toffee' 'Apollo' 'Carly' 'Asher' 'Glacier'
 'Chuck' 'Champ' 'Ozzie' 'Griswold' 'Cheesy' 'Moofasa' 'Hector' 'Goliat
h'
 'Kawhi' 'Emmie' 'Penelope' 'Willie' 'Rinna' 'Mike' 'William' 'Dwight'
 'Evy' 'Hurley' 'Rubio' 'Chompsky' 'Rascal' 'Linda' 'Tug' 'Tango' 'Griz
 'Jerome' 'Crumpet' 'Jessifer' 'Izzy' 'Ralph' 'Sandy' 'Humphrey' 'Tass
 'Juckson' 'Chuq' 'Tyrus' 'Karl' 'Godzilla' 'Vinnie' 'Kenneth' 'Herm'
 'Bert' 'Striker' 'Donny' 'Pepper' 'Bernie' 'Buddah' 'Lenny' 'Arnold'
 'Zuzu' 'Mollie' 'Laela' 'Tedders' 'Superpup' 'Rufio' 'Jeb' 'Rodman'
 'Jonah' 'Chesney' 'Kenny' 'Henry' 'Bobbay' 'Mitch' 'Kaiya' 'Acro' 'Aid
en'
 'Obie' 'Dot' 'Shnuggles' 'Kendall' 'Jeffri' 'Steve' 'Eve' 'Mac' 'Fletc
 'Kenzie' 'Pumpkin' 'Schnozz' 'Gustaf' 'Cheryl' 'Ed' 'Leonidas' 'Norma
 'Caryl' 'Scott' 'Taz' 'Darby' 'Jackie' 'Jazz' 'Frang' 'Pippin' 'Rolf'
 'Snickers' 'Ridley' 'Cal' 'Bradley' 'Bubba' 'Tuco' 'Patch' 'Mojo' 'Bat
 'Dylan' 'Mark' 'JD' 'Alejandro' 'Scruffers' 'Pip' 'Julius' 'Tanner'
 'Sparky' 'Anthony' 'Holly' 'Jett' 'Amy' 'Sage' 'Andy' 'Mason' 'Trigge
 'Antony' 'Creg' 'Traviss' 'Gin' 'Jeffrie' 'Danny' 'Ester' 'Pluto' 'Blo
 'Edd' 'Paull' 'Willy' 'Herb' 'Damon' 'Peanut' 'Nigel' 'Butters' 'Sandr
 'Fabio' 'Randall' 'Liam' 'Tommy' 'Ben' 'Raphael' 'Julio' 'Andru' 'Kloe
 'Shawwn' 'Skye' 'Kollin' 'Ronduh' 'Billl' 'Saydee' 'Dug' 'Sully' 'Kir
 'Ralf' 'Clarq' 'Jaspers' 'Samsom' 'Terrance' 'Harrison' 'Chaz' 'Jerem
 'Jaycob' 'Lambeau' 'Ruffles' 'Amélie' 'Bobb' 'Banditt' 'Kevon' 'Winifr
 'Hanz' 'Churlie' 'Zeek' 'Timofy' 'Maks' 'Jomathan' 'Kallie' 'Marvin'
 'Spark' 'Gòrdón' 'Jo' 'DayZ' 'Jareld' 'Torque' 'Ron' 'Skittles'
 'Cleopatricia' 'Erik' 'Stu' 'Tedrick' 'Shaggy' 'Filup' 'Kial' 'Naphani
el'
 'Dook' 'Hall' 'Philippe' 'Biden' 'Fwed' 'Genevieve' 'Joshwa' 'Timison'
 'Bradlay' 'Pipsy' 'Clybe' 'Keet' 'Carll' 'Jockson' 'Josep' 'Lugan'
 'Christoper'
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	tweet_id	timestamp	source	
0	892420643555336193	2017-08- 01 16:23:56	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Phi mystical b</th></a>	This is Phi mystical b
1	892177421306343426	2017-08- 01 00:17:27	<a href="http://twitter.com/download/iphone" r</a 	This is Tilly checking p
2	891815181378084864	2017-07- 31 00:18:03	<a href="http://twitter.com/download/iphone" r</a 	This is Arc rare Norwa Pouncin
3	891689557279858688	2017-07- 30 15:58:51	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Dar commence mid meal</th></a>	This is Dar commence mid meal
4	891327558926688256	2017-07- 29 16:00:24	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Fra</th></a>	This is Fra
5	891087950875897856	2017-07- 29 00:08:17	<a href="http://twitter.com/download/iphone" r<="" th=""><th>Here we had great white</th></a>	Here we had great white
6	890971913173991426	2017-07- 28 16:27:12	<a href="http://twitter.com/download/iphone" r<="" th=""><th>Meet Jax. cream so I</th></a>	Meet Jax. cream so I
7	890729181411237888	2017-07- 28 00:22:40	<a href="http://twitter.com/download/iphone" r</a 	When you owner call g
8	890609185150312448	2017-07- 27 16:25:51	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Zoe want to be</th></a>	This is Zoe want to be
9	890240255349198849	2017-07- 26 15:59:51	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Cas</th></a>	This is Cas
10	890006608113172480	2017-07- 26 00:31:25	<a href="http://twitter.com/download/iphone" r</a 	This is Koo South Aus decksha
11	889880896479866881	2017-07- 25 16:11:53	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Bru service sha</th></a>	This is Bru service sha
12	889665388333682689	2017-07- 25 01:55:32	<a href="http://twitter.com/download/iphone" r</a 	Here's a p seems to l a

	tweet_id	timestamp	source	
13	889638837579907072	2017-07- 25 00:10:02	<a href="http://twitter.com/download/iphone" r</a 	This is Ted best. Som
14	889531135344209921	2017-07- 24 17:02:04	<a href="http://twitter.com/download/iphone" r</a 	This is Stu sporting h
15	889278841981685760	2017-07- 24 00:19:32	<a href="http://twitter.com/download/iphone" r</a 	This is Olivwitnessing
16	888917238123831296	2017-07- 23 00:22:39	<a href="http://twitter.com/download/iphone" r</a 	This is Jim fren. Taugl
17	888804989199671297	2017-07- 22 16:56:37	<a href="http://twitter.com/download/iphone" r</a 	This is Zel- new stick.
18	888554962724278272	2017-07- 22 00:23:06	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Ral powering</th></a>	This is Ral powering
20	888078434458587136	2017-07- 20 16:49:33	<a href="http://twitter.com/download/iphone" r</a 	This is Gei
21	887705289381826560	2017-07- 19 16:06:48	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Jef</th></a>	This is Jef
22	887517139158093824	2017-07- 19 03:39:09	<a href="http://twitter.com/download/iphone" r</a 	I've yet to Venezuela Wiener. Th
23	887473957103951883	2017-07- 19 00:47:34	<a href="http://twitter.com/download/iphone" r</a 	This is Car attempted porch
24	887343217045368832	2017-07- 18 16:08:03	<a href="http://twitter.com/download/iphone" r</a 	You may n you neede
25	887101392804085760	2017-07- 18 00:07:08	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is a Antarctic I We</th></a>	This is a Antarctic I We
26	886983233522544640	2017-07- 17 16:17:36	<a href="http://twitter.com/download/iphone" r</a 	This is Ma

	tweet_id	timestamp	source	
27	886736880519319552	2017-07- 16 23:58:41	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Mir wonderful</td></a>	This is Mir wonderful
28	886680336477933568	2017-07- 16 20:14:00	<a href="http://twitter.com/download/iphone" r</a 	This is Der for a dog r
29	886366144734445568	2017-07- 15 23:25:31	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Ros</th></a>	This is Ros
31	886258384151887873	2017-07- 15 16:17:19	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Wa doggles ar down</th></a>	This is Wa doggles ar down
80	877316821321428993	2017-06- 21 00:06:44	<a href="http://twitter.com/download/iphone" r<="" td=""><td>Meet Dant wasn't a fa</td></a>	Meet Dant wasn't a fa
81	877201837425926144	2017-06- 20 16:29:50	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Nel graduated dogtorate.</td></a>	This is Nel graduated dogtorate.
82	876838120628539392	2017-06- 19 16:24:33	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Gin having a ru</td></a>	This is Gin having a ru
83	876537666061221889	2017-06- 18 20:30:39	<a href="http://twitter.com/download/iphone" r</a 	I can say v pupmost c the
84	876484053909872640	2017-06- 18 16:57:37	<a href="http://twitter.com/download/iphone" r</a 	This is Ber wants to thus
85	876120275196170240	2017-06- 17 16:52:05	<a href="http://twitter.com/download/iphone" r</a 	Meet Vent caffeinated
86	875747767867523072	2017-06- 16 16:11:53	<a href="http://twitter.com/download/iphone" r</a 	This is Gowomanize h*c
87	875144289856114688	2017-06- 15 00:13:52	<a href="http://twitter.com/download/iphone" r</a 	Meet Nug Nugget to bone
88	875097192612077568	2017-06- 14 21:06:43	<a href="http://twitter.com/download/iphone" r</a 	You'll get y when that man

	tweet_id	timestamp	source	
89	875021211251597312	2017-06- 14 16:04:48	<a href="http://twitter.com/download/iphone" r<="" td=""><td>Guys pleas sending pi any</td></a>	Guys pleas sending pi any
90	874680097055178752	2017-06- 13 17:29:20	<a href="http://twitter.com/download/iphone" r<="" td=""><td>Meet Cash acquired a go</td></a>	Meet Cash acquired a go
92	874296783580663808	2017-06- 12 16:06:11	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Jec the fancies</td></a>	This is Jec the fancies
93	874057562936811520	2017-06- 12 00:15:36	<a href="http://twitter.com/download/iphone" r<="" td=""><td>I can't beli happening</td></a>	I can't beli happening
94	874012996292530176	2017-06- 11 21:18:31	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Set can't see a</td></a>	This is Set can't see a
96	873580283840344065	2017-06- 10 16:39:04	<a href="http://twitter.com/download/iphone" r<="" td=""><td>We usually Deck-bour Bla</td></a>	We usually Deck-bour Bla
98	873213775632977920	2017-06- 09 16:22:42	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Sie precious p</td></a>	This is Sie precious p
99	872967104147763200	2017-06- 09 00:02:31	<a href="http://twitter.com/download/iphone" r<="" td=""><td>Here's a v He has a c</td></a>	Here's a v He has a c
100	872820683541237760	2017-06- 08 14:20:41	<a href="http://twitter.com/download/iphone" r</a 	Here are n #dogsatpc \n
102	872620804844003328	2017-06- 08 01:06:27	<a href="http://twitter.com/download/iphone" r</a 	This is Mo supporting everyw
103	872486979161796608	2017-06- 07 16:14:40	<a href="http://twitter.com/download/iphone" r<="" td=""><td>We. Only.</td></a>	We. Only.
104	872261713294495745	2017-06- 07 01:19:32	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Har activated (</td></a>	This is Har activated (
105	872122724285648897	2017-06- 06 16:07:15	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Koo baller. Wis I</td></a>	This is Koo baller. Wis I

	tweet_id	timestamp	source	
106	871879754684805121	2017-06- 06 00:01:46	<a href="http://twitter.com/download/iphone" r<="" td=""><td>Say hello t</td></a>	Say hello t
107	871762521631449091	2017-06- 05 16:15:56	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Rov</th></a>	This is Rov
108	871515927908634625	2017-06- 04 23:56:03	<a href="http://twitter.com/download/iphone" r<="" td=""><td>This is Na<sub>l</sub> Raggedy E</td></a>	This is Na <sub>l</sub> Raggedy E
110	871102520638267392	2017-06- 03 20:33:19	<a href="http://twitter.com/download/iphone" r<="" th=""><th>Never dou 14/10 https://t.ca</th></a>	Never dou 14/10 https://t.ca
111	871032628920680449	2017-06- 03 15:55:36	<a href="http://twitter.com/download/iphone" r</a 	This is Boo doing an a t
112	870804317367881728	2017-06- 03 00:48:22	<a href="http://twitter.com/download/iphone" r<="" th=""><th>Real funny in a pic wi</th></a>	Real funny in a pic wi
114	870656317836468226	2017-06- 02 15:00:16	<a href="http://twitter.com/download/iphone" r<="" th=""><th>This is Cortoo aggres</th></a>	This is Cortoo aggres
115	870374049280663552	2017-06- 01 20:18:38	<a href="http://twitter.com/download/iphone" r</a 	This is Zoe

100 rows × 12 columns

Quality Improvement - Change Source to HTML value instead of the whole html string

In [24]: # Use Extract to extract the text between the HTML markers.

df\_tae\_clean['source'].unique()
df\_tae\_clean['source'] = df\_tae\_clean.source.str.extract('>(.\*)<', expan d=True)

df\_tae\_clean.head()</pre>

Out[24]:

	tweet id	timestamp	source	text	
	tweet_id	umestamp	Source	lext	
0	892420643555336193	2017-08- 01 16:23:56	Twitter for iPhone	This is Phineas. He's a mystical boy. Only eve	https://twitter.com/dog_rates/statu
1	892177421306343426	2017-08- 01 00:17:27	Twitter for iPhone	This is Tilly. She's just checking pup on you	https://twitter.com/dog_rates/statu
2	891815181378084864	2017-07- 31 00:18:03	Twitter for iPhone	This is Archie. He is a rare Norwegian Pouncin	https://twitter.com/dog_rates/statu
3	891689557279858688	2017-07- 30 15:58:51	Twitter for iPhone	This is Darla. She commenced a snooze mid meal	https://twitter.com/dog_rates/statu
4	891327558926688256	2017-07- 29 16:00:24	Twitter for iPhone	This is Franklin. He would like you to stop ca	https://twitter.com/dog_rates/statu

#### Tidyness - Remove links from text since they have been put into expanded url section

In [27]: # Test individual value that was proving to be tricky because it had two
links
print(df\_tae\_clean['text'][6])

Meet Jax. He enjoys ice cream so much he gets nervous around it. 13/10 help Jax enjoy more things by clicking below

In [28]: df\_tae\_clean.head()

Out[28]:

	tweet_id	timestamp	source	text	
0	892420643555336193	2017-08- 01 16:23:56	Twitter for iPhone	This is Phineas. He's a mystical boy. Only eve	https://twitter.com/dog_rates/statu
1	892177421306343426	2017-08- 01 00:17:27	Twitter for iPhone	This is Tilly. She's just checking pup on you	https://twitter.com/dog_rates/statu
2	891815181378084864	2017-07- 31 00:18:03	Twitter for iPhone	This is Archie. He is a rare Norwegian Pouncin	https://twitter.com/dog_rates/statu
3	891689557279858688	2017-07- 30 15:58:51	Twitter for iPhone	This is Darla. She commenced a snooze mid meal	https://twitter.com/dog_rates/statu
4	891327558926688256	2017-07- 29 16:00:24	Twitter for iPhone	This is Franklin. He would like you to stop ca	https://twitter.com/dog_rates/statu

Quality Improvement - Remove rows where the rating denominator is not 10

```
In [29]: # Check the rows that have a rating_demoninator not equal to 10.
# option to drop those rows or to adjust the data - Decide to drop the d
    ata because it would end up with
# non-interger numbers

df_tae_clean.head()
    df_tae_clean.info()

df_rd = df_tae_clean[df_tae_clean['rating_denominator'] != 10]

print(df_rd)

df_tae_clean=df_tae_clean[df_tae_clean['rating_denominator'] == 10]
    df_tae_clean.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2096 entries, 0 to 2355
Data columns (total 12 columns):
tweet_id
                      2096 non-null int64
                      2096 non-null datetime64[ns]
timestamp
                      2096 non-null object
source
text
                      2096 non-null object
expanded urls
                      2093 non-null object
                      2096 non-null int64
rating_numerator
rating_denominator
                      2096 non-null int64
name
                      1389 non-null object
doggo
                      83 non-null object
                      10 non-null object
floofer
                      230 non-null object
pupper
                      24 non-null object
puppo
dtypes: datetime64[ns](1), int64(3), object(8)
memory usage: 292.9+ KB
                tweet id
                                   timestamp
                                                          source \
433
      820690176645140481 2017-01-15 17:52:40
                                              Twitter for iPhone
516
      810984652412424192 2016-12-19 23:06:23
                                              Twitter for iPhone
902
      758467244762497024 2016-07-28 01:00:57
                                              Twitter for iPhone
1068
     740373189193256964 2016-06-08 02:41:38
                                              Twitter for iPhone
1120
     731156023742988288 2016-05-13 16:15:54
                                              Twitter for iPhone
      722974582966214656 2016-04-21 02:25:47
1165
                                              Twitter for iPhone
1202
     716439118184652801 2016-04-03 01:36:11
                                              Twitter for iPhone
1228
     713900603437621249 2016-03-27 01:29:02
                                              Twitter for iPhone
1254
     710658690886586372 2016-03-18 02:46:49
                                              Twitter for iPhone
     709198395643068416 2016-03-14 02:04:08
1274
                                              Twitter for iPhone
1351
     704054845121142784 2016-02-28 21:25:30
                                              Twitter for iPhone
1433
     697463031882764288 2016-02-10 16:51:59
                                              Twitter for iPhone
1635
     684222868335505415 2016-01-05 04:00:18
                                              Twitter for iPhone
1662
     682962037429899265 2016-01-01 16:30:13
                                              Twitter for iPhone
1779 677716515794329600 2015-12-18 05:06:23
                                              Twitter for iPhone
1843
     675853064436391936 2015-12-13 01:41:41
                                              Twitter for iPhone
2335
     666287406224695296 2015-11-16 16:11:11
                                              Twitter for iPhone
                                                   text \
433
      The floofs have been released I repeat the flo...
516
     Meet Sam. She smiles 24/7 & amp; secretly aspir...
902
     Why does this never happen at my front door.....
1068
     After so many requests, this is Bretagne. She ...
1120
     Say hello to this unbelievably well behaved sq...
1165
     Happy 4/20 from the squad! 13/10 for all https...
1202
     This is Bluebert. He just saw that both #Final...
1228
     Happy Saturday here's 9 puppers on a bench. 99...
1254
     Here's a brigade of puppers. All look very pre...
1274
                                    From left to right:
     Here is a whole flock of puppers.
1351
                                         60/50 I'll ...
1433
     Happy Wednesday here's a bucket of pups. 44/40...
1635
     Someone help the girl is being mugged. Several...
     This is Darrel. He just robbed a 7/11 and is i...
1662
     IT'S PUPPERGEDDON. Total of 144/120 ...I think...
1779
1843
     Here we have an entire platoon of puppers. Tot...
     This is an Albanian 3 1/2 legged Episcopalian...
```

```
433
      https://twitter.com/dog rates/status/820690176...
84
516
      https://www.gofundme.com/sams-smile,https://tw...
24
902
      https://twitter.com/dog_rates/status/758467244...
                                                                         1
65
1068
      https://twitter.com/dog_rates/status/740373189...
      https://twitter.com/dog_rates/status/731156023...
                                                                         2
1120
04
      https://twitter.com/dog_rates/status/722974582...
1165
4
1202
      https://twitter.com/dog rates/status/716439118...
50
1228
      https://twitter.com/dog_rates/status/713900603...
99
      https://twitter.com/dog_rates/status/710658690...
1254
80
1274
      https://twitter.com/dog_rates/status/709198395...
45
1351
      https://twitter.com/dog_rates/status/704054845...
60
      https://twitter.com/dog_rates/status/697463031...
1433
44
1635
      https://twitter.com/dog_rates/status/684222868...
                                                                         1
21
1662
      https://twitter.com/dog_rates/status/682962037...
7
1779
      https://twitter.com/dog rates/status/677716515...
                                                                         1
44
      https://twitter.com/dog_rates/status/675853064...
1843
88
      https://twitter.com/dog rates/status/666287406...
2335
1
      rating denominator
                               name doggo floofer pupper puppo
433
                       70
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
516
                        7
                                              NaN
                                                      NaN
                                Sam
                                      NaN
                                                            NaN
902
                      150
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1068
                       11
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1120
                     170
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1165
                       20
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1202
                       50
                          Bluebert
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1228
                       90
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1254
                       80
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1274
                       50
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1351
                       50
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1433
                       40
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1635
                      110
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1662
                             Darrel
                                                      NaN
                       11
                                      NaN
                                              NaN
                                                            NaN
1779
                      120
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
1843
                       80
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
2335
                        2
                                NaN
                                      NaN
                                              NaN
                                                      NaN
                                                            NaN
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2079 entries, 0 to 2355
Data columns (total 12 columns):
tweet id
                      2079 non-null int64
```

```
2079 non-null datetime64[ns]
timestamp
source
                      2079 non-null object
                      2079 non-null object
text
expanded urls
                      2076 non-null object
rating numerator
                      2079 non-null int64
rating_denominator
                      2079 non-null int64
                      1386 non-null object
name
doggo
                      83 non-null object
                      10 non-null object
floofer
                      230 non-null object
pupper
                      24 non-null object
puppo
dtypes: datetime64[ns](1), int64(3), object(8)
memory usage: 211.1+ KB
```

#### Quality Improvement - Remove rows where there are no photos

```
Int64Index: 1882 entries, 0 to 2355
Data columns (total 12 columns):
tweet id
                       1882 non-null int64
timestamp
                       1882 non-null datetime64[ns]
source
                       1882 non-null object
text
                       1882 non-null object
expanded urls
                       1882 non-null object
rating_numerator 1882 non-null int64
rating_denominator 1882 non-null int64
                       1882 non-null int64
                       1313 non-null object
name
                       64 non-null object
doggo
floofer
                       7 non-null object
                       199 non-null object
pupper
                       23 non-null object
puppo
dtypes: datetime64[ns](1), int64(3), object(8)
memory usage: 191.1+ KB
```

Tidiness - Merging the three tables into one table

```
In [31]: # For reference:
    # df_tae_clean = pd.read_csv('twitter-archive-enhanced.csv') - cleaned u
    p
    # df_tip = pd.read_csv('image-predictions.tsv',sep='\t')
    # df_info = pd.read_csv('tweet_extra_info.csv')

# Double Checking the data

df_tae_clean.info()
df_tae_clean.head()
```

<class 'pandas.core.frame.DataFrame'> Int64Index: 1882 entries, 0 to 2355 Data columns (total 12 columns): tweet\_id 1882 non-null int64 timestamp 1882 non-null datetime64[ns] source 1882 non-null object 1882 non-null object text 1882 non-null object expanded urls 1882 non-null int64 rating\_numerator rating\_denominator 1882 non-null int64 name 1313 non-null object doggo 64 non-null object floofer 7 non-null object 199 non-null object pupper 23 non-null object puppo

dtypes: datetime64[ns](1), int64(3), object(8)

memory usage: 191.1+ KB

#### Out[31]:

	turant id	time a ataman		dov.d.	
	tweet_id	timestamp	source	text	
0	892420643555336193	2017-08- 01 16:23:56	Twitter for iPhone	This is Phineas. He's a mystical boy. Only eve	https://twitter.com/dog_rates/statu
1	892177421306343426	2017-08- 01 00:17:27	Twitter for iPhone	This is Tilly. She's just checking pup on you	https://twitter.com/dog_rates/statu
2	891815181378084864	2017-07- 31 00:18:03	Twitter for iPhone	This is Archie. He is a rare Norwegian Pouncin	https://twitter.com/dog_rates/statu
3	891689557279858688	2017-07- 30 15:58:51	Twitter for iPhone	This is Darla. She commenced a snooze mid meal	https://twitter.com/dog_rates/statu
4	891327558926688256	2017-07- 29 16:00:24	Twitter for iPhone	This is Franklin. He would like you to stop ca	https://twitter.com/dog_rates/statu

```
In [32]: # Double Checking the data

df_tip.info()
df_tip.head()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2075 entries, 0 to 2074
Data columns (total 12 columns):
tweet id 2075 non-null int64
           2075 non-null object
jpg_url
           2075 non-null int64
img_num
           2075 non-null object
р1
p1_conf
           2075 non-null float64
p1_dog
           2075 non-null bool
           2075 non-null object
p2
p2_conf
           2075 non-null float64
p2_dog
           2075 non-null bool
           2075 non-null object
p3
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
```

#### Out[32]:

	tweet_id	jpg_url	img_num	
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	1	W
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	1	re
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	1	G
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-IEu.jpg	1	R
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	1	m

# In [33]: # Double Checking the data df\_info.info() df\_info.head()

dtypes: int64(3)
memory usage: 55.0 KB

#### Out[33]:

	tweet_id	retweet_count	like_count
0	892420643555336193	8553	38673
1	892177421306343426	6287	33127
2	891815181378084864	4166	24943
3	891689557279858688	8675	42044
4	891327558926688256	9441	40193

# In [34]: # Left merge to get values from the likes and retweets table into the ma in cleaned table.

df\_merged=pd.merge(df\_tae\_clean, df\_info, on='tweet\_id', how='left')
df\_merged.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1882 entries, 0 to 1881
Data columns (total 14 columns):

tweet id 1882 non-null int64

timestamp 1882 non-null datetime64[ns]

source 1882 non-null object text 1882 non-null object expanded urls 1882 non-null object rating\_numerator 1882 non-null int64 rating\_denominator 1882 non-null int64 1313 non-null object name 64 non-null object doggo 7 non-null object floofer 199 non-null object pupper puppo 23 non-null object retweet count 1882 non-null int64 like count 1882 non-null int64 dtypes: datetime64[ns](1), int64(5), object(8)

memory usage: 220.5+ KB

```
In [35]: # Left merge to get values from the photo analysis table into the main c
         leaned table.
         df_merged=pd.merge(df_merged, df_tip, on='tweet_id', how='left')
         df merged.info()
         df merged.head()
         df cleaned = df merged.copy()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 1882 entries, 0 to 1881
         Data columns (total 25 columns):
         tweet id
                               1882 non-null int64
                               1882 non-null datetime64[ns]
         timestamp
         source
                               1882 non-null object
         text
                               1882 non-null object
         expanded_urls
                               1882 non-null object
         rating_numerator
                               1882 non-null int64
         rating denominator
                               1882 non-null int64
                               1313 non-null object
         name
         doggo
                               64 non-null object
                               7 non-null object
         floofer
         pupper
                               199 non-null object
                               23 non-null object
         puppo
         retweet_count
                               1882 non-null int64
                               1882 non-null int64
         like count
         jpg_url
                              1882 non-null object
         img num
                               1882 non-null int64
                              1882 non-null object
         р1
         p1_conf
                               1882 non-null float64
                               1882 non-null bool
         p1 dog
                               1882 non-null object
         p2
```

p3\_dog 1882 non-null bool dtypes: bool(3), datetime64[ns](1), float64(3), int64(6), object(12) memory usage: 343.7+ KB

1882 non-null float64

1882 non-null float64

1882 non-null bool 1882 non-null object

### **Storing**

```
In [36]: # Saving final cleaned data off to csv

df_cleaned.to_csv('./data/twitter_archive_master.csv',index=False)
```

## **Analyzing and Visualizing Data**

p2 conf

p2\_dog

p3 conf

p3

#### Requirements:

Analyze and visualize your wrangled data in your wrangle\_act.ipynb Jupyter Notebook. At least three (3) insights and one (1) visualization must be produced.

```
In [37]: # Reading in the cleaned data and looking at the basic info for the fina
l product.

df_t = pd.read_csv("./data/twitter_archive_master.csv")
    df_t.head()
```

Out[37]:

	tweet_id	timestamp	source	text	
0	892420643555336193	2017-08- 01 16:23:56	Twitter for iPhone	This is Phineas. He's a mystical boy. Only eve	https://twitter.com/dog_rates/statu
1	892177421306343426	2017-08- 01 00:17:27	Twitter for iPhone	This is Tilly. She's just checking pup on you	https://twitter.com/dog_rates/statu
2	891815181378084864	2017-07- 31 00:18:03	Twitter for iPhone	This is Archie. He is a rare Norwegian Pouncin	https://twitter.com/dog_rates/statu
3	891689557279858688	2017-07- 30 15:58:51	Twitter for iPhone	This is Darla. She commenced a snooze mid meal	https://twitter.com/dog_rates/statu
4	891327558926688256	2017-07- 29 16:00:24	Twitter for iPhone	This is Franklin. He would like you to stop ca	https://twitter.com/dog_rates/statu

5 rows × 25 columns

In [38]: df\_t.describe()

Out[38]:

	tweet_id	rating_numerator	rating_denominator	retweet_count	like_cou
count	1.882000e+03	1882.000000	1882.0	1882.000000	1882.000000
mean	7.350763e+17	11.691817	10.0	2466.648247	8417.953773
std	6.751862e+16	41.857148	0.0	3611.245684	11378.795038
min	6.660209e+17	0.000000	10.0	13.000000	80.000000
25%	6.753650e+17	10.000000	10.0	591.000000	1830.000000
50%	7.077175e+17	11.000000	10.0	1284.500000	3919.500000
75%	7.868996e+17	12.000000	10.0	3012.500000	10829.750000
max	8.924206e+17	1776.000000	10.0	48908.000000	142895.00000

In [39]: # Most popular in terms of retweet counts

df\_t[df\_t['retweet\_count'] == df\_t['retweet\_count'].max()]

Out[39]:

	tweet_id	timestamp	source	text	
292	822872901745569793	2017-01- 21 18:26:02	Twitter for iPhone	Here's a super supportive puppo participating	https://twitter.com/dog_rates/st

1 rows × 25 columns

In [40]: df\_t['retweet\_count'].max()

Out[40]: 48908

```
In [41]: # Most popular in terms of like counts
         df_t[df_t['like_count']==df_t['like_count'].max()]
```

Out[41]:

	tweet_id	timestamp	source	text	
292	822872901745569793	2017-01- 21 18:26:02	Twitter for iPhone	Here's a super supportive puppo participating	https://twitter.com/dog_rates/st

1 rows × 25 columns

```
df_t['like_count'].max()
In [42]:
```

Out[42]: 142895

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

```
RangeIndex: 1882 entries, 0 to 1881
Data columns (total 25 columns):
tweet id
                      1882 non-null int64
timestamp
                      1882 non-null object
source
                      1882 non-null object
text
                      1882 non-null object
expanded urls
                      1882 non-null object
rating numerator
                      1882 non-null int64
                      1882 non-null int64
rating denominator
                      1313 non-null object
name
                      64 non-null object
doggo
floofer
                      7 non-null object
pupper
                      199 non-null object
                      23 non-null object
puppo
retweet count
                      1882 non-null int64
                      1882 non-null int64
like_count
jpg url
                      1882 non-null object
                      1882 non-null int64
img_num
                      1882 non-null object
p1
                      1882 non-null float64
p1 conf
                      1882 non-null bool
p1 dog
                      1882 non-null object
p2
p2 conf
                      1882 non-null float64
                      1882 non-null bool
p2 dog
                      1882 non-null object
p3
                      1882 non-null float64
p3 conf
                      1882 non-null bool
p3 dog
dtypes: bool(3), float64(3), int64(6), object(13)
```

<class 'pandas.core.frame.DataFrame'>

memory usage: 329.1+ KB

In [44]: df\_t.sample(100)

	tweet_id	timestamp	source	text	
225	834209720923721728	2017-02- 22 01:14:30	Twitter for iPhone	This is Wilson. He's aware that he has somethi	https://twitter.com/dog_rate
751	739485634323156992	2016-06- 05 15:54:48	Twitter for iPhone	This is Kyle. He's a heavy drinker and an avid	https://twitter.com/dog_rate
1719	668979806671884288	2015-11- 24 02:29:49	Twitter for iPhone	This is Chaz. He's an X Games half pipe supers	https://twitter.com/dog_rate
1508	673240798075449344	2015-12- 05 20:41:29	Twitter for iPhone	Magical floating dog here. Very calm. Always h	https://twitter.com/dog_rate
1637	670676092097810432	2015-11- 28 18:50:15	Twitter for iPhone	This is Bloo. He's a Westminster Cîroc. Doesn'	https://twitter.com/dog_rate
1388	676098748976615425	2015-12- 13 17:57:57	Twitter for iPhone	Extremely rare pup here. Very religious.	https://twitter.com/dog_rate
1479	673709992831262724	2015-12- 07 03:45:53	Twitter for iPhone	I know a lot of you are studying for finals. G	https://twitter.com/dog_rate
1460	674082852460433408	2015-12- 08 04:27:30	Twitter for iPhone	This is a Sagitariot Baklava mix. Loves her ne	https://twitter.com/dog_rate
883	713411074226274305	2016-03- 25 17:03:49	Twitter for iPhone	Here we see an extremely rare Bearded Floofmal	https://twitter.com/dog_rate
841	719551379208073216	2016-04- 11 15:43:12	Twitter for iPhone	This is Harnold. He accidentally opened the fr	https://twitter.com/dog_rate

	tweet_id	timestamp	source	text	
380	805487436403003392	2016-12- 04 19:02:24	Twitter for iPhone	Meet Sansa and Gary. They run along the fence	https://twitter.com/dog_rate
1399	675710890956750848	2015-12- 12 16:16:45	Twitter for iPhone	This is Lenny. He was just told that he couldn	https://twitter.com/dog_rate
1053	697482927769255936	2016-02- 10 18:11:03	Twitter for iPhone	Meet Blipson. He's a Doowap Hufflepuff. That U	https://twitter.com/dog_rate
120	860524505164394496	2017-05- 05 16:00:04	Twitter for iPhone	This is Carl. He likes to dance. Doesn't care	https://twitter.com/dog_rate
939	707776935007539200	2016-03- 10 03:55:45	Twitter for iPhone	This is Sadie. She's a Bohemian Rhapsody. Rema	https://twitter.com/dog_rate
648	752917284578922496	2016-07- 12 17:27:23	Twitter for iPhone	This is Grizzie. She's a semi- submerged Bahrai	https://twitter.com/dog_rate
657	751538714308972544	2016-07- 08 22:09:27	Twitter for iPhone	This is Max. She has one ear that's always sli	https://twitter.com/dog_rate
1380	676430933382295552	2015-12- 14 15:57:56	Twitter for iPhone	Meet Duke. He's an Urban Parmesan. They know h	https://twitter.com/dog_rate
1267	681242418453299201	2015-12- 27 22:37:04	Twitter for iPhone	This is Champ. He's being sacrificed to the Az	https://twitter.com/dog_rate
390	802239329049477120	2016-11- 25 19:55:35	Twitter for iPhone	This is Loki. He'll do your taxes for you. Can	https://twitter.com/dog_rate

	tweet_id	timestamp	source	text	
960	706291001778950144	2016-03- 06 01:31:11	Twitter for iPhone	When you're just relaxin and having a swell ti	https://twitter.com/dog_rate
526	776218204058357768	2016-09- 15 00:36:55	Twitter for iPhone	Atlas rolled around in some chalk and now he's	https://twitter.com/dog_rate
1393	675878199931371520	2015-12- 13 03:21:34	Twitter for iPhone	Ok, I'll admit this is a pretty adorable bunny	https://twitter.com/dog_rate
1843	666649482315059201	2015-11- 17 16:09:56	Twitter for iPhone	Cool dog. Enjoys couch. Low monotone bark. Ver	https://twitter.com/dog_rate
1419	675135153782571009	2015-12- 11 02:08:58	Twitter for iPhone	This is Steven. He got locked outside. Damn it	https://twitter.com/dog_rate
1691	669573570759163904	2015-11- 25 17:49:14	Twitter for iPhone	This is Linda. She just looked up and saw you	https://twitter.com/dog_rate
1245	682389078323662849	2015-12- 31 02:33:29	Twitter for iPhone	Meet Brody. He's a Downton Abbey Falsetto. Add	https://twitter.com/dog_rate
125	859607811541651456	2017-05- 03 03:17:27	Twitter for iPhone	Sorry for the lack of posts today. I came home	https://twitter.com/dog_rate
1029	699423671849451520	2016-02- 16 02:42:52	Twitter for iPhone	"Don't ever talk to me or my son again." bo	https://twitter.com/dog_rate
810	727286334147182592	2016-05- 02 23:59:09	Twitter for iPhone	I swear to god if we get sent another Blue Mad	https://twitter.com/dog_rate

	tweet_id	timestamp	source	text	
416	796116448414461957	2016-11- 08 22:25:27	Twitter for iPhone	I didn't believe it at first but now I can see	https://twitter.com/dog_rate
612	759793422261743616	2016-07- 31 16:50:42	Twitter for iPhone	Meet Maggie & Lila. Maggie is the doggo, L	https://twitter.com/dog_rate
251	829501995190984704	2017-02- 09 01:27:41	Twitter for iPhone	This is Leo. He was a skater pup. She said see	https://twitter.com/dog_rate
520	777885040357281792	2016-09- 19 15:00:20	Twitter for iPhone	This is Wesley. He's clearly trespassing. Seem	https://twitter.com/dog_rate
1339	677895101218201600	2015-12- 18 16:56:01	Twitter for iPhone	Guys this was terrifying. Really spooked me up	https://twitter.com/dog_rate
976	704819833553219584	2016-03- 02 00:05:17	Twitter for iPhone	This is Chesterson. He's a Bolivian Scoop Dog	https://twitter.com/dog_rate
1398	675781562965868544	2015-12- 12 20:57:34	Twitter for iPhone	Say hello to Buddah. He was Waldo for Hallowee	https://twitter.com/dog_rate
1001	701981390485725185	2016-02- 23 04:06:20	Twitter for iPhone	This is Fiji. She's a Powdered Stegafloof. Ver	https://twitter.com/dog_rate
1266	681261549936340994	2015-12- 27 23:53:05	Twitter for iPhone	Say hello to Panda. He's a Quackadilly Shooste	https://twitter.com/dog_rate
1505	673317986296586240	2015-12- 06 01:48:12	Twitter for iPhone	Take a moment and appreciate how these two dog	https://twitter.com/dog_rate

	tweet_id	timestamp	source	text	
670	750071704093859840	2016-07- 04 21:00:04	Twitter for iPhone	Pause your cookout and admire this pupper's ni	https://twitter.com/dog_rate
1405	675501075957489664	2015-12- 12 02:23:01	Twitter for iPhone	I shall call him squishy and he shall be mine,	https://twitter.com/dog_rate
1394	675845657354215424	2015-12- 13 01:12:15	Twitter for iPhone	This is Pepper. She's not fully comfortable ri	https://twitter.com/dog_rate
1849	666430724426358785	2015-11- 17 01:40:41	Twitter for iPhone	Oh boy what a pup! Sunglasses take this one to	https://twitter.com/dog_rate
571	768473857036525572	2016-08- 24 15:43:39	Twitter for iPhone	Meet Chevy. He had a late breakfast and now ha	https://twitter.com/dog_rate
1713	668994913074286592	2015-11- 24 03:29:51	Twitter for iPhone	Two gorgeous pups here. Both have cute fake ho	https://twitter.com/dog_rate
1443	674638615994089473	2015-12- 09 17:15:54	Twitter for iPhone	This pupper is fed up with being tickled. 12/1	https://twitter.com/dog_rate
1642	670452855871037440	2015-11- 28 04:03:11	Twitter for iPhone	This dog can't see its haters. 11/10 https://t	https://twitter.com/dog_rate
303	820749716845686786	2017-01- 15 21:49:15	Twitter for iPhone	Meet Sunny. He can take down a polar bear in o	https://twitter.com/dog_rate
553	771770456517009408	2016-09- 02 18:03:10	Twitter for iPhone	This is Davey. He'll have your daughter home b	https://twitter.com/dog_rate

	tweet_id	timestamp	source	text	
299	821765923262631936	2017-01- 18 17:07:18	Twitter for iPhone	This is Duchess. She uses dark doggo forces to	https://twitter.com/dog_rate
163	848324959059550208	2017-04- 02 00:03:26	Twitter for iPhone	Meet Odin. He's supposed to be giving directio	https://twitter.com/dog_rate
1395	675822767435051008	2015-12- 12 23:41:18	Twitter for iPhone	HELLO FROM THE OTHER SIIIIIIIDE 10/10s ht	https://twitter.com/dog_rate
438	793135492858580992	2016-10- 31 17:00:11	Twitter for iPhone	Your favorite squad is looking extra h*ckin sp	https://twitter.com/dog_rate
1463	674053186244734976	2015-12- 08 02:29:37	Twitter for iPhone	This is Stanley. Yes he is aware of the spoon'	https://twitter.com/dog_rate
1255	681694085539872773	2015-12- 29 04:31:49	Twitter for iPhone	This is Bo. He's a Benedoop Cumbersnatch. Seem	https://twitter.com/dog_rate
1490	673656262056419329	2015-12- 07 00:12:23	Twitter for iPhone	This is Albert AKA King Banana Peel. He's a ki	https://twitter.com/dog_rate
460	788908386943430656	2016-10- 20 01:03:11	Twitter for iPhone	This is Lucy. She destroyed not one, but two r	https://twitter.com/dog_rate
1158	688116655151435777	2016-01- 15 21:52:49	Twitter for iPhone	Please send dogs. I'm tired of seeing other st	https://twitter.com/dog_rate
977	704761120771465216	2016-03- 01 20:11:59	Twitter for iPhone	This pupper killed this great white in an epic	https://twitter.com/dog_rate

In [45]: # What breed is the most popular to tweet about?

dog\_breed = df\_t.groupby('p1').count().sort\_values('tweet\_id',ascending=
False)

dog\_breed.head(10)

Out[45]:

	tweet_id	timestamp	source	text	expanded_urls	rating_numerator
p1						
golden_retriever	131	131	131	131	131	131
Pembroke	87	87	87	87	87	87
Labrador_retriever	87	87	87	87	87	87
Chihuahua	76	76	76	76	76	76
pug	54	54	54	54	54	54
chow	41	41	41	41	41	41
toy_poodle	37	37	37	37	37	37
Pomeranian	37	37	37	37	37	37
Samoyed	37	37	37	37	37	37
malamute	29	29	29	29	29	29

10 rows × 24 columns

```
In [46]: # Quick check to see if there might be different ways that golden retrie
    ver appears

df_t['p1'][df_t['p1'].str.contains('olden')].unique()
    df_t['p1'][df_t['p1'].str.contains('nian')].unique()
```

Out[46]: array(['Pomeranian'], dtype=object)

#### What breed is the most popular to tweet about?

#### Golden Retrievers

#### Observations:

- 1. Most of these breeds would be expected to be seen on a list of the most popular breeds. 131 seems kind of low for golden retriever given there are thousands of tweets being analyzed.
- 2. Looks like to get real information the data would need to be further cleaned. Thought the photo information would be more consistent.
- After doing a quick it looks like although the breed names could be more consistent in terms of upper case and lower case, not seeing obvious duplicates. Would recommend changing all breed names to all lower case.

```
In [47]: # What breed of dog got the highest rating?
         df_t['rating_numerator'].max()
         # What breed of dog got the highest rating?
         df_t['p1'][df_t['rating_numerator'] == df_t['rating_numerator'].max()]
                bow tie
Out[47]: 676
         Name: p1, dtype: object
In [48]: # Bow tie, what is a bow tie? Checking p2 option.
         df_t['p2'][df_t['rating_numerator'] == df_t['rating_numerator'].max()]
Out[48]: 676
                sunglasses
         Name: p2, dtype: object
In [49]: # Sunglasses. It appears that the image file is classifying things othe
         r than breed.
         df_t['tweet_id'][df_t['rating_numerator'] == df_t['rating_numerator'].ma
         x()]
         # Using tweet id visit the actual tweet: https://twitter.com/dog rate
         s/status/749981277374128128
Out[49]: 676
                749981277374128128
         Name: tweet id, dtype: int64
```

#### What breed of dog got the highest rating?

The highest rated dog was not idenfied by breed, but was correctly identified to be wearing a bow\_tie and sunglasses: <a href="https://twitter.com/dog\_rates/status/749981277374128128">https://twitter.com/dog\_rates/status/749981277374128128</a> (<a href="https://twitter.com/dog\_rates/status/749981277374128128">https://twitter.com/dog\_rates/status/749981277374128128</a>)

#### Observations:

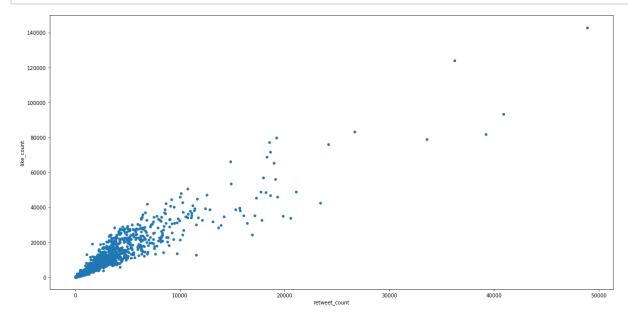
- 1. The highest rated dog was p1 = bow\_tie and p2 = sunglasses.
- 2. The information in the image predictions file contains more than breed information, so can't rely on it to get consistent breed information to compare tweets to breed.

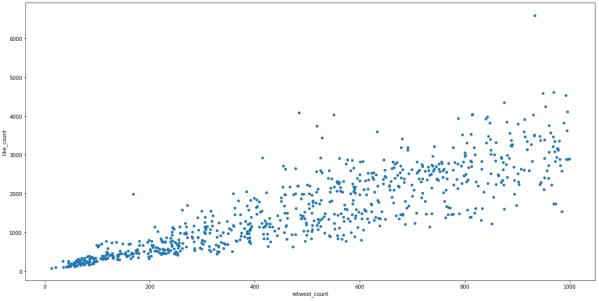
```
In [50]: # Are number of retweets and likes related?

df_t.plot('retweet_count','like_count',kind='scatter',figsize=(20,10));

df_p = df_t[df_t['retweet_count']<=1000]

df_p.plot('retweet_count','like_count',kind='scatter',figsize=(20,10));</pre>
```





In [51]: df\_t['tweet\_id'][df\_t['like\_count'] == df\_t['like\_count'].max()]

Out[51]: 292 822872901745569793 Name: tweet\_id, dtype: int64

In [52]: df\_t.nlargest(5, 'like\_count')

Out[52]:

	tweet_id	timestamp	source	text	
292	822872901745569793	2017-01- 21 18:26:02	Twitter for iPhone	Here's a super supportive puppo participating	https://twitter.com/dog_rates/st
102	866450705531457537	2017-05- 22 00:28:40	Twitter for iPhone	This is Jamesy. He gives a kiss to every other	https://twitter.com/dog_rates/st
312	819004803107983360	2017-01- 11 02:15:36	Twitter for iPhone	This is Bo. He was a very good First Doggo. 14	https://twitter.com/dog_rates/st
87	870374049280663552	2017-06- 01 20:18:38	Twitter for iPhone	This is Zoey. She really likes the planet. Wou	https://twitter.com/dog_rates/st
375	806629075125202948	2016-12- 07 22:38:52	Twitter for iPhone	"Good afternoon class today we're going to lea	https://twitter.com/dog_rates/st

5 rows × 25 columns

#### Are the number of retweets and likes related?

Looking at the plot above they do look related.

#### Observations:

- 1. Manually viewing the top 5 tweets they included the subjects:
- 2. Cute puppy
- 3. Political/Women's March dog
- 4. Obama's dog Bo (famous dog)
- 5. Environmental issue dog
- 6. Dog acting as professor

Possibility that tweets that are politically connected or tied to a major issue of the day may be more popular.

```
In [53]: # Are the number of likes and retweets related to the age of the tweet?

df_a=df_t.copy()

# Can't plot a scatterplot based on the timestamp so need to add age in days.

ndates =[]

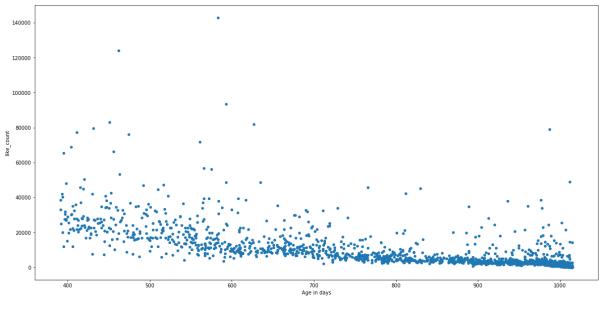
for d in df_a['timestamp']:
    ndates.append((datetime.now() - pd.to_datetime(d)).days)

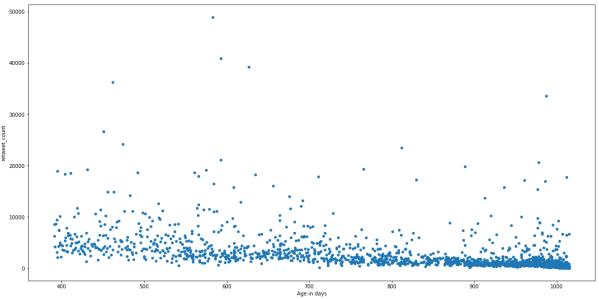
df_a['Age in days']=ndates

df_a.head()

# Plotting Age in days vs likes and retweets

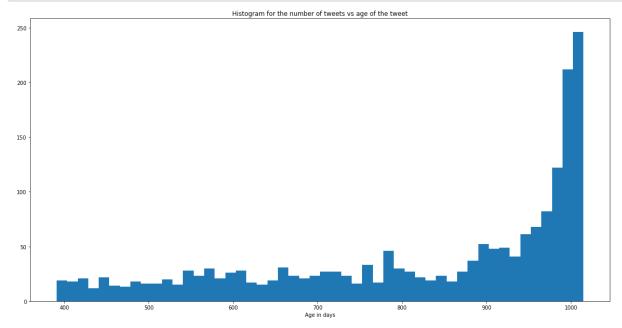
df_a.plot('Age in days', 'like_count', kind='scatter', figsize=(20,10));
    df_a.plot('Age in days', 'retweet_count', kind='scatter', figsize=(20,10));
```





```
In [54]: # Plotting histogram of Age in days to see tweet volumn over time
    plt.figure(figsize=(20,10))
    plt.hist(ndates, bins=50)
    plt.xlabel("Age in days")
    plt.title("Histogram for the number of tweets vs age of the tweet")

    plt.show();
```



#### Are the number of likes and retweets related to the age of the tweet?

Looking at the plots above they do look related.

#### Observations:

- 1. There does seem to be a relationship, but it appears that newer tweets are more likely to get more likes and retweets than older tweets.
- 2. Looks like there were tweets more frequent a few years ago than there are today. Plotted histogram and it does appear that the frequency of tweets was much higher 800 days plus ago.
- 3. Cleaned data doesn't always provide the data in the right format for graphing.