

# act\_report

June 27, 2022

## 0.1 Report: act\_report

- Create a **250-word-minimum written report** called "act\_report.pdf" or "act\_report.html" that communicates the insights and displays the visualization(s) produced from your wrangled data. This is to be framed as an external document, like a blog post or magazine article, for example.

## 1 Analyzing, and Visualizing Data

*We Rate Dogs dataset act\_report By Edwin Kihara*

```
In [1]: import matplotlib
import matplotlib.pyplot as plt
import pandas as pd
import datetime as dt
import seaborn as sns
```

```
%matplotlib inline
```

```
In [2]: # Change the style of the plots (http://tonysyu.github.io/raw\_content/matplotlib-style-g)
matplotlib.style.use('ggplot')
```

```
In [3]: # Import the clean dataset into dataframe
df_master = pd.read_csv('twitter_archive_master.csv')
df_master.info()
```

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

```
RangeIndex: 1994 entries, 0 to 1993
```

```
Data columns (total 16 columns):
```

tweet_id	1994 non-null int64
tweet_date	1994 non-null object
tweet_source	1994 non-null object
tweet_text	1994 non-null object
tweet_url	1994 non-null object
tweet_picture_predicted	1994 non-null object
tweet_favorites	1994 non-null int64
tweet_retweets	1994 non-null int64
user_followers	1994 non-null int64

```

dog_stage          1994 non-null object
dog_breed          1686 non-null object
confidence_level   1994 non-null float64
rating_numerator   1993 non-null float64
dogs_count         1994 non-null int64
dog_name           1369 non-null object
dog_gender         862 non-null object
dtypes: float64(2), int64(5), object(9)
memory usage: 249.3+ KB

```

In [4]: *# Convert columns to their appropriate types and set the tweet\_date as an index*

```

df_master['tweet_id'] = df_master['tweet_id'].astype(object)
df_master['tweet_date'] = pd.to_datetime(df_master.tweet_date)
df_master['tweet_source'] = df_master['tweet_source'].astype('category')
df_master['dog_stage'] = df_master['dog_stage'].astype('category')
df_master['dog_gender'] = df_master['dog_gender'].astype('category')

df_master = df_master.set_index('tweet_date')
df_master.info()

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 1994 entries, 2015-11-19 18:13:27 to 2016-06-16 01:25:36
Data columns (total 15 columns):
tweet_id          1994 non-null object
tweet_source      1994 non-null category
tweet_text        1994 non-null object
tweet_url         1994 non-null object
tweet_picture_predicted 1994 non-null object
tweet_favorites   1994 non-null int64
tweet_retweets    1994 non-null int64
user_followers    1994 non-null int64
dog_stage         1994 non-null category
dog_breed         1686 non-null object
confidence_level   1994 non-null float64
rating_numerator   1993 non-null float64
dogs_count        1994 non-null int64
dog_name          1369 non-null object
dog_gender        862 non-null category
dtypes: category(3), float64(2), int64(4), object(6)
memory usage: 208.4+ KB

```

### 1.0.1 Plot the correlation map to see the relationship between our variables

```

In [5]: f,ax = plt.subplots(figsize=(18, 18))
        sns.heatmap(df_master[['tweet_source', 'tweet_favorites',

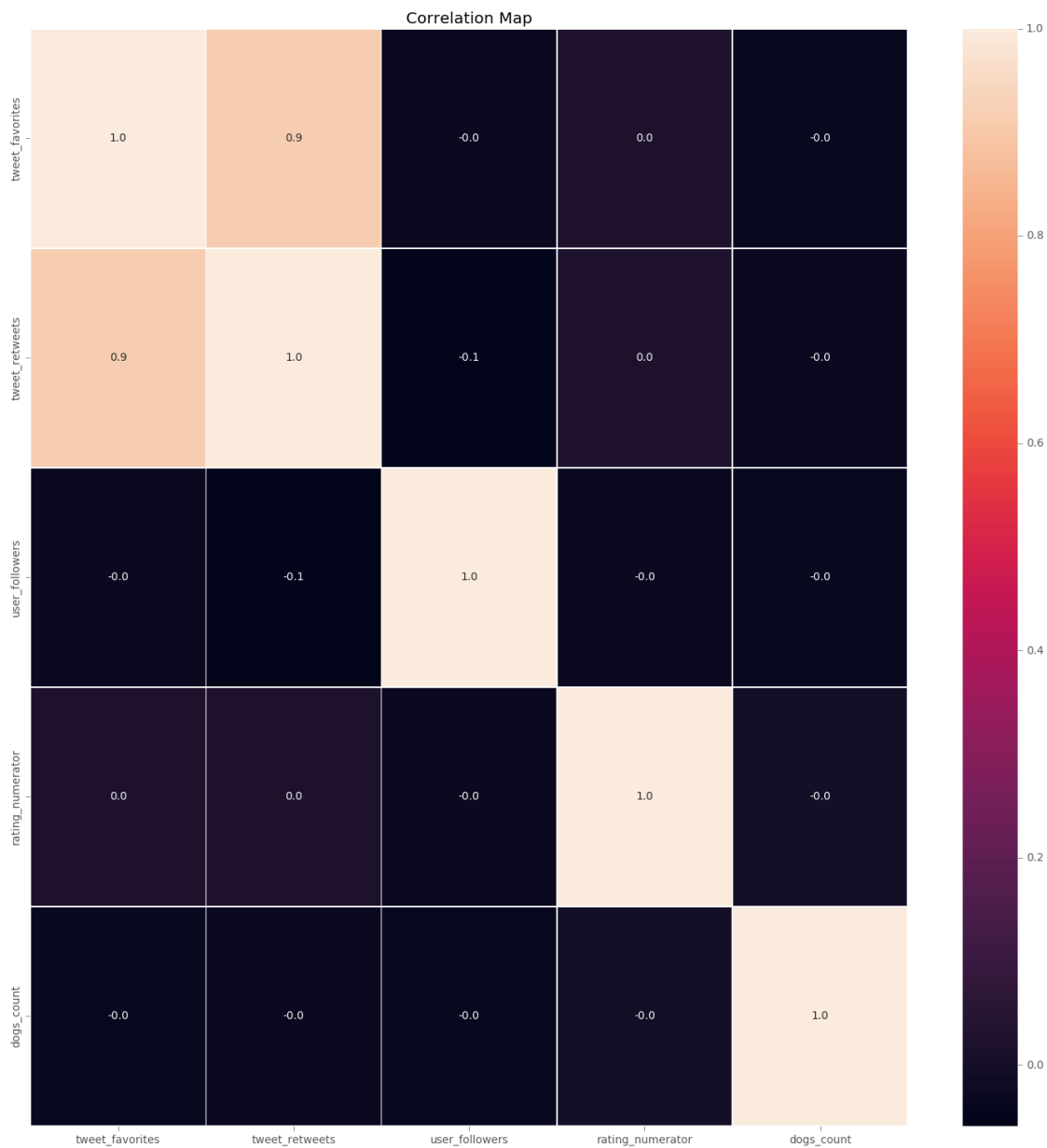
```

```

        'tweet_retweets', 'user_followers',
        'rating_numerator', 'dogs_count']].corr(), annot=True, linewidths=1)
plt.title('Correlation Map')

```

Out[5]: <matplotlib.text.Text at 0xb0661d0>



- The only strong correlation we see here is between tweet\_favorites and tweet\_retweet, this is normal (more favorites mean more retweets)
- User followers and retweet have a weak negative correlation of -0.1 (this seems the opposite of normal prediction)

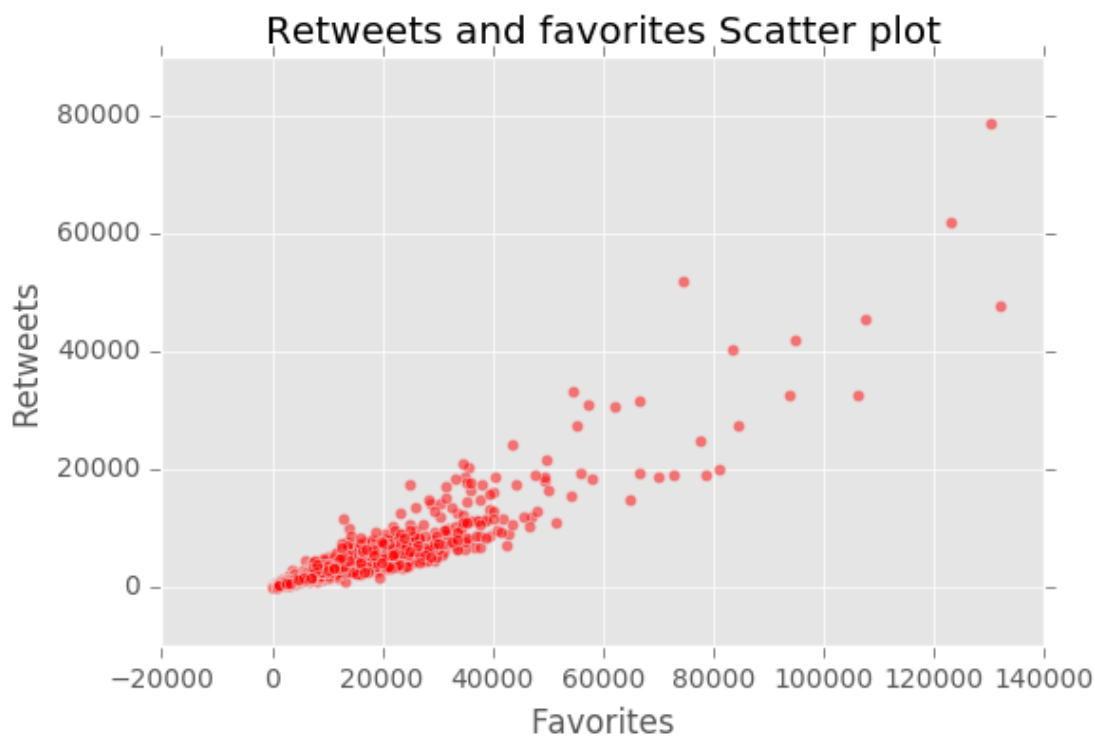
- More dogs in the picture doesn't mean high rating
- Rating don't get affected with any other variable from the ones we plotted

*let's dig more starting with the relation between tweet\_favorites and tweet retweet*

## 1.0.2 tweet\_favorites and tweet\_retweet

```
In [6]: df_master.plot(kind = 'scatter', x = 'tweet_favorites', y = 'tweet_retweets', alpha = 0.1)
plt.xlabel('Favorites')
plt.ylabel('Retweets')
plt.title('Retweets and favorites Scatter plot')
```

```
Out[6]: <matplotlib.text.Text at 0xb7d52b0>
```



- As the correlation map shows if the count of retweet is high the count of favorites go high

```
In [7]: top_retweet_count_url = df_master.tweet_url[df_master.tweet_retweets == max(df_master.tweet_retweets)]
print("The maximum number of retweet is: {}, for the tweet: {}".format(max(df_master.tweet_retweets), top_retweet_count_url[0]))

top_favorites_count_url = df_master.tweet_url[df_master.tweet_favorites == max(df_master.tweet_favorites)]
print("The maximum number of favorites is: {}, for the tweet: {}".format(max(df_master.tweet_favorites), top_favorites_count_url[0]))
```

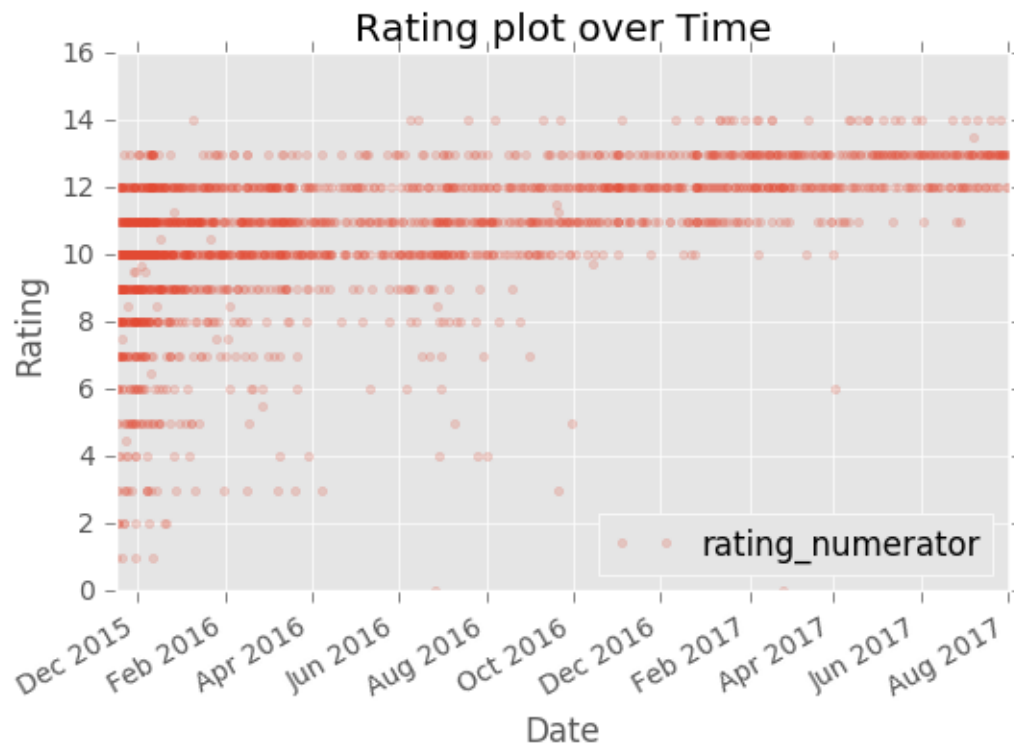
The maximum number of retweet is: 78809, for the tweet: [https://twitter.com/dog\\_rates/status/744](https://twitter.com/dog_rates/status/744)

The maximum number of favorites is: 131903, for the tweet: [https://twitter.com/dog\\_rates/status/744](https://twitter.com/dog_rates/status/744)

### 1.0.3 Rating System

```
In [8]: # Our range will be [0,16] taking of the two outliers (1776 and 420)
df_master.plot(y='rating_numerator', ylim=[0,16], style='.', alpha=.2)
plt.title('Rating plot over Time')
plt.xlabel('Date')
plt.ylabel('Rating')
```

```
Out[8]: <matplotlib.text.Text at 0xb848a20>
```



```
In [9]: df_master[df_master['rating_numerator'] <= 14]['rating_numerator'].describe()
```

```
Out[9]: count      1991.000000
mean         10.550701
std           2.178563
min           0.000000
25%          10.000000
50%          11.000000
75%          12.000000
max           14.000000
Name: rating_numerator, dtype: float64
```

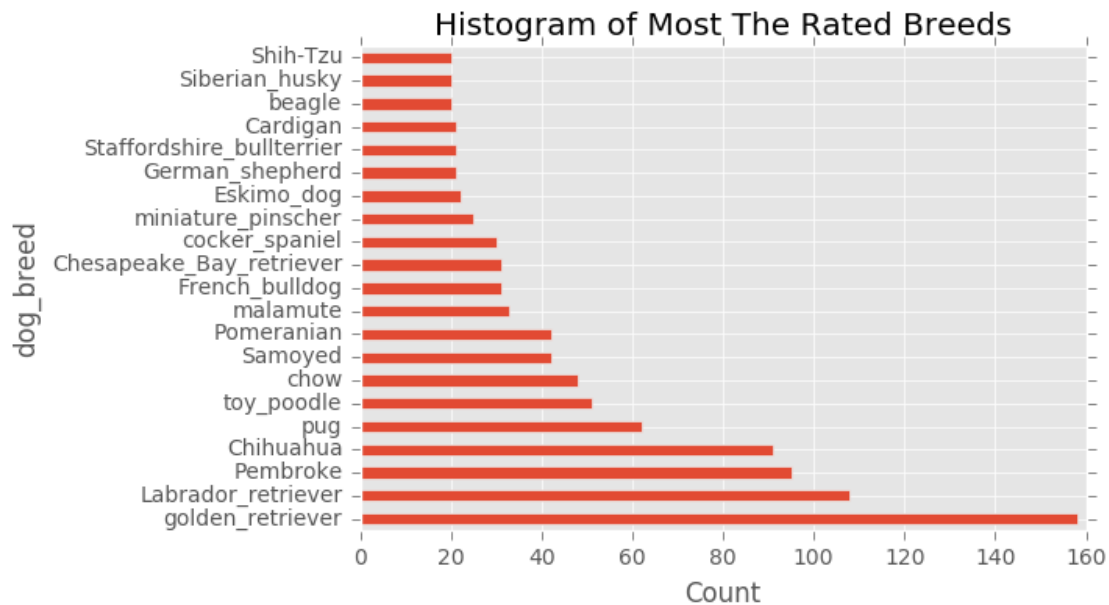
- More than 75% of the data has more than 12/10 as rating
- The page start with small rating than they adopt the system of rating numerator more than the denominator
- [Brent has all the right to get mad](#) (ratings getting higher with no specific reason)

### 1.0.4 Famous Breeds

```
In [11]: # Without specify the length we don't get good result so we will subset our data on the
df_by_breed = df_master.groupby('dog_breed').filter(lambda x: len(x) >= 20)

df_by_breed['dog_breed'].value_counts().plot(kind = 'barh')
plt.title('Histogram of Most The Rated Breeds')
plt.xlabel('Count')
plt.ylabel('dog_breed')
```

Out[11]: <matplotlib.text.Text at 0xbfb8940>

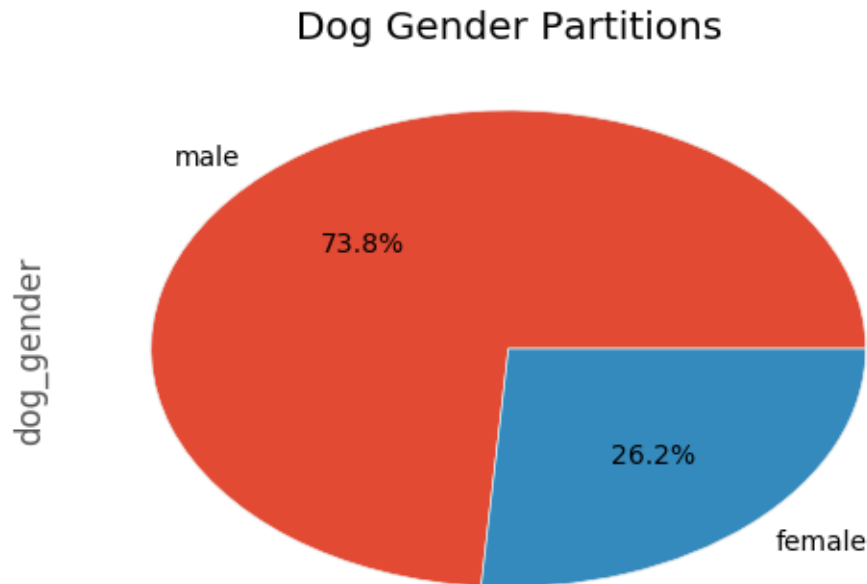


- Top two famous breeds are : Golden\_retriver and Labrador\_retriver according to a meural network that can classify breeds of dogs

### 1.0.5 Famous dog gender

```
In [12]: # Plot the data partitioned by dog gender
df_master[df_master['dog_gender'].notnull()]['dog_gender'].value_counts().plot(kind = '
plt.title('Dog Gender Partitions')
```

Out[12]: <matplotlib.text.Text at 0x978ad30>



```
In [13]: # Which gender had high ratings
df_master[['dog_gender', 'rating_numerator']][df_master.dog_gender.notnull()].groupby('dog_gender').mean()
```

```
Out[13]:
```

dog_gender	rating_numerator
female	11.353333
male	10.652123

- According to our treatment (getting the gender from the text of the tweet) we have male dogs more than female dogs in our dataset, whatever the female rating mean more than the male rating mean

### 1.0.6 Famous dog Stages

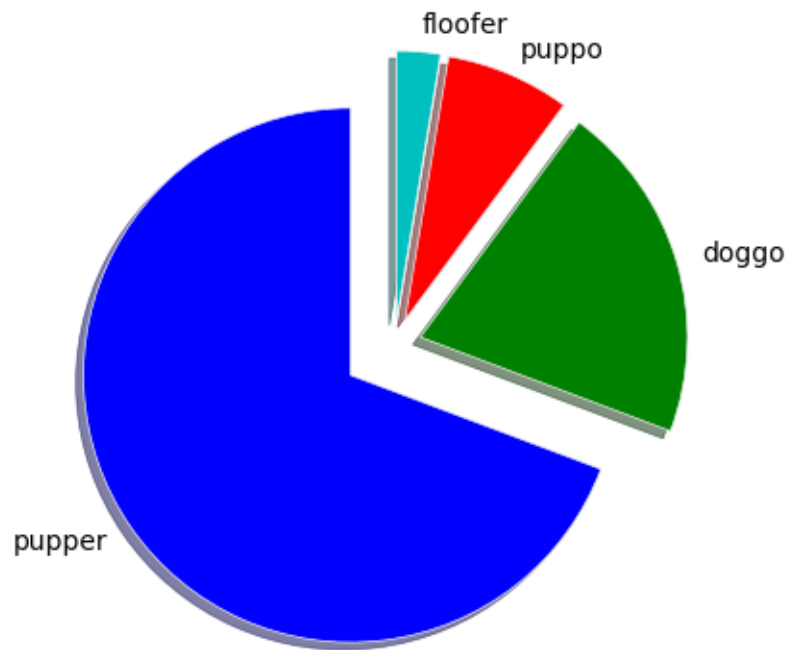
```
In [16]: # Plot the data partitioned by dog stages

dog_stage_count = list(df_master[df_master['dog_stage'] != 'None']['dog_stage'].value_counts())
dog_stages = df_master[df_master['dog_stage'] != 'None']['dog_stage'].value_counts().index
explode = (0.2, 0.1, 0.1, 0.1)

fig1, ax1 = plt.subplots()
ax1.pie(dog_stage_count, explode = explode, labels = dog_stages, shadow = True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
```

```
Out[16]: (-1.1737546111107124,
          1.116516995666861,
```

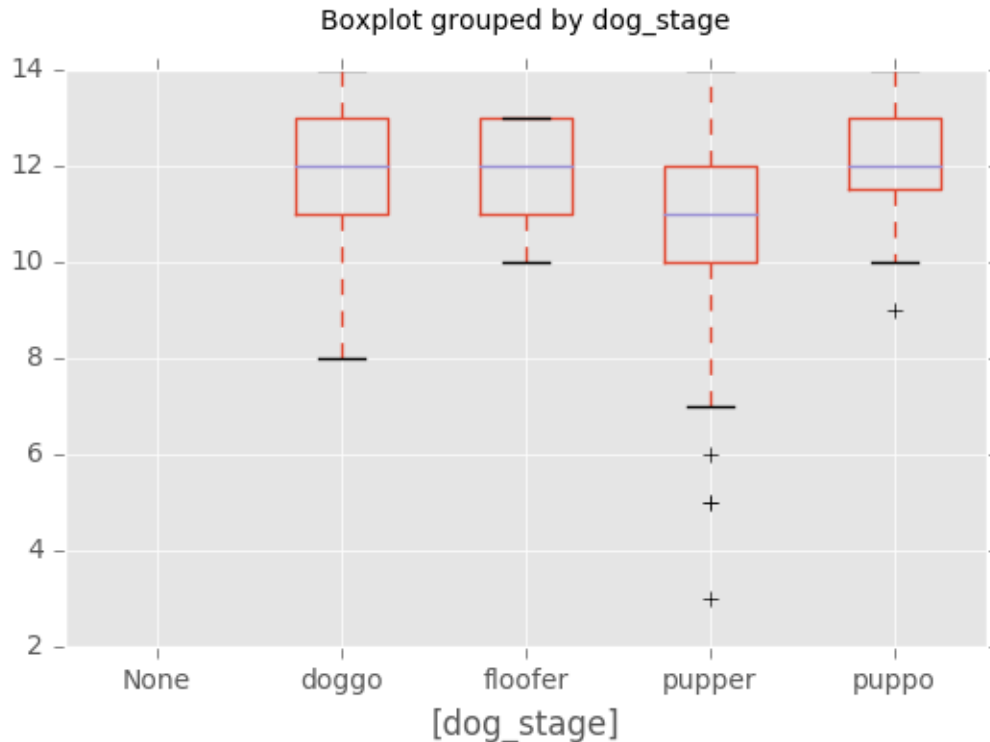
```
-1.1287190256412241,  
1.0996629023294362)
```



```
In [17]: # Plot the dog stages with ratings  
df_master[df_master['dog_stage'] != 'None'].boxplot(column = ['rating_numerator'], by =  
plt.title('')
```

```
Out[17]: <matplotlib.text.Text at 0x9aea7b8>
```





```
In [18]: #df_master[df_master['dog_stage'] == None].groupby('dog_stage')['rating_numerator'].des
df_master[df_master['dog_stage'] != 'None'].groupby('dog_stage')['rating_numerator'].me
```

```
Out[18]: dog_stage
None      NaN
doggo      11.888889
floofer     11.875000
pupper      10.645142
puppo      12.043478
Name: rating_numerator, dtype: float64
```

- Puppies represent the big number of our pie, but it has the lowest mean rating

## 2 Conclusion

The Twitter account WeRateDogs ([@dog\_rates](https://twitter.com/dog\_rates)) is devoted to humorously reviewing pictures of dogs doing adorable poses. Dogs are rated on a scale of one to ten, but are invariably given ratings in excess of the maximum, such as "13/10". It has acquired over 4.50 million followers since its debut.