## **Analysis report**

Once we combine all the cleaned dataframes into one, we have the following variables

```
tweet_id
in_reply_to_status_id
in_reply_to_user_id
timestamp
source
text
expanded_urls
rating_numerator
rating_denominator
name
stage
jpg_url
img_num
dog_species
retweet_count
float 64

1685 non-null float 64

1685 non-null object
1685 non-null object
1685 non-null object
1685 non-null int 64
1685 non-null int 64
1685 non-null object
1685 non-null int 64
```

Of these, the following are quantitative variables

- retweet\_count
- 2. favorite count
- 3. rating\_numerator (discreet)

## Categorical variables

- 1. dog\_species
- 2. name
- 3. stage

Timestamp is also a continuous numeric variable.

First, I looked at the ratings over time. A simple scatter plot as shown in figure 1 can help us look at this relationship. From the plot it is clear that *after 2016/11 most of the ratings were >10, and before that, there were non-negligible ratings less than 10.* However, most of the ratings are generally over 10.

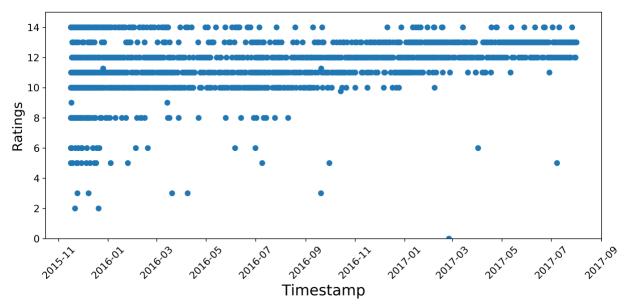


Figure 1. Ratings over Time

Next, I look at the relationship between retweet and favorite counts. There is a clear positive linear relationship. However, we see that generally the higher ratings have higher retweets and higher favorite counts.

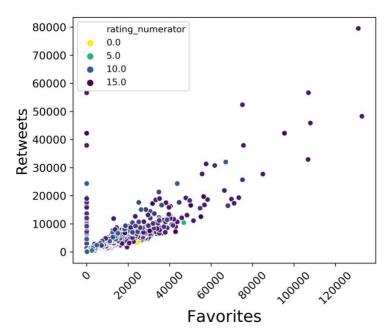


Figure 2. Retweets vs Favorite Counts

After that I analyzed the counts of the ratings and dog stages. As we can see in the ratings plot, most ratings are over 10, and the higher number around 10-12. There are very few ratings less than 10. Analyzing the stage counts, we see the most common stage is pupper followed by doggo, while floofer and blep are the least common stages.

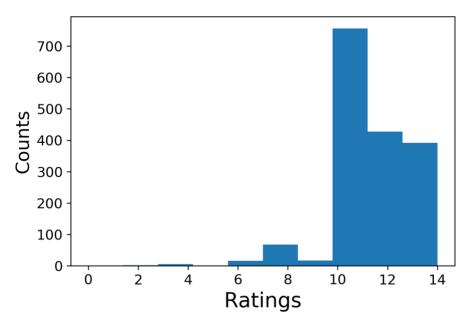


Figure 3. Ratings Count

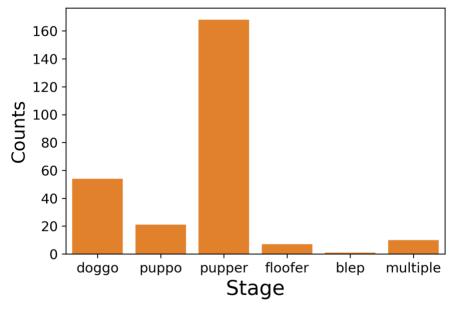


Figure 4. Stage Count

Lastly, to analyze which species and names are most popular, I used word clouds, because there are several names and species, and using count plots would get too cluttered.



Figure 5. Species Word Cloud



Figure 6. Name Word Count