

# Wrangle Report

The purpose of this report is to document and describe my wrangling efforts.

As stated in the Read Me file, we had to obtain data in 3 different ways.

1. We were given a 'csv' file which can resemble data given to us by a colleague or a company.
2. We had to download another file 'image-predictions.tsv' which can be compared to extracted a file from a website.
3. Finally, we had to extract the last piece of information from Twitter themselves. This was facilitated by the Tweepy wrapper available for python.

This is what the files contained:

-twitter-archive-enhanced.csv : a compiled list of approximately 2300 tweets, with their ID's. This table contained dog ratings and data on dog stages.

-image\_predictions.tsv : this is a file that was created by Udacity.com it contains the outputs after each image from X to Y date was ran through a neural network (image recognition algorithm) in an attempt to determine the dog's breed.

-information gathered through Twitter's API: When I ran the script for this section, I obtained JSON objects with 310 keys for every tweet.

Wrangling and cleaning the data:

The goal here was to combine all 3 datasets and create a cleaner table. I had to spend a lot of time investigating the datasets to identify issues. This meant printing out specific results and analyzing specific columns and the corresponding tweet. Below I will explain the algorithms used for some of the tougher issues.

### **Problem 3: Some of the ratings in twitter-archive-enhanced.csv were not accurate.**

I solved this issue by analyzing the text content of the tweets. I separated the text using the `.split('/10')[0]` string method. This separates the string in portions depending on where the `'/10'` (the denominator) is found. Since we are only interested in the numerator here, I only considered the first part with the `[0]` argument. The final step was extracting all the numbers using the 'Regular expression' library which I imported with `'import re'`. Since the result can output multiple numbers, I made sure to select the last element in the list with `[-1]`. This element is the numerator we are looking for because the string section ends right before the `'/10'` for the denominator. The `.apply()` function lets us work on columns row by row in a quick and efficient manner. The `set` function lets me work on a specific set of indexes. I made sure not to include the ratings solved in 'Problem 2' as those included ratings like 84/70 and they did not contain `'/10'`

```
df_clean.loc[set(df_clean.index) - set(ratings_with_denom_not_10), 'rating_numerator'] =  
df_clean.loc[set(df_clean.index) - set(ratings_with_denom_not_10)].split_text.apply(lambda x:  
re.findall(r"\d*\.\d+|\d+", x)).apply(lambda x: ".join(x[-1])).apply(lambda x: float(x))
```

### **Problem 5: The dog\_stages in 'twitter-archive-enhanced.csv' contained many errors:**

I decided here to re-calculate everything by re-extracting the dog stages from the text column.

```
floofer_rows = df_clean[df_clean['text'].str.contains("floof", case=False)].index
```

This was the code used for floofer. The reason I chose floof was because I wanted to consider 'floof', 'floofer' 'floofs' or any other combination of the word 'floof'. Once the index was found I set the value in the column with the `.loc` command. I repeated this for every dog stage.

```
df_clean.loc[floofer_rows, 'floofer'] = 'floofer'
```

### **Problem 8: In certain cases, the algorithm used to predict the dog breed did not predict a dog at all.**

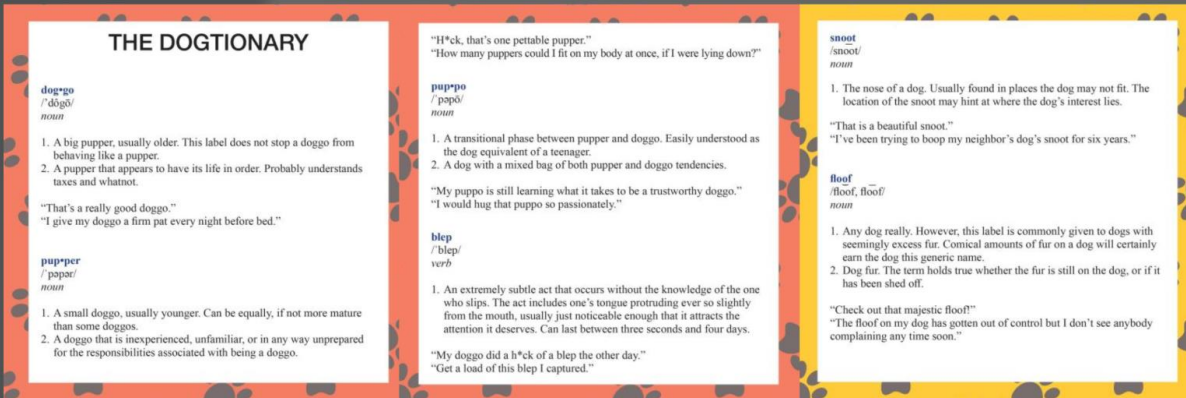
I took the decision here to remove these rows (324 in total) from the table. This was a quality issue and not doing so could have given us erroneous results since we are only interested in ratings associated with dogs. I discovered this was a problem with the following command:

```
image_predictions_clean.query('p1_dog ==0 and p2_dog==0 and p3_dog==0').tweet_id.count()
```

## **Problem 10: Combining all the dataframes and creating a master dataframe:**

I dedicated a small section in the code for this part. I trimmed out the columns which I didn't need anymore to obtain a smaller and cleaner dataframe. I used inner merges to make sure the final database only contained dogs as confirmed by the image\_prediction algorithm. Also, a few tweets have since been deleted so I didn't want to include those for the retweet and favorite/like portion of the work.

See appendix for dog stages:



*The Dogtionary explains the various stages of dog: doggo, pupper, puppo, and floof(er) (via the #WeRateDogs book on Amazon)*

Referenced book:

<https://www.amazon.com/WeRateDogs-Most-Hilarious-Adorable-Youve/dp/1510717145>