Introduction

tweet archive of Twitter user @dog_rates, also known as WeRateDogs. WeRateDogs is a Twitter account that rates people's dogs with a humorous comment about the dog. These ratings almost always have a denominator of 10. The numerators, though? Almost always greater than 10. 11/10, 12/10, 13/10, etc. Why? Because "they're good dogs Brent." WeRateDogs" has over 4 million followers and has received international media coverage

After gathering data from different sources for the required analysis and do the required cleaning of data, this report describe wrangling efforts to clean the data as much as possible

1. Data Gathering

We gathering data from 3 different sources as the following: -

- ✓ Gathering data from downloadable file in the Resources tab which hosted at udacity server, file name twitter_archive_enhanced.csv, this file imported in tabular data using pandas' package at data frame named df_archive_clean
- ✓ <u>Gathering data from the following URL</u>
 https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image-predictions/image-predictions.tsv using requests library, and the data imported into tabular data using pandas' package at data frame named df image predictions
- ✓ <u>Gathering data from Twitter API</u> for each tweet's JSON data using Python's Tweepy library, in fact till now my twitter developer account don't approved, therefore I used the JASON file which hosted at UDACITY server, and the data imported into tabular data using pandas' package at data frame named df-api

2. Data Assessing

After gathering data, we made assessment visually and programmatically for quality and tidiness issues to detect quality issues tidiness issues as the following: -

√ Visual assessment

We made quick visual assessment for the three data frames, using the pandas library for example we use the head () , tail () , sample ()etc for each data frame to detect the issues

√ programmatically assessment

We made programmatically assessment for the three data frames, using the pandas library for example we use value_counts (), nunique(), dtypes()....etc For each column to detect the issues, then clean all issues as per the below table which describe the issue and the action done as a solution

3. Data Cleaning

We do the best effort for cleaning all issues which appeared in the assessment step, we found 10 quality issues and 2 tidiness issues as mentioned in the below table

Serial	Issue	Type of issue	Data frame	Solution
1	Column 'name' has 745 cell None instead of NAN	Quality	df_archive_clean	we replaced all None vlaues by Nan to be compitable in analysis
2	Data type of column 'timestamp' is object instead of datetime	Quality		we convert the type of data in column to timestamp instead of object using built in function
3	Column 'doggo' has 2259 cells , column 'floofer' has 2346 cells , column 'pupper' has 2099 cells and column 'puppo' has 2326 cells None instead of NAN	Quality		we replaced all None vlaues by Nan to be compitable in analysis
4	rating_numerator column data type must be float not int64	Quality	df_archive_clean	we convert the type of data in column to float instead of int64 using built in function to conclude the right rating from the text column

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5	rating_numerator column for dog Almost always greater than 10 , but there are some outlier observation, for exmaple 666 , 143 , 182 , 204 , 961 , 1776 , 172 this values may affect the analysis	Quality		we extracted the correct rating_numerator from 'text' column using python functions
6	rating_denominator for dog have values less than 10 and values more than 10 for some observations, we will replace all values at this column by 10	Quality		we changed all rating_denominator to 10
7	column 'text' has a combined content of tweet text , link and ratings will not used in analysis , therefore we can delete it	Quality		we dropped the column from the data frame after extracting the correct rating_numerator
8	As per Key points in Project Motivation, we need only original ratings (no retweets) that have images, consequently we can delete all rows which mean for Retweet or Reply for rows which have Nan in 'in_reply_to_status_id', 'in_reply_to_user_id', 'retweeted_status_id', 'retweeted_status_user_id', 'retweeted_status_timestamp', will considered original tweet	Quality	df_archive_clean	we dropped the columns('in_reply_to_status_id', 'in_reply_to_user_id' , 'retweeted_status_id',

9	expanded_urls columns have 59 NAN values before deletion of some rows in the previous cleaning steps , we can drop the remaining rows from data fram	Quality		After cleaninig data from serial 1 to 8 , onlt two rows with NaN at expanded_urls column, we dropped this rows from data frame
10	Column 'img_num' not important at analysis	Quality	df_image_predictions	we dropped the column from the data framw due to the useless in analysis
11	name of column 'id' in df_api is diffrent than the two other data frame , the name for the same column in df_archine and df_image_predictions is tweet_id	Quality	df-api	we changed the name of column 'id' in df_api to "tweet_id", to be the same name in the other data frame to be the same column name in merging data frame step
12	Types of dogs during their age in columns 'doggo', 'floofer', 'pupper', 'puppo' may be combined in one column	Tidness	df_archive	we add new column called dog_type and quarying the values from the four columns, then delete the four columns
13	we need only three columns in the analysis 'id', 'favorite_count' and ' retweet_count' only	Tidness	df-api	we modified the data frame to include only the required rows ('id' , 'favorite_count' and ' retweet_count')