## WeRateDogs' Wrangling Report

A full report of my Data Wrangling process on the WeRateDogs' data.

My notebook included these sections:

- 1. Data Wrangling:
  - Gathering the data from multiple sources
  - Assessment of the data
  - Cleaning the issues
- 2. Storing the data
- 3. Analyzing and visualizing

## 1- Data Gathering

Here, I imported the data from different sources using different methods and libraries. Here is a brief description of my data gathering process :

- Twitter Archive Enhanced CSV: I downloaded the data manually from the classroom and then used Pandas library to read the data and store it in a variable as a DataFrame object.
- Image Predictions TSV: I used the Requests library to perform a *GET* request on the resource url for the data provided in the classroom. After writing response's content to a file, I loaded it as a Pandas dataframe
- Tweets JSON: Unfortunately, I couldn't use the Tweepy api for getting the data due to multiple obstacles (limited posts per month, 15 minutes rate limit and limited endpoints) which were all caused by the new access levels of the X api.
  As a result, I manually downloaded the text file containing ison data
  - Secondly, I imported the json data using pandas and assigned it to a DataFrame Lastly, I assigned the required columns of the DataFrame to a variable for usage in the notebook.

## **Data Assessment & Cleaning**

In these two sections, I firstly assessed the data thoroughly using both Visual and Programmatic assessment and made a conclusion for all the issues I noticed in the data

In Data Cleaning, I fixed all issues mentioned in Data Assessment and ensured that I have clean data that can be depended on in analysis.

Below is a table of the addressed issues in Data Assessment and how I fixed them in Data Cleaning

| Issue  | Fix  |
|--|--|
| Presence of duplicated rows ( retweets )                                   | I filtered the rows which are retweets using the 3 conditions I made and dropped them from the data  |
| Missing / Inaccurate dog names   | Since all the inaccurate names were lowercase, I replaced all missing names and lowercase names with 'Unknown' which indicates absence of name in tweet's text |
| Missing Expanded URLs  | Got automatically fixed in issue 1 after removing retweets   |
| ID columns interpreted as integer in all DataFrames                        | Transforming the Tweet_id in the 3 DataFrames into a string and dropping other ID columns  |
| Timestamp isn't in date format   | Transforming the column using pandas   |
| Inaccurate dog ratings   | Re-assign the ratings columns with the extracted rating from the text. In addition to manually fixing some inaccurate ratings.                                 |
| Missing Dog Stages   | It got already fixed in issue 12 where I replaced all dog stages columns with one  |
| Inconsistency of predictions   | Making all predictions in title casing   |
| Underscores in predictions   | Replacing underscores with a space   |
| 66 duplicate images in predictions   | Dropping the 66 duplicates   |
| Missing values in likes / retweets   | Filling with the mean of the column  |
| All dog stages columns are one variable and must be in one column          | Replacing all dog stages columns with one column for the dog stage extracted from the tweet's text   |
| Timestamps can be splitted into other values (Year,month,day and day name) | Using .dt accessor and making 4 new columns for these values. Then, removing the timestamp   |
| Existence of tweet's link inside its text                                  | Filtering the tweet's link and assigning it to its own column, then removing it from the text  |
| The img_num column isn't needed  | Dropping the column from the predictions   |
| All of our DataFrames are on one topic and must be in one dataframe        | Merging the 3 DataFrames on the tweet_id using left method for keeping matches in the archive only   |