



# Wrangle Report

**This document describes the steps taken to wrangle the #WeRateDogs Dataset. The Project is part of Udacity Data Analyst Nanodegree.**

**By Gwiza Bonhomme Maryse**

11.02.2019

---

## Data Wrangling Steps:

In order to wrangle the data, I followed the 3 steps explained in class.

1. Collect the data
2. Assess the data
3. Clean the data

After the mentioned above 3 steps, I could proceed to Data Analysis and visualization. This document is only about Data Wrangling. The Analysis findings and visualization are in the document `act_report.pdf`.

I will explain each the actions taken through each data wrangling steps:

### 1. Collecting the Data

For this project we worked with data from 3 different sources.

- a. The dataset `twitter_archive_enhanced.csv` that we manually downloaded from the Udacity. We used `pandas read_csv` to import the dataset in jupyter notebook. In the file `wrangle_act.ipynb` as requested in the project details. I named the dataframe containing this data `df_archive`.
- b. We also programmatically downloaded the file containing image and breed predictions. The file was downloaded and saved in a file called `image_predictions.tsv`. The file was imported in a pandas dataframe as `image_df`.
- c. The last piece of data came from twitter API. we used the python library Tweepy as recommended in the project details. I first created an account on twitter developer platform. I then created an app. And I could finally get a `consumer_key`, `consumer_secret`, `access_token` and `access_secret` that I used for authentication so that I could download data from twitter API. I made a query to twitter api for data based on the list of `tweet_id` that I collected from `twitter_archive_enhanced.csv` `tweet_id` column. 19 `tweet_ids` failed out of 2356 entries. The data was stored in a file named `tweet_json.txt`. The data collected was in json format. And then stored

in an array `api_data = []` . using `pd.dataframe` converted to a pandas dataframe that we named `api_data_df`. This new dataframe had 32 columns, but we only needed 2 as explained in the project details. Those columns are favorite and retweet counts. This new diminished dataframe name was `twitter_api_df`.

Now since we have the data collected from 3 different sources already collected and imported in pandas dataframe. The next step is to assess the data for quality and tidiness issues.

## 2. Assessing the data

Now we are done collecting our data, we are going to visually and programatically assess our 3 pandas dataframes. The goal of the assessment is to find issues in our data. As specified in the project details, we had to select 8 quality issues and 2 tidiness issues in our data.

We are going to assess the 3 previously collected dataframes:

a. `archive_df`

I first visually assessed this dataframe. I could notice a lot of column with None value. But also NA. I then programatically assessed the dataframe with the command `archive_df.info()` command. The command indicated that most columns like the following had most values marked as non null. This is misleading because it might lead to think that the dataset is more full that it's actually is.

b. `image_df`

With this dataset I also did a visual assessment. Visually, The dataset looked “cleaner ” compared to `archive_df`. I could see that the dataset hold a column named `jpg_url` with an image url for each tweet. Another important information was the breed prediction. In the project motivation, we could see that the first prediction could be trusted with a 95% confidence. That is a very good confidence level so I considered those breed prediction in the analysis.

c. `Twitter_api_df`

For this dataset in order to visually conduct the assessment, I used the command `twitter_api_df.sample(500)`. I ran the command several times so that the values could shift. This is the smallest dataset that we have. We checked for duplicates. Null values. And we used `pandas info()` to conduct programmatic assessment.

In addition to the steps described above, I also used `pandas shape()` to compare the 3 datasets Sizes. I took a screenshots that is below. It was somehow expected that all 3 datasets wouldn't hold the exact same of tweets. This is a tidiness issue. And the cleaned dataset to be used for analysis should hold the same tweets.

```

In [31]: image_df.shape
Out[31]: (2075, 12)

In [32]: twitter_api_df.shape
Out[32]: (2337, 3)

In [33]: df_archive.shape
Out[33]: (2356, 17)

```

## Conclusions on Assessment:

### Data Quality Issues:

1. In the twitter archive file, (from the csv) I doubt that 745 dogs's name is None. I would tend to consider this as missing value. this is a quality issue and should be replaced by NA for missing value.
2. The columns doggo , floofer, pupper and puppo which are dogs stage have the value None (python equivalent of NA). which is an issue because when doing programmatic assessment with pandas, its shows no null values
3. `in_reply_to_status_id` with 78 non-null float64 has too many missing values, it would be hard to do any analysis with so many missing values.
4. `in_reply_to_user_id` with only 78 non-null float64 has too many missing values, it would be hard to do any analysis with this column, to drop

5. `retweeted_status_user_id` has too many missing values 181 non-null float64
6. `retweeted_status_timestamp` has too many missing values 181 non-null object
7. `retweeted_status_timestamp` with only 181 non-null object has too many missing values to be useful. The column will be dropped.
8. 19 `tweet_id` failed while taking data from the twitter api.
9. I also doubt that 55 dogs are really called "a". This might be a typo.

#### Tidiness Issues:

1. rename the name in the columns for image predictions. from `p1` to `breed_prediction_1`
2. merge the archive dataset and api dataset based on `tweet_ids`
3. move the image and `breed_prediction` columns from `image_df` to `new_df`

### 3. Cleaning the data

Each time, I will address the issue and then test that the solutions I applied actually worked.

#### Fixing Quality Issues:

To avoid unnecessary lengthy explanations, when possible issues identified in the previous sections are combined in the cleaning phase below:

1. In the `df_archive` file, in the column name, replace value `None` by `NA`. In this way, while performing programmatic assessment, we won't find that the most common name in the dataset is "None". This `None` might have represented python `None`.
2. The columns `doggo`, `floofer`, `pupper` and `puppo` which are dogs stage have the value `None`. I am going to replace `None` with `NA`. I used `pandas replace()`. `Pandas info()` was used to verify the value in each of the columns.
3. Deleting the ids that were present in the `twitter_archive_enhanced.csv` but not in twitter API. It is not necessary to keep those ids for the analysis. I created a new dataframe called `df_archive_new`. In that dataframe I stored `tweet_ids` from `twitter_archive_enhanced.csv` which were successfully downloaded from twitter

API. To test the solution I subtracted the counts from `df_archive` and `api_df`. As expected the result was 0. Which proves that the solution was successful. Now both datasets hold the same tweets.

4. In `df_archive`, 55 dog's name were "a", which seems very unlikely. And seemed more like a typo. So I replaced a by NA for missing value. I tested the solution by making a query to find all names "a" and of course the query came back empty. Which is what we wanted

## Fixing Tidiness Issues:

1. In `image_df`, the column name `p1` is not saying much. To make that dataset more tidy and readable, we renamed the column `breed_prediction_1`. We tested this by using the following codes: `image_df.columns.tolist()`. The name `breed_prediction_1` appeared as one of the column. Good.
2. merge the archive dataset and `twitter_api_df` dataset based on common `tweet_ids`. I created `new_df`.
3. move the image and `breed_prediction` columns from `image_df` to `new_df`. The new dataframe created is called `final_df`

I concluded the wrangling efforts by saving `final_df` to `twitter_archive_master.csv` as in the project details. The analysis and insights will be described in a different document, `act_report.pdf`. Thank you