

Reporting: Wrangle Report

Methods used in data wrangling for this analysis are gathering, assessing, and cleaning. This wrangling process was based on data gathered from the Twitter user **WeRateDogs**. Three different datasets were required to be obtained during the gathering stage. These datasets had to be imported to the workspace (Jupyter Notebook) using different methods.

Gathering:

The first dataset was provided on the Udacity project page to be downloaded manually to my local machine and read it to my notebook using the pandas function `.read_csv()`. The second dataset had to be downloaded programmatically using python's requests library with the URL that was provided. After downloading it programmatically, I read it to dataframe specifying the delimiter as a tab because it is a tsv file.

The third dataset had to be obtained using the tweepy library. I created a Twitter developer account, which gave me some access codes to help me query WeRateDogs page for a json file collecting the tweet id, favorite count, and retweet count but unfortunately it didn't work out. So I proceeded to use the json file provided by Udacity, downloading it manually. I created an empty list to hold the three attributes (columns). Using a loop, I run through the json file line by line to add the tweet id, favorite count, and retweet count to the empty list created earlier. After that, I converted the list to a dataframe, using the pandas function and specifying the column names.

Assessing and Cleaning:

After gathering all the datasets, I assessed all three individually manually, and programmatically. In the process of assessing, I looked out for quality and tidiness issues. I discovered that they all had quality issues by only two had tidiness issues.

Quality

| Assessing | Cleaning |
|---|---|
| Tweet ids in all data frames were integers instead of it being a string. | Convert the column <code>tweet_id</code> to strings |
| There are columns in both <code>twitter_archive</code> and <code>image_prediction</code> dataframes that have a lot of missing data and are not needed. | Use pandas' <code>.drop()</code> method to drop columns |
| The name column in <code>twitter_archive</code> df does not have a consistent letter casing. | Use the <code>.capitalize()</code> function in pandas to iterate through the whole column to make changes |
| Rename the name column in the <code>twitter_archive</code> to a more specific name to help with column identification | Change column name to <code>dog_name</code> using the <code>rename</code> function. |

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|---|---|
| Values in p1, p2, and p3 are not consistent in terms of letter casing | Change values in columns to lowercase using <code>str.lower()</code> |
| Timestamp column is an object datatype instead of a datetime datatype | Convert the timestamp column to datetime |
| The project description describes the ratings in the data to have a denominator of 10, but the dataset has 18 different denominators. | Change the existing values in <code>rating_denominator</code> column by assigning the column to 10 |
| Twitter_archive df should contain only original tweets. | Remove rows where <code>retweets</code> and <code>replies</code> are not null. This leaves only original tweets in our df. Drop <code>retweets</code> and <code>replies</code> columns. |

Tidiness

| Assessing | Cleaning |
|---|--|
| Doggo, floofer, pupper and puppo should be in one column. | Convert none values in all four columns to NaN. Create a new column <code>dog_type</code> to combine all columns by filling in nan values. |
| The dataframes are related to tweets and should be merged. Favorite count and retweet count should be part of the <code>twitter_archive</code> df which follows the rule each type of observational unit forms a table. | Join the two tables using the merge function with their common attribute <code>'tweet_id'</code> . |

After finishing my assessing and cleaning phase, I saved my final dataframes as a csv file.