wrangle_and_analyse_a_dataset-report

September 6, 2022

1 Wrangle and Analyse WeRateDogs Twitter Data

Udacity alx Data Analyst Nanodegree

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1.1 Introduction

This project invetigates the 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.

The data contains each tweet's retweet count and favorite ("like") count collected from the Twitter API. The data also contains dog image predictions generated from the neural network.

For our analysis, the specific questions being investigated here are:

- What is the distribution of tweets overtime?
- What is the most popular dog name?
- How are retweets and favorites couts related?

```
[1]: # load required libraries
  import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  %matplotlib inline
[2]: ## load dataset
```

```
[2]: ## load dataset

df_master = pd.read_csv('twitter_archive_master.csv')
```

```
[5]: list(df_master.columns)
```

```
'rating_numerator',
'rating_denominator',
'name',
'dog_stage',
'jpg_url',
'img_num',
'p1',
'p1_conf',
'p1_dog',
'p2',
'p2_conf',
'p2_dog',
'p3',
'p3_conf',
'p3_dog',
'retweets_count',
'favorite_count']
```

[4]: df_master.info()

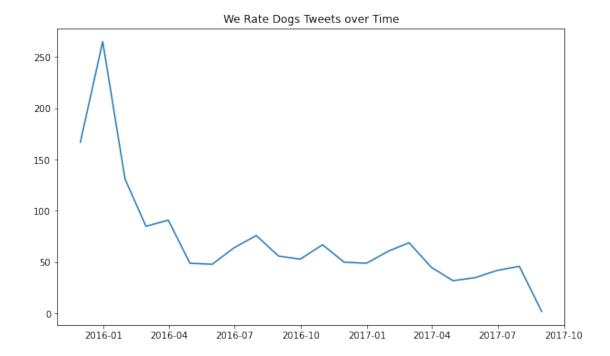
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1583 entries, 0 to 1582
Data columns (total 20 columns):

| # | Column | Non-Null Count | Dtype |
|--|--------------------|----------------|----------|
| | | | |
| 0 | tweet_id | 1583 non-null | int64 |
| 1 | timestamp | 1583 non-null | object |
| 2 | text | 1583 non-null | object |
| 3 | rating_numerator | 1583 non-null | float64 |
| 4 | rating_denominator | 1583 non-null | float64 |
| 5 | name | 1583 non-null | object |
| 6 | dog_stage | 1583 non-null | object |
| 7 | jpg_url | 1583 non-null | object |
| 8 | img_num | 1583 non-null | float64 |
| 9 | p1 | 1583 non-null | object |
| 10 | p1_conf | 1583 non-null | float64 |
| 11 | p1_dog | 1583 non-null | bool |
| 12 | p2 | 1583 non-null | object |
| 13 | p2_conf | 1583 non-null | float64 |
| 14 | p2_dog | 1583 non-null | bool |
| 15 | p3 | 1583 non-null | object |
| 16 | p3_conf | 1583 non-null | float64 |
| | p3_dog | 1583 non-null | bool |
| 18 | retweets_count | 1583 non-null | float64 |
| 19 | favorite_count | 1583 non-null | float64 |
| dtypes: bool(3), float64(8), int64(1), | | | bject(8) |
| memory usage: 215.0+ KB | | | |
| momory abago. 210.0. ID | | | |

```
df_master.describe()
  [3]:
                  tweet id
                            rating numerator
                                              rating denominator
                                                                        img num \
       count.
              1.583000e+03
                                  1583.000000
                                                      1583.000000
                                                                    1583.000000
                                    11.984075
              7.446632e+17
                                                                       1.216677
       mean
                                                        10.120025
       std
              6.841215e+16
                                    44.683597
                                                         4.150396
                                                                       0.585691
      min
              6.660519e+17
                                     1.000000
                                                         2.000000
                                                                       1.000000
       25%
              6.805966e+17
                                    10.000000
                                                        10.000000
                                                                       1.000000
       50%
              7.271754e+17
                                    11.000000
                                                        10.000000
                                                                       1.000000
       75%
              8.011151e+17
                                    12,000000
                                                        10,000000
                                                                       1,000000
       max
              8.924206e+17
                                 1776.000000
                                                       170.000000
                                                                       4.000000
                                p2_conf
                                               p3_conf
                                                        retweets_count
                                                                         favorite_count
                  p1_conf
              1583.000000
                           1.583000e+03
                                          1.583000e+03
                                                           1583.000000
                                                                            1583.000000
       count
                 0.603551
                           1.337078e-01
                                          5.960947e-02
                                                           2801.403664
                                                                            9508.711939
       mean
       std
                 0.273676 1.018113e-01
                                          5.145330e-02
                                                           4071.597886
                                                                           11917.426789
      min
                 0.044333
                          1.011300e-08 1.740170e-10
                                                             23.000000
                                                                             107.000000
       25%
                 0.369509 4.977860e-02 1.519165e-02
                                                            735.000000
                                                                            2417.000000
       50%
                 0.605851
                           1.159840e-01
                                          4.859160e-02
                                                           1595.000000
                                                                            4995.000000
       75%
                 0.864201
                           1.978790e-01 9.203645e-02
                                                           3369.000000
                                                                           12261.500000
                 1.000000 4.880140e-01 2.734190e-01
                                                          56625.000000
                                                                          107956.000000
       max
  []:
      What is the distribution of tweets overtime?
[346]: # re-format timestamp
       df_master.timestamp = pd.to_datetime(df_master['timestamp'], format='%Y-%m-%d_
        # group by month
       tweets_by_month = df_master.groupby(pd.Grouper(key = 'timestamp', freq = "M")).

count().reset_index()[['timestamp', 'tweet_id']]

       tweets_by_month.head()
[346]:
                         timestamp
                                    tweet_id
       0 2015-11-30 00:00:00+00:00
                                          167
       1 2015-12-31 00:00:00+00:00
                                          265
       2 2016-01-31 00:00:00+00:00
                                          131
       3 2016-02-29 00:00:00+00:00
                                           85
       4 2016-03-31 00:00:00+00:00
                                           91
[347]: plt.subplots(figsize=(10, 6))
       plt.plot(tweets_by_month.timestamp, tweets_by_month.tweet_id);
       plt.title('We Rate Dogs Tweets over Time');
```

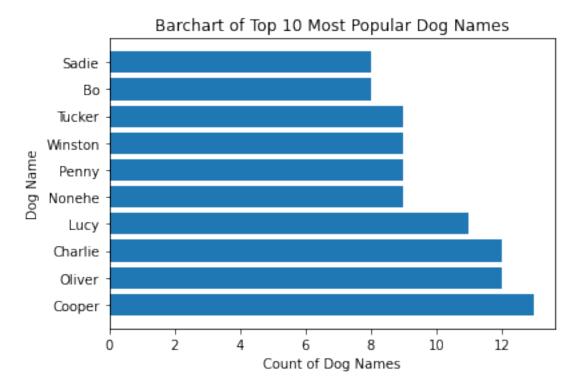


- Although WeRateDogs is very popular, we noticed that for the given data, it peaked in December 2015
- The number of tweets has since fallen until 2017 where our data ends

```
What is the most popular dog name?
[348]: df_dog_name = df_master.groupby("name")["name"].count().
        ⇔sort_values(ascending=False)[0:10]
      df_dog_name
```

```
[348]: name
       Cooper
                  13
       Oliver
                  12
       Charlie
                  12
       Lucy
                  11
       Nonehe
                   9
       Penny
                   9
       Winston
                   9
       Tucker
                   9
       Во
                   8
       Sadie
       Name: name, dtype: int64
[349]: plt.barh(df_dog_name.index, df_dog_name)
       plt.title('Barchart of Top 10 Most Popular Dog Names')
```

```
plt.xlabel('Count of Dog Names')
plt.ylabel('Dog Name');
```



- The Top 10 Dog names are generally as expected
- 70% of the our top 10 dog names appear on 'dogtime.com' top 100 most popular dog names

```
How are retweets and favorites couts related?
```

```
[350]: # scatter plot of profit vs vote average

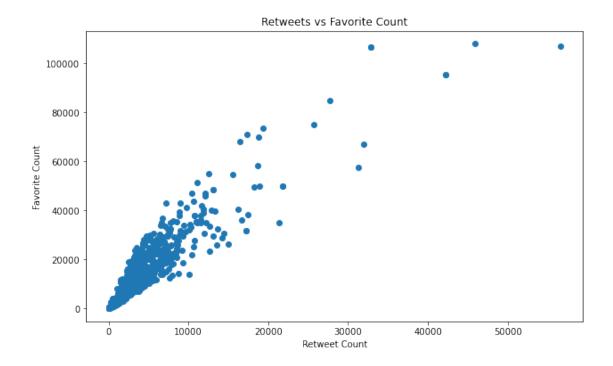
plt.figure(figsize = (10, 6))

plt.scatter(df_master['retweets_count'], df_master['favorite_count'])

plt.title('Retweets vs Favorite Count')

plt.xlabel("Retweet Count")

plt.ylabel('Favorite Count');
```



- The plot shows there is positive correlation between retweets and favorite counts
- This implies favorited tweets are most likely to get more retweets and even more favorites

[]: