# **Analyzing and Visualizing Data**

In [35]:

import matplotlib import matplotlib.pyplot as plt import pandas as pd import seaborn as sns import numpy as np

%matplotlib inline

## Introduction and the History in a Nutshell

The Twitter account, <u>WeRateDogs</u> (<u>https://twitter.com/dog\_rates</u>) was started in 2015 to rate users' dogs and make funny comments about the pictures. The dogs are rated from 0 to 10 but a lot of ratings excess the maximum as an inner joke. The account rapidly became popular and after only two years, in October 2017, the account had about 3.75 million followers. It also created some viral memes such as "They are good dogs, Brent" and described the rated dogs as "puppers" and "doggos".

During my project I analyzed the trends, related dog breeds, dog categories, dog rates and dog names, based on the number of favorites and the number of the account's tweets. I was also curios about the account's own categorical system and analyzed the proportions of the dog stages.

#### I. Possible Correlations

I created a correlation matrix to examine the trends between the rating, favorite count and retweet count. Can rating influence popularity or the number of retweets?

Based on my result, there is no correlation between rating, favorite count and retweet count, which is absolutely makes sense as rating is "only" part of the site's joke.

In [51]:

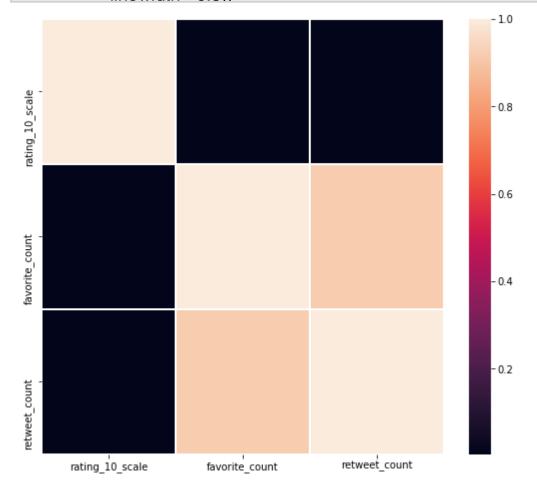
data\_master = pd.read\_csv('twitter\_archive\_master.csv', date\_parser =
pd.datetoo ls.to\_datetime)

/opt/conda/lib/python3.6/site-packages/ipykernel\_launcher.py:1: FutureWarnin g: pandas.core.datetools.to\_datetime is deprecated. Please use pandas.core.to ols.datetimes.to\_datetime instead.

"""Entry point for launching an IPython kernel.

In [55]:

data\_master.timestamp = pd.to\_datetime(data\_master.timestamp)

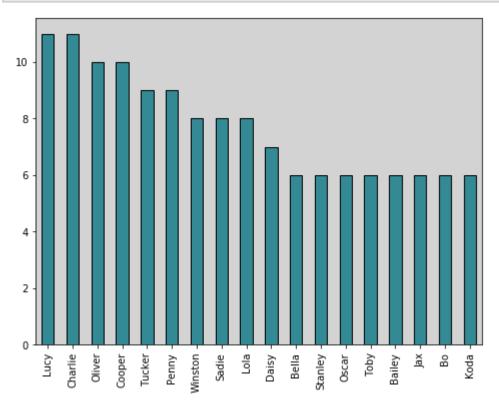


## II. Most Popular Dog Names on the Account

I would like to find the most popular dog names. As rating has not anything to do with popularity, in this case.

The most often tweeted dognames are Charlie, Lucy, Oliver and Cooper. Other names such as Lola, Daisy and Bella are also quiet popular. This trend corresponds my personal experience as I also consider these names popular among dog owners.

```
In [38]: data_name = data_master.groupby('dog_name').filter(lambda x: len(x)
>= 6) fig, ax = plt.subplots(figsize=(8,6))
data_name['dog_name'].value_counts().plot(kind = 'bar', color = "#348A94", ax
= ax, edgecolor = ['Black']*len(data_name))
ax.set_facecolor('#D3D3D3')
plt.show();
```

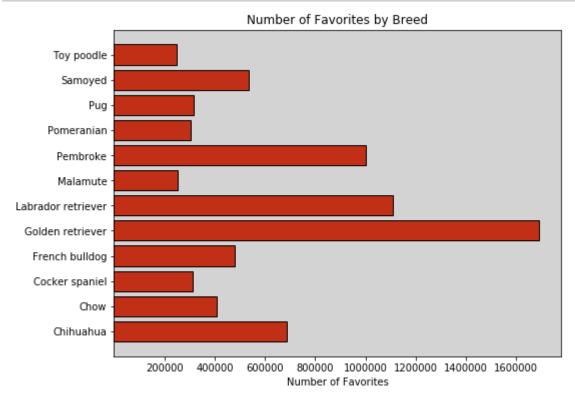


#### **II. Most Beloved Breeds**

After that, I analyzed the relationship between the numbers of favorite and dog breeds to find out which dog breeds are the most beloved by the users. According to my results, golden retrievers are the all-time favorites among the users with more than 1600000 favorites, the second and third breeds are Labrador retrievers (about 1100000 favorites) and Pembroke's (about 1000000 favorites).

```
In [39]: data_breed = data_master.groupby('p1').filter(lambda x: len(x) >= 25)
data_favorite =
data_breed.groupby('p1')['favorite_count'].sum().reset_index() ser_fav =
data_favorite['favorite_count']
ser_breed = data_favorite['p1']

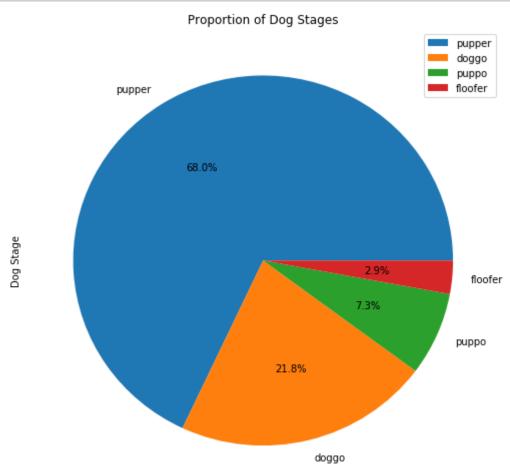
fig, ax = plt.subplots(figsize=(8,6))
fav = plt.barh(ser_breed, ser_fav, color = "#C13016", edgecolor = ['Black']*le
n(ser_breed))
ax.set_facecolor('#D3D3D3')
plt.xlabel('Number of Favorites')
plt.title('Number of Favorites by Breed')
plt.xticks(np.arange(200000, 1800000, 200000))
```



### III. Dog Stages

I also found interesting the unique language and expressions the account used over the times to talk about the dogs. I analyzed the specific "dog stages", excluding the missing values to make a pie chart about the proportions of the categories. The most often used stage is "pupper" with more than 65%, the second one is "doggo" (about 22%). The account owners use the words puppo and floofer in 10% of the cases altogether.

```
In [40]: data_stage = data_master[data_master['dog_stage'] !=
"None"] fig, ax = plt.subplots(figsize=(8.5,8.5))
data_stage['dog_stage'].value_counts().plot(kind = 'pie', ax = ax, label =
'Dog Stage', autopct='%1.1f%%')
plt.title('Proportion of Dog Stages')
plt.legend();
```



## IV. Ratings

I analyzed the portions of the ratings. I decided to use the cut function to make the parts of the pie chart almost equal and better readable. I gave the categories funny names to emphasize that higher rating often means cuter dogs. According to the results, most of the dogs are rated between 0 and 13, only 3% of the dogs are rated higher than 13. However, the highest rate is 1776! The numerators exceed the nominators in 40% of the cases, 40% of the dogs have higher ratings that 10.

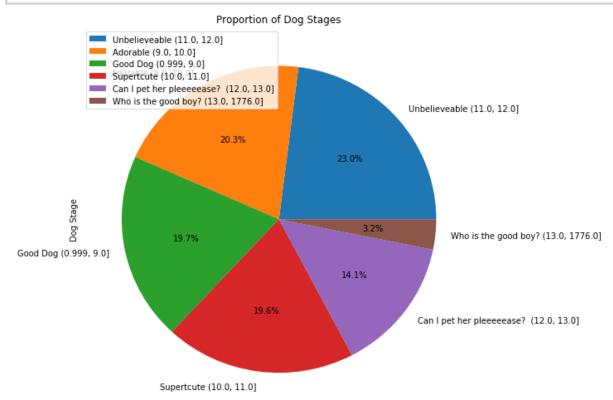
```
In [41]: # Exclude the negative or null values from the dataframe data_cuteness = data_master[data_master['rating_10_scale'] > 0]
```

4/23/2018 act\_report

In [42]: cute\_label = ['Good Dog (0.999, 9.0]', 'Adorable (9.0, 10.0]', 'Supertcute (10.0, 11.0]', 'Unbelieveable (11.0, 12.0]', 'Can I pet her pleeeeease? (12.0, 13.0]', 'Who is the good boy? (13.0, 1776.0]']

cute\_bins = pd.qcut(data\_cuteness['rating\_10\_scale'], 6, labels = cute\_label)

In [43]: fig, ax = plt.subplots(figsize=(8.5,8.5))
cute\_bins.value\_counts().plot(kind = 'pie', ax = ax, label = 'Dog Stage', auto
pct='%1.1f%%')
plt.title('Proportion of Dog Stages')
plt.legend();



## Conclusion

We got some interesting finding related to the Twitter account WeRateDogs. Some conclusions supported our assumptions and correspond our daily experiences, for example golden retrievers are very popular all over the world and a lot of dog owner name his dog Lucy. But there are other results, telling the own unique story of this rapidly growing account - who wouldn't say that dogs are just too good to be strictly rated on a scale 0-10?