task-2

December 22, 2020

```
[1]: import matplotlib.pyplot as plt
import seaborn as sns

plt.style.use("seaborn")
plt.style.use("fivethirtyeight")
```

Loading the data

```
[2]: import json
import pandas as pd

with open("tweets-dataset.json", 'r') as f:
    raw_data = json.load(f)

# Converting to dataframe for easy handling of data
df = pd.json_normalize(raw_data, max_level=1)
```

What can we say about the users who tweeted this hashtag?

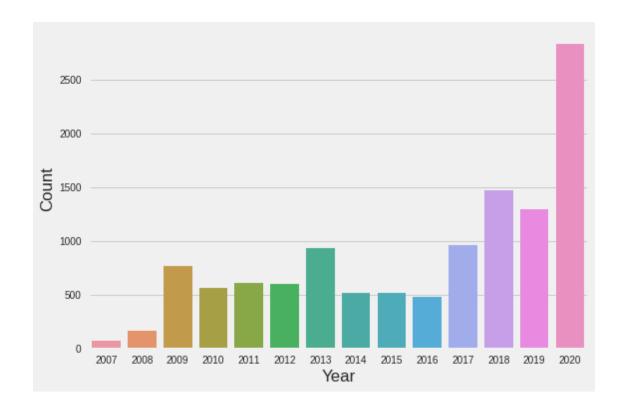
Let's begin by checking all the information about a user that is available to us

```
[3]: print(*raw_data[0]['user'].keys(), sep="\n")
```

```
id
id_str
name
screen_name
location
description
url
entities
protected
followers_count
friends_count
listed_count
created_at
favourites_count
utc_offset
time_zone
```

```
geo_enabled
    verified
    statuses_count
    lang
    contributors enabled
    is_translator
    is translation enabled
    profile_background_color
    profile_background_image_url
    profile_background_image_url_https
    profile_background_tile
    profile_image_url
    profile_image_url_https
    profile_banner_url
    profile_link_color
    profile_sidebar_border_color
    profile_sidebar_fill_color
    profile_text_color
    profile_use_background_image
    has extended profile
    default_profile
    default_profile_image
    following
    follow_request_sent
    notifications
    translator_type
    We will first see when were these accounts created
[4]: import datetime
     # Utility function to get year from the timestamp string
     get_year = lambda x: datetime.datetime.strptime(x, "%a %b %d %H:%M:%S %z %Y").
      →year
[5]: user_years = [*map(get_year, df['user.created_at'])]
     ax = sns.countplot(
         x=user_years
     );
```

ax.set(xlabel="Year", ylabel="Count");



It seems that a good chunk of accounts that were tweeting about #Bitcoin were created in this year

```
[6]: # Percentage of accounts that were made this year

print("Total Accounts:", len(user_years))

print("2020 Accounts:", user_years.count(2020))

print("Percentage : {:.2f}%".format(user_years.count(2020)/len(user_years) *

→100))
```

Total Accounts: 11772 2020 Accounts: 2827 Percentage : 24.01%

Almost a quater of accountes were created just this year

Lets now see the distribution of months for our 2020 accounts

```
[7]: # Get a list of months
months = []
for i in range(1, 13):
    time = datetime.date(2020, i, 1)
    months.append(time.strftime("%B"))

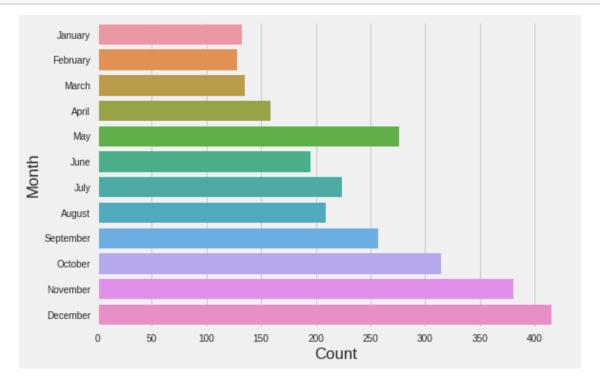
# Utility function to get month
```

```
get_month = lambda x: datetime.datetime.strptime(x, "%a %b %d %H:%M:%S %z %Y").

→month
```

```
[8]: # Filtering out accounts not created in 2020
new_users = [*filter(lambda x: get_year(x) == 2020, df["user.created_at"])]
new_user_months = [*map(get_month, new_users)]
```

```
[9]: # Changed chart orientation for better readabilty of labels
ax = sns.countplot(
    y=new_user_months,
);
ax.set(xlabel="Count", ylabel="Month")
ax.set(yticklabels=months);
```



It seems that out of all the months in 2020, December saw the highest account creation

```
[10]: print("2020 Accounts :", len(new_user_months))

print("Accounts made in last 4 months :", len([*filter(lambda x: x>=9, □

→new_user_months)]))

print("Percentage : {:.2f}%".format(len([*filter(lambda x: □

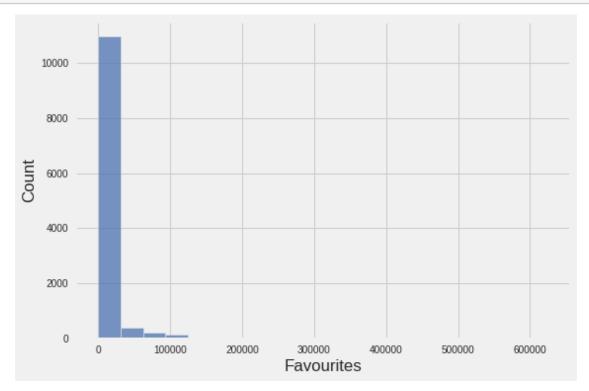
→x>=9, new_user_months)])/len(new_user_months) * 100))
```

2020 Accounts : 2827 Accounts made in last 4 months : 1369

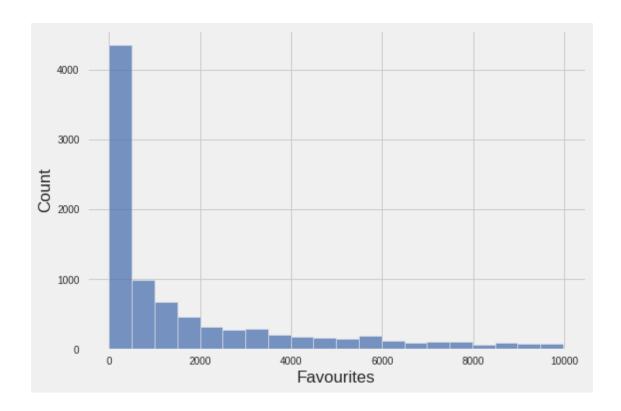
Percentage : 48.43%

Almost half of the accounts created in 2020 were created in just the last four months.

Now lets examine the number of tweets each user has favourited (liked)

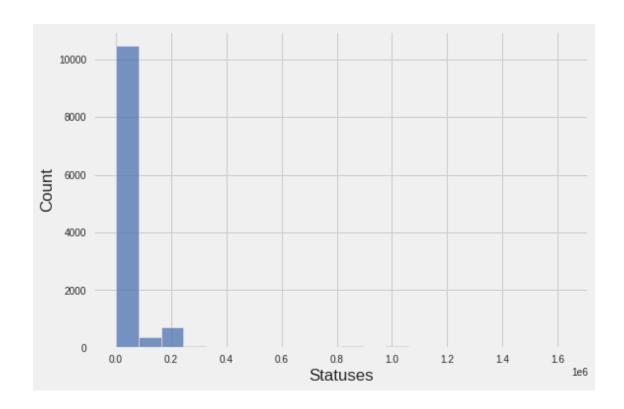


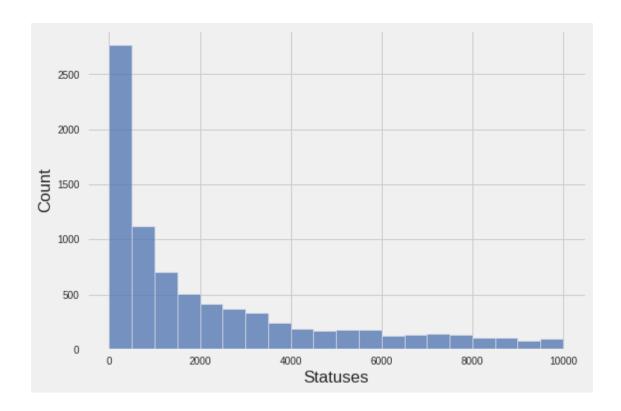
Not very helpful, it just tells us that most of these people has liked less than 50000 tweets. Let's take into consideration, less than 10000 likes



That's better. The likes from users seem to be close to a half normal distribution.

Same approach can be used for the status counts from the user. Twitter count each tweet and retweet as part of counting their statuses





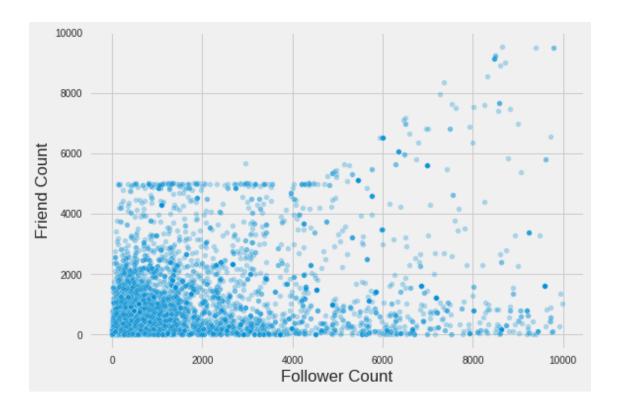
It seems to be the case that most users don't use their accounts a lot after creating them

Can we draw any insights about their user's followers and friends?

The only relevant information available in the dataset are the follower and friends count.

Let's draw a scatter plot using these two attributes with consideration for the limts on the axes

```
[15]: ax = sns.scatterplot(
    x="user.followers_count",
    y="user.friends_count",
    data=df[
        (df["user.followers_count"] <= 10000) &
        (df["user.friends_count"] <= 10000)
    ],
        alpha=0.3,
        s=30
    );
    ax.set(xlabel="Follower Count", ylabel="Friend Count");</pre>
```



It is clear that most of the users are concentrated around the lower end for followers and friends count

An interesting phenomenon to observe is that there is a sharp line at 5000 at the friends count axis for relatively smaller accounts

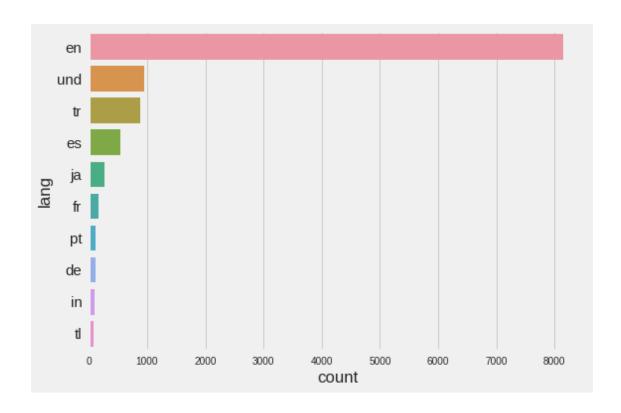
Upon further digging it turns out that twitter imposes a limit of 5000 for following people. This limit is pushed up when an account gains followers

More info here

Can we comment on the language aspect of the corpus collected?

A chart showing the top 10 languages in the collected data

```
[16]: ax = sns.countplot(
    y="lang",
    data=df,
    order=df["lang"].value_counts().iloc[:10].index
)
ax.yaxis.set_tick_params(labelsize=15)
```



As is common for any global social media website, most tweets are in English (en) The second place is occupied by the "undefined" tag i.e. the language could not be detected They are followed by Turkish, Spanish, Japanese, etc.