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1 Final Project Phase 4

1.1 Team Information

- Group 2: Pied Piper
- President Trump Twitter Analysis & Election Simulat
- Alex Bzdel - abzdel@bryant.edu (<mailto:abzdel@bryant.edu>)
- Zach Galante - zgalante@bryant.edu (<mailto:zgalante@bryant.edu>)
- Robert Mitchell - rmitchell2@bryant.edu (<mailto:rmitchell2@bryant.edu>)

2 Abstract

We plan to explore Donald Trump's approval well as tweets regarding coronavirus. We w n conjunction with these tweets. Our goal d sentiment) of tweets with a keyword and ump Administration's coronavirus timeline is analysis and see how they correlate to

Part 1 will deal with Trump's approval rat
Part 2 will produce an election simulation

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2.1 Keywords & hashtags:

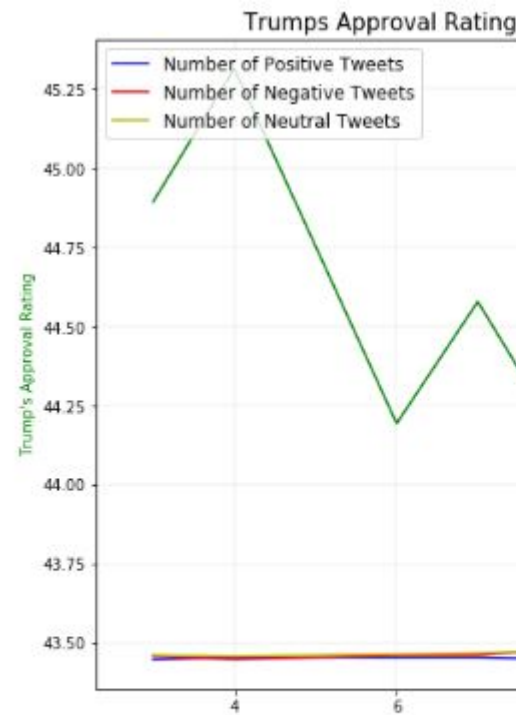
- keywords: Trump, Stimulus Check, \$1200, covid19,
- hashtags: #trump, #stimuluscheck, #money

2.2 Hypotheses

- 1) President Trump's approval rating has i ut the stimulus checks distributed by the hat have highly rated words as determined al rating will be.
- 2) Sentiment ratings for the tweets will s be distributed.
- 3) That Trump's approval rating is going t also be reflected in the results of the 20

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Week 12 - 6.6 Million people file for unemployment, stay at home orders announced

One of our final plots on Trump's Approval Rating vs Stim

```

Trump Wins Penns
Trump Vote Count
-----
Biden Wins Illin
Biden Vote Count
-----
Trump Wins Ohio
Trump Vote Count
-----
Biden Wins Michi
Biden Vote Count
-----
Trump Wins Georg
Trump Vote Count
-----
Biden Wins North
Biden Vote Count
-----
*****Biden win
-----
Trump's final sc
Biden's final sc
-----

```

Final simulation example above

3 Part 1: Tracking Trump's Approval

```
In [146]: ▶ # imports
import numpy as np
import pandas as pd
#import tweepy
#import twython

import json
import csv
import os
import codecs
import time
import gender_guesser.detector as gender
from textblob import TextBlob
from wordcloud import WordCloud, STOPWORDS
import seaborn as sns
import matplotlib.pyplot as plt
import re
%matplotlib inline
```

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3.1 Description of the Data - Part

3.1.1 Scraper(s) used

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In [147]:  *#----- NOTE THAT THIS CELL IS NOT SUPP*

```
import time
from random import seed
from random import randint
# seed random number generator
seed(1)
from datetime import date, datetime, time

def datespan(startDate, endDate, delta=timedelta(days=1)):
    currentDate = startDate
    while currentDate < endDate:
        yield currentDate
        currentDate += delta

text_query = "Stimulus Check"
start_of_period = date(2020, 1, 4)
end_of_period = date(2020, 5, 2)
start_of_week = start_of_period
week_plus_1_start = start_of_period + timedelta(days=7)

for i, week_plus_1_start in enumerate(datespan(start_of_period, end_of_period, delta=timedelta(days=7))):
    print(f"week {i}, start of week: {start_of_week}")
    get_tweet_result(text_query = text_query, until_date = f"{week_plus_1_start}")
    time.sleep(randint(20, 90))

    start_of_week = week_plus_1_start
```

week 0, start of week: 2020-01-04 2020-01-

```
-----
NameError
<ipython-input-147-6a8d3fb3b268> in <module>
    21 for i, week_plus_1_start in enumerate(datespan(start_of_period, end_of_period, delta=timedelta(days=7))):
    22     print(f"week {i}, start of week: {start_of_week}")
--> 23     get_tweet_result(text_query = text_query, until_date = f"{week_plus_1_start}")
    24         until_date = f"{week_plus_1_start}"
    25     time.sleep(randint(20, 90))
```


NameError: name 'get_tweet_result' is not defined

3.1.2 Attempt at pulling user data (failed)

Citation for scraper: <https://github.com/taspinar/twitterscraper>

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In [148]:  `#----- NOTE THAT THIS CELL IS NOT SUPP`
`from twitterscraper.query import query_use`
`import pandas as pd`
`from multiprocessing import Pool`
`import time`
`from IPython.display import display`
`import random`

```
global twitter_user_info
twitter_user_info=[]
```

```
def get_user_info(twitter_user):
    """
```

```
    An example of using the query_user_inf
    :param twitter_user: the twitter user
    :return: twitter_user_data: returns a
    """
```

```
    user_info = query_user_info(user= twit
    twitter_user_data = {}
    twitter_user_data["user"] = user_info.
    twitter_user_data["fullname"] = user_i
    twitter_user_data["location"] = user_i
    twitter_user_data["blog"] = user_info.
    twitter_user_data["date_joined"] = use
    twitter_user_data["id"] = user_info.ic
    twitter_user_data["num_tweets"] = user
    twitter_user_data["following"] = user_
    twitter_user_data["followers"] = user_
    twitter_user_data["likes"] = user_infc
    twitter_user_data["lists"] = user_infc
```

```
    return twitter_user_data
```

```
def main():
```

```
    start = time.time()
    # users = list(df.username.values) # pc
    #users = ['A3Patriot']
    users = ['A3Patriot', 'whaley1212', 'J
    'JLaroc318', 'MamaR130', 'cjstocktc
    'VBrown13245591', 'bradyswenson', '
    'cajunvincent', 'grumpy_idiot', 'fv
    'LMenssen', 'AmericanEarlR', 'Frank
    'mtrwf11', 'Isaac_Visage', None]
    pool = Pool(8)
    for user in pool.map(get_user_info,use
        if True: # if/else meant to fix th
            twitter_user_info.append(user)
            time.sleep(random.randint(1,10
        elif False:
            print('an error has occurred')
            break
```

```
    # while true else break
```

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```
cols=['id','fullname','date_joined','location']
#cols=['id','date_joined','location']

data_frame = pd.DataFrame(twitter_user_info)
# save data_frame as a global variable
data_frame.index.name = "Users"
data_frame.sort_values(by="followers", ascending=False)
elapsed = time.time() - start
print(f"Elapsed time: {elapsed}")
display(data_frame)

if __name__ == '__main__':
    main()
```

```
-----
ModuleNotFoundError
<ipython-input-148-72ca092c163d> in <module>
      1 #----- NOTE THAT THIS CELL IS
----> 2 from twitterscraper.query import query
      3 import pandas as pd
      4 from multiprocessing import Pool
      5 import time
```

ModuleNotFoundError: No module named 'twit

3.1.3 Data Description

Our first analysis involves data with tweets involving key

- * meant to analyze tweet counts and sentiment
- * drawback: no user data available (follow

- **We scraped the following tweet characteristics v**
 - Formatted Date of tweet
 - Author ID (user ID)
 - Username
 - User Location
 - Tweet Text
 - Hashtags
 - Mentions
 - to
 - who the tweet is addressed to - if anyone
 - Tweet URL
 - Replies
 - Retweets
 - Favorites
 - Permalink
- **We then added the following:**

- filtered_text
- retweet_flags
- Polarity_Score
- SubjectivityScore
- sentimentLabel

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3.1.3.1 Third-Party Data

For our third party data, we will be using Donald Tru will primarily be using this to compare the first datas fivethirtyeight.com

3.1.3.2 Data Processing Tasks

For both sets of data, a big step in processing is sentime different keywords in the context of sentiment. Additional depict how sentiments (of both certain keywords and tota creating three sub-DataFrames (one for each sentiment) tweets vs. Trump's Approval Rating

3.1.3.3 When and How Long You Scraped Twitter

- we scraped twitter on 4/30 and 5/1 for an hour each

3.1.4 Load in Data

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```
In [1]: import pandas as pd
import numpy as np
hashtagTrump = pd.read_csv("data_files/Gr
hashtagTrump.columns = ['formatted_date',
new_money = pd.read_csv("data_files/Group
new_money.columns = ['formatted_date', "a
a = new_money.append(hashtagTrump, ignore_
hash_stimCheck = pd.read_csv("data_files/C
hash_stimCheck.columns = ['formatted_date'
b = a.append(hash_stimCheck, ignore_index
coronavirus = pd.read_csv("data_files/Grou
coronavirus.columns = ['formatted_date', '
c = b.append(coronavirus, ignore_index = 1
COVID19 = pd.read_csv("data_files/Group_2
COVID19.columns = ['formatted_date', "auth
d = c.append(COVID19, ignore_index = True)
stim_check = pd.read_csv("data_files/Group
stim_check.columns = ['formatted_date', "a
e = d.append(stim_check, ignore_index = Tr
Trump = pd.read_csv("data_files/Group_2_P
Trump.columns= ['formatted_date', "author_
df = e.append(Trump, ignore_index = True)
df.head()
```

Out[1]:

	formatted_date	author_id	username	
	Fri Apr 10			
0	23:59:57 +0000 2020	2.716171e+08	ForeverMe_MsB	
	Fri Apr 10			
1	23:59:48 +0000 2020	4.970938e+07	_carceexoxo	
	Fri Apr 10			
2	23:59:45 +0000 2020	1.239146e+08	bitcoinization	
	Fri Apr 10			
3	23:59:44 +0000 2020	2.826924e+09	4rdaSquad	
	Fri Apr 10			
4	23:59:39 +0000 2020	5.381119e+08	Felipe__	

3.1.4.1 Approval rating data

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In [150]:

approval_df = pd.read_csv('data_files/Gro...
approval_df.head()

Out[150]:

	president	subgroup	modeldate	startdate	en
0	Donald Trump	Voters	4/30/2020	5/28/2019	4/30
1	Donald Trump	Voters	4/30/2020	5/29/2019	4/30
2	Donald Trump	Voters	4/30/2020	5/30/2019	4/29
3	Donald Trump	Voters	4/30/2020	5/31/2019	4/29
4	Donald Trump	Voters	4/30/2020	6/1/2019	4/29

5 rows × 22 columns

3.2 Data Cleaning

In [151]:

!pip install nltk

Requirement already satisfied: nltk in /usr...
Requirement already satisfied: six in /usr...

In [152]:

deal with notes
!pip install TextBlob

Requirement already satisfied: TextBlob in ...
Requirement already satisfied: nltk>=3.1 i... 5)
Requirement already satisfied: six in /usr... (1.14.0)

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```
In [153]: ► import numpy as np
import pandas as pd
#import tweepy
#import twython

import json
import csv
import os
import codecs
import time
import gender_guesser.detector as gender
from textblob import TextBlob
from wordcloud import WordCloud, STOPWORDS
import seaborn as sns
import matplotlib.pyplot as plt
import re
%matplotlib inline
```

```
In [154]: ► import nltk
nltk.download('stopwords')
nltk.download('punkt')

[nltk_data] Downloading package stopwords
[nltk_data] Package stopwords is already
[nltk_data] Downloading package punkt to /
[nltk_data] Package punkt is already up-
```

Out[154]: True

```
In [155]: ► from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
import re
stop_words = set(stopwords.words('english'))

def remove_url(txt):
    """Replace URLs found in a text string
    (i.e. it will remove the URL from the
    """
    return re.sub("([^\0-9A-Za-z \t])|(\w+:\w+)", " ", txt)

def remove_stop_words(txt):
    """lower case the text, and DROP stop
    words"""
    return " ".join([w for w in word_tokenize(txt) if w not in stop_words])

def preprocess_tweet_text(txt):
    return remove_stop_words(remove_url(txt))
```

```
In [156]: ► df.text.isnull().sum()
df.text = df.text.fillna(' ') # fill nulls
```

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```
In [157]: # Remove URLs and Stop Words
df['filtered_text'] = df.text.apply(preproc

#retweet
df['retweet_flags'] = df.text.str.startswi
```

3.2.1 Create week bins

```
In [158]: approval_df.enddate = pd.to_datetime(approval_df['week_bins'] = approval_df.enc
```

```
In [159]: approval_df = approval_df.where(approval_c
```

```
In [160]: df.formatted_date = pd.to_datetime(df.formatted_date)
df['week_bins'] = df.formatted_date.dt.wei
```

3.2.2 Calculate Polarity

```
In [161]: df['PolarityScore'] = df.filtered_text.apply(lambda x: sentiment_bin(x))
df['SubjectivityScore'] = df.filtered_text.apply(lambda x: subjectivity_bin(x))
```

```
In [162]: def sentiment_bins(data):
    if data <= -0.5:
        grouping = 'Strong-Negative'
    elif data > -0.5 and data < 0.0:
        grouping = 'Mild-Negative'
    elif data > 0.0 and data < 0.5:
        grouping = 'Mild Positive'
    elif data >= 0.5:
        grouping = 'Strong-Positive'
    else:
        grouping = 'Neutral'
    return grouping

df['sentimentLabel'] = df['PolarityScore']
#These bins should be used later for graphing
```

In [163]: ▶

df['sentiment'] = df.PolarityScore.apply(s
df.head()

Out[163]:

	formatted_date	author_id	username	
0	2020-04-10 23:59:57+00:00	2.716171e+08	ForeverMe_MsB	↗
1	2020-04-10 23:59:48+00:00	4.970938e+07	_carceexoxo	↗
2	2020-04-10 23:59:45+00:00	1.239146e+08	bitcoinization	↗
3	2020-04-10 23:59:44+00:00	2.826924e+09	4rdaSquad	↗
4	2020-04-10 23:59:39+00:00	5.381119e+08	Felipe__	↗

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3.3 Workflow Diagrams

First, let's look at our planned workflow for how we will us

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Regex to pull out specific
keywords, assign to new
df

Group by week bins,
visualize with approval
rating (two y-axes)

Here's how we plan to use our third party data about Tru

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3.4 Visual EDA Prep - Part 1

3.4.1 Stimulus check tweets (regexp)

```
In [164]: ► import re
           df.text = df.text.fillna(" ")
           stimuluscheck_df = df[df['text'].str.contains
```

```
In [165]: ► stimuluscheck_df.groupby('week_bins')['te
```

```
Out[165]: array([ 2,  5, 16, 16, 18, 24,
                  747, 980, 963, 1030, 1031, 477])
```


3.4.2 Coronavirus tweets

```
In [166]: corona_df = df[df['text'].str.contains("12
arrcorona = corona_df.groupby('week_bins')

arrcorona = np.insert(arrcorona, 0, 0) #ins
```

3.4.3 Count of tweets for each sentiment

```
In [167]: df_pos = df.where((df['sentimentLabel']=='
df_neg = df.where((df['sentimentLabel']=='
df_neut = df.where((df['sentimentLabel']=='
```

```
In [168]: df_pos = df_pos.dropna(subset=['sentimentL
df_neg = df_neg.dropna(subset=['sentimentL
df_neut = df_neut.dropna(subset=['sentimer
```

```
In [169]: df_pos.groupby('week_bins')['sentimentLabe
df_neg.groupby('week_bins')['sentimentLabe
df_neut.groupby('week_bins')['sentimentLabe
```

```
Out[169]: week_bins
1.0      2
2.0    1101
3.0    1108
4.0    1179
5.0    1164
6.0    1200
7.0    1137
8.0    1138
9.0    1372
10.0   1333
11.0   1967
12.0   2002
13.0   1984
14.0   2069
15.0   2115
16.0   2022
17.0   1597
Name: sentimentLabel, dtype: int64
```

```
In [170]: arrneg = df_neg.groupby('week_bins')['text
arrneg = np.insert(arrneg, 0, 0)

arrpos = df_pos.groupby('week_bins')['text
arrneut = df_neut.groupby('week_bins')['te
```

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3.4.4 Stimulus check by sentiment

```
In [171]: ► stimuluscheck_pos_df = df_pos[df_pos['text']
stimuluscheck_neg_df = df_neg[df_neg['text']
stimuluscheck_neut_df = df_neut[df_neut['t
```

Tweets with positive sentiment are missing week bins 1,

```
In [172]: ► approval_temp_y = approval_df.groupby('wee
approval_temp_y = approval_temp_y.drop(la

approval_temp_x = approval_df.groupby('wee
approval_temp_x = approval_temp_x.drop(la

stimcheckneg = stimuluscheck_neg_df.groupt
stimcheckneg = stimcheckneg.drop(labels=[5

stimcheckneut = stimuluscheck_neut_df.grou
stimcheckneut = stimcheckneut.drop(labels=
```

3.4.5 Coronavirus tweets by sentiment

```
In [173]: ► # search for coronavirus tweets by sentime

corona_pos_df = df_pos[df_pos['text'].str.
corona_neg_df = df_neg[df_neg['text'].str.
corona_neut_df = df_neut[df_neut['text'].s
```

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```
In [174]: corona_pos_df.groupby('week_bins')['sentiment'].mean()
corona_neg_df.groupby('week_bins')['sentiment'].mean()
corona_neut_df.groupby('week_bins')['sentiment'].mean()
```

```
Out[174]: week_bins
2.0      169
3.0      253
4.0      323
5.0      274
6.0      283
7.0      301
8.0      310
9.0      592
10.0     526
11.0     650
12.0     694
13.0     708
14.0     719
15.0     668
16.0     687
17.0     541
Name: sentimentLabel, dtype: int64
```

```
In [175]: corona_temp_y = approval_df.groupby('week_bins')['approval'].mean()
corona_temp_y = corona_temp_y.drop(labels=['1.0', '2.0', '3.0', '4.0', '5.0', '6.0', '7.0', '8.0', '9.0', '10.0', '11.0', '12.0', '13.0', '14.0', '15.0', '16.0', '17.0'])

corona_temp_x = approval_df.groupby('week_bins')['approval'].mean()
corona_temp_x = corona_temp_x.drop(labels=['1.0', '2.0', '3.0', '4.0', '5.0', '6.0', '7.0', '8.0', '9.0', '10.0', '11.0', '12.0', '13.0', '14.0', '15.0', '16.0', '17.0'])
```

3.5 Visuals and EDA - Part 1

3.5.1 Number of Stimulus Check Tweets

First, let's explore how Trump's approval rating has changed over the year

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```
In [176]: ► import numpy as np
import matplotlib.pyplot as plt
x = approval_df.groupby('week_bins')['appr
y1 = approval_df.groupby('week_bins')['app
y2 = stimuluscheck_df.groupby('week_bins')
#y3 = arr1200

fig, ax1 = plt.subplots(figsize=(15,8))

ax1.set_title('Trumps Approval Rating vs.
ax1.grid(color='grey', linestyle='-', line

#duplicate ax1
ax2 = ax1.twinx())

#ax1 will plot Trump's approval rating, ax
ax1.plot(x, y1, 'g-')
ax2.plot(x, y2, 'b-')
#ax2.plot(x, y3, 'r-')

# set legends
ax1.legend(["Trump's Approval Rating"], fc
ax2.legend(["Number of Tweets about Stimul
#ax2.legend(bbox_to_anchor=(0.04, 0.82, 1.
#handlelength=0.1, handletextp

ax1.set_xlabel('Week of Year')
ax1.set_ylabel("Trump's Approval Rating",
ax2.set_ylabel("Number of tweets", color='

# plt.text(6.4,3.3,"""Note on our visualiz
#         null value we replace it with t
#         on that platform's top 10 they
plt.text(14,200, """"Weeks 15 and 16:
Talks of reopening the US""", fontsize = 1

plt.annotate("""Week 12 - 6.6 Million peop
stay at home orders announced""",
            xy = (12,15), xytext = (2,15)
            arrowprops=dict(facecolor='bl

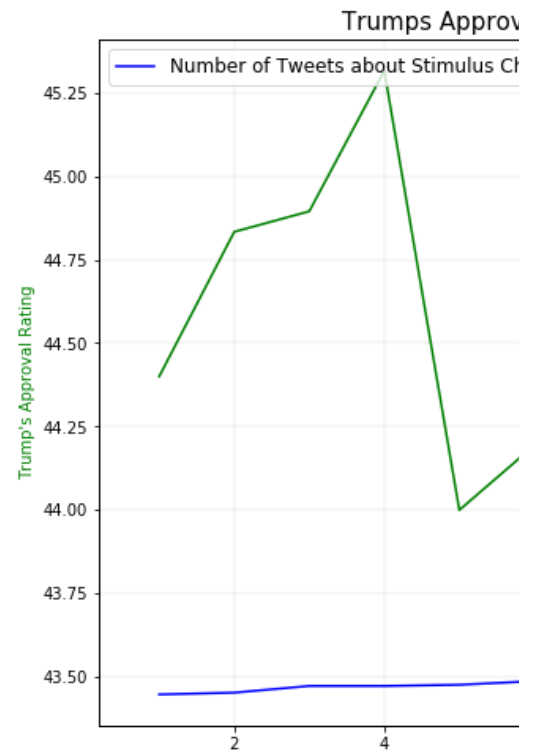
plt.text(12,-200, """"End of week 12:
Trump signs stimulus check bill""", fontsi

plt.show()

# print('notable dates:')
# print('Week 12 - 6.6 Million people file
```

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Week 12 - 6.6 Million people file for unemployment, stay at home orders announced

3.5.2 Number of Coronavirus Tweets

Now, let's look into how his rating changes with number c

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```
In [177]: ► import numpy as np
import matplotlib.pyplot as plt
x = approval_df.groupby('week_bins')['appr
y1 = approval_df.groupby('week_bins')['app
y2 = arrcorona

fig, ax1 = plt.subplots(figsize=(14,7))

ax1.set_title('Trumps Approval Rating vs.
ax1.grid(color='grey', linestyle='-', line

#duplicate ax1
ax2 = ax1.twinx()

#ax1 will plot Trump's approval rating, ax
ax1.plot(x, y1, 'g-')
ax2.plot(x, y2, 'b-')

# set legends
ax1.legend(["Trump's Approval Rating"], fc
ax2.legend(["Number of Tweets about Corona

plt.text(14,200, ""Weeks 15 and 16:
Talks of reopening the US"", fontsize = 1

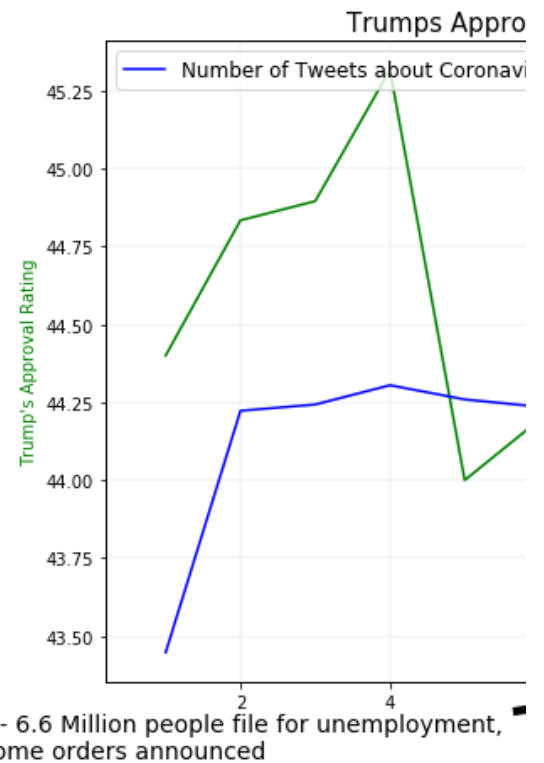
plt.annotate("""Week 12 - 6.6 Million peop
stay at home orders announced""",
            xy = (12,15), xytext = (2,15)
            arrowprops=dict(facecolor='bl

plt.text(12,-200, ""End of week 12:
Trump signs stimulus check bill"", fontsi

ax1.set_xlabel('Week of Year')
ax1.set_ylabel("Trump's Approval Rating",
ax2.set_ylabel("Number of tweets", color='
#ax.legend((y1, y2), ('label1', 'label2'))
plt.show()
```

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3.5.3 Number of Total Tweets by Sentime

These give some interesting insights into how these figur how this plot changes when we break it down by sentime

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```
In [178]: ▶ import numpy as np
import matplotlib.pyplot as plt
x = approval_df.groupby('week_bins')['appr
y1 = approval_df.groupby('week_bins')['app
y2 = arrpos
y3 = arrneg
y4 = arrneut

fig, ax1 = plt.subplots(figsize=(15,8))

ax1.set_title('Trumps Approval Rating vs.
ax1.grid(color='grey', linestyle='-', line

#duplicate ax1
ax2 = ax1.twinx()

#ax1 will plot Trump's approval rating, ax
ax1.plot(x, y1, 'g-')
ax2.plot(x, y2, 'b-')
ax2.plot(x, y3, 'r-')
ax2.plot(x, y4, 'y-')

# set legends
ax1.legend(["Trump's Approval Rating"], fc
ax2.legend(["Number of Positive Tweets", '

plt.text(14,300, ""Weeks 15 and 16:
Talks of reopening the US"", fontsize = 1

plt.annotate("""Week 12 - 6.6 Million peop
stay at home orders announced""",
            xy = (12,15), xytext = (2,15)
            arrowprops=dict(facecolor='bl

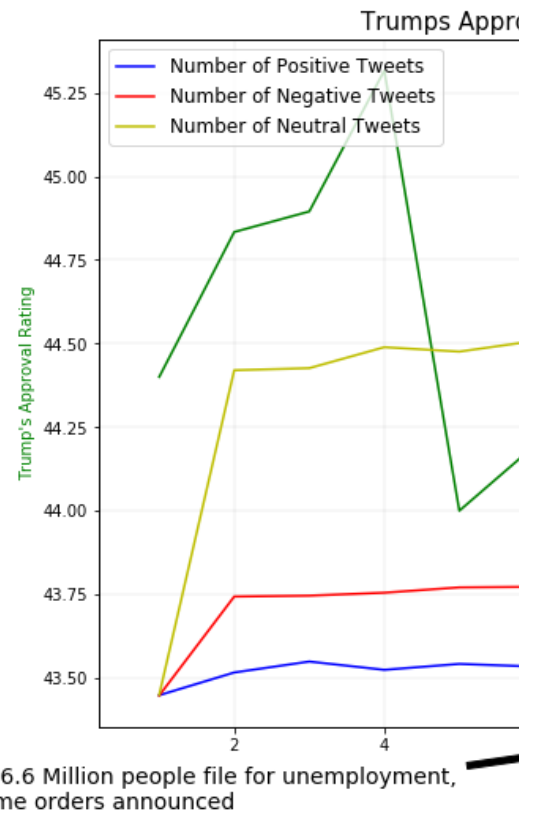
plt.text(12,-400, ""End of week 12:
Trump signs stimulus check bill"", fontsi

ax1.set_xlabel('Week of Year')
ax1.set_ylabel("Trump's Approval Rating",
ax2.set_ylabel("Number of tweets", color='
#ax.legend((y1, y2, y3), ('label1', 'label

plt.show()
```


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3.5.4 Number of Stimulus Check Tweets

Let's take our positive, negative, and neutral tweets and :

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```
In [179]: ► # import numpy as np
import matplotlib.pyplot as plt
x = approval_temp_x
y1 = approval_temp_y
y2 = stimuluscheck_pos_df.groupby('week_bi
y3 = stimcheckneg
y4 = stimcheckneut

fig, ax1 = plt.subplots(figsize=(15,8))

ax1.set_title('Trumps Approval Rating vs.
ax1.grid(color='grey', linestyle='-', line

#duplicate ax1
ax2 = ax1.twinx()

#ax1 will plot Trump's approval rating, ax
ax1.plot(x, y1, 'g-')
ax2.plot(x, y2, 'b-') # POSITIVE TWEETS
ax2.plot(x, y3, 'r-') # NEGATIVE TWEETS
ax2.plot(x, y4, 'y-') # NEUTRAL TWEETS

# set legends
ax1.legend(["Trump's Approval Rating"], fc
ax2.legend(["Number of Positive Tweets", '

plt.text(14,80, ""Weeks 15 and 16:
Talks of reopening the US"", fontsize = 1

plt.annotate("""Week 12 - 6.6 Million peop
stay at home orders announced""",
            xy = (12,15), xytext = (2,15)
            arrowprops=dict(facecolor='bl

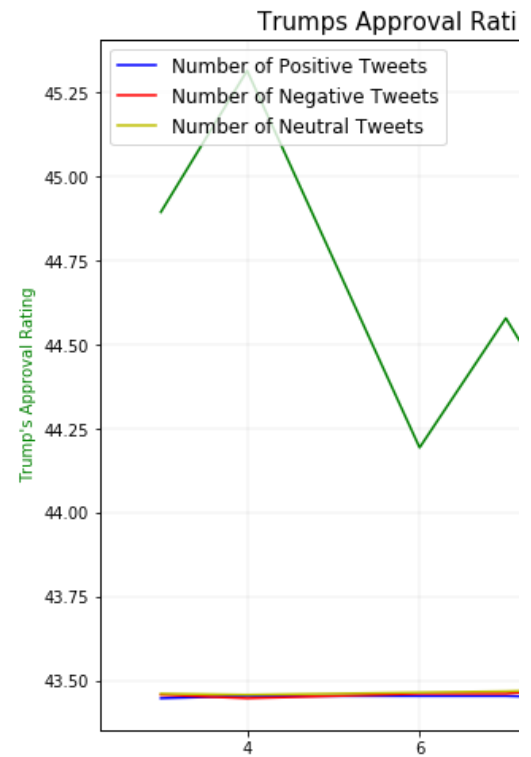
plt.text(12,-150, ""End of week 12:
Trump signs stimulus check bill"", fontsi

ax1.set_xlabel('Week of Year')
ax1.set_ylabel("Trump's Approval Rating",
ax2.set_ylabel("Number of tweets", color='

plt.show()
```

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Week 12 - 6.6 Million people file for unemployment, stay at home orders announced

3.5.5 Number of Coronavirus Tweets by

Let's take our positive, negative, and neutral tweets and

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```
In [180]:  # import numpy as np
import matplotlib.pyplot as plt
x = corona_temp_x
y1 = corona_temp_y
y2 = corona_pos_df.groupby('week_bins')['s
y3 = corona_neg_df.groupby('week_bins')['s
y4 = corona_neut_df.groupby('week_bins')['s

fig, ax1 = plt.subplots(figsize=(15,8))

ax1.set_title('Trumps Approval Rating vs.
ax1.grid(color='grey', linestyle='-', line

#duplicate ax1
ax2 = ax1.twinx()

#ax1 will plot Trump's approval rating, ax
ax1.plot(x, y1, 'g-')
ax2.plot(x, y2, 'b-') # POSITIVE TWEETS
ax2.plot(x, y3, 'r-') # NEGATIVE TWEETS
ax2.plot(x, y4, 'y-') # NEUTRAL TWEETS

# set legends
ax1.legend(["Trump's Approval Rating"], fc
ax2.legend(["Number of Positive Tweets", '

ax1.set_xlabel('Week of Year')
ax1.set_ylabel("Trump's Approval Rating",
ax2.set_ylabel("Number of tweets", color='

plt.text(14,115, ""Weeks 15 and 16:
Talks of reopening the US"", fontsize = 1

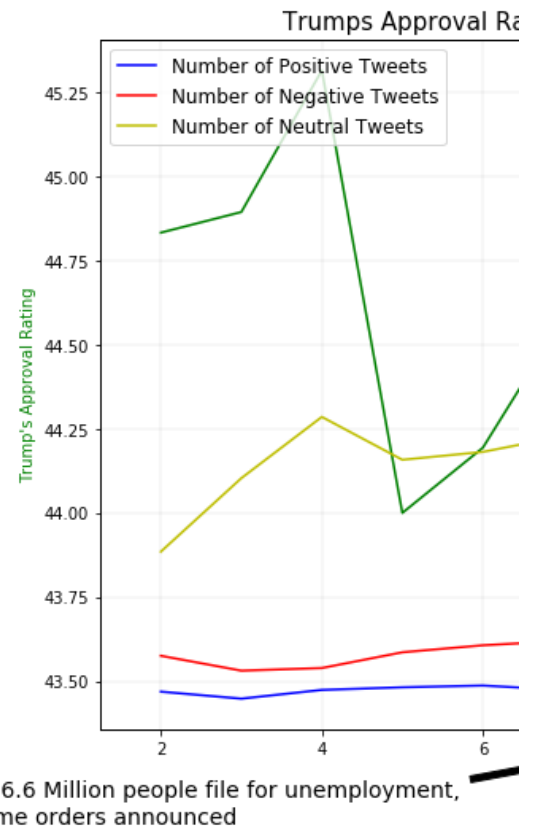
plt.annotate("""Week 12 - 6.6 Million peop
stay at home orders announced"",
            xy = (12,15), xytext = (2,15)
            arrowprops=dict(facecolor='bl

plt.text(12,-150, ""End of week 12:
Trump signs stimulus check bill"", fontsi

plt.show()
```

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Further analysis of graph will be discussed in the "discus

4 Part 2: Predicting the 2020

4.1 Description Of The Data - Par

4.2 Scraper(s) Used

For this part of the project, we scraped for the same keyw users on current tweets.

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```
In [181]: ► from twython import TwythonStreamer
import csv
import json
import codecs
import time
from random import seed
from random import randint
# seed random number generator
seed(1)

tweets_filename = "twitter_output_Trump"

consumer_key = "ND4wRUZRSgRS1NBtb9vRbm96v"
consumer_secret = "z0zxb6Q0bq4AUU2bA8BBclr
access_token = "1250793114274598913-1mA5Lz
access_token_secret = "cqNiCEcr50sbS5G1t3l

tweetabbr = []

# In this session we are using the Twitter
AllPDs = ['561106229', '34296669', '974277
          '35871927', '304847225']
# Filter out unwanted data for the CSV fil
def process_tweet(tweet):
    d = {}
    d['hashtags'] = [hashtag['text'] for hashtag in tweet['hashtags']]
    d['id'] = tweet['id']
    d['text'] = tweet['text']
    d['name'] = tweet['user']['name']
    d['user'] = tweet['user']['screen_name']
    d['user_loc'] = tweet['user']['location']
    d['user_desc'] = tweet['user']['description']
    d['user_followers'] = tweet['user']['followers_count']
    d['user_friends'] = tweet['user']['friends_count']
    d['user_listed'] = tweet['user']['listed_count']
    d['user_created'] = tweet['user']['created_at']
    d['user_favs'] = tweet['user']['favorite_count']
    d['user_statuses'] = tweet['user']['statuses_count']

    return d

# Create a class that inherits TwythonStreamer
class MyStreamer(TwythonStreamer):

    # Received data
    def on_success(self, data):

        # Save full JSON to file
        # TODO : save properly so we can l
        # A tweet JSON record per line
        with open(f'{tweets_filename}.json', 'a') as jsonfile:
            json.dump(data, jsonfile)
            jsonfile.write("\n")

        # Only save tweets in English
```

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```

if data['lang'] == 'en':
    tweet_data = process_tweet(data)
    self.save_to_csv(tweet_data)

# Problem with the API
def on_error(self, status_code, data):
    print(status_code, data)
    self.disconnect()

# Save each tweet to csv file
def save_to_csv(self, tweet):
    with open(f'{tweets_filename}.csv')
        writer = csv.writer(file)
        writer.writerow(list(tweet.values()))

while True:
    try:
        # Instantiate from our streaming c
        stream = MyStreamer(consumer_key,
                            access_token, access_t

        # Start the stream - this would co
        # There are online tools to get th
        # stream.statuses.filter(follow=17

        # Start the stream - this would co
        #stream.statuses.filter(follow=All

        # Start the stream - This stream l
        #stream.statuses.filter(track='@V
        stream.statuses.filter(track='Trun
        #track="Trump, Stimulus Check,stimulus che
    except (KeyboardInterrupt):
        print("Exiting")
        break
    except Exception as e:
        print("error - sleeping " + str(e))
        time.sleep(randint(30, 90)) #suspe
        continue

```

```

File "<ipython-input-181-c91607ce482b>",
    stream.statuses.filter(track='Trump',
    timuluscheck', '#money' )

```

SyntaxError: positional argument follows k

4.3 Data Description

Our second analysis involves data scraped from twitter A

- We scraped the following tweet characteristics v
 - Hashtags

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```
In [183]: ► length = len(df)
new_length = len(df.drop_duplicates())
print(f" The current length of our Datafra
print(f" The length of our Dataframe after
```

The current length of our Dataframe is 28
The length of our Dataframe after droppin

5 Simulation Data Cleaning/

Importing libraries and functions given to us from Profess

```
In [184]: ► # Remove URLs and Stop Words
df['filtered_text'] = df.Tweet_Text.apply(

#retweet
df['retweet_flags'] = df.Tweet_Text.str.st

# TODO Add code analyse URLs, their domair
```

```
In [185]: ► # deal with notes
!pip install TextBlob
```

Requirement already satisfied: TextBlob in
Requirement already satisfied: nltk>=3.1 i
5)
Requirement already satisfied: six in /usr
(1.14.0)

```
In [186]: ► from textblob import TextBlob
df['PolarityScore'] = df.filtered_text.app
df['SubjectivityScore'] = df.filtered_text
```

```
In [187]: ► def sentiment_bins(data):
    if data <= -0.5:
        grouping = 'Strong-Negative'
    elif data > -0.5 and data < 0.0:
        grouping = 'Mild-Negative'
    elif data > 0.0 and data < 0.5:
        grouping = 'Mild Positive'
    elif data >= 0.5:
        grouping = 'Strong-Positive'
    else:
        grouping = 'Neutral'
    return grouping

df['sentimentLabel'] = df['PolarityScore']
```

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In [188]:

user locations
df.User_Location.value_counts(dropna = True)

Out[188]:

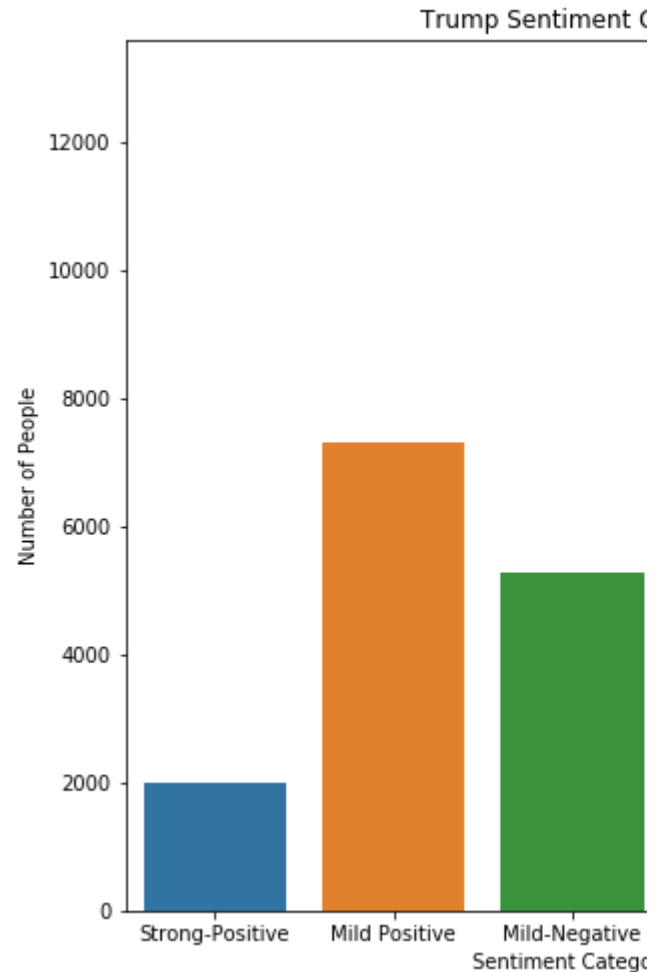
United States716
USA253
California, USA253
Texas, USA172
India148
...
Time is a man made concept!!!1
From NYC. Live in La Jolla, CA1
Arizona USA1
Paddington, London1
society1
Name: User_Location, Length: 8696, dtype: object

6 Visual EDA- Part 2

6.1 Further Filtering Neutral Values

```
In [189]: ▶ plt.figure(figsize = (8,8))
ax = sns.countplot(x = "SentimentLabel", c
ax.set(title = "Trump Sentiment Counts", >
plt.text(4.7, 10000, "We clearly see that w
```

Out[189]: Text(4.7, 10000, 'We clearly see that we h
r down shortly')



From a basic level, this is how we are filtering neutral val

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Strong Negative

Strong Negative

6.2 Creating the Neutral Datafram

```
In [190]: Neutral = df[df['sentimentLabel'].str.contains(
#Creating a dataframe of all the Neutral tweets
Neutral['User_Description'] = Neutral.User
```

/usr/local/lib/python3.6/site-packages/ipy
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =

See the caveats in the documentation: <http://ml#returning-a-view-versus-a-copy> (<https://returning-a-view-versus-a-copy>)

This is separate from the ipykernel pack

So here we are taking our dataframe of Neutral tweets at

further filter them to be able to get rid of the number of N

After looking at some of the User_Descriptions we found political affiliation (ex. Patriot Pro-Trump Christian Englis

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```
In [191]: Neutral['PolarityScore'] = Neutral.User_De
Neutral['SubjectivityScore'] = Neutral.User
#Actually creating the Polarity and Subjectivity
Neutral['sentimentLabel'] = Neutral['PolarityScore']
#Now putting them into bins
```

```
/usr/local/lib/python3.6/site-packages/ipython/
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =
```

See the caveats in the documentation: <http://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy> (<https://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy>)

```
"""Entry point for launching an IPython shell
/usr/local/lib/python3.6/site-packages/ipython/
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =
```

See the caveats in the documentation: <http://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy> (<https://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy>)

```
/usr/local/lib/python3.6/site-packages/ipython/
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =
```

See the caveats in the documentation: <http://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy> (<https://ipython.org/ipython-doc/3/faq.html#returning-a-view-versus-a-copy>)
after removing the cwd from sys.path.

We now have the neutral values with an assigned sentiment

In [192]:

Neutral.head()

Out[192]:

Hashtags	ID	Tweet_Te
3	1254972505367994368	@RedWingGrip BREAKIN President Trun ha
4	1254972505497808897	RT @mog754 Health exper from the Trum Ob
5	1254972505544105984	@costarepori "U.S. intelligen agencies
9	1254972505661616129	@soledadobrie That 'somebod is you, bab
10	1254972505837764610	Trump Blame Plummeting P Numbers c People

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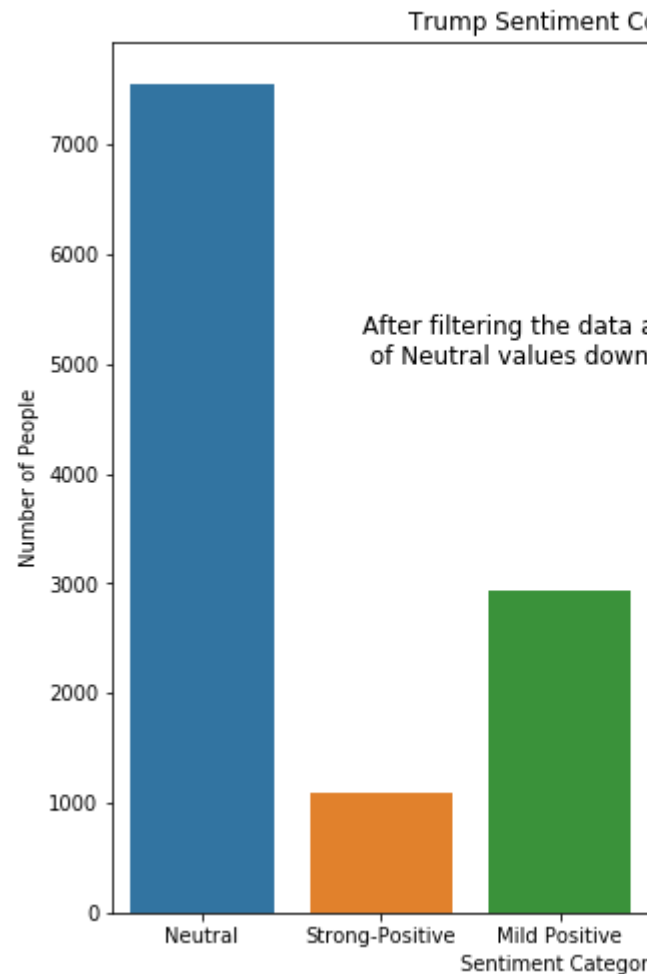
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```
In [193]: ▶ plt.figure(figsize = (8,8))
ax = sns.countplot(x = "SentimentLabel", c
ax.set(title = "Trump Sentiment Counts", >
plt.text(0.9,5000,"After filtering the dat
```

Out[193]: Text(0.9, 5000, 'After filtering the data 40%')

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 - 7.1 2020 E

7 EDA Part 2

7.1 2020 Election Simulation

This section includes an elaborate simulation of the 2020 We then looked at the outcomes of the previous election was lower than 5% then we determined that the state wo and belong to the other party in the upcoming election. T race, and could ultimately be a 50/50 chance for each pa small simulation of the actual election, and is determinig

```
In [194]: ▶ topStates = pd.read_csv("data_files/Group_
topStates #A BLUE STATE= WON BY THE DEMOCR
```

```
Out[194]:
```

	State	Electoral Votes	Difference in % of V
0	California	53	28
1	Texas	36	9
2	New York	27	21
3	Florida	27	1
4	Pennsylvania	18	1
5	Illinois	18	
6	Ohio	16	8
7	Michigan	14	0
8	Georgia	14	5
9	North Carolina	13	3

7.2 How the simulation actually w

In summary, the simulation will follow these steps to dete

- PLEASE KEEP IN MIND THAT THIS IS FROM TRU Trump)
- Add up all of the positive and negative tweets for the label.
- Do the exact same for the original DataFrame to get
- Take the number of remaining Neutral values for eac

- It is now determined if a state is a swing state or not
- If the state is not a swing state, then the remaining n election. But if the state is indeed a swing state, it er chance of winning one of those states, a random nu are given to Joe Biden, but if that number comes ou
- Then based off of that determination, we calculate w

Below is a screenshot for just one state, this then replica

```
Neutral['User_Location'] = df.User_Location.fillna('')
#Creating positive and negative counts for the fi
FLN = Neutral[Neutral['User_Location'].str.contains("FLORIDA")]
FLN1= FLN[FLN['sentimentLabel'].str.contains("Mild-Negative")]
FLN2= FLN[FLN['sentimentLabel'].str.contains("Strong-Negative")]
FLN3 = FLN[FLN['sentimentLabel'].str.contains("Mild-Positive")]
FLN4 = FLN[FLN['sentimentLabel'].str.contains("Strong-Positive")]
NegativeN = len(FLN1)+len(FLN2)
PositiveN = len(FLN3)+len(FLN4)
#Now creating positive and negative counts for th
df['User_Location'] = df.User_Location.fillna('')
FL = df[df['User_Location'].str.contains("FLORIDA")]
FL1= FL[FL['sentimentLabel'].str.contains("Mild-Negative")]
FL2= FL[FL['sentimentLabel'].str.contains("Strong-Negative")]
FL3 = FL[FL['sentimentLabel'].str.contains("Mild-Positive")]
FL4 = FL[FL['sentimentLabel'].str.contains("Strong-Positive")]
Negative = len(FL1)+len(FL2)
Positive = len(FL3)+len(FL4)
Only_Neutral = FLN[FLN['sentimentLabel'].str.contains("Neutral")]
Neutral_Final = len(Only_Neutral)
#This for loop is entered becuase the state is a
#from the election before
import numpy as np
random = np.random.randint(low = 1, high = 10)
if (random%2) == 0:
    Negative_Final = Negative + NegativeN + Neutral_Final
    Positive_Final = Positive + PositiveN + Neutral_Final
else:
    Positive_Final = Positive + PositiveN + Neutral_Final
    Negative_Final = Negative + NegativeN + Neutral_Final
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+27
    print("Biden Wins Florida")
    print(f"Biden Vote Count: {Biden_vote_count}")
    print('-'*50)
#Updating the list to later analyze in our final
trump_win.append(0)
biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+27
    print("Trump Wins Florida")
    print(f"Trump Vote Count: {Trump_vote_count}")
    print('-'*50)
    trump_win.append(1)
    biden_win.append(0)
```

7.3 The Actual Simulation

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Please note this is a live simulation, so each time this output we got from this simulation.

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```
In [195]: ► Biden_vote_count = 0
Trump_vote_count = 0
trump_win = []
biden_win = []
# CALIFORNIA-----
Neutral['User_Location'] = df.User_Location
CALN = Neutral[Neutral['User_Location'].str.contains('CALIFORNIA')]
CALN1= CALN[CALN['sentimentLabel'].str.contains('positive')]
CALN2= CALN[CALN['sentimentLabel'].str.contains('neutral')]
CALN3 = CALN[CALN['sentimentLabel'].str.contains('negative')]
CALN4 = CALN[CALN['sentimentLabel'].str.contains('other')]
NegativeN = len(CALN1)+len(CALN2)
PositiveN = len(CALN3)+len(CALN4)
df['User_Location'] = df.User_Location.fillna('')
CA = df[df['User_Location'].str.contains('CALIFORNIA')]
CAL1= CA[CA['sentimentLabel'].str.contains('positive')]
CAL2= CA[CA['sentimentLabel'].str.contains('neutral')]
CAL3 = CA[CA['sentimentLabel'].str.contains('negative')]
CAL4 = CA[CA['sentimentLabel'].str.contains('other')]
Negative = len(CAL1)+len(CAL2)
Positive = len(CAL3)+len(CAL4)
Only_Neutral = CALN[CALN['sentimentLabel'].str.contains('neutral')]
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN
Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+5
    print("Biden Wins California")
    print(f"Biden Vote Count: {Biden_vote_count}")
    print('-'*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+5
    print("Trump Wins California")
    print(f"Trump Vote Count: {Trump_vote_count}")
    trump_win.append(1)
    biden_win.append(0)
#TEXAS-----
Neutral['User_Location'] = df.User_Location
TXN = Neutral[Neutral['User_Location'].str.contains('TEXAS')]
TXN1= TXN[TXN['sentimentLabel'].str.contains('positive')]
TXN2= TXN[TXN['sentimentLabel'].str.contains('neutral')]
TXN3 = TXN[TXN['sentimentLabel'].str.contains('negative')]
TXN4 = TXN[TXN['sentimentLabel'].str.contains('other')]
NegativeN = len(TXN1)+len(TXN2)
PositiveN = len(TXN3)+len(TXN4)
df['User_Location'] = df.User_Location.fillna('')
TX = df[df['User_Location'].str.contains('TEXAS')]
TX1= TX[TX['sentimentLabel'].str.contains('positive')]
TX2= TX[TX['sentimentLabel'].str.contains('neutral')]
TX3 = TX[TX['sentimentLabel'].str.contains('negative')]
TX4 = TX[TX['sentimentLabel'].str.contains('other')]
Negative = len(TX1)+len(TX2)
Positive = len(TX3)+len(TX4)
```

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```

Only_Neutral = TXN[TXN['sentimentLabel']].s
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN+Neut
Negative_Final = Negative + NegativeN
Neutral_Final = len(Only_Neutral)
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+36
    print("Biden Wins Texas")
    print(f"Biden Vote Count: {Biden_vote_
    print('- '*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+36
    print("Trump Wins Texas")
    print(f"Trump Vote Count: {Trump_vote_
    print('- '*50)
    trump_win.append(1)
    biden_win.append(0)

# NEW YORK-----
Neutral['User_Location'] = df.User_Locatic
NYN = Neutral[Neutral['User_Location']].str
NYN1= NYN[NYN['sentimentLabel'].str.contai
NYN2= NYN[NYN['sentimentLabel'].str.contai
NYN3 = NYN[NYN['sentimentLabel'].str.conta
NYN4 = NYN[NYN['sentimentLabel'].str.conta
NegativeN = len(NYN1)+len(NYN2)
PositiveN = len(NYN3)+len(NYN4)
df['User_Location'] = df.User_Location.fil
NY = df[df['User_Location'].str.contains('
NY1= NY[NY['sentimentLabel'].str.contains(
NY2= NY[NY['sentimentLabel'].str.contains(
NY3 = NY[NY['sentimentLabel'].str.contains
NY4 = NY[NY['sentimentLabel'].str.contains
Negative = len(NY1)+len(NY2)
Positive = len(NY3)+len(NY4)
Only_Neutral = NYN[NYN['sentimentLabel']].s
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN
Negative_Final = Negative + NegativeN+Neut
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+27
    print("Biden Wins New York")
    print(f"Biden Vote Count: {Biden_vote_
    print('- '*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+27
    print("Trump Wins New York")
    print(f"Trump Vote Count: {Trump_vote_
    print('- '*50)
    trump_win.append(1)
    biden_win.append(0)

# FLORIDA-----
Neutral['User_Location'] = df.User_Locatic
FLN = Neutral[Neutral['User_Location']].str

```

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```

FLN1= FLN[FLN['sentimentLabel'].str.contai
FLN2= FLN[FLN['sentimentLabel'].str.contai
FLN3 = FLN[FLN['sentimentLabel'].str.conta
FLN4 = FLN[FLN['sentimentLabel'].str.conta
NegativeN = len(FLN1)+len(FLN2)
PositiveN = len(FLN3)+len(FLN4)
df['User_Location'] = df.User_Location.fil
FL = df[df['User_Location'].str.contains('
FL1= FL[FL['sentimentLabel'].str.contains(
FL2= FL[FL['sentimentLabel'].str.contains(
FL3 = FL[FL['sentimentLabel'].str.contains
FL4 = FL[FL['sentimentLabel'].str.contains
Negative = len(FL1)+len(FL2)
Positive = len(FL3)+len(FL4)
Only_Neutral = FLN[FLN['sentimentLabel'].s
Neutral_Final = len(Only_Neutral)
import numpy as np
random = np.random.randint(low = 1, high =
if (random%2) == 0:
    Negative_Final = Negative + Negati
    Positive_Final = Positive + Positi
else:
    Positive_Final = Positive + PositiveN
    Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+27
    print("Biden Wins Florida")
    print(f"Biden Vote Count: {Biden_vote_
    print('- '*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+27
    print("Trump Wins Florida")
    print(f"Trump Vote Count: {Trump_vote_
    print('- '*50)
    trump_win.append(1)
    biden_win.append(0)
#Pennsylvania-----
Neutral['User_Location'] = df.User_Locatic
PNN = Neutral[Neutral['User_Location'].str
PNN1= PNN[PNN['sentimentLabel'].str.contai
PNN2= PNN[PNN['sentimentLabel'].str.contai
PNN3 = PNN[PNN['sentimentLabel'].str.conta
PNN4 = PNN[PNN['sentimentLabel'].str.conta
NegativeN = len(PNN1)+len(PNN2)
PositiveN = len(PNN3)+len(PNN4)
df['User_Location'] = df.User_Location.fil
PN = df[df['User_Location'].str.contains('
PN1= PN[PN['sentimentLabel'].str.contains(
PN2= PN[PN['sentimentLabel'].str.contains(
PN3 = PN[PN['sentimentLabel'].str.contains
PN4 = PN[PN['sentimentLabel'].str.contains
Negative = len(PN1)+len(PN2)
Positive = len(PN3)+len(PN4)
Only_Neutral = PN[PN['sentimentLabel'].str
Neutral_Final = len(Only_Neutral)

```

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```
import numpy as np
random = np.random.randint(low = 1, high =
if (random%2) == 0:
    Negative_Final = Negative + Negati
    Positive_Final = Positive + Positi
else:
    Positive_Final = Positive + PositiveN
    Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins Pennsylvania")
    print(f"Biden Vote Count: {Biden_vote_
    print('- '*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins Pennsylvania")
    print(f"Trump Vote Count: {Trump_vote_
    print('- '*50)
    trump_win.append(1)
    biden_win.append(0)

#ILLINOIS-----
Neutral['User_Location'] = df.User_Locatic
ILN = Neutral[Neutral['User_Location'].str
ILN1= ILN[ILN['sentimentLabel'].str.contai
ILN2= ILN[ILN['sentimentLabel'].str.contai
ILN3 = ILN[ILN['sentimentLabel'].str.conta
ILN4 = ILN[ILN['sentimentLabel'].str.conta
NegativeN = len(ILN1)+len(ILN2)
PositiveN = len(ILN3)+len(ILN4)
df['User_Location'] = df.User_Location.fil
IL = df[df['User_Location'].str.contains('
IL1= IL[IL['sentimentLabel'].str.contains(
IL2= IL[IL['sentimentLabel'].str.contains(
IL3 = IL[IL['sentimentLabel'].str.contains
IL4 = IL[IL['sentimentLabel'].str.contains
Negative = len(IL1)+len(IL2)
Positive = len(IL3)+len(IL4)
Only_Neutral = ILN[ILN['sentimentLabel'].s
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN
Negative_Final = Negative + NegativeN+Neut
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins Illinois")
    print(f"Biden Vote Count: {Biden_vote_
    print('- '*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins Illinois")
    print(f"Trump Vote Count: {Trump_vote_
    trump_win.append(1)
    biden_win.append(0)

#OHIO-----
```


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```
Neutral['User_Location'] = df.User_Location
OHN = Neutral[Neutral['User_Location'].str.contains('Neutral')]
OHN1= OHN[OHN['sentimentLabel'].str.contains('Neutral')]
OHN2= OHN[OHN['sentimentLabel'].str.contains('Neutral')]
OHN3 = OHN[OHN['sentimentLabel'].str.contains('Neutral')]
OHN4 = OHN[OHN['sentimentLabel'].str.contains('Neutral')]
NegativeN = len(OHN1)+len(OHN2)
PositiveN = len(OHN3)+len(OHN4)
df['User_Location'] = df.User_Location.fillna('')
OH = df[df['User_Location'].str.contains('Neutral')]
OH1= OH[OH['sentimentLabel'].str.contains('Neutral')]
OH2= OH[OH['sentimentLabel'].str.contains('Neutral')]
OH3 = OH[OH['sentimentLabel'].str.contains('Neutral')]
OH4 = OH[OH['sentimentLabel'].str.contains('Neutral')]
Negative = len(OH1)+len(OH2)
Positive = len(OH3)+len(OH4)
Only_Neutral = OHN[OHN['sentimentLabel'].str.contains('Neutral')]
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN
Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins Ohio")
    print(f"Biden Vote Count: {Biden_vote_count}")
    print('-'*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins Ohio")
    print(f"Trump Vote Count: {Trump_vote_count}")
    print('-'*50)
    trump_win.append(1)
    biden_win.append(0)

#MICHIGAN-----
Neutral['User_Location'] = df.User_Location
MIN = Neutral[Neutral['User_Location'].str.contains('Neutral')]
MIN1= MIN[MIN['sentimentLabel'].str.contains('Neutral')]
MIN2= MIN[MIN['sentimentLabel'].str.contains('Neutral')]
MIN3 = MIN[MIN['sentimentLabel'].str.contains('Neutral')]
MIN4 = MIN[MIN['sentimentLabel'].str.contains('Neutral')]
NegativeN = len(MIN1)+len(MIN2)
PositiveN = len(MIN3)+len(MIN4)
df['User_Location'] = df.User_Location.fillna('')
MI = df[df['User_Location'].str.contains('Neutral')]
MI1= MI[MI['sentimentLabel'].str.contains('Neutral')]
MI2= MI[MI['sentimentLabel'].str.contains('Neutral')]
MI3 = MI[MI['sentimentLabel'].str.contains('Neutral')]
MI4 = MI[MI['sentimentLabel'].str.contains('Neutral')]
Negative = len(MI1)+len(MI2)
Positive = len(MI3)+len(MI4)
Only_Neutral = MIN[MIN['sentimentLabel'].str.contains('Neutral')]
Neutral_Final = len(Only_Neutral)
import numpy as np
random = np.random.randint(low = 1, high = 10)
if (random%2) == 0:
    Negative_Final = Negative + NegativeN
```

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```

Positive_Final = Positive + Positi
else:
    Positive_Final = Positive + PositiveN
    Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins Michigan")
    print(f"Biden Vote Count: {Biden_vote_
    print('-'*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins Michigan")
    print(f"Trump Vote Count: {Trump_vote_
    print('-'*50)
    trump_win.append(1)
    biden_win.append(0)

#GEORGIA-----
Neutral['User_Location'] = df.User_Locatic
GAN = Neutral[Neutral['User_Location'].str
GAN1= GAN[GAN['sentimentLabel'].str.contai
GAN2= GAN[GAN['sentimentLabel'].str.contai
GAN3 = GAN[GAN['sentimentLabel'].str.conta
GAN4 = GAN[GAN['sentimentLabel'].str.conta
NegativeN = len(GAN1)+len(GAN2)
PositiveN = len(GAN3)+len(GAN4)
df['User_Location'] = df.User_Location.fil
GA = df[df['User_Location'].str.contains('
GA1= GA[GA['sentimentLabel'].str.contains(
GA2= GA[GA['sentimentLabel'].str.contains(
GA3 = GA[GA['sentimentLabel'].str.contains
GA4 = GA[GA['sentimentLabel'].str.contains
Negative = len(GA1)+len(GA2)
Positive = len(GA3)+len(GA4)
Only_Neutral = GAN[GAN['sentimentLabel'].s
Neutral_Final = len(Only_Neutral)
Positive_Final = Positive + PositiveN+Neut
Negative_Final = Negative + NegativeN

if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins Georgia")
    print(f"Biden Vote Count: {Biden_vote_
    print('-'*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins Georgia")
    print(f"Trump Vote Count: {Trump_vote_
    print('-'*50)
    trump_win.append(1)
    biden_win.append(0)

#North Carolina-----
Neutral['User_Location'] = df.User_Locatic

```


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NCN = Neutral[Neutral['User_Location']].str
NCN1= NCN[NCN['sentimentLabel'].str.contai
NCN2= NCN[NCN['sentimentLabel'].str.contai
NCN3 = NCN[NCN['sentimentLabel'].str.conta
NCN4 = NCN[NCN['sentimentLabel'].str.conta
NegativeN = len(NCN1)+len(NCN2)
PositiveN = len(NCN3)+len(NCN4)
df['User_Location'] = df.User_Location.fil
NC = df[df['User_Location'].str.contains('
NC1= NC[NC['sentimentLabel'].str.contains(
NC2= NC[NC['sentimentLabel'].str.contains(
NC3 = NC[NC['sentimentLabel'].str.contains
NC4 = NC[NC['sentimentLabel'].str.contains
Negative = len(NC1)+len(NC2)
Positive = len(NC3)+len(NC4)
Only_Neutral = NCN[NCN['sentimentLabel'].s
Neutral_Final = len(Only_Neutral)
import numpy as np
random = np.random.randint(low = 1, high =
if (random%2) == 0:
    Negative_Final = Negative + Negati
    Positive_Final = Positive + Positi
else:
    Positive_Final = Positive + PositiveN
    Negative_Final = Negative + NegativeN
if Negative_Final > Positive_Final:
    Biden_vote_count = Biden_vote_count+1
    print("Biden Wins North Carolina")
    print(f"Biden Vote Count: {Biden_vote_
    print('-'*50)
    trump_win.append(0)
    biden_win.append(1)
else:
    Trump_vote_count = Trump_vote_count+1
    print("Trump Wins North Carolina")
    print(f"Trump Vote Count: {Trump_vote_
    print('-'*50)
    trump_win.append(1)
    biden_win.append(0)

#-----
if Trump_vote_count > Biden_vote_count:
    print(" *****Trump wins the 2020 elec
else:
    print(" *****Biden wins the 2020 elec
print('-'*50)
print(f"Trump's final score was {Trump_vot
print(f"Biden's final score was {Biden_vot
print('-'*50)

```

/usr/local/lib/python3.6/site-packages/ipy
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =

See the caveats in the documentation: <http://ml#returning-a-view-versus-a-copy> (<https://returning-a-view-versus-a-copy>)

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Try using .loc[row_indexer,col_indexer] =
```

See the caveats in the documentation: <http://ml#returning-a-view-versus-a-copy> (<https://returning-a-view-versus-a-copy>)

```
Biden Wins California
Biden Vote Count: 53
```

```
-----
Trump Wins Texas
Trump Vote Count: 36
```

```
-----
Biden Wins New York
Biden Vote Count: 80
```

```
-----
Trump Wins Florida
Trump Vote Count: 63
-----
```

```
/usr/local/lib/python3.6/site-packages/ipy
A value is trying to be set on a copy of a
Try using .loc[row_indexer,col_indexer] =
```

See the caveats in the documentation: <http://ml#returning-a-view-versus-a-copy> (<https://returning-a-view-versus-a-copy>)

```
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```
/usr/local/lib/python3.6/site-packages/ipy
```

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A value is trying to be set on a copy of a
Try using `.loc[row_indexer,col_indexer] =`

See the caveats in the documentation: [http
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returning-a-view-versus-a-copy\)](https://pandas.pydata.org/pandas-docs/stable/10min/returning-a-view-versus-a-copy.html)
/usr/local/lib/python3.6/site-packages/ipy
A value is trying to be set on a copy of a
Try using `.loc[row_indexer,col_indexer] =`

See the caveats in the documentation: [http
ml#returning-a-view-versus-a-copy \(https://
returning-a-view-versus-a-copy\)](https://pandas.pydata.org/pandas-docs/stable/10min/returning-a-view-versus-a-copy.html)

Biden Wins Pennsylvania
Biden Vote Count: 98

Biden Wins Illinois
Biden Vote Count: 116

Trump Wins Ohio
Trump Vote Count: 79

Biden Wins Michigan
Biden Vote Count: 130

Trump Wins Georgia
Trump Vote Count: 93

Trump Wins North Carolina
Trump Vote Count: 106

*****Biden wins the 2020 election*****

Trump's final score was 106
Biden's final score was 130

/usr/local/lib/python3.6/site-packages/ipy
A value is trying to be set on a copy of a
Try using `.loc[row_indexer,col_indexer] =`

See the caveats in the documentation: [http
ml#returning-a-view-versus-a-copy \(https://
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See the caveats in the documentation: [http
ml#returning-a-view-versus-a-copy \(https://
returning-a-view-versus-a-copy\)](https://pandas.pydata.org/pandas-docs/stable/10min/returning-a-view-versus-a-copy.html)

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```
In [196]: topStates = pd.read_csv("data_files/Group_topStates
topStates["Trump's States"] = trump_win
topStates["Trump's States"].replace(to_replace="Trump's States", value="Biden's States")
topStates["Biden's States"] = biden_win
topStates["Biden's States"].replace(to_replace="Biden's States", value="Trump's States")
topStates
```

Out[196]:

	State	Electoral Votes	Difference in % of V
0	California	53	28
1	Texas	36	9
2	New York	27	21
3	Florida	27	1
4	Pennsylvania	18	1
5	Illinois	18	
6	Ohio	16	8
7	Michigan	14	0
8	Georgia	14	5
9	North Carolina	13	3

7.5 Geo-Map

```
In [197]: !pip install geopy
import geopy
from geopy.geocoders import Nominatim
from geopy.geocoders import Nominatim
from mpl_toolkits.basemap import Basemap
nom = Nominatim()
```

Requirement already satisfied: geopy in /usr/local/lib/python3.6/site-packages/ipython/notebook/ (1.50)

/usr/local/lib/python3.6/site-packages/ipython/notebook/ the default "geopy/1.21.0" `user_agent` is operations.osmfoundation.org/policies/nominatim/) and may possibly cause 403 and 429 HTTP errors (user_agent="my-application")` or by overwriting `user_agent` to "my-application". In geopy

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```
In [198]: location = df["User_Location"].value_count
new_loc = location.to_string()
location #Taking the frequency of each loc
locc = pd.DataFrame(location)
locc
e = df.User_Location.dropna()
ee = pd.DataFrame(e)
zg = ee.User_Location.value_counts().head(
ZG = zg.index
ZLG= pd.DataFrame(ZG)
ZLG.columns = ["Location"]
ZLG
e = df.User_Location.dropna()
ee = pd.DataFrame(e)
zg = ee.User_Location.value_counts().head(
count = pd.DataFrame(zg)
count
ZLG["Coordinates"] = ZLG["Location"].apply
ZLG.dropna()
```

Out[198]:

	Location	
1	United States	(United States
2	USA	(United States
3	California, USA	(California, Unite
4	Texas, USA	(Texas, United Sta
5	India	(भारत - Inc
...	...	
295	Islamabad, Pakistan	(اسلام آباد, Pa
296	Amsterdam, The Netherlands	(Amsterdam, Noord-H
297	Bay Area, CA	(San Francisco Bay
298	Montréal, Québec	(Montréal, Agglomé
299	Connecticut	(Connecticut, Unit

297 rows × 2 columns

```
In [199]: ZLG["Latitude"] = ZLG["Coordinates"].apply
ZLG["Longitude"] = ZLG["Coordinates"].app
Lat = ZLG["Latitude"]
Long = ZLG["Longitude"]
import numpy as np # linear algebra
import pandas as pd # data processing, CSV
import seaborn as sns #Statistical Data Vi
import matplotlib.pyplot as plt
%matplotlib inline
```

Contents

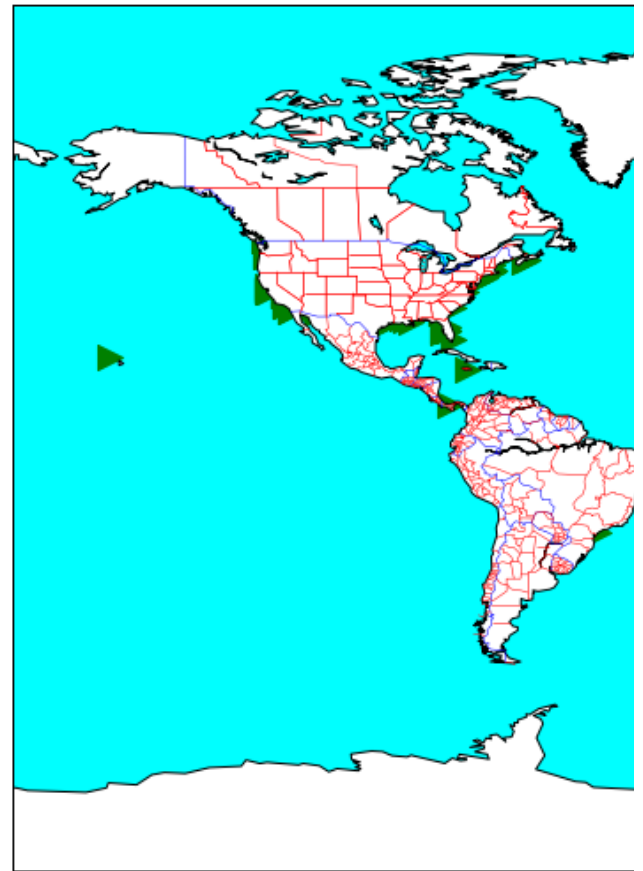
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```
In [200]: ► from mpl_toolkits.basemap import Basemap
import matplotlib.pyplot as plt
fig = plt.figure(figsize=(20,9))
m = Basemap(projection='gall',
            resolution = 'c')
m.drawcoastlines()
m.drawcountries(color= 'blue')
m.drawstates(color= 'red')
m.drawmapboundary(fill_color='aqua')
m.fillcontinents(color='white',lake_color=
plt.title("Tweet Location", fontsize = 30)
Lat = list(ZLG["Latitude"])
Long = list(ZLG["Longitude"])
m.scatter(Long,Lat,latlon=True,color="green")
plt.show()
```

```
/usr/local/lib/python3.6/site-packages/mpl
ue encountered in greater
    lonsin = np.where(lonsin > lon_0+180, lc
/usr/local/lib/python3.6/site-packages/mpl
ue encountered in less
    lonsin = np.where(lonsin < lon_0-180, lc
/usr/local/lib/python3.6/site-packages/mpl
ue encountered in greater_equal
    itemindex = len(lonsin)-np.where(londiff
/usr/local/lib/python3.6/site-packages/mpl
ue encountered in less
    mask = np.logical_or(lonsin<lon_0-180,lc
/usr/local/lib/python3.6/site-packages/mpl
ue encountered in greater
    mask = np.logical_or(lonsin<lon_0-180,lc
```

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8 Discussions

Starting off with this project, we ran into quite a few roadl metric and the user location to test our second and third ways to pull user info from our tweets and joining it to ou direction. However, we were still very interested in our fir we went ahead and worked on this part without followers ratings by week (our third party data) and spent most of t polarity scores, or a combination of the two. We were the be discussed below.

Another issue we ran into had to do with the number of t per week for 17 weeks with 9 keywords, which gave us a size in our virtual containers proved to be extremely ineff keywords and scrape less tweets per week to ensure we

9 Conclusions

Let's explore how each hypothesis turned out

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1) President Trump's approval rating has increased with by the US government. The more tweets involving stimu the higher we believe Trump's approval rating will be. Th

- Given that the TextBlob tool pooled most of our twee negative sentiment tweets specifically had a correlat seems as if the more tweets there were about stimu Trump's approval rating was. They seemed to move rating, but they were just correlated. We did some re coronavirus actions and added them to our plots as
- Despite our issues with the sentiment tool, we were about stimulus checks and Trump's approval rating. (approval rating) seem to generally move together. T coronavirus (shown in section 3.5.5) as there appea inversely with approval rating in this figure.
- Given the results from the simulation, we conclude t he is still winning his historically Republican states, t saw in Part 1, it appears that his approval rating is o

2) Sentiment ratings for the tweets will show an increase

- We were able to explore this via sections 3.5.3, 3.5.4