

```
!pip install tweepy
!pip install vaderSentiment
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```
from google.colab import files
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from datetime import datetime
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from pprint import pprint
import six
import tweepy
```

```
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
```

```
↳ Requirement already satisfied: tweepy in /usr/local/lib/python3.6/dist-packages (3.6.0)
Requirement already satisfied: requests>=2.11.1 in /usr/local/lib/python3.6/dist-packa
Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.6/di
Requirement already satisfied: six>=1.10.0 in /usr/local/lib/python3.6/dist-packages (
Requirement already satisfied: PySocks>=1.5.7 in /usr/local/lib/python3.6/dist-package
Requirement already satisfied: urllib3<1.23,>=1.21.1 in /usr/local/lib/python3.6/dist-
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-pac
Requirement already satisfied: idna<2.7,>=2.5 in /usr/local/lib/python3.6/dist-package
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-
Requirement already satisfied: oauthlib>=0.6.2 in /usr/local/lib/python3.6/dist-packag
Requirement already satisfied: vaderSentiment in /usr/local/lib/python3.6/dist-package
```

```
def df_creator(tweets):
    # WANTED List to hold wanted information from tweets
    date = []
    favorite_count = []
    name = []
    retweet = []
    text = []
    id_ = []

    # Loop through the list of tweets to grab needed info
    for tweet in tweets:
        date.append(tweet['created_at'])
        favorite_count.append(tweet['favorite_count'])
        retweet.append(tweet['retweet_count'])
        text.append(tweet['full_text'])
        id_.append(tweet['id'])

    # Create DF based on WANTED lists
    df = pd.DataFrame({
        'Created': date,
        'Likes': favorite_count,
        'Retweet': retweet,
        'Text': text,
        'ID': id_,
    })

    # Convert date to datetime dtype
    df['Created'] = [datetime.strptime(date, "%a %b %d %H:%M:%S %z %Y") for date in df['Created']]

    return df

# Analyze Pulled Tweets and get Compound, Positive, Negative, & Neutral Scores
def sentiment_analyzer(df):
    # Setup sentiment analyzer
    analyzer = SentimentIntensityAnalyzer()
```

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# Variables for holding sentiments
compound_list = []
positive_list = []
negative_list = []
neutral_list = []

# Loop through Tweets
for text in df['Text']:

    # Run Vader Analysis on each tweet
    results = analyzer.polarity_scores(text)
    compound = results["compound"]
    pos = results["pos"]
    neu = results["neu"]
    neg = results["neg"]

    # Add each value to the appropriate list
    compound_list.append(compound)
    positive_list.append(pos)
    negative_list.append(neg)
    neutral_list.append(neu)

df['Compound Score'] = compound_list
df['Positive Score'] = positive_list
df['Negative Score'] = negative_list
df['Neutral Score'] = neutral_list

return df

def user_tweets(user, api, consumer_key, consumer_secret, access_token, access_token_secret, ei):
    target_user = ('@'+user)

    # List to store dictionaries of tweets
    tweets = []

    # Loop through 25 pages of tweets and grab 500 tweets
    for x in range(1, endpage):

        for tweet in api.user_timeline(target_user, page=x, tweet_mode='extended'):
            tweets.append(tweet)

    # Convert list of dictionary tweets into a dataframe
    tweet_df = df_creator(tweets)

    # Reset index to date created for Group By purposes
    tweet_df = tweet_df.set_index('Created')

    tweet_df = sentiment_analyzer(tweet_df)

    final_df = tweet_df.resample('Y').mean()

    final_df['Handle'] = user

    final_df.index = [date.year for date in final_df.index]

    return final_df[['Handle', 'Likes', 'Retweet', 'Compound Score', 'Positive Score', 'Negative !',
                    'Neutral Score']]

# Convert Pandas DF to Png format
def render_mpl_table(data, col_width=3.0, row_height=0.625, font_size=14,
                    header_color='#40466e', row_colors=['#f1f1f2', 'w'], edge_color='w',
                    bbox=[0, 0, 1, 1], header_columns=0,
                    ax=None, **kwargs):
    if ax is None:
        size = (np.array(data.shape[:: -1]) + np.array([0, 1])) * np.array([col_width, row_height])
        fig, ax = plt.subplots(figsize=size)
        ax.axis('off')

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mpl_table = ax.table(cellText=data.values, bbox=bbox, colLabels=data.columns, **kwargs)

mpl_table.auto_set_font_size(False)
mpl_table.set_fontsize(font_size)

for k, cell in six.iteritems(mpl_table._cells):
    cell.set_edgecolor(edge_color)
    if k[0] == 0 or k[1] < header_columns:
        cell.set_text_props(weight='bold', color='w')
        cell.set_facecolor(header_color)
    else:
        cell.set_facecolor(row_colors[k[0]%len(row_colors) ])
return ax

```

```

response = user_tweets('Speedhunters',api=api,consumer_key=consumer_key,consumer_secret=consui
response

```



	Handle	Likes	Retweet	Compound Score	Positive Score	Negative Score	Neutral Score
2015	Speedhunters	34.714286	13.333333	0.088795	0.114143	0.055048	0.830810
2016	Speedhunters	51.619982	20.306869	0.108966	0.100009	0.032875	0.867113
2017	Speedhunters	73.459651	21.303422	0.118024	0.082337	0.031057	0.886607
2018	Speedhunters	127.321691	40.393382	0.146217	0.089213	0.031583	0.879200

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# Render response to a time series chart
ax = response[['Compound Score','Positive Score','Negative Score', 'Neutral Score']].plot()
fig = ax.get_figure()
fig.savefig('time_plot.png')

# Render response to PNG
render_mpl_table(response).get_figure().savefig('table.png')

api.update_with_media('table.png', 'test table')
api.update_with_media('time_plot.png', 'test plot')

```



```
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```

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