#### **Example from**

 https://github.com/RodolfoFerro/pandas\_twitter/blob/master/01-extracting-data.md (https://github.com/RodolfoFerro/pandas\_twitter/blob/master/01-extracting-data.md)
 Added some notes and explanation

# Extracting twitter data (tweepy + pandas)

```
In [6]: # General:
    import tweepy  # To consume Twitter's API
    import pandas as pd  # To handle data
    import numpy as np  # For number computing

# For plotting and visualization:
    from IPython.display import display
    import matplotlib.pyplot as plt
    import seaborn as sns
%matplotlib inline
```

```
In [11]: #import libraries that you need
         import tweepy
         from tweepy import OAuthHandler, Stream
         # Go to https://developer.twitter.com/en/apps to create an app and get valu
         # for these credentials, which you'll need to provide in place of these
         # empty string values that are defined as placeholders.
         # See https://developer.twitter.com/en/docs/basics/authentication/overview/
         # for more information on Twitter's OAuth implementation.
         from credentials import *
         #create a file called credentials.py make sure it is in the same folder as
         # the credentials file will look like this
         #ACCESS TOKEN = 'xxx'
         \#ACCESS SECRET = 'xx'
         \#CONSUMER KEY = 'xx'
         #CONSUMER SECRET = 'xx'
         auth = tweepy.OAuthHandler(CONSUMER KEY, CONSUMER SECRET)
         auth.set access token(ACCESS TOKEN, ACCESS SECRET)
         twitter api = tweepy.API(auth, wait on rate limit=True)
         # Nothing to see by displaying twitter api except that it's now a
         # defined variable
         print(twitter api)
```

```
In [12]: # We import our access keys:

# API's setup:
def twitter_setup():
    """

    Utility function to setup the Twitter's API
    with our access keys provided.
    """

# Authentication and access using keys:
    auth = tweepy.OAuthHandler(CONSUMER_KEY, CONSUMER_SECRET)
    auth.set_access_token(ACCESS_TOKEN, ACCESS_SECRET)

# Return API with authentication:
    api = tweepy.API(auth)
    return api
```

#### **Tweets Extraction**

```
In [15]: # We create an extractor object:
    extractor = twitter_setup()

# We create a tweet list as follows:
    tweets = extractor.user_timeline(screen_name="bojangles", count=200)
    print("Number of tweets extracted: {}.\n".format(len(tweets)))

# We print the most recent 5 tweets:
    print("5 recent tweets:\n")
    for tweet in tweets[:5]:
        print(tweet.text)
        print()
```

Number of tweets extracted: 198.

5 recent tweets:

When you get the text from your coworker "I'm running late. What you want from Bojangles?" It's Sausage & Biscu... https://t.co/BtOtJG41tj (https://t.co/BtOtJG41tj)

@CharlotteFC WINS! https://t.co/2pqW8Rm0dW (https://t.co/2pqW8Rm0dW)

Tee off in Augusta or tea off from the couch? https://t.co/Fuk2o9YE5x (https://t.co/Fuk2o9YE5x)

And they're off! Let's go @CharlotteFC #forthecrown https://t.co/e2yYKNHrCm (https://t.co/e2yYKNHrCm)

RT @CharlotteFC\_CFO: Supporters Tailgate is already poppin off. @Bojangle s handing out breakfast biscuits and prizes too! Supporters march...

#### Creating a (pandas) DataFrame

```
In [16]: # We create a pandas dataframe as follows:
    data = pd.DataFrame(data=[tweet.text for tweet in tweets], columns=['Tweets
# We display the first 10 elements of the dataframe:
    display(data.head(10))
```

#### **Tweets**

- **0** When you get the text from your coworker "I'm ...
- 1 @CharlotteFC WINS! https://t.co/2pqW8Rm0dW
- **2** Tee off in Augusta or tea off from the couch? ...
- 3 And they're off! Let's go @CharlotteFC #forthe...
- 4 RT @CharlotteFC\_CFO: Supporters Tailgate is al...
- 5 pov: you're a scratch-made Cajun Filet Biscuit...
- 6 Next #cltfc star? Come take your shot! https:/...
- 7 https://t.co/s7p1TrlY0k
- 8 We're back serving up scratch-made biscuits (w...
- **9** A bucket of golf balls or Bojangles biscuits? ...

# In [17]: # Internal methods of a single tweet object: print(dir(tweets[0]))

['\_class\_', '\_delattr\_', '\_dict\_', '\_dir\_', '\_doc\_', '\_eq\_',
'\_format\_', '\_ge\_', '\_getattribute\_', '\_getstate\_', '\_gt\_', '\_
hash\_', '\_init\_', '\_init\_subclass\_', '\_le\_', '\_lt\_', '\_module\_
\_', '\_ne\_', '\_new\_', '\_reduce\_', '\_reduce\_ex\_', '\_repr\_', '\_se
tattr\_', '\_sizeof\_', '\_slots\_', 'str\_', '\_subclasshook\_', '\_we
akref\_', 'api', 'json', 'author', 'contributors', 'coordinates', 'crea
ted\_at', 'destroy', 'entities', 'favorite', 'favorite\_count', 'favorite
d', 'geo', 'id', 'id\_str', 'in\_reply\_to\_screen\_name', 'in\_reply\_to\_status\_
id', 'in\_reply\_to\_status\_id\_str', 'in\_reply\_to\_user\_id', 'in\_reply\_to\_us
er\_id\_str', 'is\_quote\_status', 'lang', 'parse', 'parse\_list', 'place', 'p
ossibly\_sensitive', 'retweet', 'retweet\_count', 'retweeted', 'retweets',
'source', 'source\_url', 'text', 'truncated', 'user']

```
In [18]: # We print info from the first tweet:
         print(tweets[0].id)
         print(tweets[0].created_at)
         print(tweets[0].source)
         print(tweets[0].favorite_count)
         print(tweets[0].retweet_count)
         print(tweets[0].geo)
         print(tweets[0].coordinates)
         print(tweets[0].entities)
         1514263192918274057
         2022-04-13 15:24:13+00:00
         Sprinklr Publishing
         27
         6
         None
         None
         {'hashtags': [], 'symbols': [], 'user_mentions': [], 'urls': [{'url': 'ht
         tps://t.co/BtOtJG41tj', 'expanded_url': 'https://twitter.com/i/web/statu
         s/1514263192918274057', 'display_url': 'twitter.com/i/web/status/1...', 'in
         dices': [121, 144]}]}
```

## Adding relevant info to our dataframe

```
In [19]: # We add relevant data:
    data['len'] = np.array([len(tweet.text) for tweet in tweets])
    data['ID'] = np.array([tweet.id for tweet in tweets])
    data['Date'] = np.array([tweet.created_at for tweet in tweets])
    data['Source'] = np.array([tweet.source for tweet in tweets])
    data['Likes'] = np.array([tweet.favorite_count for tweet in tweets])
    data['RTs'] = np.array([tweet.retweet_count for tweet in tweets])
```

In [20]: # Display of first 10 elements from dataframe:
 display(data.head(10))

	Tweets	len	ID	Date	Source	Likes	RTs
0	When you get the text from your coworker "I'm	144	1514263192918274057	2022-04-13 15:24:13+00:00	Sprinklr Publishing	27	6
1	@CharlotteFC WINS! https://t.co/2pqW8Rm0dW	42	1513244822542426120	2022-04-10 19:57:35+00:00	Twitter for iPhone	11	2
2	Tee off in Augusta or tea off from the couch?	69	1513207690838917135	2022-04-10 17:30:02+00:00	Sprinklr Publishing	19	3
3	And they're off! Let's go @CharlotteFC #forthe	75	1513187788878925825	2022-04-10 16:10:57+00:00	Twitter for iPhone	19	2
4	RT @CharlotteFC_CFO: Supporters Tailgate is al	139	1513178005681721346	2022-04-10 15:32:04+00:00	Twitter for iPhone	0	4
5	pov: you're a scratch-made Cajun Filet Biscuit	117	1513175568401018884	2022-04-10 15:22:23+00:00	Twitter for iPhone	96	6
6	Next #cltfc star? Come take your shot! https:/	62	1513173566417686538	2022-04-10 15:14:26+00:00	Twitter for iPhone	4	0
7	https://t.co/s7p1TrlY0k	23	1513170923519713291	2022-04-10 15:03:56+00:00	Twitter for iPhone	2	0
8	We're back serving up scratch- made biscuits (w	139	1513169376685568003	2022-04-10 14:57:47+00:00	Twitter for iPhone	56	7
9	A bucket of golf balls or Bojangles biscuits?	69	1512860418410270721	2022-04-09 18:30:05+00:00	Sprinklr Publishing	60	7

## Visualization and basic statistics

# Averages and popularity

```
In [21]: # We extract the mean of lenghts:
    mean = np.mean(data['len'])
    print("The lenght's average in tweets: {}".format(mean))
```

The lenght's average in tweets: 82.20202020202021

• To extract more data, we will use some pandas' functionalities:

```
In [22]: # We extract the tweet with more FAVs and more RTs:
         fav_max = np.max(data['Likes'])
         rt_max = np.max(data['RTs'])
         fav = data[data.Likes == fav max].index[0]
         rt = data[data.RTs == rt_max].index[0]
         # Max FAVs:
         print("The tweet with more likes is: \n{}".format(data['Tweets'][fav]))
         print("Number of likes: {}".format(fav_max))
         print("{} characters.\n".format(data['len'][fav]))
         # Max RTs:
         print("The tweet with more retweets is: \n{}".format(data['Tweets'][rt]))
         print("Number of retweets: {}".format(rt_max))
         print("{} characters.\n".format(data['len'][rt]))
         The tweet with more likes is:
         Ready for halftime... https://t.co/vNHaMVqky1 (https://t.co/vNHaMVqky1)
         Number of likes: 346
         45 characters.
         The tweet with more retweets is:
         Bracket busted? You can still own the watch party game when you enter to
          win a Bo's gift card! We're giving away 2... https://t.co/lfYeR1ZxTN (http
         s://t.co/lFYeR1ZxTN)
         Number of retweets: 239
         139 characters.
```

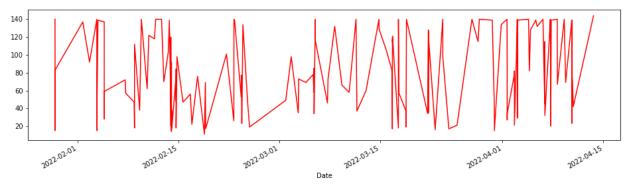
#### Time series

- Pandas has its own object for time series. Since we have a whole vector with creation dates, we can construct time series respect tweets lengths, likes and retweets.
- The way we do it is:

```
In [23]: # We create time series for data:
    tlen = pd.Series(data=data['len'].values, index=data['Date'])
    tfav = pd.Series(data=data['Likes'].values, index=data['Date'])
    tret = pd.Series(data=data['RTs'].values, index=data['Date'])
```

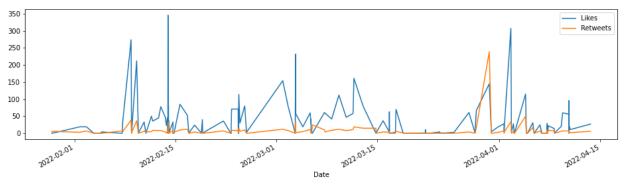
 And if we want to plot the time series, pandas already has its own method in the object. We can plot a time series as follows:

# In [24]: # Lenghts along time: tlen.plot(figsize=(16,4), color='r');



• And to plot the likes versus the retweets in the same chart:

```
In [25]: # Likes vs retweets visualization:
    tfav.plot(figsize=(16,4), label="Likes", legend=True)
    tret.plot(figsize=(16,4), label="Retweets", legend=True);
```



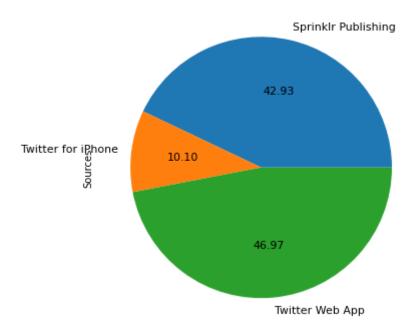
· Pie charts of sources

```
In [26]: # We obtain all possible sources:
    sources = []
    for source in data['Source']:
        if source not in sources:
            sources.append(source)

# We print sources list:
    print("Creation of content sources:")
    for source in sources:
        print("* {}".format(source))
```

Creation of content sources:

- \* Sprinklr Publishing
- \* Twitter for iPhone
- \* Twitter Web App
  - With the following output, we realize that basically this twitter account has two sources: Creation of content sources: \* Twitter for iPhone \* Media Studio
  - We now count the number of each source and create a pie chart. You'll notice that this code cell is not the most optimized one.



# Sentiment analysis

#### Importing textblob

- textblob will allow us to do sentiment analysis in a very simple way. We will also use the re library from Python, which is used to work with regular expressions.
- For this, I'll provide you two utility functions to: a) clean text (which means that any symbol distinct to an alphanumeric value will be remapped into a new one that satisfies this condition), and b) create a classifier to analyze the polarity of each tweet after cleaning the text in it. I won't explain the specific way in which the function that cleans works, since it would be extended and it might be better understood in the official redocumentation.

```
In [28]: from textblob import TextBlob
          import re
          def clean_tweet(tweet):
              1.1.1
              Utility function to clean the text in a tweet by removing
              links and special characters using regex.
              return ' '.join(re.sub("(@[A-Za-z0-9]+)|([^0-9A-Za-z \t])|(\w+:\/\/\S+)
          def analize_sentiment(tweet):
              Utility function to classify the polarity of a tweet
              using textblob.
              \mathbf{1} \cdot \mathbf{1} \cdot \mathbf{1}
              analysis = TextBlob(clean_tweet(tweet))
              if analysis.sentiment.polarity > 0:
                   return 1
              elif analysis.sentiment.polarity == 0:
                   return 0
              else:
                   return -1
```

In [29]: # We create a column with the result of the analysis:
 data['SA'] = np.array([ analize\_sentiment(tweet) for tweet in data['Tweets'

# We display the updated dataframe with the new column:
 display(data.head(10))

	Tweets	len	ID	Date	Source	Likes	RTs	SA
0	When you get the text from your coworker "I'm	144	1514263192918274057	2022-04-13 15:24:13+00:00	Sprinklr Publishing	27	6	-1
1	@CharlotteFC WINS! https://t.co/2pqW8Rm0dW	42	1513244822542426120	2022-04-10 19:57:35+00:00	Twitter for iPhone	11	2	1
2	Tee off in Augusta or tea off from the couch?	69	1513207690838917135	2022-04-10 17:30:02+00:00	Sprinklr Publishing	19	3	0
3	And they're off! Let's go @CharlotteFC #forthe	75	1513187788878925825	2022-04-10 16:10:57+00:00	Twitter for iPhone	19	2	0
4	RT @CharlotteFC_CFO: Supporters Tailgate is al	139	1513178005681721346	2022-04-10 15:32:04+00:00	Twitter for iPhone	0	4	0
5	pov: you're a scratch- made Cajun Filet Biscuit	117	1513175568401018884	2022-04-10 15:22:23+00:00	Twitter for iPhone	96	6	0
6	Next #cltfc star? Come take your shot! https:/	62	1513173566417686538	2022-04-10 15:14:26+00:00	Twitter for iPhone	4	0	0
7	https://t.co/s7p1TrlY0k	23	1513170923519713291	2022-04-10 15:03:56+00:00	Twitter for iPhone	2	0	0
8	We're back serving up scratch-made biscuits (w	139	1513169376685568003	2022-04-10 14:57:47+00:00	Twitter for iPhone	56	7	1
9	A bucket of golf balls or Bojangles biscuits?	69	1512860418410270721	2022-04-09 18:30:05+00:00	Sprinklr Publishing	60	7	0

# **Analyzing the results**

• To have a simple way to verify the results, we will count the number of neutral, positive and negative tweets and extract the percentages.

```
In [30]: # We construct lists with classified tweets:
    pos_tweets = [ tweet for index, tweet in enumerate(data['Tweets']) if data[
    neu_tweets = [ tweet for index, tweet in enumerate(data['Tweets']) if data[
    neg_tweets = [ tweet for index, tweet in enumerate(data['Tweets']) if data[
```

• Now that we have the lists, we just print the percentages:

```
In [31]: # We print percentages:
    print("Percentage of positive tweets: {}%".format(len(pos_tweets)*100/len(d print("Percentage of neutral tweets: {}%".format(len(neu_tweets)*100/len(da print("Percentage de negative tweets: {}%".format(len(neg_tweets)*100/len(d) Percentage of positive tweets: 32.8282828282828283%
    Percentage of neutral tweets: 47.4747474747474%
    Percentage de negative tweets: 19.696969696969695%
In []:
```