Social Media Analytics - Twitter

Signing up for using Twitter APIs

Using Twitter API to access twitter data.

We are creating a twitter app using https://apps.twitter.com. Once we create an app we make a file named auth.k which contains keys/tokens in the order Consumer_key, Consumer_secret, Access_token, Access_token_secret. These are copied from the twitter app which we created.

Get the Twitter search Python script

We write script in Python to use the authentication keys to get data from twitter. The file tw_search.py takes the auth.k as input which calls the twitter and get us the data.

We run tw_search.py python script using the configuration python ./twitter_search.py brexit -c 180 where the third word (brexit)is the search term we want to run on Twitter. This searches brexit tweets and retrieve 180 tweets. The script writes the data in result.csv file having 6 columns – created time, retweet count, hashtag, followers count, friends count. We take different topics/politicians and re-run the python script to collect 180 tweets for each topics/politicians.

Regression Analysis

We use the dataset to see if there are any relationships among number of followers, number of friends, and number of retweets. We create a new python script (Twitter_FriendsFollowers.py) for analysing the relationships. We find correlations among variables in the dataset. We create regression model, plot and line for variables having medium to high level positive/negative correlation. We check the R-square value and the p-value. Based on the regression plot and model, we state the hypothesis such as more followers, more number of retweets.

Sentiment Analysis on Twitter data

People express all kinds of opinions and sentiments on Twitter, so we analyse those sentiments. We write a script in python twitter_sentiments.py script to perform Sentiment Analysis for a topic/politician.

We use a package called TexBlob, which has a number of very useful functions for processing textual data. To use those functions, we need to convert a string (text) to an object of TextBlob type.

First, we collect some data as we did before. Next, we use the TextBlob package to go through the dataframe one row at a time and find the text – in this case a tweet stored in a variable/column named 'text'. Once we have that tweet, convert it into a TextBlob object, and then we can ask it to analyze that string for subjectivity and polarity.

This script calculates the polarity and the subjectivity and adds in the data. The resulted file from this contains 10 columns: username, author id, created, text, retwc, hashtag, followers, friends, polarity, subjectivity. We change the query to different topics/politicians and re-run the python script to collect tweets and calculate polarity and subjectivity(sentiments) for each topics/politicians (Gun control).

We create a new python script(TwitterGun.py) for analysing the above resulted file. We again find the correlation between variables in the resulted file. We create regression model, plot and line for variables having medium to high level positive/negative correlation. We check the R-square value and the p-value. Based on the regression plot and model, we state the hypothesis such as More the followers, more the tweets sentiment is becoming objective, as the subjectivity is getting close to zero.

I have selected Steve Smith, Kim Jong, and Trump. So 180 tweets are collected for each personalities using tw_search.py and around 50 tweets using twitter_sentiments.py.

For Steve Smith:

We find the correlation between variables in the dataset created by twitter_sentiments.py Correlation of Followers and Friends= -0.178951077396 Correlation of Followers and Polarity= 0.0921856495584 Correlation of Followers and subjectivity= -0.137120912875 Correlation of Friends and Polarity= 0.201494410746 Correlation of Friends and subjectivity= 0.159399656592

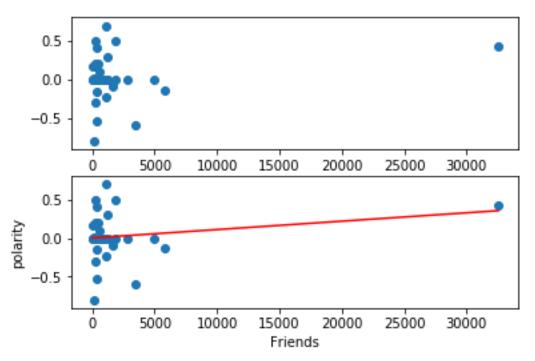
So we first take the variables Friends and Polarity as it has the highest correlation among the variables (0.2014) which is low positive correlation. We try to predict Polarity using Friends.

		OLS Regre	ssion Re	sults		
Dep. Varia Model: Method: Date: Time: No. Observ Df Residua Df Model: Covariance	rations: als:	polarity OLS Least Squares Fri, 30 Mar 2018 13:42:55 45 43 nonrobust	Adj. F-sta Prob Log-L AIC: BIC:	ared: R-squared: tistic: (F-statistic ikelihood:	:):	0.041 0.018 1.820 0.184 -2.3156 8.631 12.24
	coef	std err	t	P> t	[0.025	0.975]
const friends	0.0010 1.091e-05	0.041 8.09e-06	0.024 1.349	0.981 0.184	-0.082 -5.4e-06	0.084 2.72e-05
Omnibus: Prob(Omnib Skew: Kurtosis:	ous):	8.661 0.013 -0.419 5.556	Jarqu Prob(1.800 13.566 0.00113 5.36e+03

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified. [2] The condition number is large, 5.36e+03. This might indicate that there are strong multicollinearity or other numerical problems. const 0.000970 friends 0.000011 dtype: float64

We get a R-square value of 0.041 for friends and polarity. The equation of the regression line is Polarity = 0.000011*Friends + 0.000970 Regression plot :

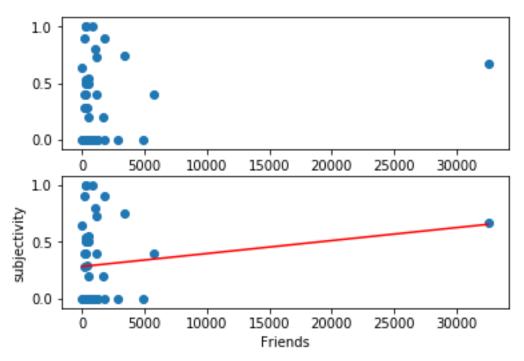


Based on the regression plot and model, We can say that More the friends, more the polarity is becoming from neutral (0) to positive. (0.5), so the tweets sentiments is becoming positive.

We then take the variables Friends and subjectivity which has (0.1593) low positive correlation. We try to predict subjectivity using Friends.

	0LS Regres		ults				
Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model: Covariance Type:	subjectivity OLS Least Squares Fri, 30 Mar 2018 20:32:18 45 43 1	Adj. F F-stat Prob (-squared:	Lc):	0.025 0.003 1.121 : 0.296 -15.429 34.86 38.47		
coe	f std err	t	P> t	[0.025	0.975]		
const 0.282 friends 1.146e-0		5.146 1.059	0.000 0.296	0.172 -1.04e-05	0.393 3.33e-05		
======================================	6.217 0.045 0.771 2.258			:	1.767 5.493 0.0641 5.36e+03		
Warnings: [1] Standard Errors specified. [2] The condition nustrong multicollinea const 0.282629 friends 0.000011 dtype: float64	mber is large, 5.3	86e+03. T	his might				

We get a R-square value of 0.025 for friends and subjectivity. The equation of the regression line is Subjectivity = 0.000011*Friends + 0.282629 Regression plot :

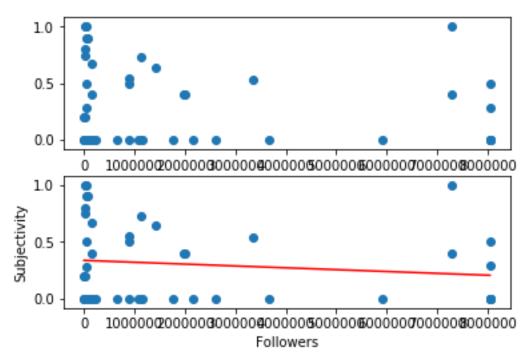


Based on the regression plot and model, We can say that More the friends, more the tweets sentiment is becoming subjective, as the subjectivity is becoming close to 1.

We then take the variables Followers and subjectivity which has (-0.1371) low negative correlation. We try to predict subjectivity using Followers.

		OLS Rec	gressio	n Resul	ts		
Dep. Variable: Model: Method: Date: Time: No. Observations Df Residuals: Df Model: Covariance Type:	Fri	subjectivity OLS Least Squares Fri, 30 Mar 2018 20:32:18 45 43 1 nonrobust			Adj. R-squared: F-statistic: Prob (F-statistic):		
	coef	std err		t	P> t	[0.025	0.975]
	3366 4e-08	0.065 1.8e-08	5.1 -0.9		0.000 0.369	0.205 -5.26e-08	0.468 2e-08
Omnibus: Prob(Omnibus): Skew: Kurtosis:		0.6 0.6	042 J 662 P	urbin-W arque-B rob(JB) ond. No	era (JB) :) : 	1.806 4.719 0.0945 4.50e+06
	n number	is large, or other r 1	4.5e+0	6. This	might i		

We get a R-square value of 0.025 for friends and subjectivity. The equation of the regression line is Subjectivity = -1.634069e-08*Followers + 3.365781e-01 Regression plot :



Based on the regression plot and model, We can say that More the followers, more the tweets sentiment is becoming objective, as the subjectivity is getting close to zero.

For Kim Jong:

We find the correlation between variables in the dataset using tw_search.py Correlation of Followers and Friends= 0.132067668457 Correlation of Followers and retweets= -0.0644412837818 Correlation of Friends and retweets= -0.14146123666 OLS Regression Results

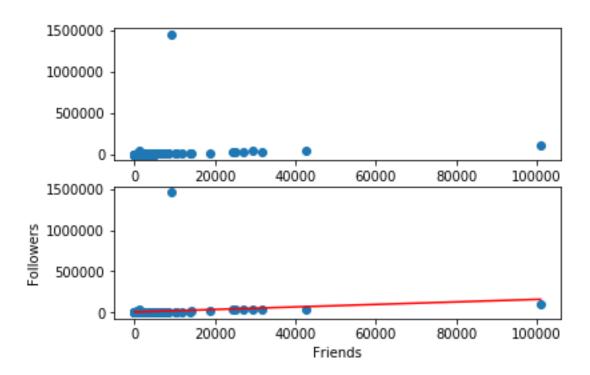
			221011 14			
Dep. Variable: Model: Method: Date: F Time: No. Observations: Df Residuals: Df Model: Covariance Type:		followers OLS Least Squares Fri, 30 Mar 2018 21:28:44 180 178 1	Adj. F-sta Prob Log-I AIC:	uared: R-squared: atistic: (F-statisti ikelihood:	c):	0.017 0.012 3.160 0.0772 -2340.4 4685. 4691.
	coef	std err	t	P> t	[0.025	0.975]
const friends	6562.2557 1.5252	8569.933 0.858		0.445 0.077	-1.03e+04 -0.168	2.35e+04 3.219
Omnibus: Prob(Omnibu: Skew: Kurtosis:	s):	395.486 0.000 13.244 176.942	Jarqı Prob		:	2.009 232181.559 0.00 1.06e+04

Warnings:
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 1.06e+04. This might indicate that there are strong multicollinearity or other numerical problems.

const 6562.255706
friends 1.525247
dtype: float64

We then take the variables Followers and friends which has (0.32067) low negative correlation.. We try to predict Followers using Friends.

We get a R-square value of 0.017 for friends and followers. The equation of the regression line is Followers = 1.5252*Friends + 6562.255706 Regression plot :



We find the correlation between variables in the dataset using twitter-sentiments.py

```
Correlation of Followers and Friends= -0.118811265005

Correlation of Followers and Polarity= -0.134251652549

Correlation of Followers and subjectivity= -0.00225257682909

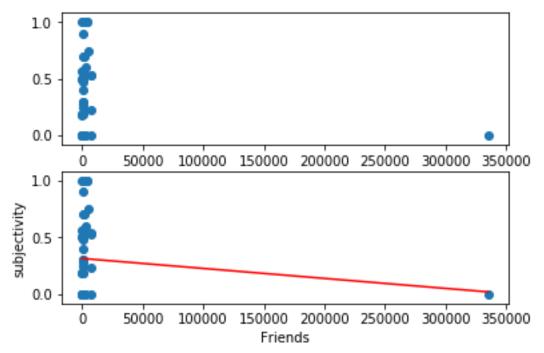
Correlation of Friends and Polarity= -0.02922533351

Correlation of Friends and subjectivity= -0.125836955729
```

We then take Friends and subjectivity variables which has (-0.1258) low negative correlation and we try to predict subjectivity using Friends.

We get a R-square value of 0.016 for friends and subjectivity. The equation of the regression line is Subjectivity = -8.820022e-07*Friends + 3.142514e-01

```
OLS Regression Results
0.016
Dep. Variable:
                        subjectivity
                                       R-squared:
                                 OLS
Model:
                                       Adj. R-squared:
                                                                       -0.005
                       Least Squares
                                                                       0.7723
Method:
                                       F-statistic:
                    Fri, 30 Mar 2018
                                                                        0.384
                                       Prob (F-statistic):
Date:
                                       Log-Likelihood:
Time:
                            21:51:13
                                                                       -14.765
No. Observations:
Df Residuals:
                                  50
                                       AIC:
                                                                        33.53
                                  48
                                       BIC:
                                                                        37.35
Df Model:
Covariance Type:
                           nonrobust
                 coef
                         std err
                                                P>|t|
                                                           [0.025
                                                                       0.9751
                                     6.598
                                                                        0.410
                                                            0.218
const
              0.3143
                           0.048
                                                0.000
friends
            -8.82e-07
                           1e-06
                                    -0.879
                                                0.384
                                                         -2.9e-06
                                                                     1.14e-06
Omnibus:
                               5.346
                                       Durbin-Watson:
                                                                        1.964
Prob(Omnibus):
                               0.069
                                       Jarque-Bera (JB):
                                                                        4.902
                                       Prob(JB):
Skew:
                               0.696
                                                                       0.0862
Kurtosis:
                               2.354
                                       Cond. No.
                                                                     4.82e+04
Warnings:
[1] Standard Errors assume that the covariance matrix of the errors is correctly
specified.
[2] The condition number is large, 4.82e+04. This might indicate that there are
strong multicollinearity or other numerical problems.
const
          3.142514e-01
friends
         -8.820022e-07
dtype: float64
```



Based on the regression plot and model, We can say that More the Friends, more the tweets sentiment is becoming objective, as the subjectivity is getting close to zero.

For Trump:

```
We find the correlation between variables in the dataset using twitter_sentiments.py Correlation of Followers and Friends= -0.130915199181 Correlation of Followers and Polarity= 0.024184757149 Correlation of Followers and subjectivity= -0.14211631085 Correlation of Friends and Polarity= -0.0170203212416 Correlation of Friends and subjectivity= -0.14998220444
```

So we first take the variables Friends and subjectivity as it has the highest correlation among the variables (-0.1499) which is low negative correlation. We try to predict subjectivity using Friends.

Dep. Variable:		subjecti	vitv	R-squ	ared:		0.022
Model:		542,551	OLS		R-squared:		0.002
Method:		Least Squ	ares		tistic:		1.105
Date:		Fri, 30 Mar	2018	Prob	(F-statisti	.c):	0.299
Time:		23:38:34		Log-L	-8.9812		
No. Observation	s:		50	AIC:			21.96
Df Residuals:			48	BIC:			25.79
Df Model:			1				
Covariance Type	:	nonro	bust				
	coef	std err		t	P> t	[0.025	0.975]
const	 0.3114	0.044	-	 7.005	0.000	0.222	0.401
friends -4.8	59e-06	4.62e-06	-1	1.051	0.299	-1.42e-05	4.44e-06
Omnibus:			 .017	Durbi	======= n-Watson:		2.664
Prob(Omnibus):			.081		e-Bera (JB)	:	4.818
Skew:		6	.754	Prob(0.0899
Kurtosis:		2	.797	Cond.	No.		1.02e+04

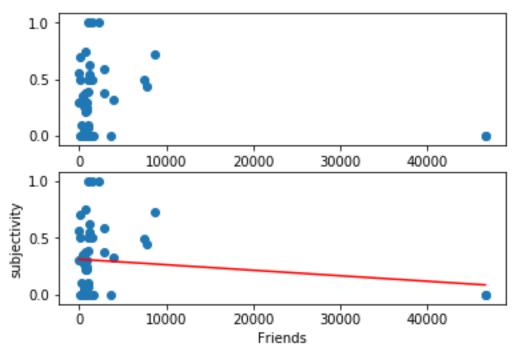
Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.02e+04. This might indicate that there are strong multicollinearity or other numerical problems.

const 0.311353 friends -0.000005 dtype: float64

We get a R-square value of 0.022 for friends and subjectivity. The equation of the regression line is Subjectivity = -0.000005*Friends + 0.311353

Regression plot:



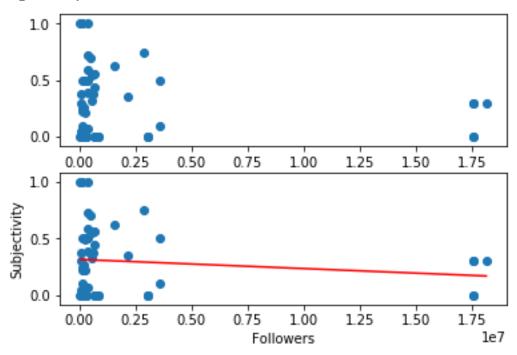
Based on the regression plot and model, We can say that More the Friends, more the tweets sentiment is becoming objective, as the subjectivity is getting close to zero.

We take the variables Followers and subjectivity as it has the second highest correlation among the variables (-0.1421) which is low negative correlation. We try to predict subjectivity using Followers.

		egression	=====	========			
Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model: Covariance Type:	subjectivity OLS Least Squares Fri, 30 Mar 2018 23:38:34 50 48 1 nonrobust		Adj. R-squared: F-statistic: Prob (F-statistic):			0.020 -0.000 0.9894 0.325 -9.0399 22.08 25.90	
co	ef std err	======	t	P> t	[0.025	0.975]	
const 0.31 followers -8.069e-		6.7 -0.9	799 995	0.000 0.325	0.222 -2.44e-08	0.409 8.24e-09	
Omnibus: Prob(Omnibus): Skew: Kurtosis:	0	.117 .685			:	2.591 4.156 0.125 6.34e+06	
Warnings: [1] Standard Errors specified. [2] The condition n strong multicolline const 3.1536 followers -8.0693 dtype: float64	number is large earity or other 65e-01	, 6.34e-	+06.	This might			

We get a R-square value of 0.020 for followers and subjectivity. The equation of the regression line is Subjectivity = -8.069399e-09*Followers+3.153665e-01

Regression plot:



Hypothesis:

Based on the regression plot and model, We can say that More the Followers, more the tweets sentiment is becoming objective, as the subjectivity is getting close to zero.