

Christopher Feliz August 2020 ~ Springboard Data Science Fellow

TWEETS CAN'T HURT RIGHT...?



Tesla shares tank after Elon Musk tweets the stock price is 'too high'

PUBLISHED FRI, MAY 1 2020-11:16 AM EDT | UPDATED FRI, MAY 1 2020-4:01 PM EDT



NEGATIVE SENTIMENT

- Tesla stock price fell 11%.
- Loss of investors capital.
- Raises concerns if
 Musk is mentally fit
 to run the company.



Replying to @elonmusk

Dude...l just lost \$10k because of this tweet. Wtf is wrong with u







POSITIVE SENTIMENT

- Retweeted over 3.4 millions times, 3rd place all time.
- Free marketing for Wendy's Restaurant.
- Great publicity for the brand.
- Created an entire movement with over a million people in order to get free nuggets for a year.



IS THERE A WAY TO INTERPRET TWEETS?

- Twitter is a platform where individuals are allowed to post their thoughts and opinions as they see fit.
- But for some it can do more damage than good, such as having the ability to drop a company's stock price.
- On the other hand Wendy's tweet went viral with a total of 3.4million retweets. This tweet generated massive amounts of free marketing and good publicity for Wendy's.

THE GOAL

In this project I plan to build a model that can classify a tweet as containing positive, negative or neutral sentiments.

This model can help companies understand the public's perception of them, both good and bad. As well as help brands who would like the benefits of free marketing from going viral on Twitter.

METHODOLOGY

- Tweets are limited to 280 characters, which means users are more likely to use acronyms.
- In order to provide as much info to my machine learning algorithms I
 iterated through each tweet and replaced any acronyms with their full
 meaning.
- Continued with basic NLP preprocessing removing numbers, special characters, and links.

DATA PREPROCESSING SAMPLE

clean_tweets	sentiment	text
few grilled mushrooms and olives feta cheese a	neutral	few grilled mushrooms and olives, feta cheese
more days till bh comes back to laughing aloud	neutral	94 more days till bh comes back to la
laughing out loud i know and hahadid you fall	negative	lol i know and hahadid you fall asleep?? o
wanted to visit the animals but whatever wer	negative	http://twitpic.com/663vr - wanted to visit the
in spoke to you yesterday and you didnt respon	neutral	in spoke to you yesterday and u didnt respond
so i get up early and i feel good about the da	positive	so i get up early and i feel good about the da
enjoy you are night	positive	enjoy ur night
wish whatever could come see you on denver hus	negative	wish we could come see u on denver husband I
ive wondered about rake to the client has made	negative	i've wondered about rake to. the client has
yay good for both of you enjoy the break you	positive	yay good for both of you. enjoy the break - y
but it was worth it	positive	but it was worth it ****.

MODEL SELECTION + PERFORMANCE

	Models	CountVectorizer	TFIDFVectorizer	Accuracy	F1 Score ▼	Difference
1	Convolutional NN*			0.70	0.70	94% ▲
2	Support Vector Classifier	~	×	0.70	0.69	92% ▲
3	Basic NN*			0.69	0.69	92% ▲
4	Random Forest Classifier	~	×	0.68	0.68	89% ▲
5	Naive Bayes Multinominal	~	×	0.65	0.64	78% ▲
6	DummyClassifier	~	×	0.35	0.36	0% =

*Neural Network

The best performing classification model was the Convolutional Neural Network with a accuracy and F1 score of .70%.

The differences in accuracy between our baseline model and the Convolutional Neural Network was 94%.

FINAL CONCLUSIONS

- After looking at the performances of all the models it's no surprise that most of them struggled with predicting the 'neutral' class. It's easier to predict 'positive' or 'negative' sentiments because they'll usually have a couple of words that lean towards one sentiment or another.
- In regards to using either
 CountVectorizer or TFIDF for data
 transformation there wasn't a significant
 difference in overall accuracy.
- For future iterations I plan on collecting more data, this will especially help in determining 'neutral' sentiments. We could also implement pretrained embedding layers such as word2vec or glove.