# Bitcoin Price Influence based on Sentiment Analysis of Relevant Tweets

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# Introduction

#### **Motives**

- Retail investors and social media as a medium for investing
- How is the price influenced by retail sentiment and vice-versa
- Inflow of investors, as market cap increased ~8 times from late 2020 to peak 2021



### Datasets

### **Training dataset**

- Kaggle: Bitcoin tweets sentiment analysis
  - 1.5 million BTC tweets from 2021
  - Sentiment tagged with VADER Sentiment
  - Used to create our own sentiment-tagging model

- API data
  - 500,000 tweets from 17/05/2022 23/05/2022
  - Collected using Twitter's Developer API

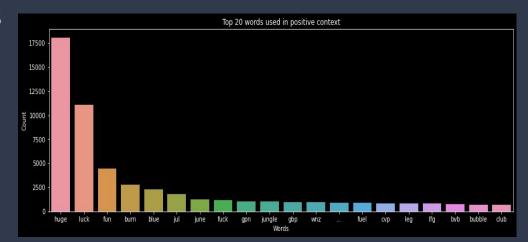
	date	tweets	score
0	2021-02-05 10:52:04	AT_USER AT_USER AT_USER right here w/ AT_USER	0.0000
1	2021-02-05 10:52:04	AT_USER AT_USER please donate bitcoin19 donate	0.6597
2	2021-02-05 10:52:06	\$sos market cap is 308 million. if they're min	0.0000
3	2021-02-05 10:52:07	bitcoin btc current price (gbp): £34,880 like	0.3612
4	2021-02-05 10:52:26	AT_USER right here w/ AT_USER URL referral cod	0.0000

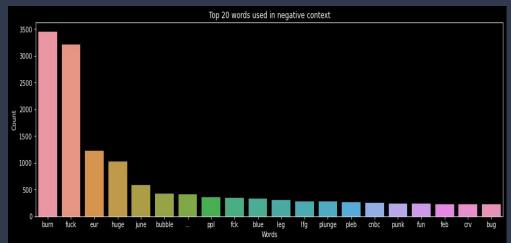
	Tweet	Date
0	current stats of delegatedonthate block find p	2022-05-19 23:59:59+00:00
1	bbcworld for all those who be new to this work	2022-05-19 23:59:58+00:00
2	smilingpunks floor price no gas fee polygon b	2022-05-19 23:59:56+00:00
3	i be claim my free lightning sat from bitcoine	2022-05-19 23:59:56+00:00
4	washingtonpost for all those who be new to thi	2022-05-19 23:59:54+00:00
(5	00000, 2)	

### In-depth Kaggle dataset properties

#### **Vader Sentiment**

- "gold-standard" lexicon
- Tailored for social media content
- Lexical features such as acronyms, initialisms and slang with sentiment value
- ~40% of training data comes from tweets
- Sensitive to both the polarity and the intensity of a sentence (



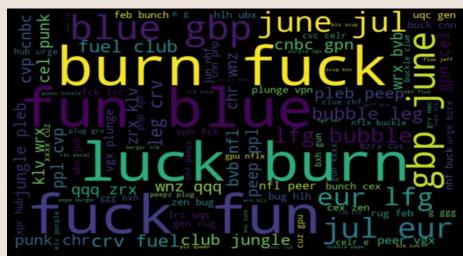


### 2018 Dataset



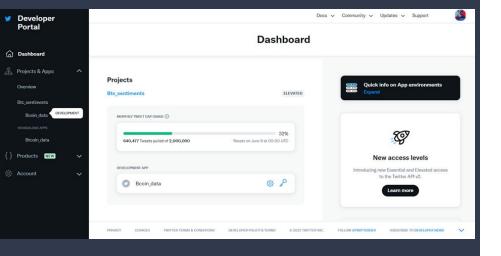
- ~ 50k tweets
- 0.0079 MSE on native test
- 0.1035 MSE on interpolated datasets

#### 2021 Dataset



- ~ 1.5m tweets
- 0.0411 MSE on native test
- Cannot perform interpolated testing because of limited vocabulary of the training set
- 0.0606 MSE on native test with sliced training

### API web crawling



- The Twitter API can be used to retrieve and analyze Twitter data.
- We crawled 500,000 tweets from the period between the 17th of May to the 24th of May 2022.
- We selected Tweets that contained any of the following words:
  - Bitcoin
  - o bitcoin
  - o Btc
  - o btc
  - o #Bitcoin
  - o #bitcoin
  - o #Btc
  - o #btc

# Preprocessing

### **Filtering**

- Stop words, punctuation, links, @s

### RegExp Tokenizer

Highly customizable

### **NLTK POS-tagging**

Solid choice out of the box

#### **WordNet Lemmatizer**

With WN converted POS-tag

## Model Training Specifications

**TF-IDF** (term frequency inverse document frequency) is used as our **baseline** for word representation.

Term frequency **TF** (number of times a term appears in a document) multiplied by Inverse Document Frequency **IDF** (log of the number of documents over the document frequency of term).

**Regression** model, as we are trying to predict sentiment on a **scale** of -1 to 1.

Ridge Regression from SKLearn excels for this type of regression

Estimates coefficients of **multiple-regression models** in scenarios where **linearly independent variables** are highly correlated.

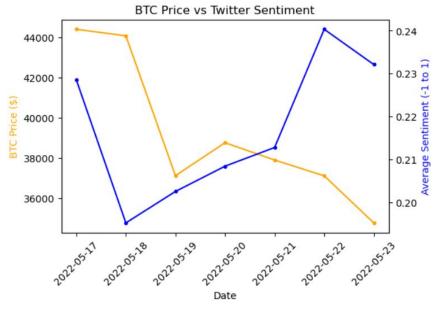
In this model, the **loss** function is a *linear least squares* and **regularization** is given by the *l2-norm* 

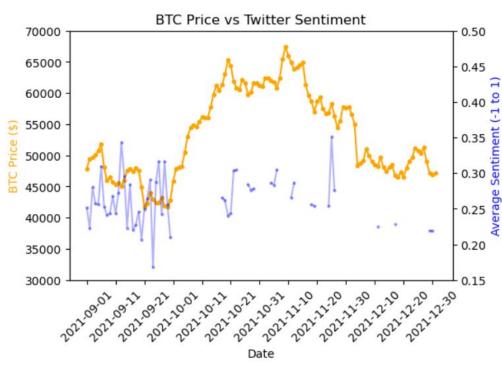
### Bert-Model

 We also decide to try train a model based on BERT (Bidirectional Encoders Representations from Transformers) which is both a contextual and bidirectional language model.

### Methodology

- We did text preprocessing (special tokens, padding, and attention masks) and built a Sentiment Classifier using BERT.
- The model was supposed to be Trained on the 1.5 Million Tweets Dataset:
  - Our features X are sequences of tweets.
  - Our target y is: the Sentiment tag for each Tweet.
- Finally, our model would be tested on the Tweets we Web Crawled through the Twitter Api and Evaluated by Accuracy score.





Graphs - 7 day vs 3 month

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Alexandrayuliu. (2021). *Bitcoin tweets sentiment analysis dataset*. Kaggle. <a href="https://www.kaggle.com/code/alexandrayuliu/bitcoin-tweets-sentiment-analysis/data">https://www.kaggle.com/code/alexandrayuliu/bitcoin-tweets-sentiment-analysis/data</a>

## References