

An abstract background image featuring a complex network graph. It consists of numerous small black dots (nodes) connected by thin, light gray lines (edges). The connections form a dense, interconnected web that fills the left side of the frame, with some areas appearing more clustered than others. The overall effect is a sense of digital connectivity and data flow.

# Using Reddit and Markov Chains to Time Bitcoin

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# Motivation

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- Camou (2022) shows that a sentiment analysis based on Reddit posts about cryptocurrencies has **predictive power** over crypto **volatility**. However, the results are **mixed** for predicting **returns**.
- Poyser (2018) tests the hypothesis that cryptocurrency prices are driven by **herding**. For that, he studies herding behavior under different conditions, including the **Markov-Regime-Switching** approach.
- In this work, we use **sentiment analysis** and the **Hidden Markov Model** (Baum & Petri (1966)) to get more (less) exposed to Bitcoin when there is a higher (lower) chance of the market being more bullish (bearish) in the next period.

# Data & Metodology

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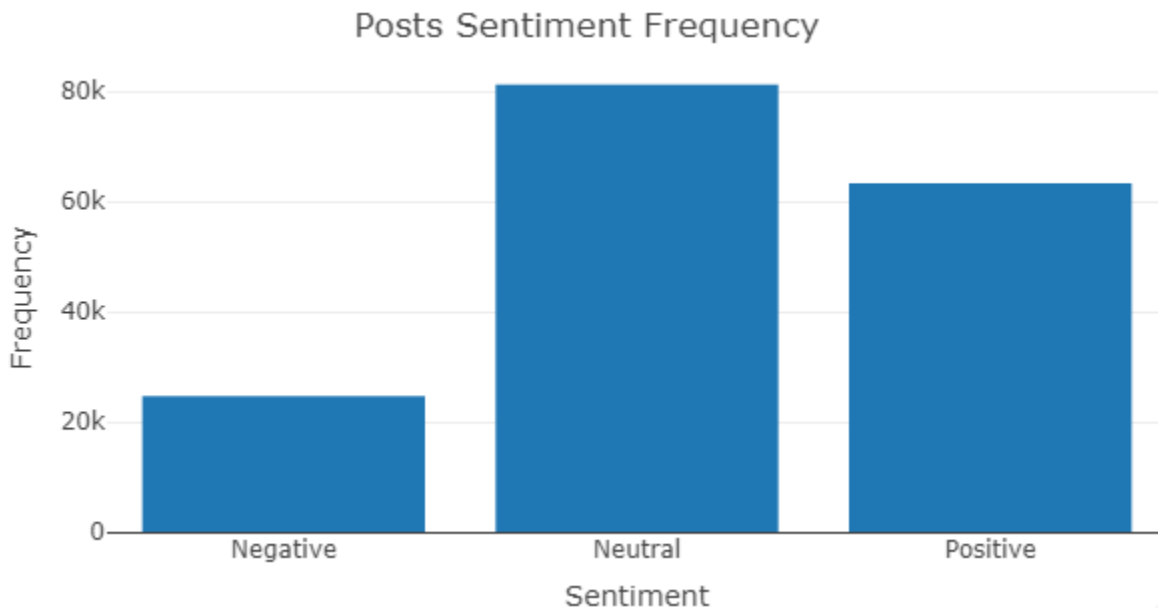
- We collect data regarding **Reddit posts** and comments from **2021-01-01** to **2022-10-01**. In total, we have **169155 text documents**, as well as information about comments and upvotes.
- First, we **preprocess** our text data, removing stopwords, punctuations, URLs, numbers, emojis, etc.
- After that, as in Camou (2022), we apply **VADER analysis** (Hutto & Gilbert (2014)) to assign an **intensity score** to every observation, indicating if it is positive or negative.

# Data & Metodology

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- Following, for **every week** in our sample, we take the **weighted average** of the **VADER score** based on the **post score** (difference between upvotes and downvotes).
- Thenceforth, we use this weekly score to model the Bitcoin returns' **hidden states**. We specify that there are **two hidden states**: a **bullish** and a **bearish** one.
- Finally, we go **long X%** in **Bitcoin** and **1 – X% long** in **cash** or the **risk-free** rate. In this case, X is the **probability of being in the bullish state** in the next week.

# NLP Analysis



# Results



# Results

Table 1 – Portfolio Return Statistics

	BTC-CDI	BTC-TBill	BTC-Cash	BTC	T-bill
Annualized Return	-89.6%	-90.8%	-90.8%	-96.4%	2.2%
Annualized Volatility	93.9%	93.9%	93.9%	125.3%	
Modified Sharpe Ratio	-0.84	-0.85	-0.85	-1.21	
Max. Drawdown	59.5%	60.5%	60.5%	71.3%	
CVaR	-13.1%	-13.2%	-13.2%	-17.7%	
Skewness	-0.25	-0.25	-0.25	-0.24	
Kurtosis	-0.32	-0.32	-0.32	-0.23	

# Conclusion and Future Developments

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- Since the **scrapping** process (API limits) and **NLP** analysis (preprocessing) are **slow**, we had to limit our sample to just under **two years**. On account of this fact, the **inferences** are **limited**. Despite this warning, using sentiment analysis and the HMM, we were able to form **better portfolios** in comparison to the plain Bitcoin one.
- Interested researchers can improve the model by adding **more data** (longer time horizon) and by exploring **different structures**. The work could also be expanded by adding **other cryptocurrencies** and getting data from **other social platforms**, like Twitter.



# Open Science

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# References

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