

# Reddit Cryptocurrency Comments Analysis

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May 31, 2022

Social Media Analytics

# Agenda

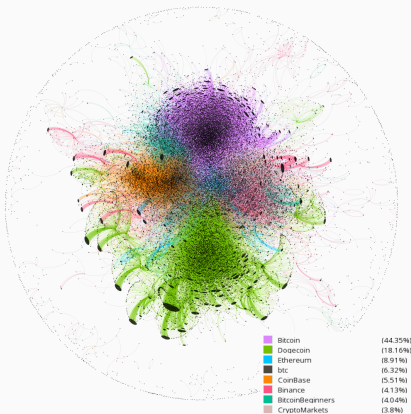
1. Overview
2. Trials and Errors
3. Data collection
4. Data Models
5. Analysis
6. Homemade implementation
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# Overview

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# Overview

- The main idea is to find a link between the activity in cryptocurrency subreddits and the price of **Bitcoin**.
- For that we modeled the data in an appropriate way and ran several algorithms seen in class.



## **Trials and Errors**

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The original idea was to perform Sentimental Analysis (SA) on Tweets.

- The Twitter API is unfortunately very limited.
- ⇒ We took Reddit as a Social Network instead.
- Reddit is less suitable to perform Sentimental analysis.
- ⇒ We decided to analyse the network and find interesting results as well as correlation between subreddit's activity and Bitcoin price.

## Data collection

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- We used Reddit API for data collection.
- API calls are performed by a bot running every 20 minutes on several Cryptocurrency related subreddits.
- Storage is optimal in a JSON format because of Reddit comment structure.





# Data Models

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# Data Models

We have tried four types of data models.

1. Unique Cartesian Link: all commenters from a specific post are linked together with an undirected edge.
2. Next Link: only main commenters of the post are linked to the Original Poster (OP) with a directed edge.

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User 0 (OP, depth=1, score=30)

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User 1 (d=1, s=10)

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User 2 (d=2, s=10)

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User 3 (d=2, s=2)

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User 4 (d=1, s=5)

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User 5 (d=1, s=3)

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User 6 (d=2, s=2)

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User 3 (d=3, s=1)

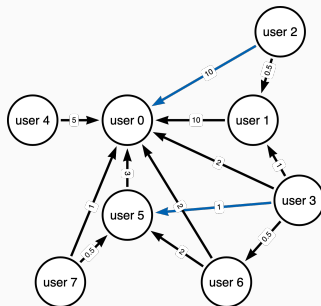
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User 7 (d=2, s=1)

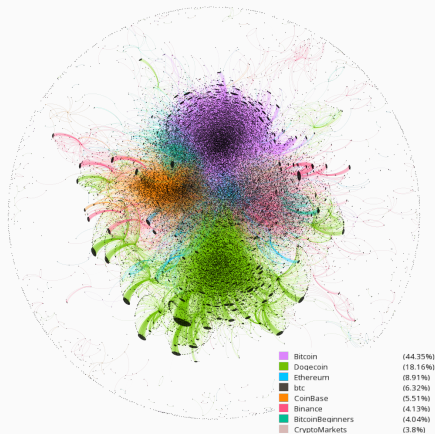
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3. Deep Link: commenters are recursively linked together with a directed edge (i.e. capturing the relation “has commented”) with a weight of relevance according to the depth of comments.
4. Deep Link No merge: same as Deep Link without merging the different edges. This results in a directed multigraph.

User 0 (OP, depth=1, score=30)
User 1 (d=1, s=10)
User 2 (d=2, s=10)
User 3 (d=2, s=2)
User 4 (d=1, s=5)
User 5 (d=1, s=3)
User 6 (d=2, s=2)
User 3 (d=3, s=1)
User 7 (d=2, s=1)



- Deep Link No merge model represented on the right.
- Tree-like structure may be observed for large comments.
- Most imposing subreddits are Bitcoin, Dogecoin and Ethereum.

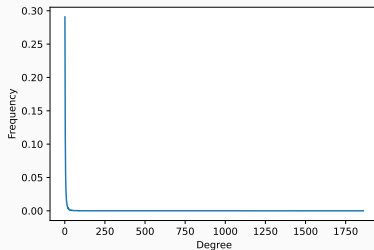


# Analysis

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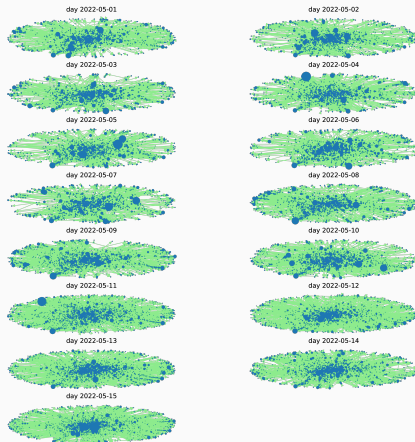
# Power law distribution

- The graph data model follows a Power law distribution.
- This most likely means that the data model is well thought out.



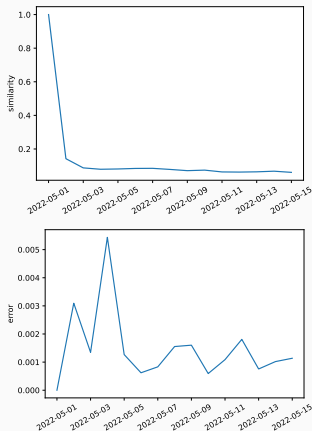
# PageRank

- The PageRank in the network can be interpreted as being the probability of commenting one user's post when browsing the given subreddits.
- Yet, the PageRank alone does not yield very surprising results: users with the most upvoted and commented comments have the highest PageRank.



# PageRank

- We therefore took interest in evaluating the **steadiness** of the PageRank along the days.
- We concluded two things.
  1. There is a tiny similarity between user bases along the days. Two days do not have a lot of users in common.
  2. There is a tiny error between PageRank intersections. Two days share about the same PageRank values for the same users.
- These two observations help us conclude that the topics and posts are lasting in time and this is consistent with other works [1, 4].



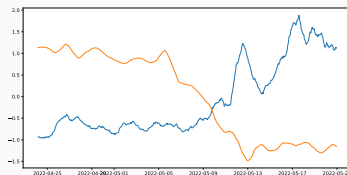
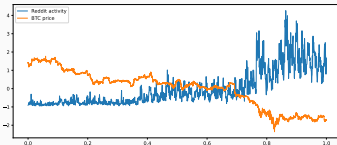


# Correlation to price development

- We made an inclusion sequence of graph with a time window of 1h.

$$G_1 \subset G_2 \subset G_3 \subset \dots \subset G_n$$

- We used degree centrality to compare each Graph  $i$  with its subsequent graph  $i + 1$ .
- We match these data with the ones of the Bitcoin price rate.
- We applied some smoothing with the window average method.

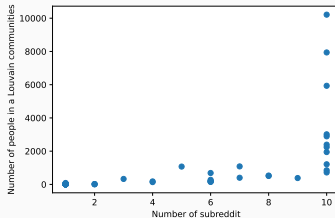
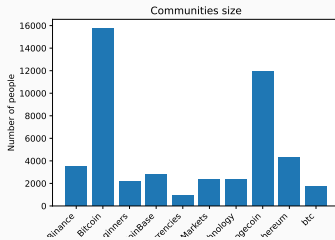


- The correlation we've found is of  $-0.85$ .

- Yet, correlation does not capture causal link.
- Causal link is difficult to assess between activity in subreddits and price rate.
- “Extracting Cryptocurrency Price Movements from the Reddit Network Sentiment.” shows with Granger causality tests, that there exists indeed a causal link between the two.
- Correlations have also been proven in “Cryptocurrency Return Prediction Using Investor Sentiment Extracted by BERT-Based Classifiers from News Articles, Reddit Posts and Tweets.”

# Louvain community detection method

- We observed that Louvain communities are formed from people belonging to the same subreddit.
- Some communities spread over multiple subreddits.
- The larger the community the more it spreads over different subreddits.
- Communities have a foothold in one unique subreddit but also spread on the nine other subs.



# Homemade implementation

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# PageRank

1. “NetworkX” and homemade implementation yields the same results.
2. Yet, the benchmarks show a night and day difference.
3. Using a dense matrix is the main reason behind that.
4. It is a direct consequence of the Power Law distribution.

implementation	nodes	edges	time	RAM
NetworkX	15537	58150	0.78s	0.128G
homemade (dense)	15537	58150	91.21s	5.81G

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NetworkX	15537	58150	0.78s	0.128G
homemade (dense)	15537	58150	91.21s	5.81G
homemade (sparse)	15537	58150	2.18s	0.215G

**Table 1:** Benchmarks of PageRank algorithm on Fedora Linux 36 with Intel i7-8550U (8) @ 4.000GHz and 16GB of RAM

- Our implementation was very slow but yielded the same result as the one from “NetworkX”.
- Cause of that is the computation of modularity gain, which could have been optimised.
- We only ran the benchmark on a subset of the total graph.

implementation	nodes	edges	time	RAM
"NetworkX"	1000	4852	0.37s	3.478MB
homemade	1000	4852	18.76s	3.235MB

**Table 2:** Benchmarks of Louvain algorithm on macOS Monterey with Apple M1 and 8Go of RAM



## Discussion

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Any questions?

### References

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- [1] Maria Glenski, Emily Saldanha, and Svitlana Volkova. “Characterizing Speed and Scale of Cryptocurrency Discussion Spread on Reddit.”. In: *The World Wide Web Conference, WWW 2019, San Francisco, CA, USA, May 13-17, 2019*. 2019, pp. 560–570. DOI: 10.1145/3308558.3313702. URL: <https://doi.org/10.1145/3308558.3313702>.
- [2] Aric A Hagberg, Daniel A Schult, and Pieter J Swart. “Exploring Network Structure, Dynamics, and Function Using NetworkX”. In: (2008), p. 5.

- [3] Duygu Ider. “Cryptocurrency Return Prediction Using Investor Sentiment Extracted by BERT-Based Classifiers from News Articles, Reddit Posts and Tweets.”. In: *CoRR* abs/2204.05781 (2022). DOI: 10.48550/arXiv.2204.05781. URL: <https://doi.org/10.48550/arXiv.2204.05781>.
- [4] Xiao Li and Linda Du. “A Multi-window Bitcoin Price Prediction Framework on Blockchain Transaction Graph.”. In: *Algorithmic Aspects in Information and Management - 15th International Conference, AAIM 2021, Virtual Event, December 20-22, 2021, Proceedings*. 2021, pp. 317–328. DOI: 10.1007/978-3-030-93176-6\_27. URL: [https://doi.org/10.1007/978-3-030-93176-6\\_27](https://doi.org/10.1007/978-3-030-93176-6_27).

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