CSE 544 - Class Project

Cryptocurrency Analysis

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Motivation

Cryptocurrencies? All aboard the hype train!

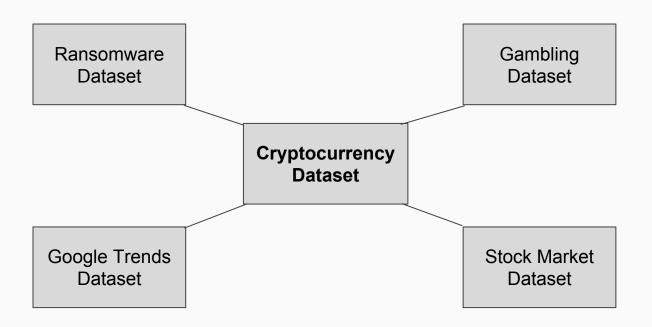
Affects multiple domains

Potential to revolutionize currencies

 Volatile in nature, hence challenging to analyze



About the Dataset



Main Dataset: bitcoin dataset

Information about the price, market cap, mining difficulty, etc of bitcoin

| Date | btc_market_price (USD) | btc_total_bitcoins (Count) | btc_n_transactions (Count) | btc_market_cap (USD) |
|------------|---------------------------|-------------------------------|-------------------------------|-------------------------|
| 2018-02-19 | 11110.965 | 16875062.5 | 187367 | 187498228810 |
| 2018-02-20 | 11390.3916 | 16876825 | 198455 | 192233646840 |

- Data size
 - Daily bitcoin data from 2010 to 2018
 - o 2920 rows and 24 feature columns
- Data Preprocessing
 - Normalization for analysis

Additional Dataset 1 - Google "Bitcoin" keyword search dataset

Information about Search frequency (adjusted), bitcoin daily closing price, and Date

| Date | btc_market_price (USD) | Searches |
|------------|---------------------------|----------|
| 2013-04-28 | 0.701172 | 3 |
| 2018-02-18 | 55.127267 | 26 |

- Data Size
 - Daily bitcoin data and Google Trends data from 2013 to 2018
 - 261 rows and 3 feature columns
- Data Preprocessing
 - Scaled btc_market_price to 0 -100 (in the range of Searches)

Additional Dataset 2 - Stocks

Stock details of ASIC companies - NVIDIA, AMD, TSM

| Date | Open (USD) | High (USD) | Low (USD) | Close (USD) | Volume (USD) |
|------------|---------------|---------------|--------------|----------------|-----------------|
| 2017-11-09 | 205.27 | 206.33 | 200.370 | 205.32 | 23895006 |
| 2017-11-10 | 213.08 | 218.67 | 211.630 | 216.14 | 31300857 |

- Data Size (from 2010 to 2018)
 - NVIDIA 4734 rows
 - o AMD 8738 rows
 - o TSM 3202 rows
- Data Preprocessing
 - Normalization

Additional Dataset 3 - GDP of Greece

 Contains the year quarter and the increase or decrease in GDP for that quarter in percentage.

| Time | Value |
|---------|-----------|
| 2013-Q1 | -2.192256 |
| 2013-Q2 | 0.077293 |

Data Size: 18 rows

Quarter 1 of 2013 - Quarter 1 of 2017

Additional Dataset 4 - Ransomware

Number of different attacks that took place per month

| Year | Month | Count |
|------|-------|-------|
| 2016 | 4 | 14 |
| 2016 | 5 | 13 |

- Data Size
 - o Data from Jan 2015 to July 2017: 31 rows
- Data Preprocessing
 - Normalization

Additional Dataset 5 - Gambling

Information about player expenditure in 5 different gambling

| Month | Year | Casino (USD) | EGM (USD) | Keno (USD) | Lottery (USD) | Wagering (USD) |
|-------|------|-----------------|--------------|---------------|------------------|-------------------|
| 7 | 2004 | 45662132.51 | 145766780.77 | 6836926.58 | 32451660.78 | 27298552.41 |
| 8 | 2004 | 36652961.84 | 143627917.85 | 4215418.49 | 24712127.1 | 26147748.54 |

- Data
 - Monthly player expenditure in USD from July 2004 till April 2017
- Data Processing and Cleaning
 - Refactored Game Stream as features (Casino, ..., Wagering)
 - Normalization

List of Topics

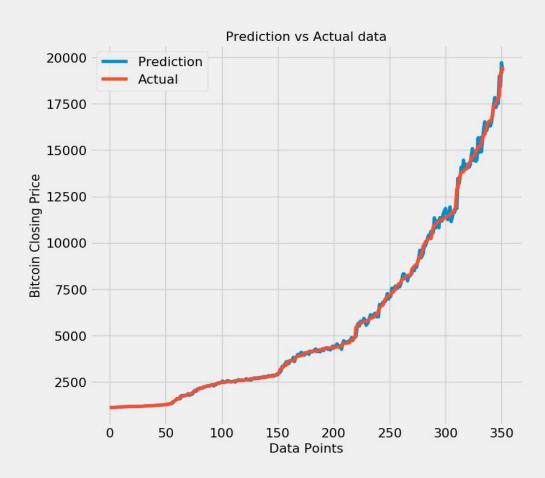
- Bitcoin price analysis and effect on other Cryptocurrencies
- Effect of Google Search on Bitcoin price
- Bitcoin and Finance
- Cryptocurrency and Ransomware
- Cryptocurrency and Gambling

1. Bitcoin price analysis and effect on other Cryptocurrencies

- Time series analysis of Bitcoin closing price
 - EWMA alpha = 0.5, 0.8
 - \circ AR with p = 352
 - Seasonal with p = 352
- Metric: Mean Absolute Percentage Error

| Method | MAPE |
|----------|---------|
| EWMA | 0.9990 |
| AR | 1.5037 |
| Seasonal | 89.6630 |

EWMA gives a better estimate for the bitcoin closing price

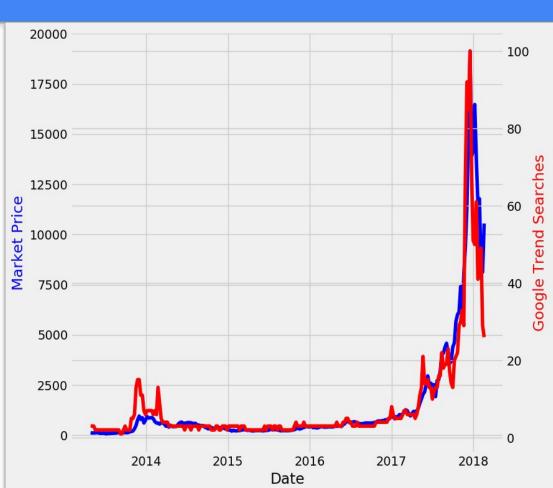


2. Effect of Google Search on Bitcoin price



Regression Analysis

 Simple Linear Regression to predict bitcoin market price using frequency of "Bitcoin" searches on Google

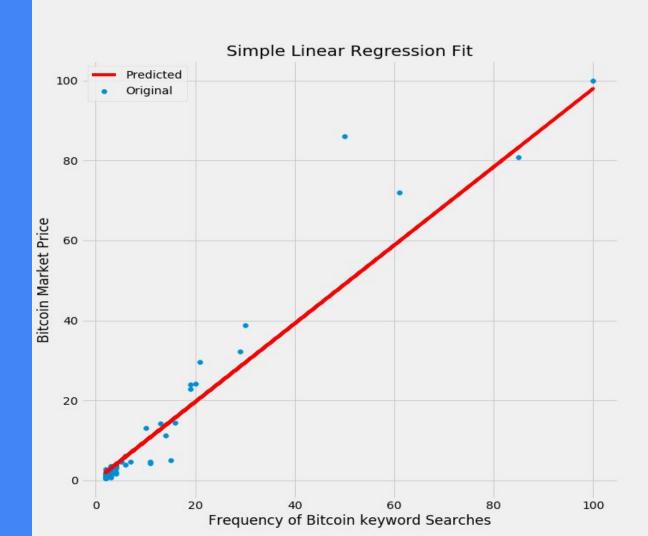


Metrics

- SSE 2051.36
- MAPE 52.93

Conclusion

The SSE and MAPE values are very bad, and hence we cannot predict bitcoin price from trends



3. Bitcoin and Finance



Predicting Stocks with Bitcoin

- Multiple linear regression
- Prediction
 - Stock price of an ASIC and GPU making company
- Features used
 - btc_market_price
 - btc_total_bitcoin
 - btc_trade_volume
 - btc_n_transaction

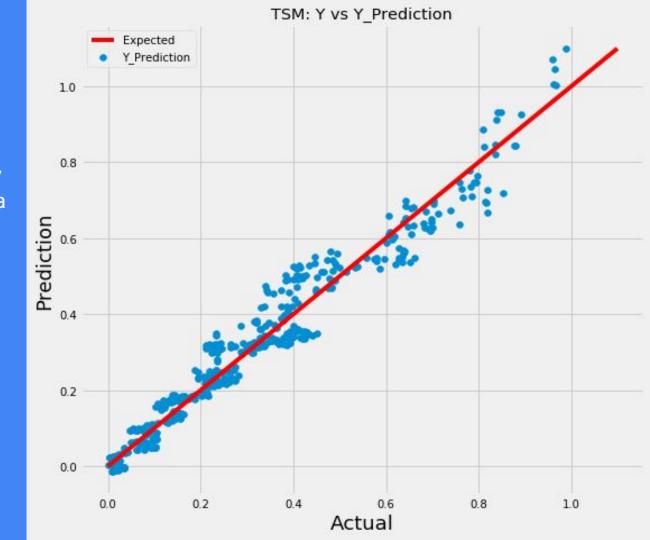
Results

| Company | SSE | MAPE |
|---------|--------|----------|
| NVIDIA | 1.4475 | 126.0607 |
| AMD | 9.2806 | 103.6849 |
| TSM | 1.0901 | 23.0741 |

Given bitcoin features linearly fit the stock market price data with the highest weight for btc_market_price

Weights For TSM:

btc_market_price: 0.6518 btc_total_bitcoins: 0.2934 btc_trade_volume: -0.1899 btc_n_transactions: 0.3536



4. Cryptocurrency and Ransomware attacks



Comparing the distribution of number of attacks and bitcoin per transaction cost

- Average data per month:
 - Number of attacks vs bitcoin cost per transaction
- Wald's 2 population test for alpha = 5%

| Statistic | P-Value | Confidence Interval | Output | Conclusion |
|-------------|---------|---------------------|--------|------------|
| Wald's Test | 0.5767 | [-0.0796, 0.1430] | 0.5581 | Accepted |

5. Cryptocurrency and Gambling





Use of bitcoin has steadily risen with increase in gambling expenditure

- Monthly averaged of amount (in USD) spent in 5 different gambling types
- Wald's two population test for alpha 5%
- Permutation test for alpha 5%

| Gambling Type | Wald's Test (P-Value) | Confidence Interval | Permutation Test (P-Value) | Conclusion |
|---------------|--------------------------|---------------------|-------------------------------|------------|
| Casino | 0.3965 | [-0.0473, 0.1194] | 0.4026 | Accepted |
| EGM | 0.1755 | [-0.0276, 0.1515] | 0.1802 | Accepted |
| Lottery | 0.0165 | [0.0194, 0.1945] | 0.0185 | Rejected |
| Keno | 0.0063 | [0.0375, 0.2292] | 0.0082 | Rejected |
| Wagering | 0.4054 | [0.0492, 0.1218] | 0.4111 | Accepted |

1. Bitcoin price analysis and effect on other Cryptocurrencies

- a. Techniques Used:
 - i. (Parametric Inference) Estimated the mean of bitcoin price using MME.
 - ii. Time Series Analysis
 - iii. Multiple Linear Regression to predict ethereum closing price based on bitcoin dataset (miners_revenue, hashing_difficulty, bitcoin_market_price)
- b. Conclusion:
 - i. lambda MME = 0.3580
 - ii. EWMA performs better than Seasonal and AR model for TSA
 - iii. Given features are closely correlated with ethereum closing value hence give good prediction for ethereum value

2. Effect of Google Search on Bitcoin price

- a. Techniques Used:
 - i. Simple Linear Regression to predict bitcoin closing price based on frequency of keywords searched
 - ii. (Non parametric) KS Test Statistic, P value, Wald's 2 population Test Statistic, P value
 - iii. Multiple Linear Regression find out which feature of bitcoin is influenced most by google searches
- b. Conclusion:
 - i. Linear relationship exists between frequency of keywords searched and bitcoin_closing_price
 - ii. Frequency of keywords searched and bitcoin closing price come from same the distribution
 - iii. Number of transactions was influenced the most.

3. Bitcoin and Finance

- a. Techniques Used:
 - i. Multiple Linear Regression
 - ii. Permutation Test
 - iii. Simple Linear Regression
- b. Conclusion
 - i. Bitcoin features linearly fit stock prices with highest dependency on bitcoin market price
 - ii. The distribution of TSM and bitcoin market price are not same
 - iii. There is no relation between GDP of Greece and prices of bitcoin

4. Cryptocurrency and Ransomware

- a. Techniques Used:
 - i. (Non parametric) Wald's 2 population test Statistic, P value, 95% Confidence Interval
 - ii. (Parametric Inference) Find the MLE parameter for Normal Distribution
 - iii. (Bayesian Inference) Posterior Distribution analysis
- b. Conclusion:
 - i. Two samples come from the same distribution
 - ii. Found mu and sigma parameters.
 - iii. The posterior distribution follows Normal distribution given Normal prior

5. Cryptocurrency and Gambling

- a. Techniques Used:
 - i. Simple and Multiple Linear Regression
 - ii. (Non parametric) Wald's 2 population test Statistic, P value, 95% Confidence Interval
 - iii. (Non parametric)Permutation Test P value
- b. Conclusion:
 - i. There is linear relationship between expenditure in gambling and total bitcoins per year and price.
 - ii. Wagering contributes most to increase in number of bitcoins per year
 - iii. EGM contributes most to increase in price of bitcoins per year

References

1. Cryptocurrency Dataset: https://www.kaggle.com/sudalairajkumar/cryptocurrencypricehistory

 Google Trends Dataset: https://trends.google.com/trends/explore?q=Bitcoin

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Ransomware Dataset:
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 6_Ransomware_and_Businesses.pdf

5. All gambling dataset Queensland: https://data.qld.gov.au/dataset/all-gambling-data-queensland

6. Wagering and bitcoin news article: https://news.bitcoin.com/bitcoin-gamblers-wagered-4-5-billion-btc-2014/

 Quarterly GDP of Greece <u>https://data.oecd.org/gdp/quarterly-gdp.htm</u>