| Paper | yy | method | Tweets | Twitterfeatures | BitcoinFeatures | Output | acc/loss | Comments |
|---|---|---|---|---|---|---|---|---|
| KryptoOracle | 2019 | Decision trees **XGBoost** | #BTC#Bitcoin; 30 days 25000twts per day; English filtering; removing media # links | Twitter Sentiment **vader** with influential persons weights; Tweet volume | Previous price; previousClosePrice; MeanAvgClosePrice;MovingAvgScores Per Minute | BTC price of next minute | Visual on graph No Explicit numbers | Not neural |
| Predicting Fluctuations in Cryptocurrencies' Price using users' Comments and Real-time Prices | 2018 | Bi-directional LSTM | #Bitcoin#CoinListingNews, #cryptocurrency#cryptopumpdump#exchangehacked; per minute? **Feb 2016 →Dec 2017** | **Twitter** sentiment (technique not described) | Several price related features, Several bitcoin related features 10 sec and 10 minute interval | Price Next day? | PR = 60% Acc = 50% | Ill Defined |
| Recurrent Neural Network Based Bitcoin Price Prediction by Twitter Sentiment Analysis | Oct. 2018 | RNN | @BTCTN@cryptocurrency@CryptoYoda1338(@BitcoinMagazine@BitcoinForums@coindesk@rogerkver 1 Jan '15 → 31 dec '17 1 Jan '18 → 30 June '18 | **Twitter** sentiment score from accounts | Normalized price from Coinmarketcap | Price prediction next day | Acc 77.62%. | Own sentiment analyzer implemented |
| Short-Term Bitcoin Price Fluctuation Prediction Using Social Media and Web Search Data | Aug. 2019 | Polynomial regression | #bitcoin Filtered on English 16 Dec 17 → 20 Feb 19 7.5 mil twts Filter #,capitals,.. | Twitter sentiment VADER; Tweet Volume; **Google** search patterns google Trend Data(SVI); | Weighted Price and Volume of Transactions per day (bitcoincharts) 9 April 14 → 07 January 19 per | Price prediction | Google: acc 77.01% Twitter vol: acc 66.66% | Cocnlusion: **Sentiment not good Neural not good** |
| Deep Learning Approach to Determine the Impact of Socio Economic Factors on Bitcoin Price Prediction | 2019 | CNN & LSTM & GRU | Twitter sentiment; Prediction is 3 months | Sentiment, number of followers, likes and the number of tweets in a time range. | Bitcoin price per 5 minute mapped to per day; Gold price per day (datahub.io) | Bitcoin price | RMSE 47 | The features are tested **individually** LSTM BEST with bitcoin and twitter sentiment |
| Forecasting Bitcoin Price Fluctuation by Twitter Sentiment Analysis | 2020 | Random Forest Regression | 60 day tweets 92550 per minute 12 March '18 → 12 May filtered | Vader sentiment score | Bitcoin price from "BITSTAMP," "COINBASE," "ITBIT," "KRAKEN" summed and avg; Close Price; Opening Price; Closing Price; High price; Low price; | Bitcoin price per minute | 43.83% error minimum 21.84% error accuracy | |
| CNN-based multivariate data analysis for bitcoin trend prediction | 2020 | 1D CNN | "Bitcoin" 28 April '13 → 15 Feb 2020 | twitter sentiment + volume | 10 bitcoin price features 4 blockchain based features | **average bitcoin trend in the next k days.** | ACC = 75% | Too much data, Not sure if I can handle this Only 2 output outcomes |
| Forecasting the Early Market Movement in Bitcoin Using Twitter's Sentiment Analysis: An Ensemblebased Prediction Model | 2021 | ENSEMBLE various instances of the XGBoost classifiers | Stemmming May 22, '20 → June 22, '20 Keywords: bitcoin, Bitcoin, bitcoins, Bitcoins, BTC, XBT, satoshi, #bitcoin, #bitcoins, #XBT, #BTC, $XBT, $BTC | Vader Sentiment Score followers, likes, and retweets associated with each tweet. | closing price of Bitcoin, the final sentiment score, and the moving average of the last 100 data points | | | ILL defined |

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|---|---|---|---|---|---|---|---|---|
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