

A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light greenish-blue. They are positioned diagonally, with the blue one partially covering the green one.

Cryptocurrency and Seasonality

David Adoni



Problem Statement

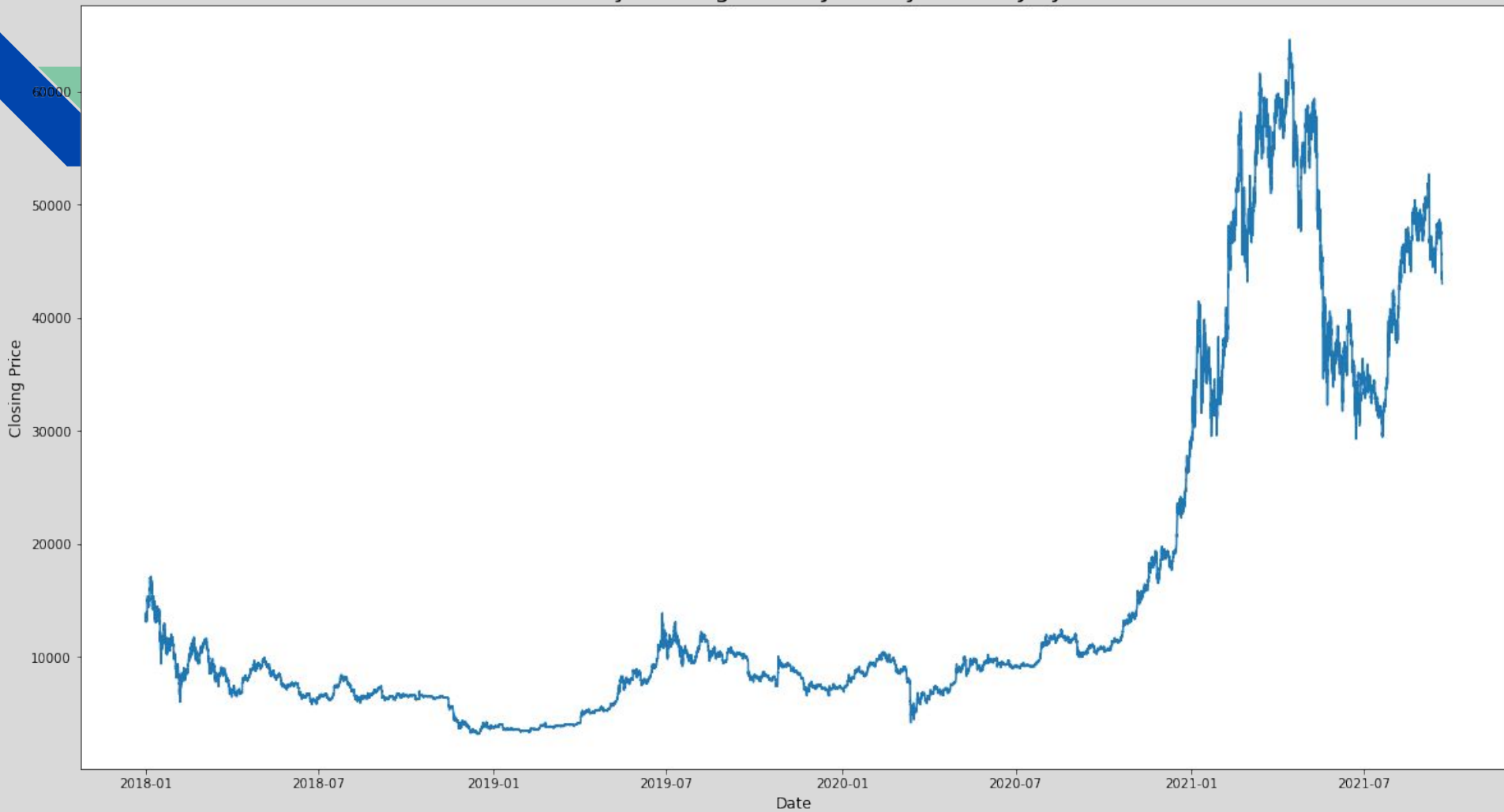
- Tasked by Coinbase to find if and when there are seasonal patterns in cryptocurrency markets
- Potential tool for forecasting crypto trade opportunities



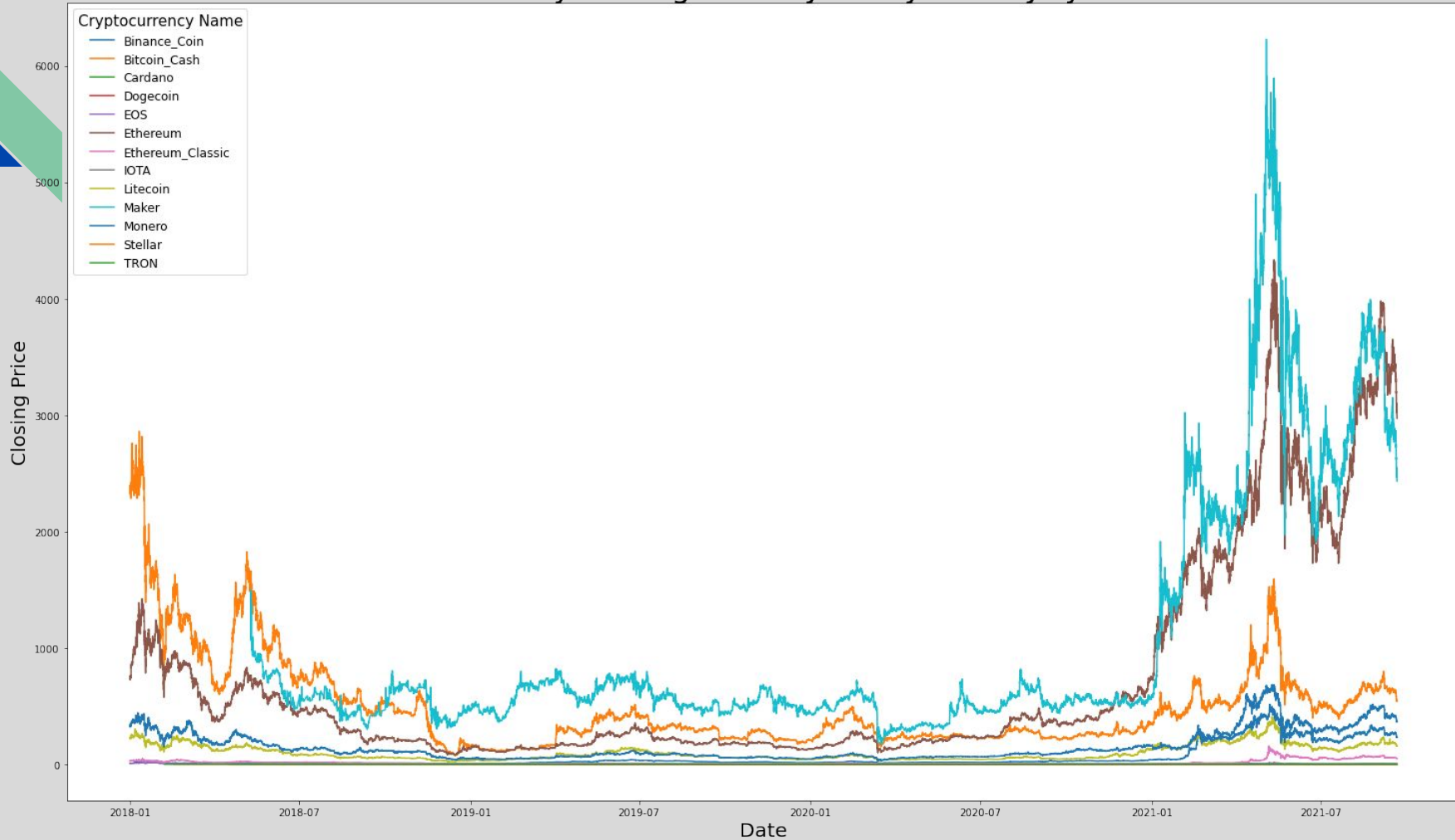
Data Collection/Cleaning

1. [Crypto Data](#) from Kaggle
2. Remove NaNs
3. Drop irrelevant columns
4. Create time series index
5. Separate data by cryptocurrency

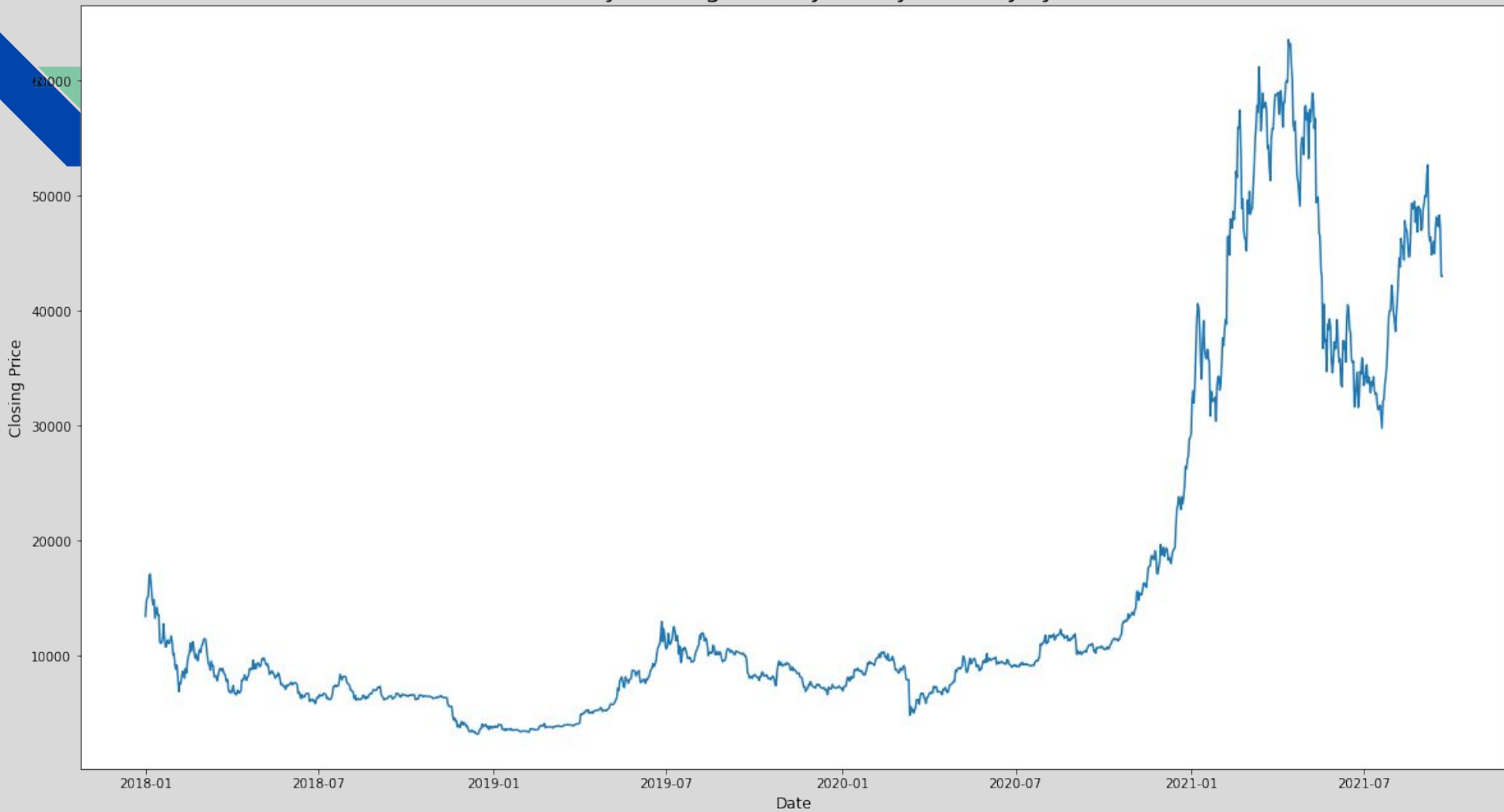
Bitcoin Hourly Closing Prices January 2018 - July 2021



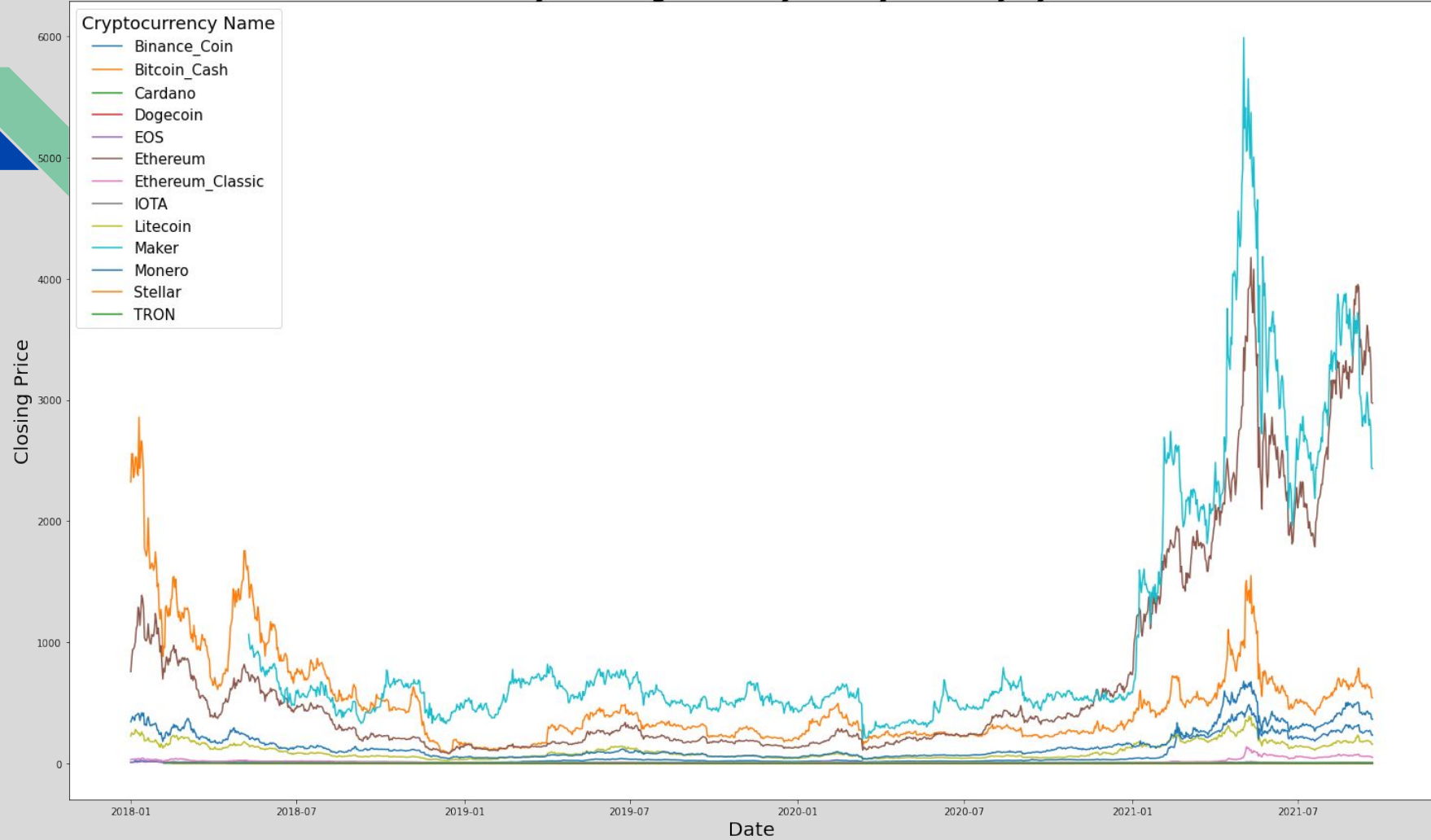
Altcoin Hourly Closing Prices January 2018- July 2021



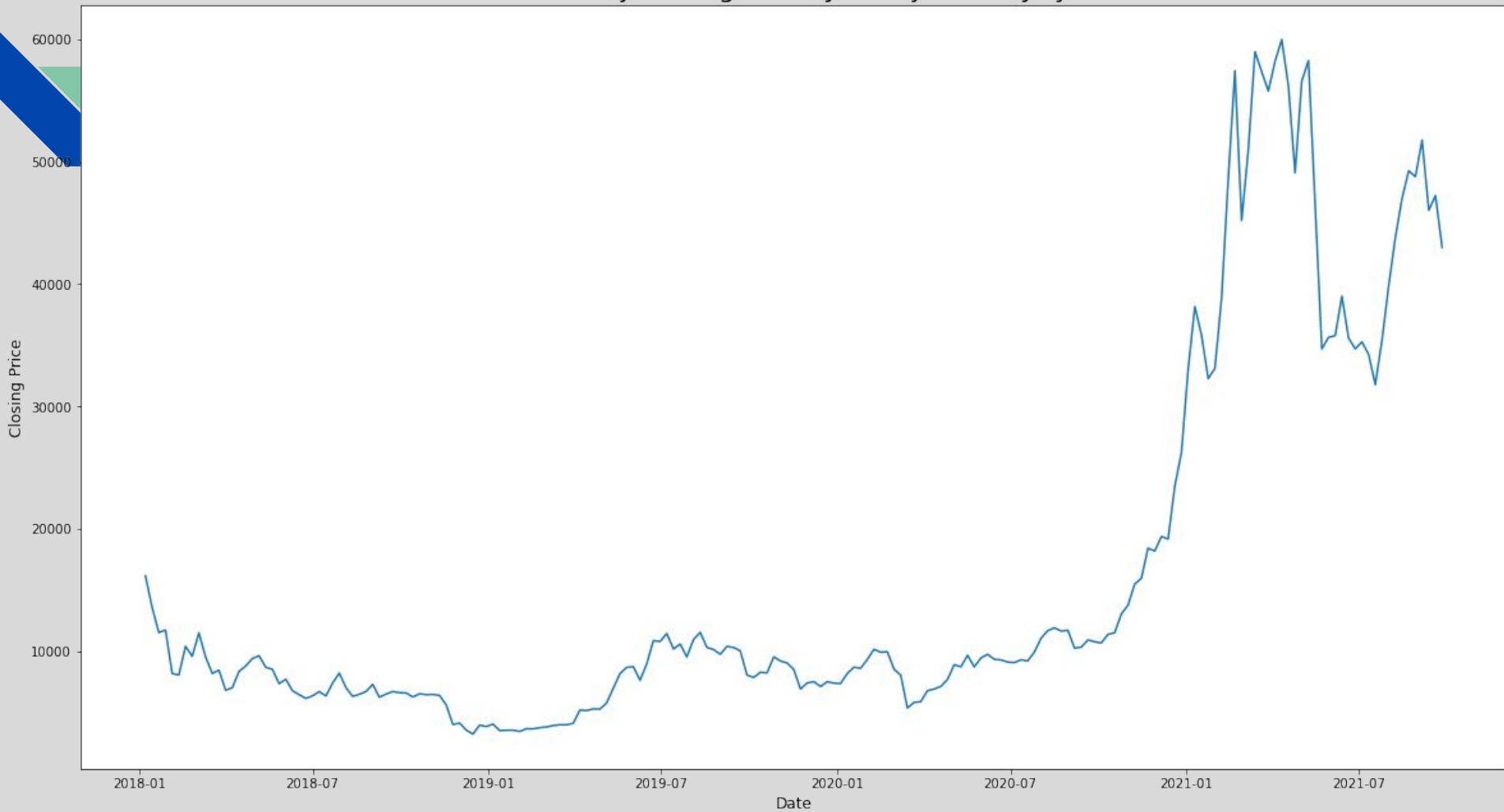
Bitcoin Daily Closing Prices January 2018 - July 2021



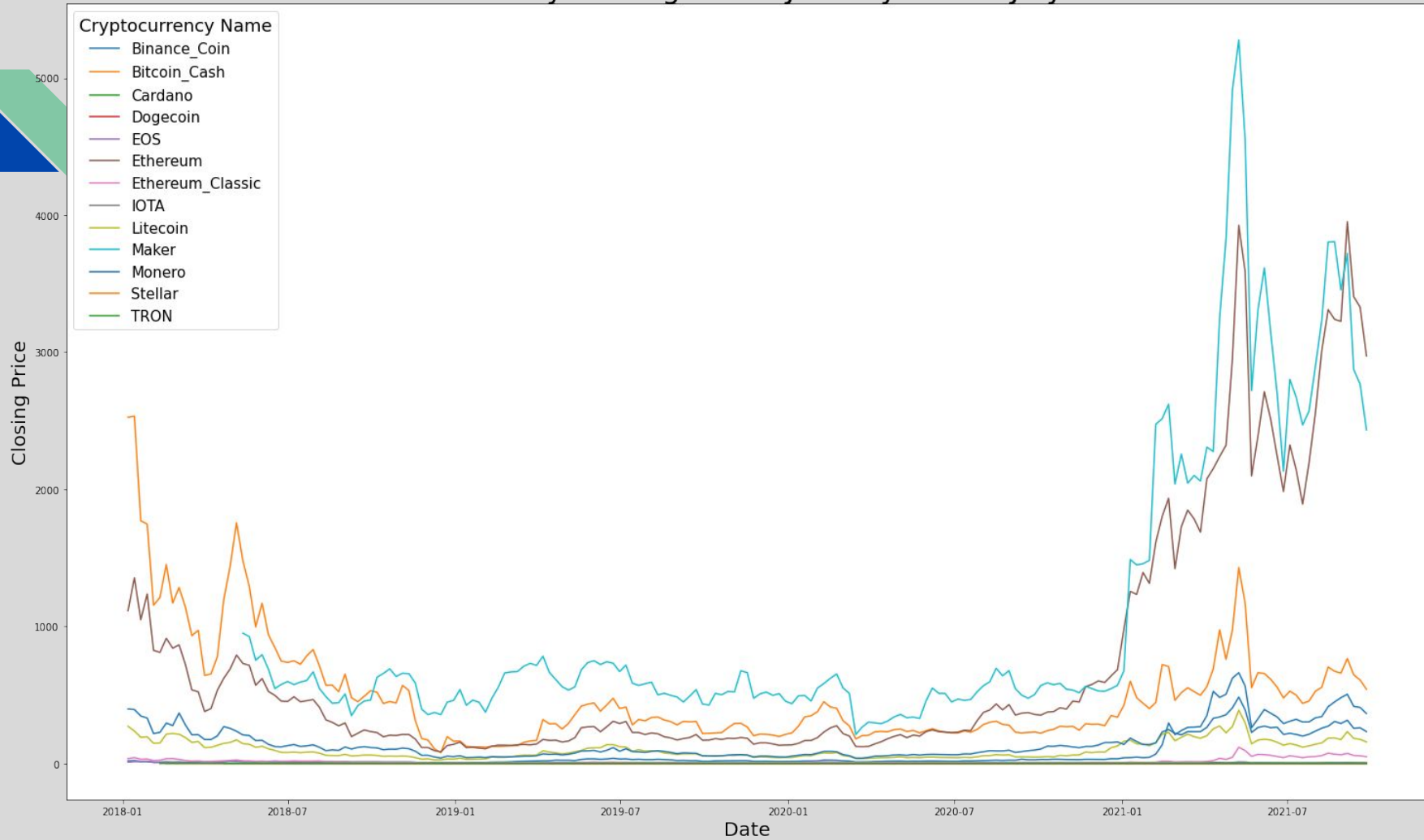
Altcoin Daily Closing Prices January 2018- July 2021



Bitcoin Weekly Closing Prices January 2018 - July 2021

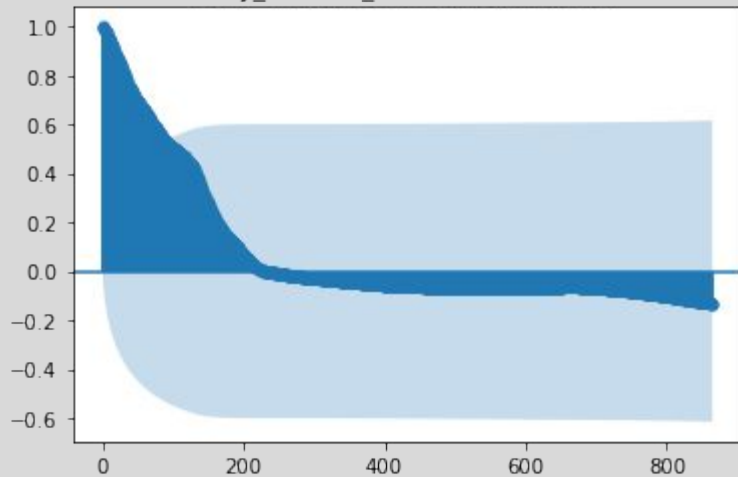


Altcoin Weekly Closing Prices January 2018 - July 2021

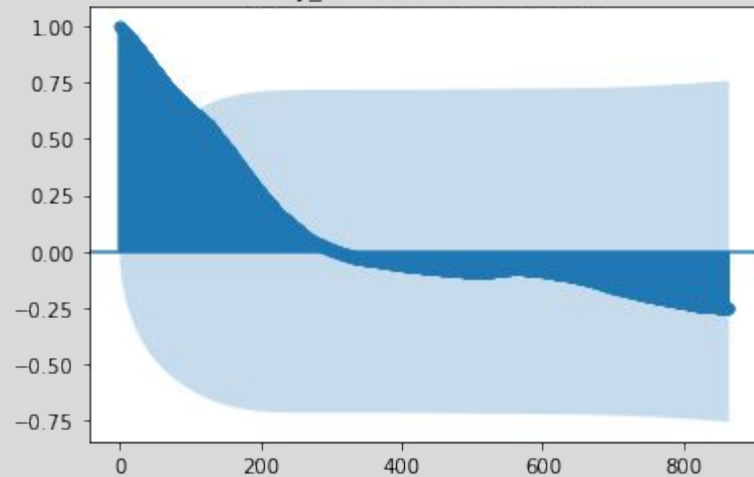




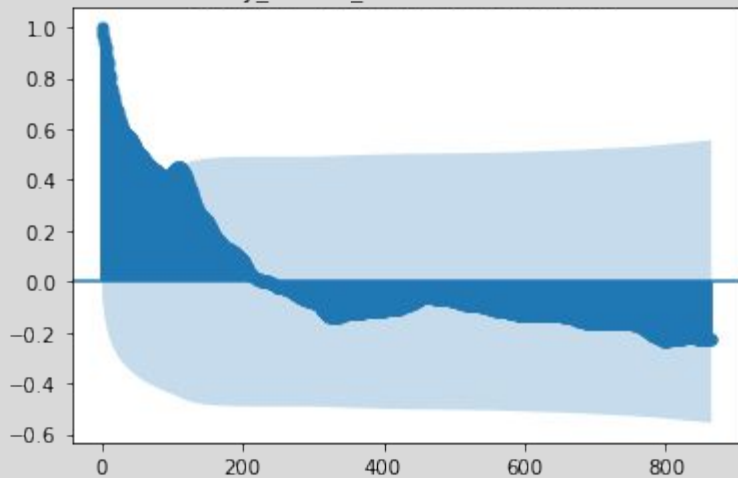
daily_Binance_Coin Autocorrelation



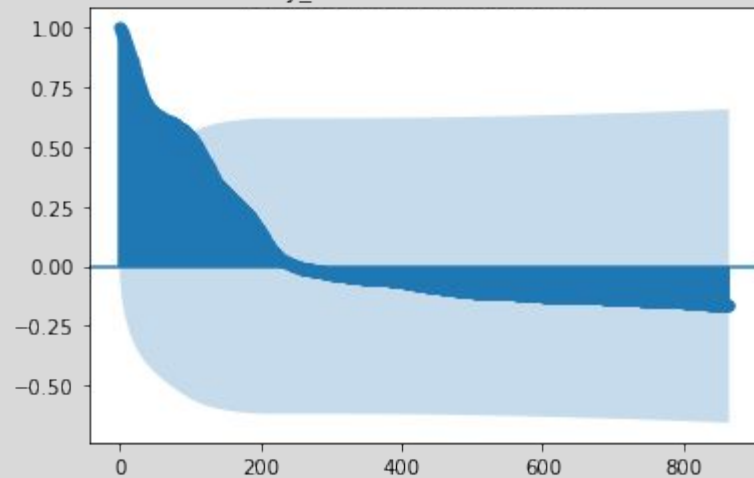
daily_Bitcoin Autocorrelation



daily_Bitcoin_Cash Autocorrelation

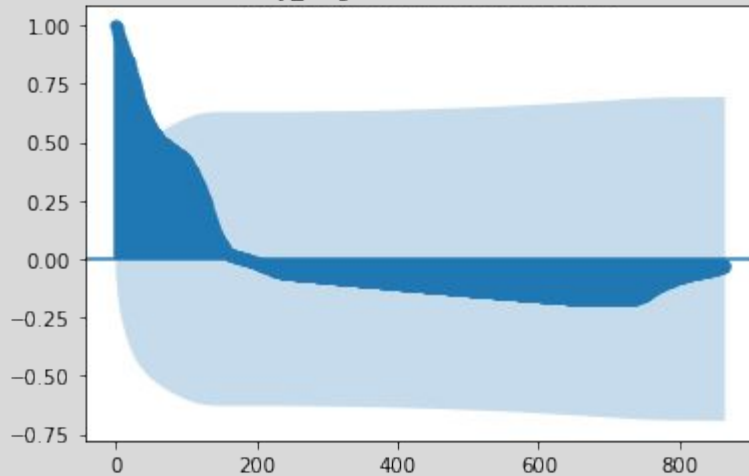


daily_Cardano Autocorrelation

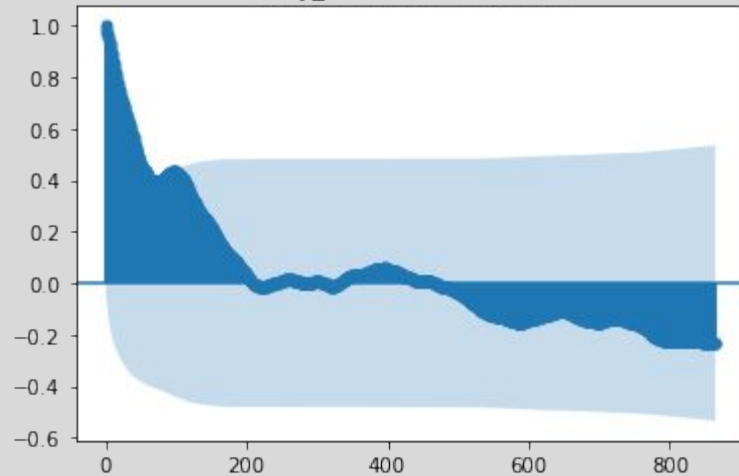




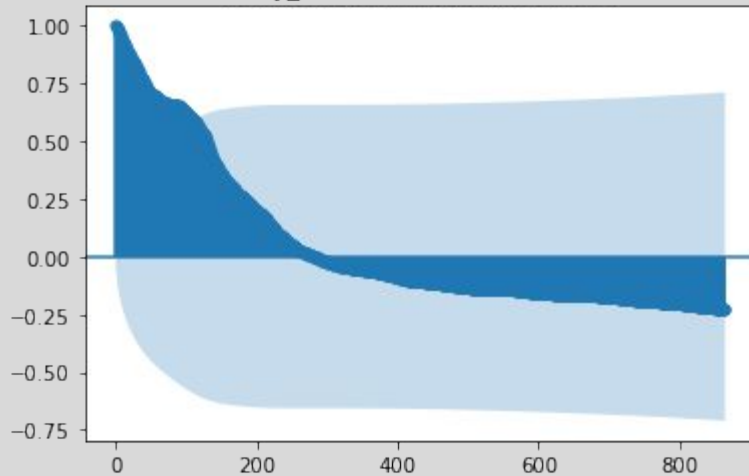
daily_Dogecoin Autocorrelation



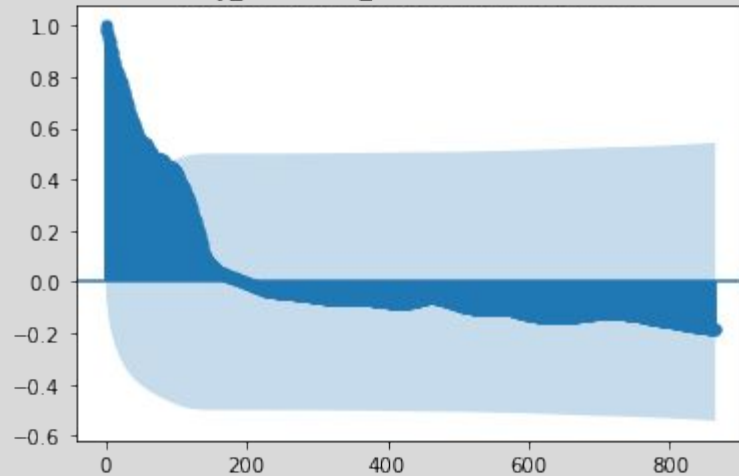
daily_EOS Autocorrelation



daily_Ethereum Autocorrelation

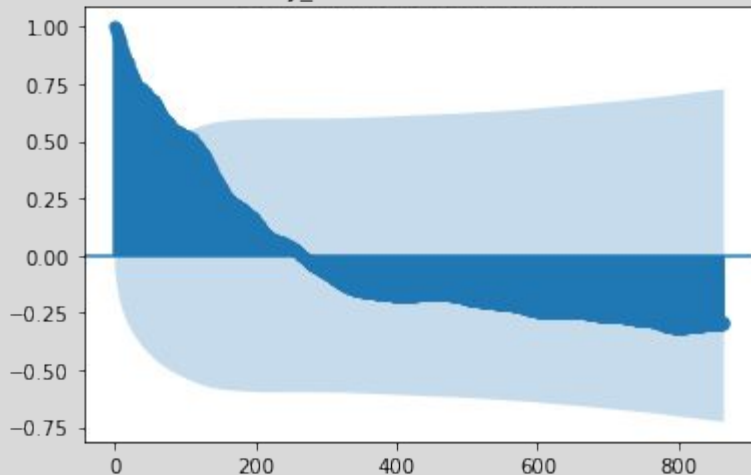


daily_Ethereum_Classic Autocorrelation

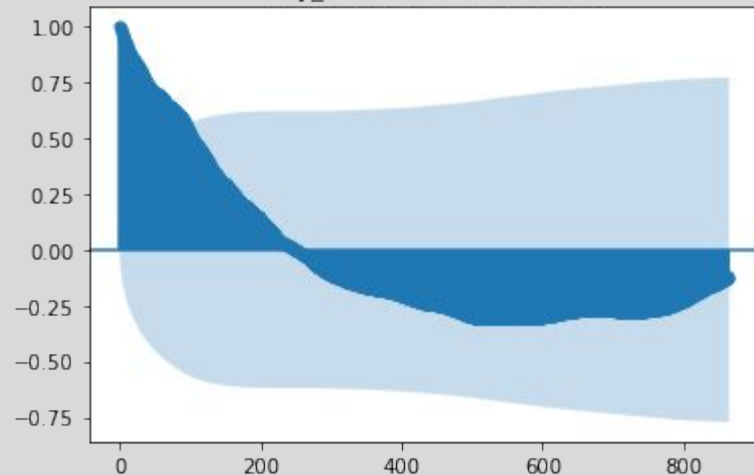




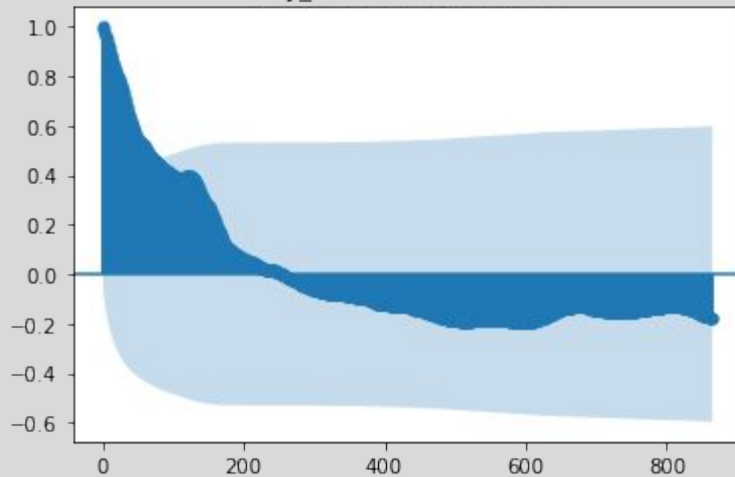
daily_Monero Autocorrelation



daily_Stellar Autocorrelation

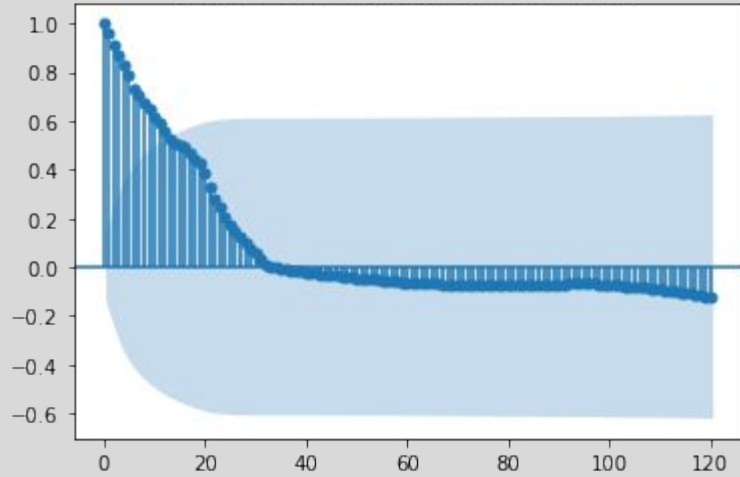


daily_TRON Autocorrelation

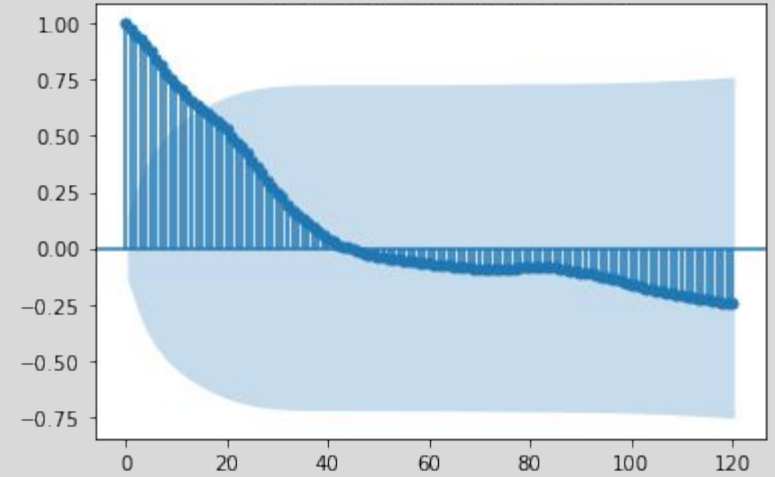




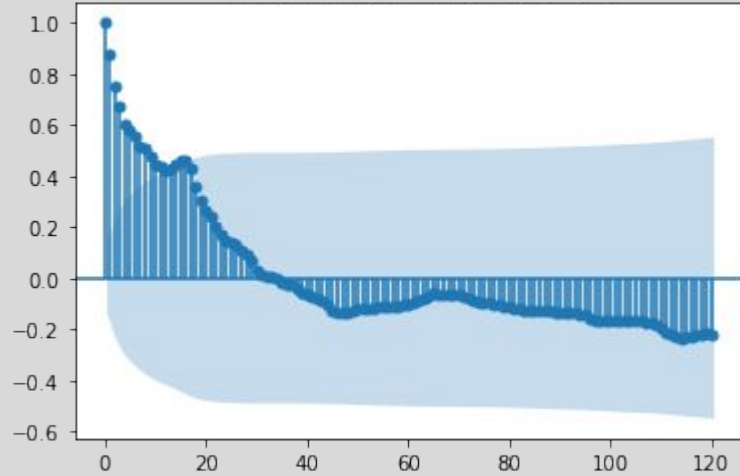
weekly_Binance_Coin Autocorrelation



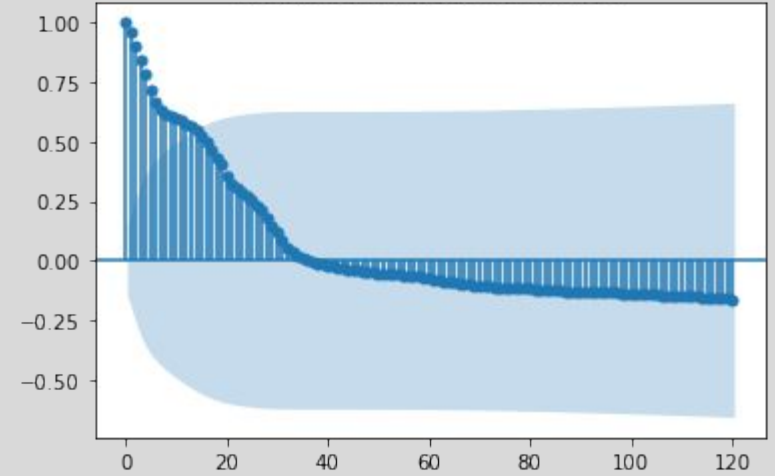
weekly_Bitcoin Autocorrelation



weekly_Bitcoin_Cash Autocorrelation

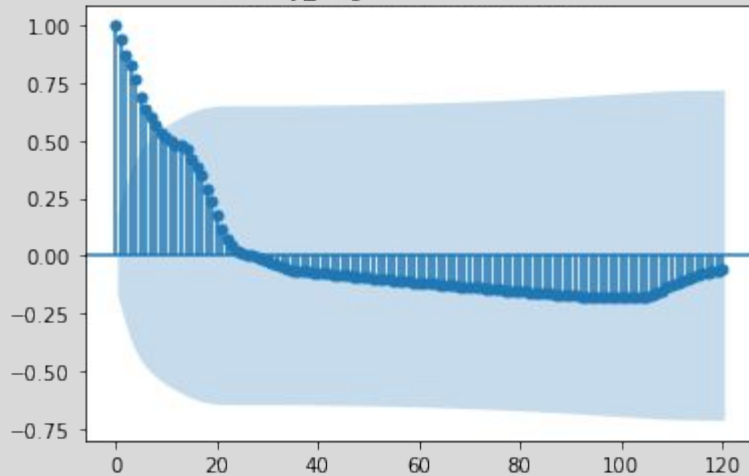


weekly_Cardano Autocorrelation

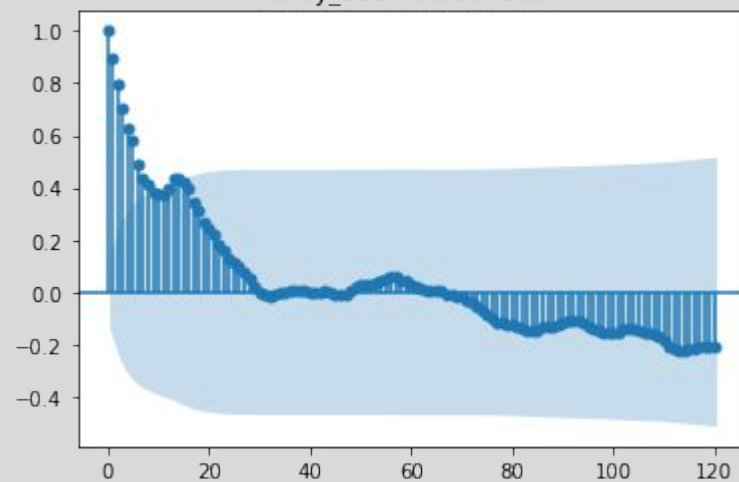




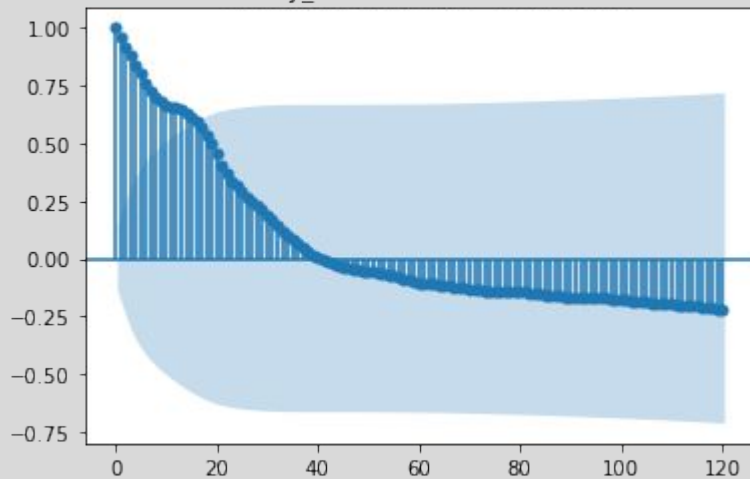
weekly_Dogecoin Autocorrelation



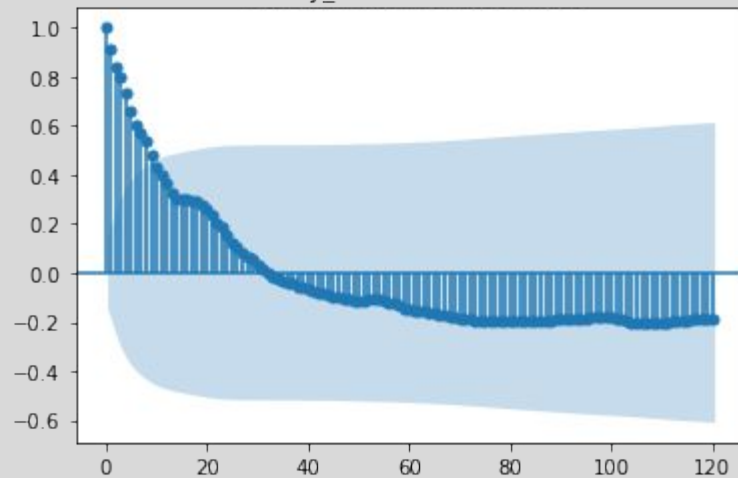
weekly_EOS Autocorrelation



weekly_Ethereum Autocorrelation

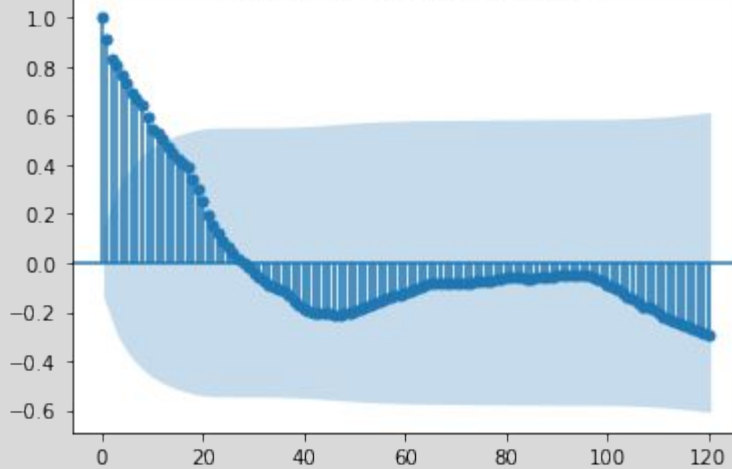


weekly_IOTA Autocorrelation

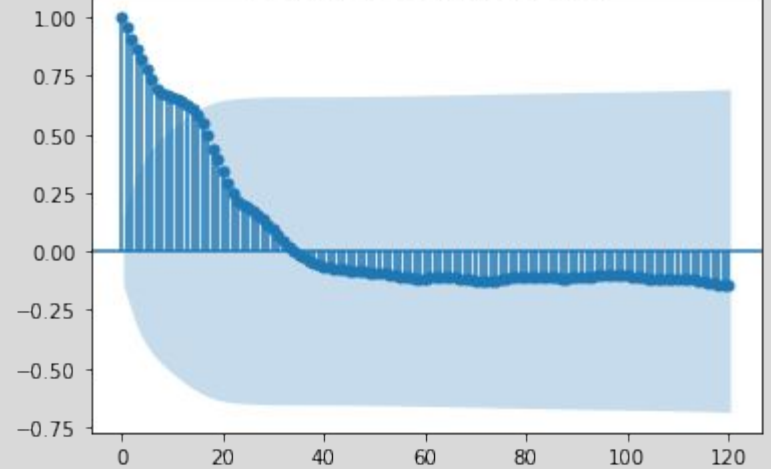




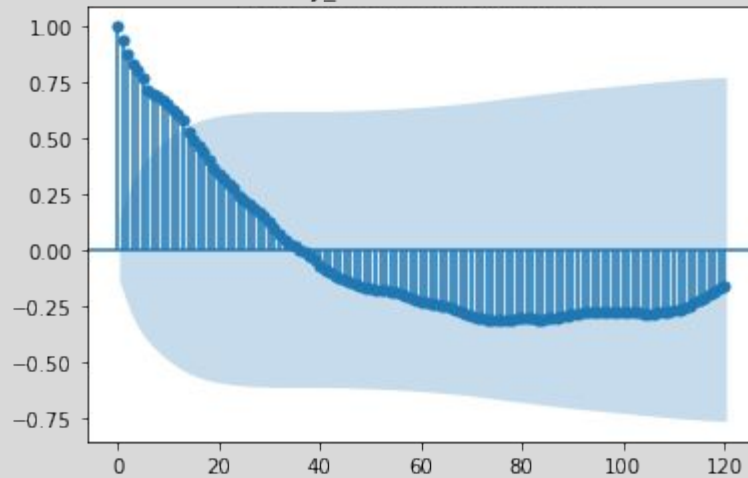
weekly_Litecoin Autocorrelation



weekly_Maker Autocorrelation

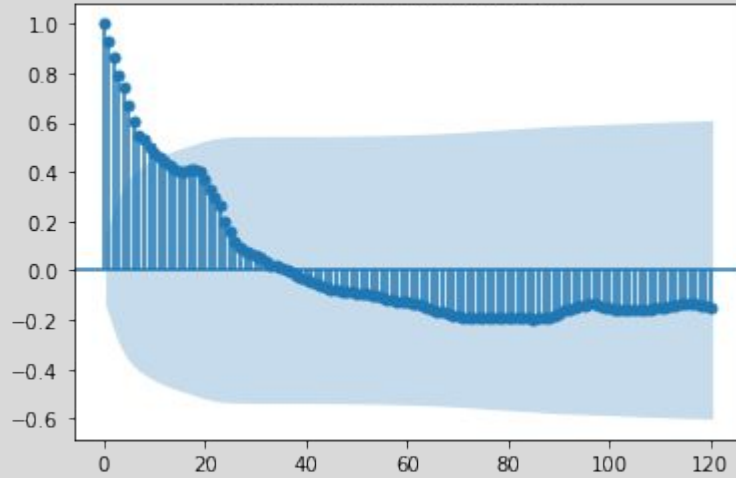


weekly_Stellar Autocorrelation

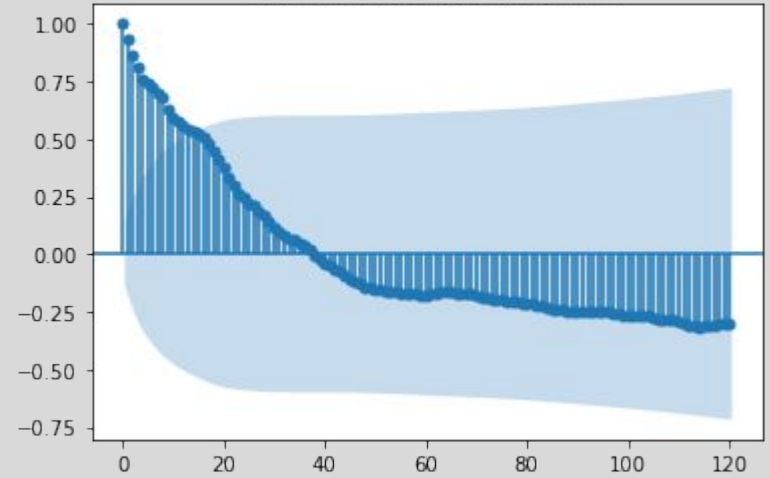




weekly_TRON Autocorrelation



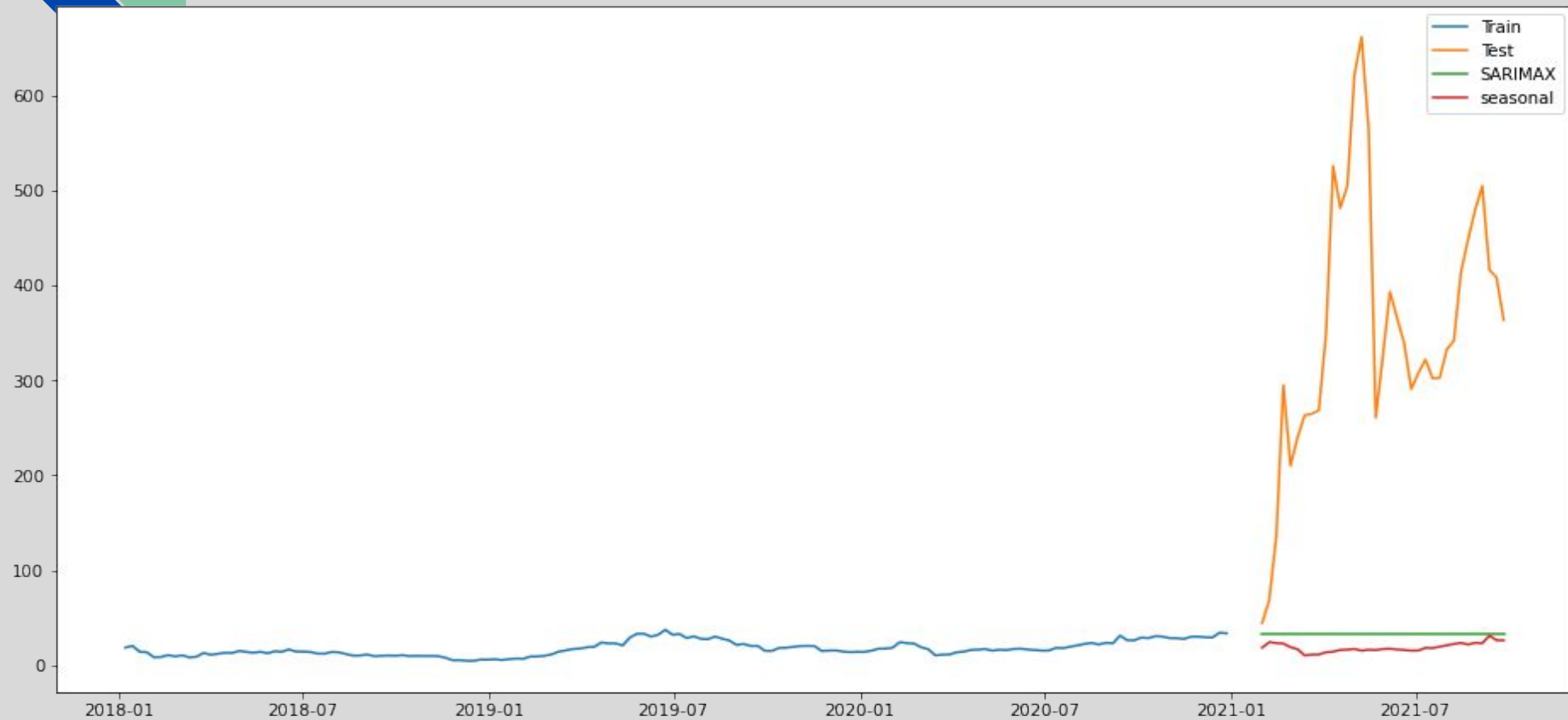
weekly_Monero Autocorrelation



Binance Coin Weekly Model

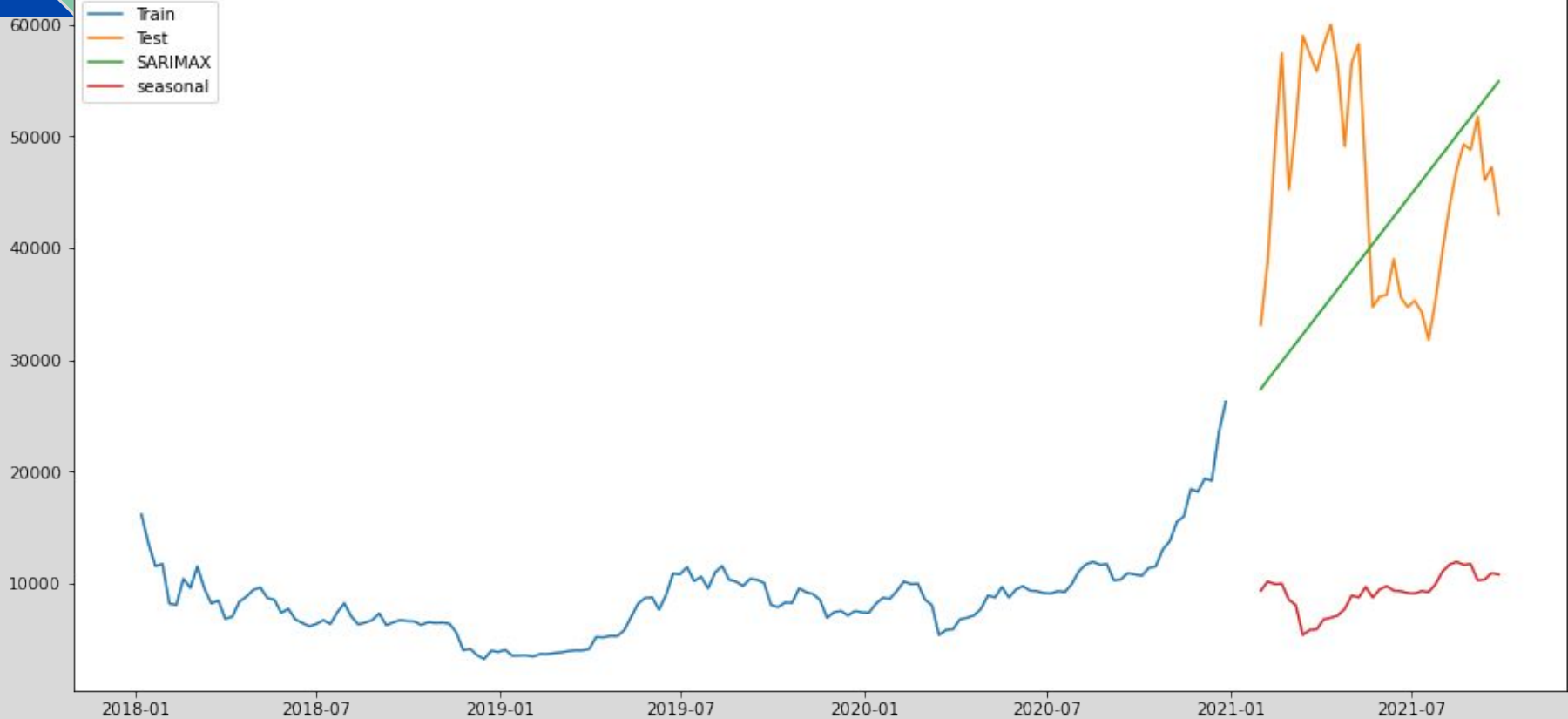
The baseline RMSE is 363

The model RMSE is 349.0



Bitcoin Weekly Model

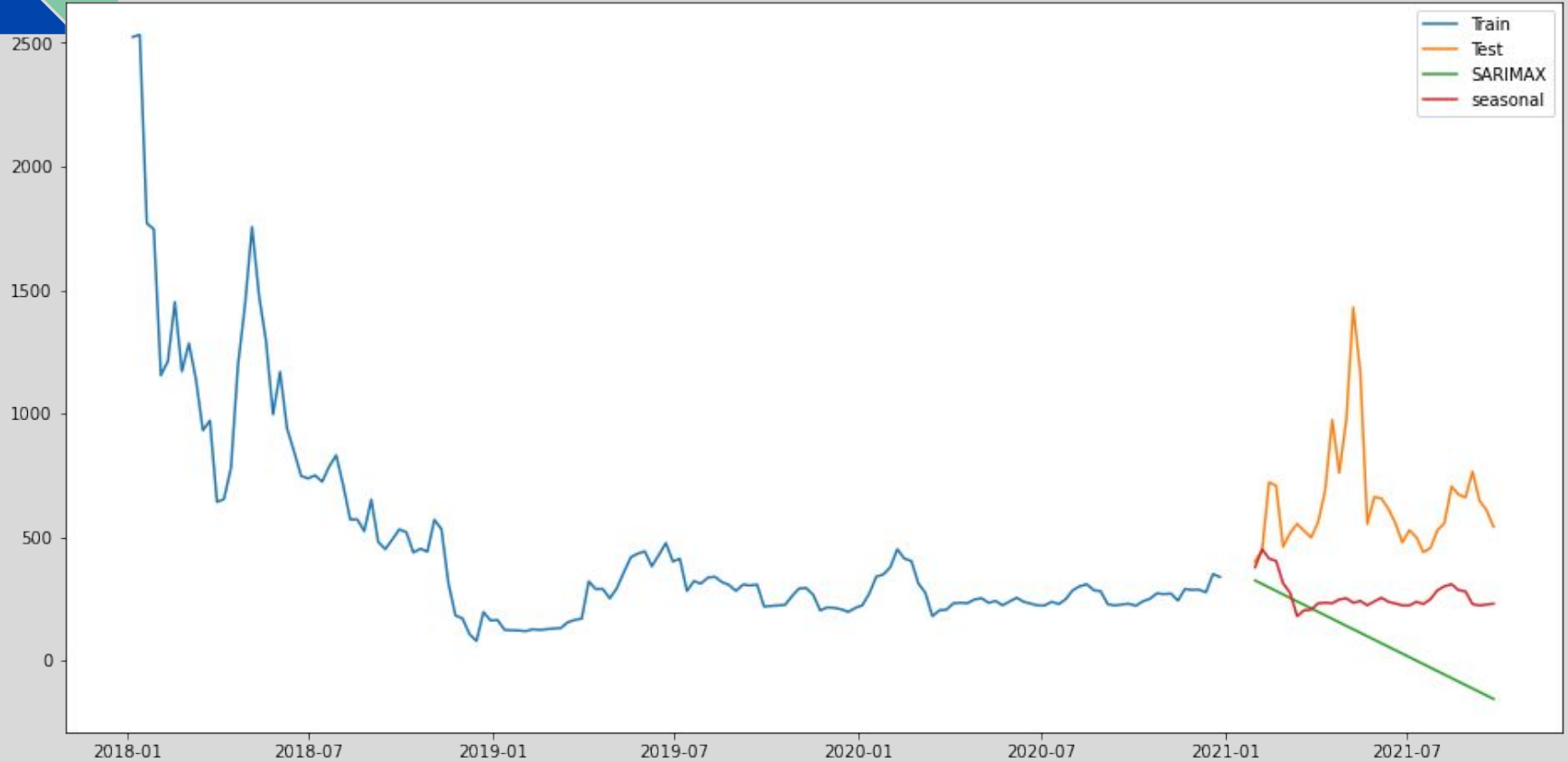
The baseline RMSE is
37794
The model RMSE is
14527.0



Bitcoin Cash Model

The baseline RMSE is 442

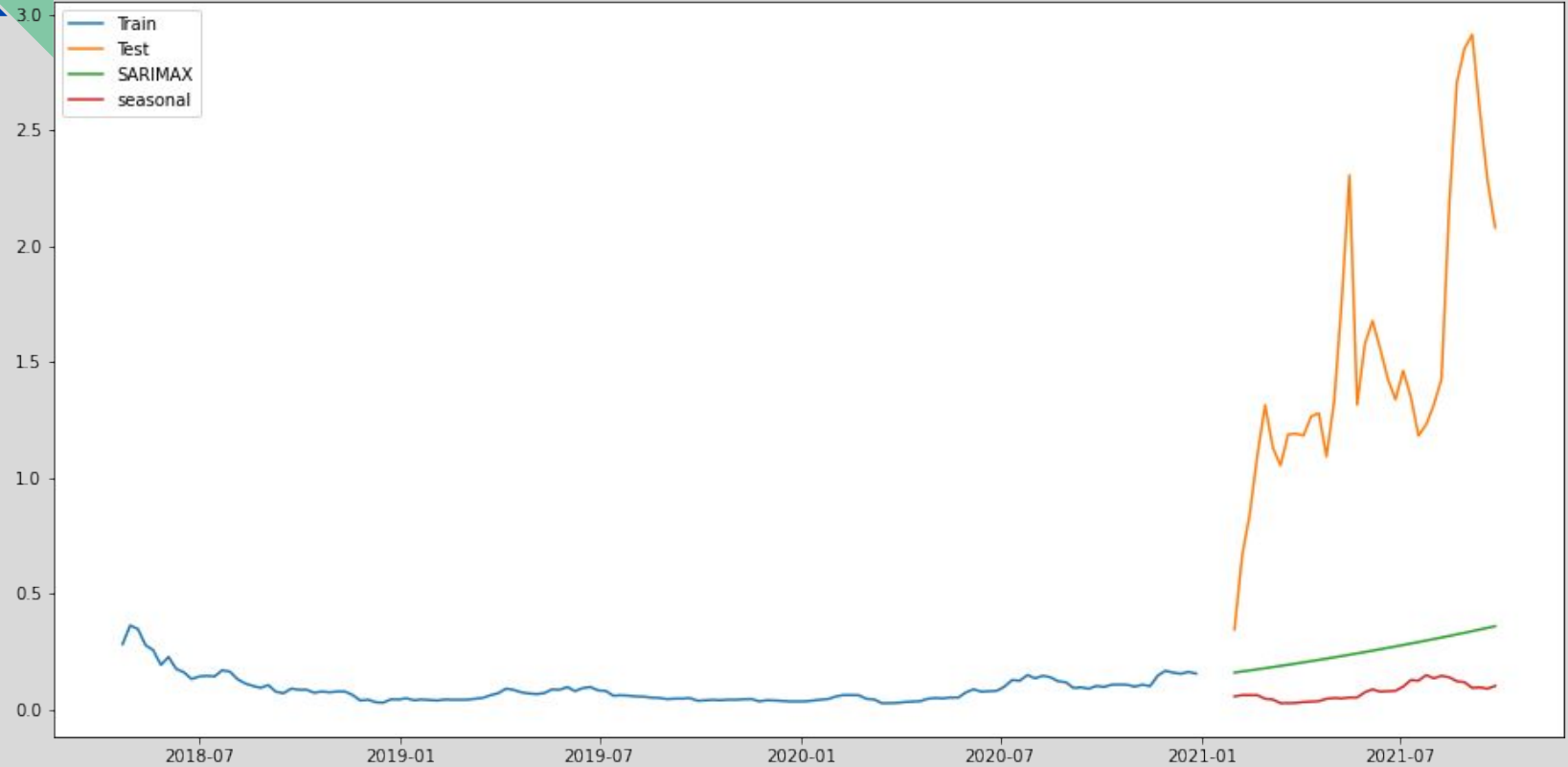
The model RMSE is 613.0



The baseline RMSE is 2

Cardano Weekly Model

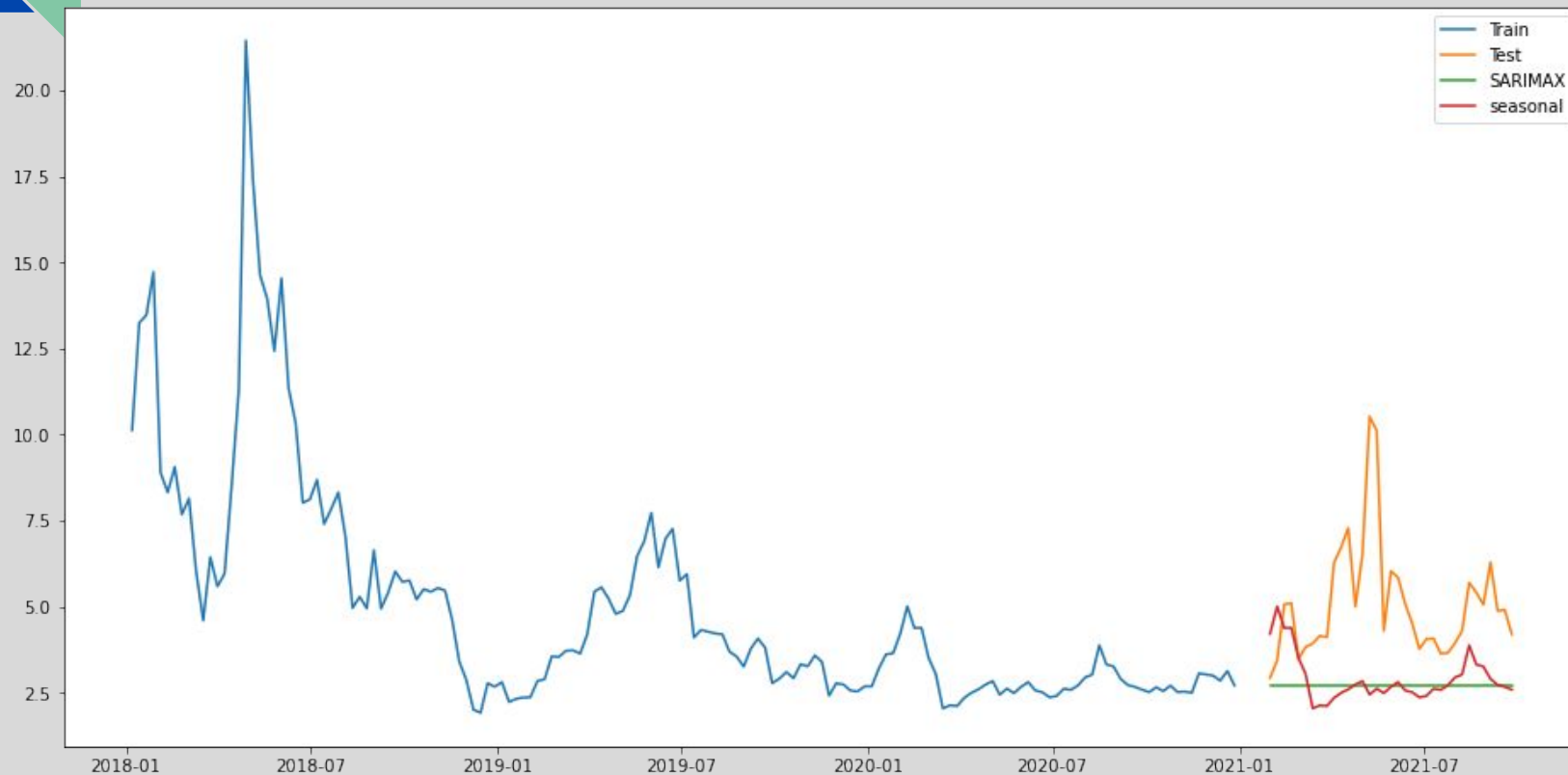
The model RMSE is 1.0



The baseline RMSE is 3

EOS Weekly Model

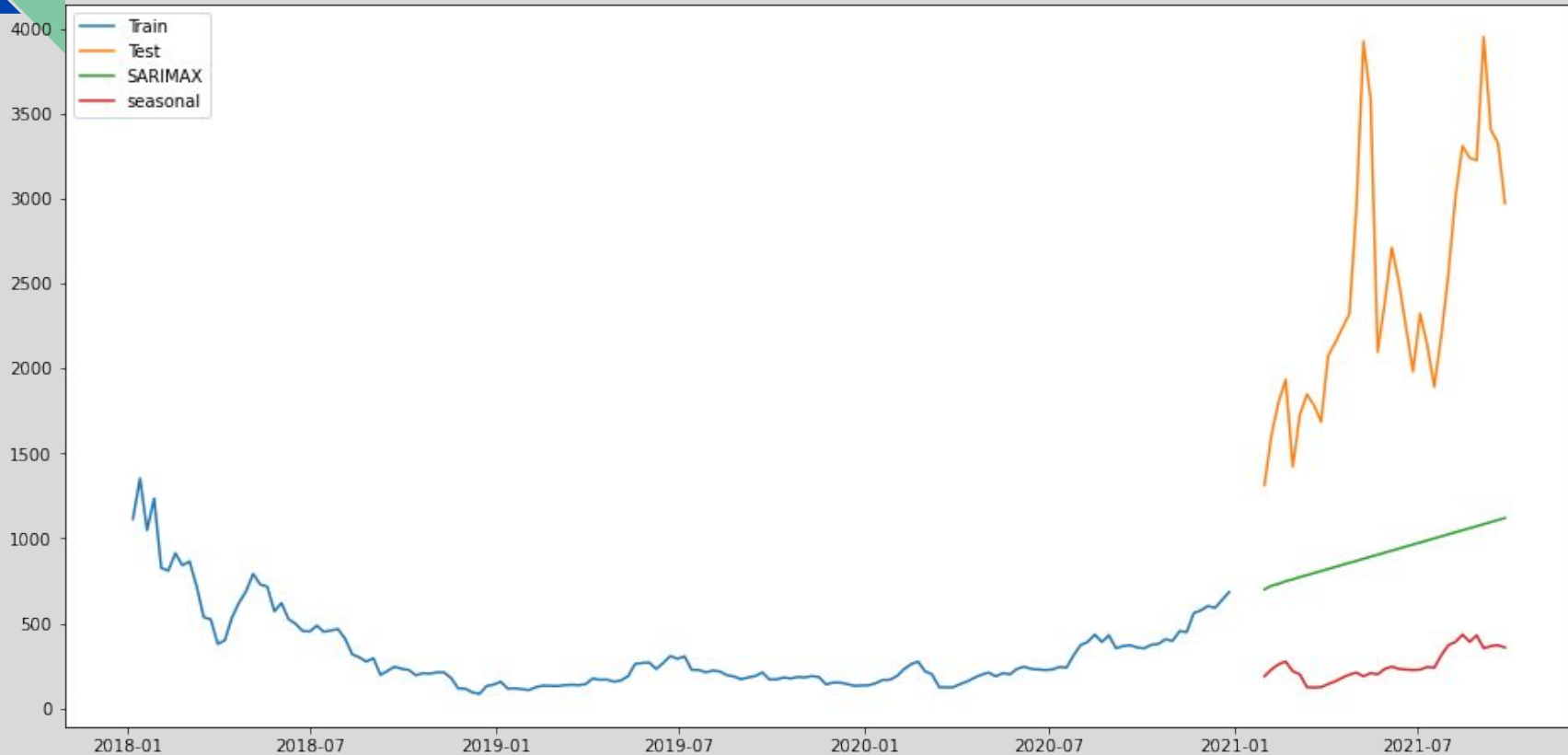
The model RMSE is 3.0



Ethereum Weekly Model

The baseline RMSE is 2294

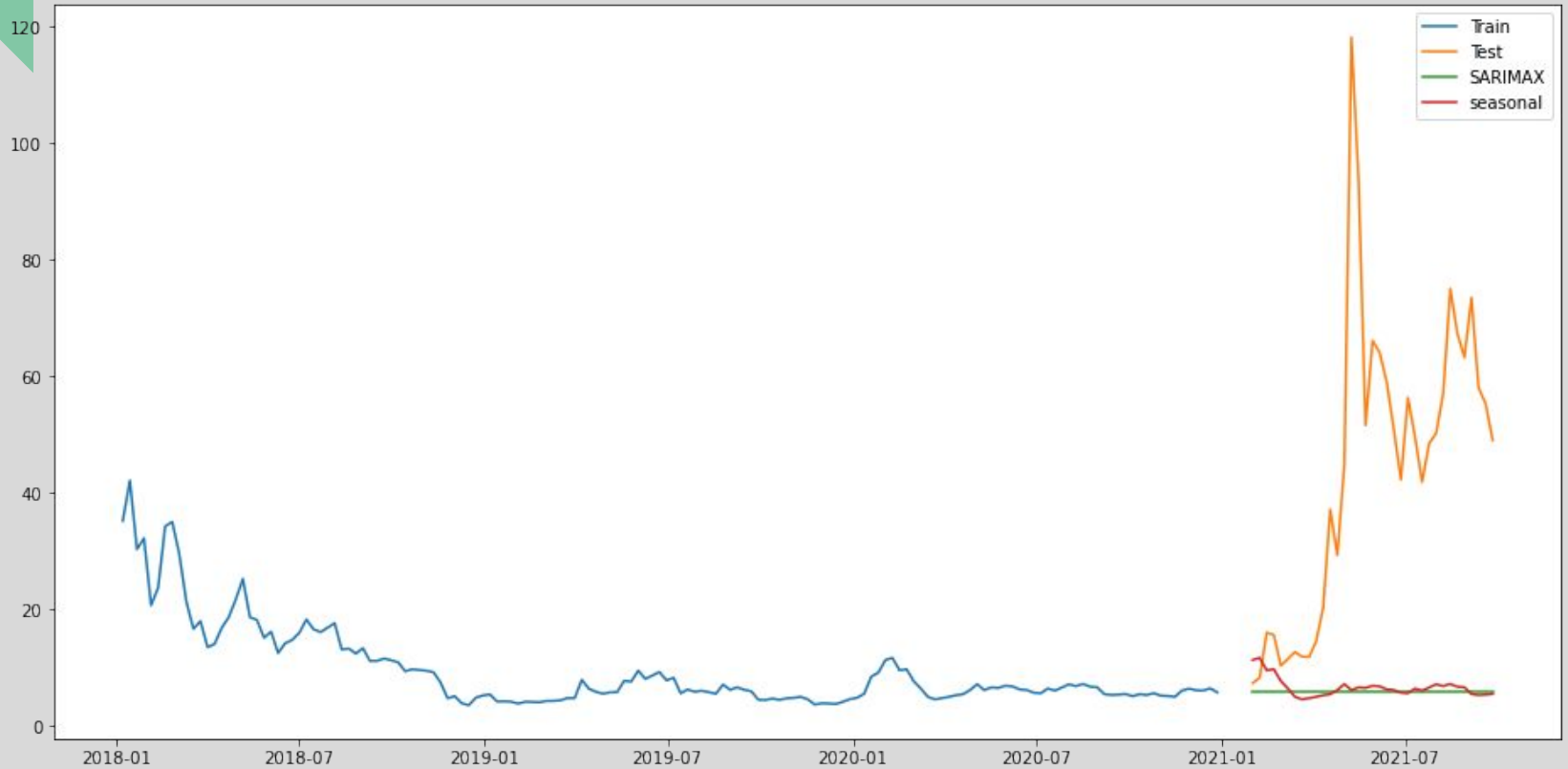
The model RMSE is 1658.0



The baseline RMSE is 46

Ethereum Classic Weekly Model

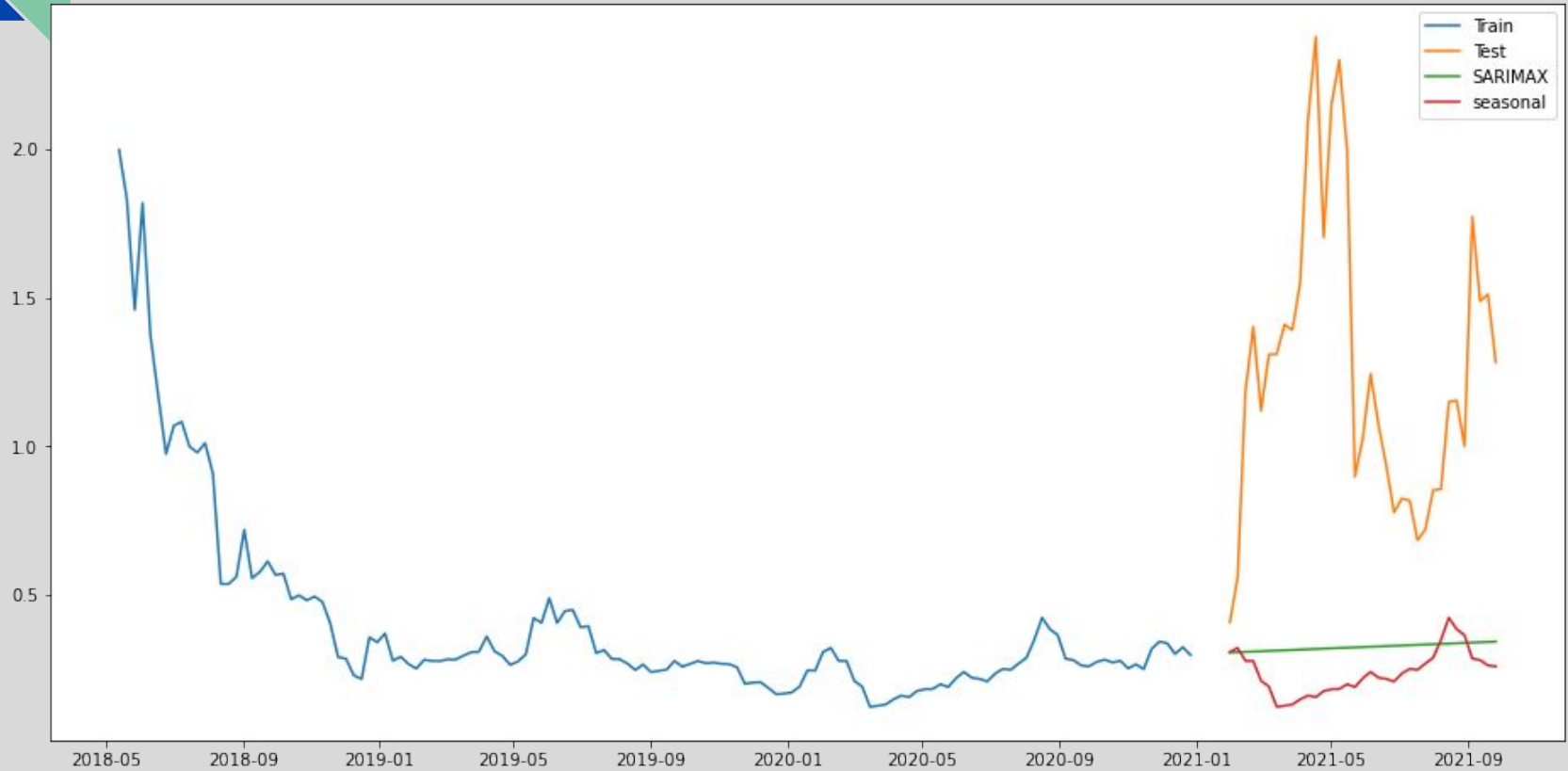
The model RMSE is 46.0



The baseline RMSE is 1

IOTA Weekly Model

The model RMSE is 1.0



The baseline RMSE is 151

Litecoin Weekly Model

The model RMSE is 70.0

400

350

300

250

200

150

100

50

Train
Test
SARIMAX
seasonal

2018-01

2018-07

2019-01

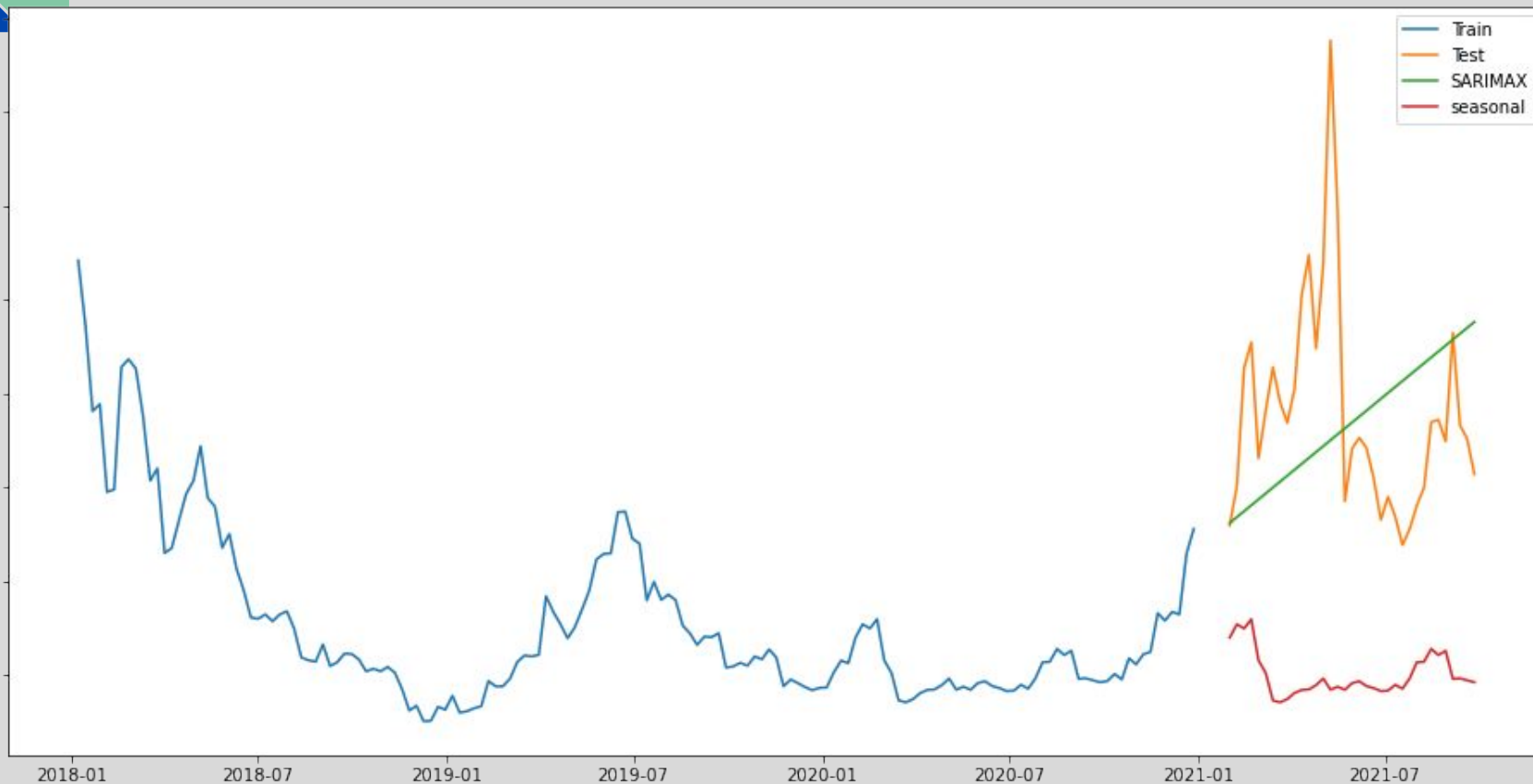
2019-07

2020-01

2020-07

2021-01

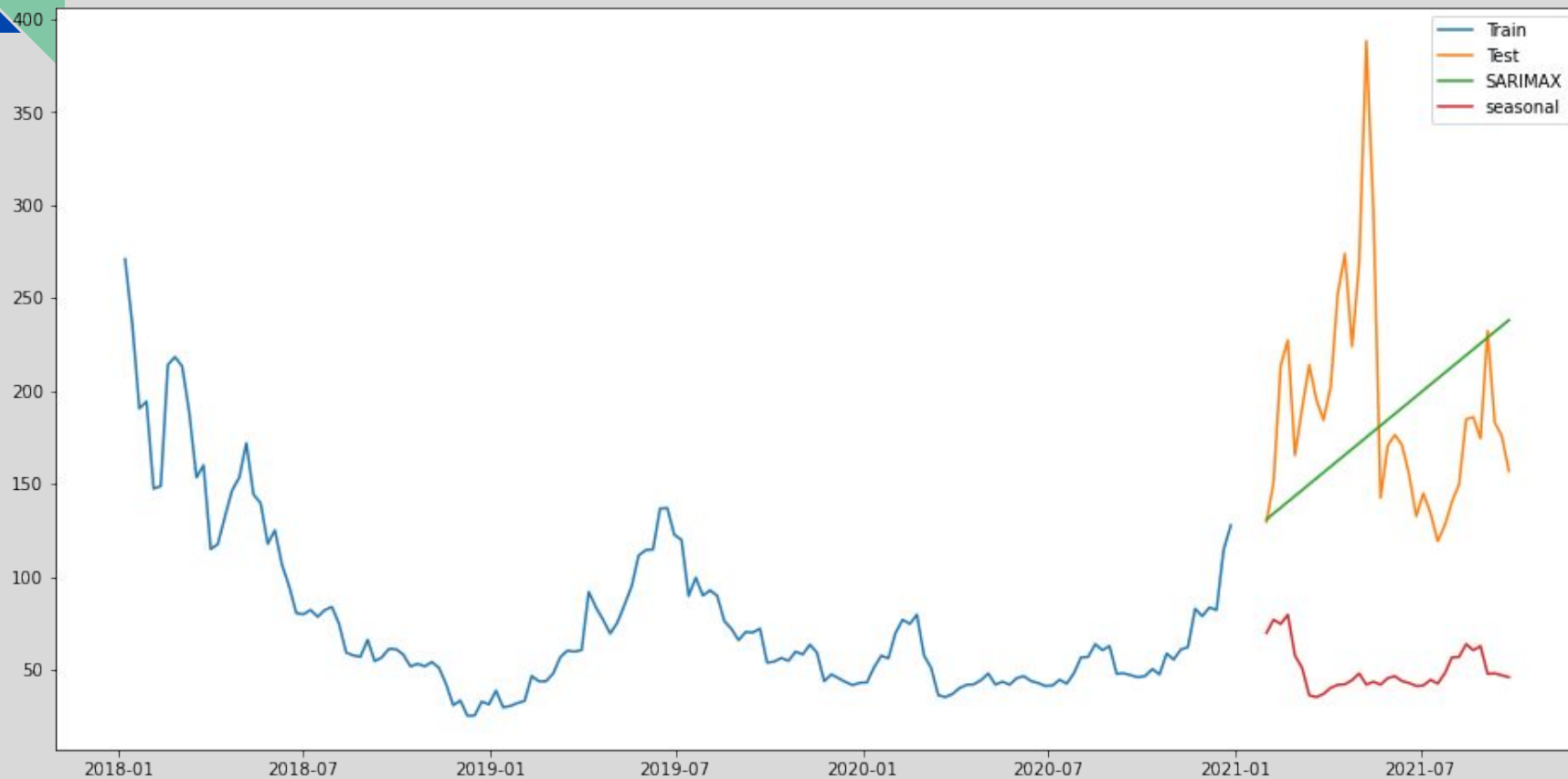
2021-07



The baseline RMSE is 2616

Maker Weekly Model

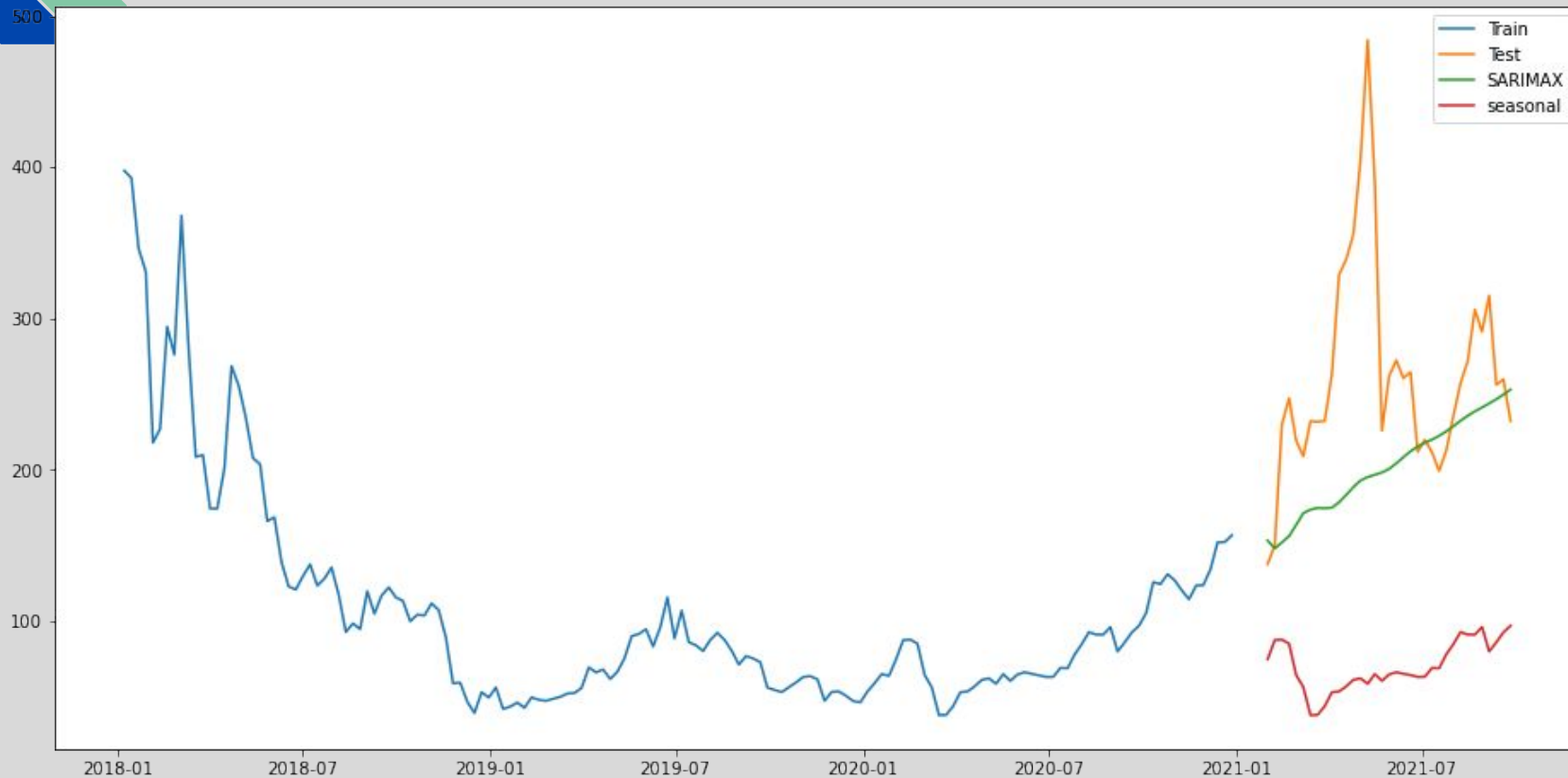
The model RMSE is 2525.0



The baseline RMSE is 206

Monero Weekly Model

The model RMSE is 94.0





Conclusion

Seasonality was not evident in closing price data

Some models performed better than others

More hyperparameter tuning could help

Trying other models may work as well

No easy route to crypto millions



Thank You

Questions?