



Developing a Control System to achieve a Profitable Strategy in the crypto market

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Outline



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Introduction and Problem Statement

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Enhancing System
Performance

02

Methodology and System Development

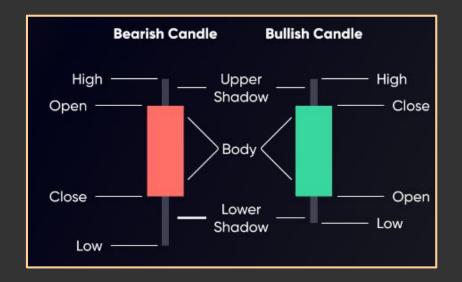
04

Results and Conclusion

Trading Concepts

Candles and Charts

- Color
- OHLC
- Timeframe







Cryptocurrency Market

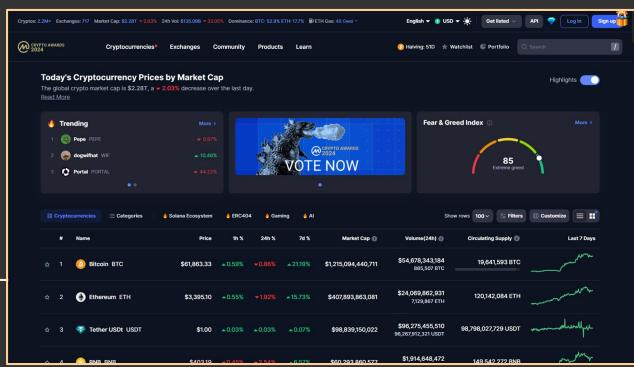


- Coin Market Cap
- Tradingview
- Binance

+340M

Monthly Visitors

Largest market aggregator and price tracking platform



Cryptocurrency Market



- Coin Market Cap
- Tradingview
- Binance

+50M

Traders and Investors

Charting Platform



Cryptocurrency Market

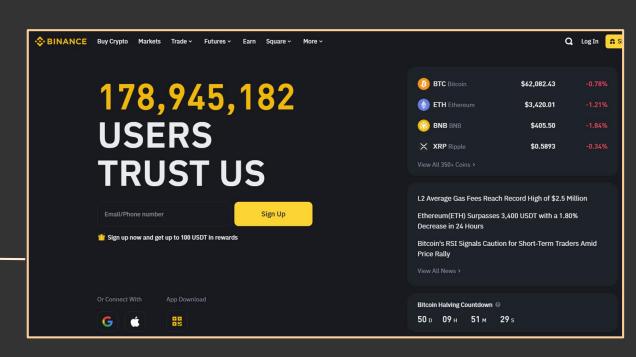


- Coin Market Cap
- Tradingview
- Binance

\$76 billion

Daily trading volume

World's largest crypto exchange



Literature Review









O1

Price Prediction

Predicting the exact price of a coin

02

Price Direction Prediction

Binary classification of candles

03

Control Based

Use of PI and LQR

Methodology



System Design

Defining system modules



Minimizing the cost function

Using control methods



Equation Determination

System identification methods

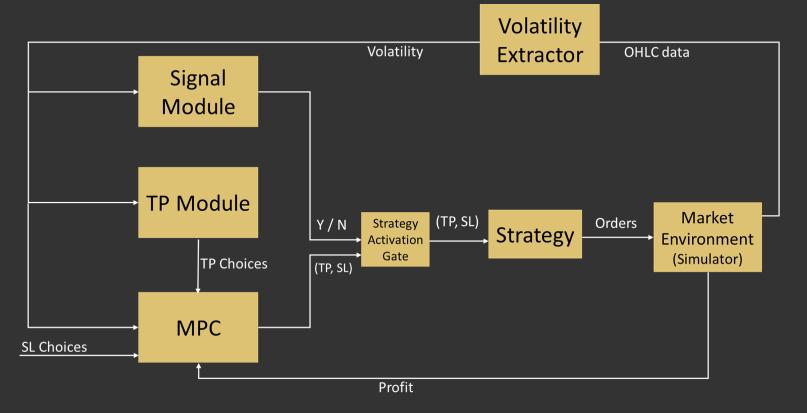


Validating results

Comparing with previous results







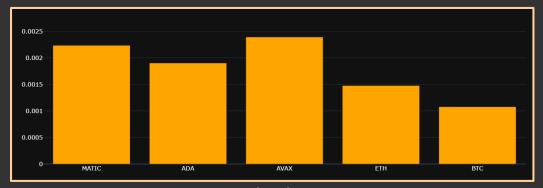


Choosing the Coin

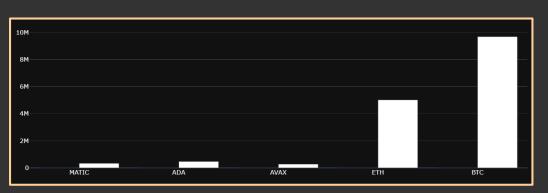
☐ Comparing coins

☐ MATIC is 15th coin

Based on the Market Cap



Volatility



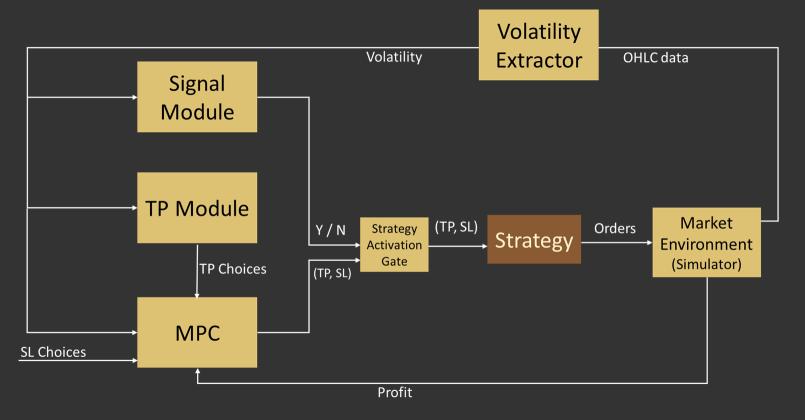


Features

Date	Take Profit	Stop Loss	TP category	Ratio	Difference	Volatility	Profit
2020-10-27 7:01:00	0.93%	1.50%	35	0.62041	2.4306	3	\$46.79
2020-10-28 7:01:00	0.91%	1.50%	35	0.60665	2.4100	4	\$128.46
2020-10-29 7:01:00	0.91%	1.50%	35	0.60959	2.4144	3	\$187.26
2020-10-30 7:01:00	0.94%	1.50%	35	0.62417	2.4363	3	\$226.93
2020-10-31 7:01:00	1.04%	1.50%	35	0.69275	2.5391	3	\$132.67
2023-07-28 7:01:00	0.20%	6.00%	75	0.03411	6.2047	2	-\$215.32
2023-07-29 7:01:00	0.20%	6.00%	75	0.03411	6.2047	1	-\$215.32
2023-07-30 7:01:00	0.20%	6.00%	75	0.03411	6.2047	1	-\$215.32

System Schematic







Strategy Module

Inputs:

- Take Profit (Dynamic)
- Stop Loss
- Time Limit
- Signal

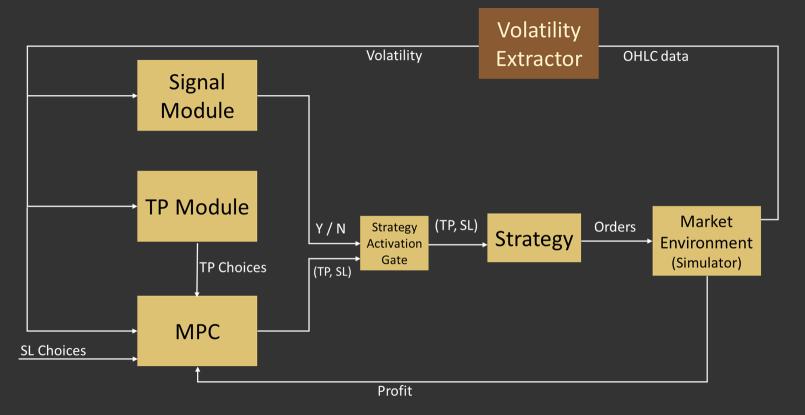
Output:

- Orders (Buy/Sell)
- Profit



System Schematic







Volatility Extractor

$$len = \frac{high-low}{open \le close \cdot low + open > close \cdot high}$$

$$m inc100 = rac{max_{100}(high_{-100})-close}{close}$$

To TP module

$$ext{len}_{w1} = rac{1}{w1} \sum_{i=-w1+1}^{0} ext{len}_i$$



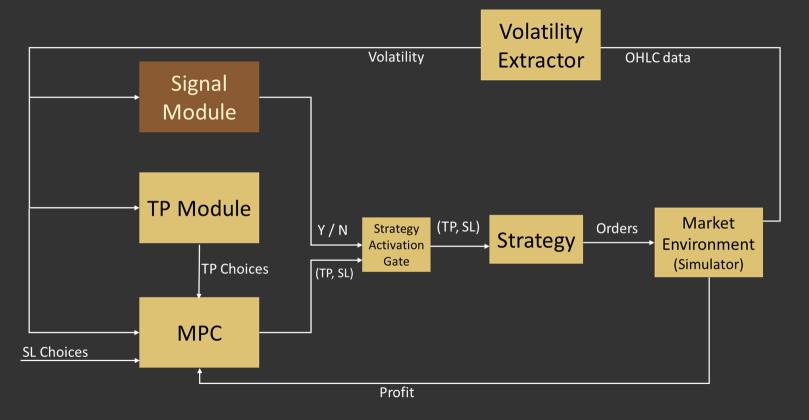
To Signal Module



To MPC Module

System Schematic



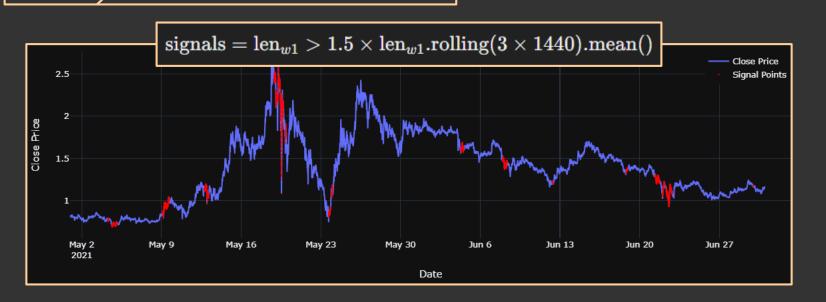




Signal Module

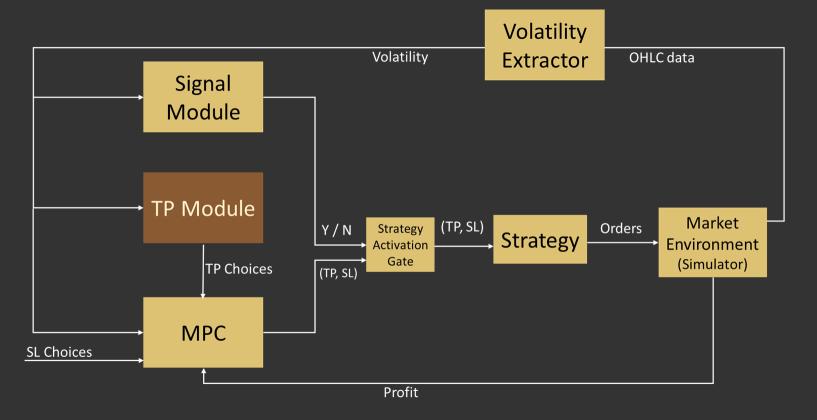
Inputs OHLC data → Volatility Configs





System Schematic







Take Profit (TP) Module

Inputs Backward Volatility List Configs

Output TP % of the volatility





Take Profit (TP) Equations

$$1 \qquad len = \frac{ \underset{\mathrm{open} \leq \mathrm{close} \cdot \mathrm{low} + \mathrm{open} > \mathrm{close} \cdot \mathrm{high}}{\mathrm{high}}$$

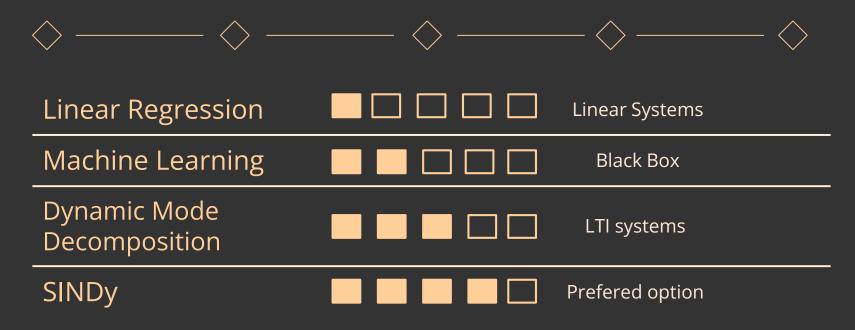
$$\log 2$$
 $\operatorname{len}_{10} = rac{1}{10} \sum_{i=-9}^{0} \operatorname{len}_{i}$

inc100 =
$$\frac{\max_{100}(\text{high}_{-100}) - \text{close}}{\text{close}}$$

$$\boxed{4} \qquad TP = \mathrm{inc} 100 (10,\mathrm{ds}) \cdot P\%$$



System Identification Methods





System Identification Methods

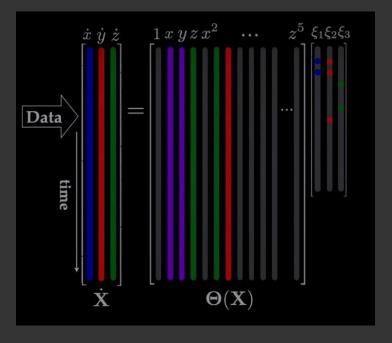
SINDy algorithm (Sparse Identification of Nonlinear Dynamics)

Lasso Regression: $Min(sum\ of\ squared\ residuals + \alpha * |slope|)$

$$\mathbf{X} = \begin{bmatrix} \mathbf{x}^{T}(t_1) \\ \mathbf{x}^{T}(t_2) \\ \vdots \\ \mathbf{x}^{T}(t_m) \end{bmatrix} = \begin{bmatrix} x_1(t_1) & x_2(t_1) & \cdots & x_n(t_1) \\ x_1(t_2) & x_2(t_2) & \cdots & x_n(t_2) \\ \vdots & \vdots & \ddots & \vdots \\ x_1(t_m) & x_2(t_m) & \cdots & x_n(t_m) \end{bmatrix} \downarrow \text{time}$$

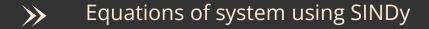
$$\Theta(\mathbf{X}) = \begin{bmatrix} \begin{vmatrix} & & & & \\ 1 & \mathbf{X} & \mathbf{X}^{P_2} & \mathbf{X}^{P_3} & \cdots & \sin(\mathbf{X}) & \cos(\mathbf{X}) & \cdots \end{bmatrix}$$

$$\dot{\mathbf{X}} = \Theta(\mathbf{X})\Xi.$$



System Equations







$$p_{1(k+1)} = 4.5p_{1(k)}.sl - 0.1p_{1(k)} + 165061.5\ sl^2 + 42.5sl.tp - 10938.6sl - 0.1tp + 98.8$$



$$p_{2(k+1)} = 0.1p_{2(k)}.sl - 71275.5sl^2 - 188.5sl.tp + 14002.3sl + 4.1tp - 27$$



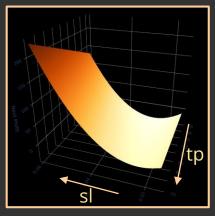
$$p_{3(k+1)} = 50777.6 \, sl^2 - 39.3 sl.tp + 136.3 sl + 0.7 tp - 23.$$

$$p_{4^{(k+1)}} = 21.2p_{4^{(k)}}.sl - 0.6p_{4^{(k)}} - 2493396.46\,sl^2 \\ - 1798.8sl.tp + 245808.8sl - 0.1tp^2 + 53.7tp - 4340.8sl - 0.1tp^2 + 53.7tp - 0.1tp^2 + 0.$$

System Equations

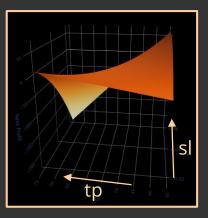
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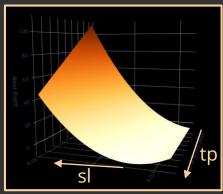
Relationship between TP, SL, Next Profit



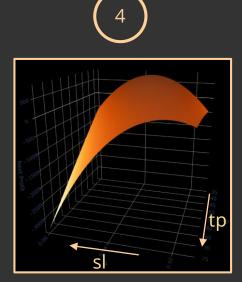








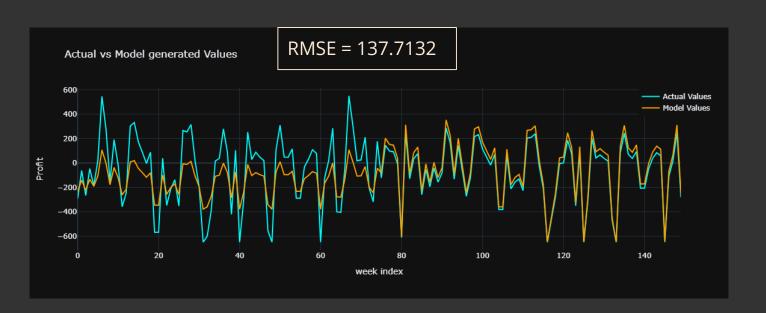




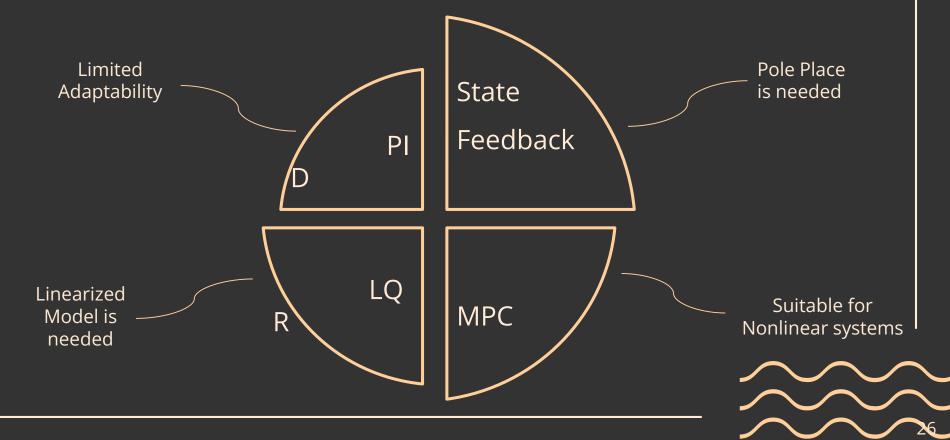
System Identification



SINDy (Sparse Identification of Nonlinear Dynamics)

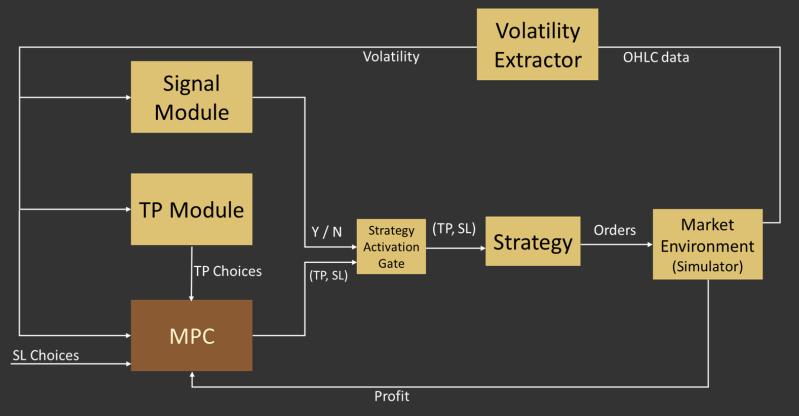


Control & Optimization Methods



System Schematic

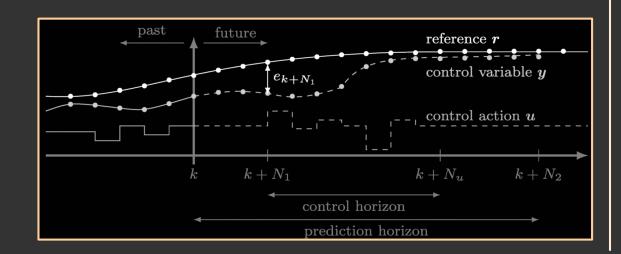




Optimization Method

MPC (Model Predictive Control)

- Handling Nonlinear Systems
- Handles Constraints
- Online Optimization
- High Flexibility

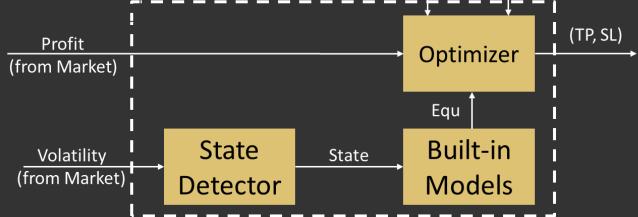


Objective Function = Profit



>> MPC

Objective Function = Profit



SL choices

TP choices

```
volatility = \frac{Rank \ of \ the \ rolling \ window \ of \ 'len\_w1'}{rolling \ window \ size}
```

 $v_{label} = Categorize (volatility, labels = [1, 2, 3, 4])$

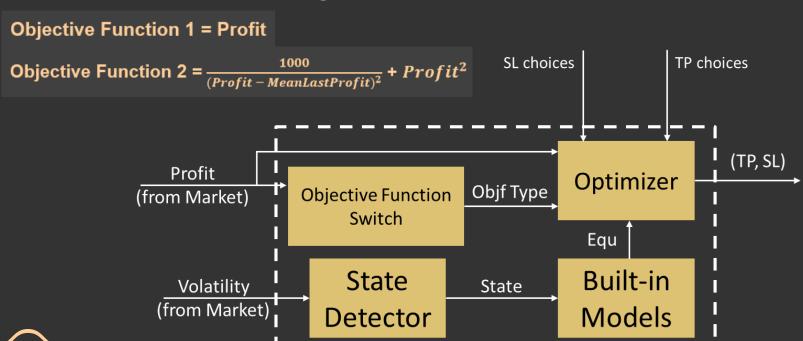


Results

>>> Improvement by using MPC (compared to random)



MPC with Risk Management





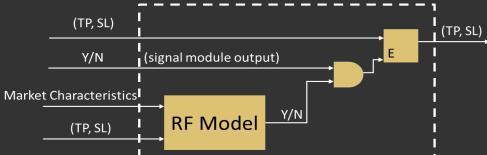


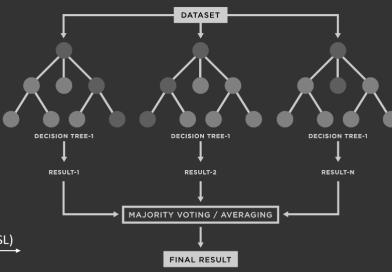
>>> Improvement by using Risk Manager



>> Random Forest

Strategy Activation Gate







Results

Reducing great loss by using Random Forest



Future Work

Compounding (reinvestment)

Reinvesting the profits earned rather than withdrawing them

> Adding Other Coins

Handling multiple coins in portfolio

Increased Position Size

Exponential growth

Withdrawal Challenges

_ Risk Management

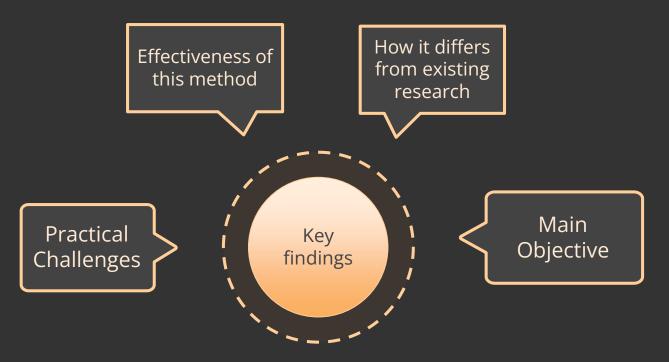
Market Correlations

Complex Implementation



Discussion







References

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- Steven L. Brunton, Joshua L. Proctor, and J. Nathan Kutz; "Discovering governing equations from data by sparse identification of nonlinear dynamical systems", 10.1073/pnas.1517384113, Princeton University, Aug. 2015.
- Steve Brunton; Sparse Identification of Nonlinear Dynamics: Sparse ML Models; https://www.youtube.com/watch?v=NxAn0oglMVw.



Thank You for Your Time