Capstone Project: Predicting the Stock Market

___Problem Statement__

Is it more profitable for a trading/investing strategy to predict the current market regime (bear or bull market) or predict the forward price returns in advance?

Agenda

- 1. The extreme unpredictability of stock markets
- 2. A short intro to Hidden Markov Model
- 3. Part 1: HMM for Daily Market Regime Prediction
- 4. Part 2: HMM for Daily Price Return Prediction
- 5. Conclusion & Recommendations
- 6. Limitations & Caveats

Stock markets are extremely hard to predict due to "Reflexivity"

"In situations that have thinking participants, there is a two-way interaction between the participants' thinking and the situation in which they participate."

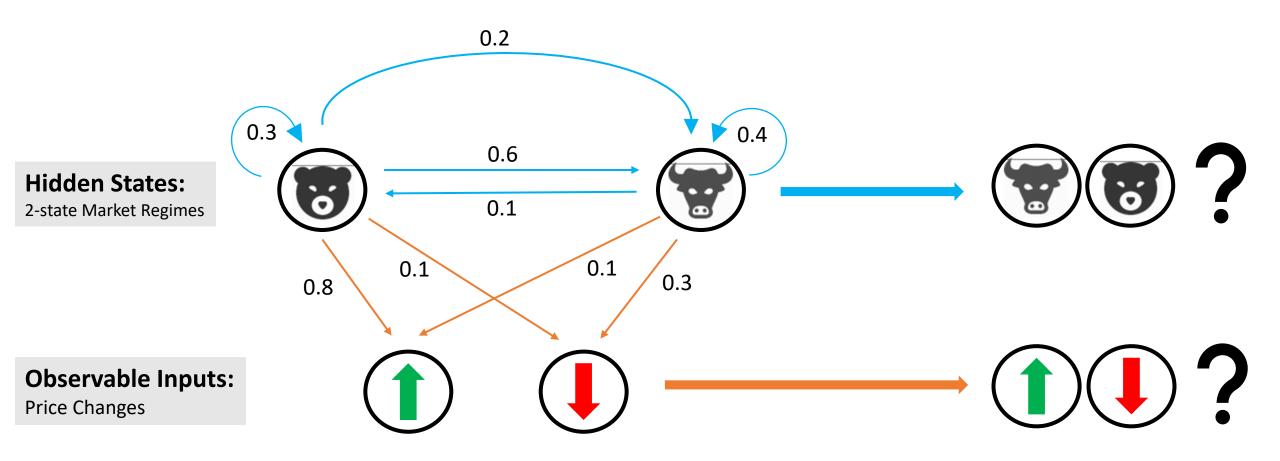
~ George Soros, billionaire hedge fund manager who nearly bankrupted the Bank of England







Hidden Markov Model



Part 1

Hidden Markov for Market Regime Prediction

Research Methodology

Asset	SPY ETF (underlying index: S&P500 Index)
Model Training Period	Since inception 1993-2009
Model Testing Period	2010-2023/1
Backtest Period	2018-2023/1
Model Evaluation Metric	Cumulative %Returns during Backtest Period

Backtesting Strategy:

- Daily trading strategy
- Long SPY when <u>last</u> regime=BULL; Short SPY when <u>last</u> regime=BEAR

Converting Price Observations into Technical Indicators

Hidden States: Market Regimes Observable Inputs: Mass T1M T3M Donchian RSI^* ADX[^] Index Channel Returns Returns **Technical Indicators** ^ Average Directional Index * Relative Strength Index **Observable Output:** CLOSE **Prices**



Strategy Backtest during 2018-2023/1



Part 2

Hidden Markov for Daily Price Returns Prediction

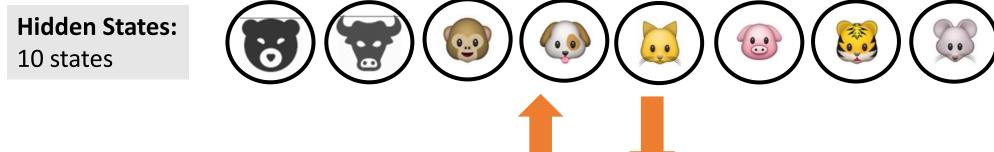
Research Methodology

Asset	SPY ETF (underlying index: S&P500 Index)
Model Training Period	Since inception 1993-2009
Sampling Period for "Possible_Outcomes"	2010-2017
Model Testing Period	2018-2023/1
Backtest Period	2018-2023/1
Model Evaluation Metric	Cumulative %Returns during Backtest Period

Backtesting Strategy:

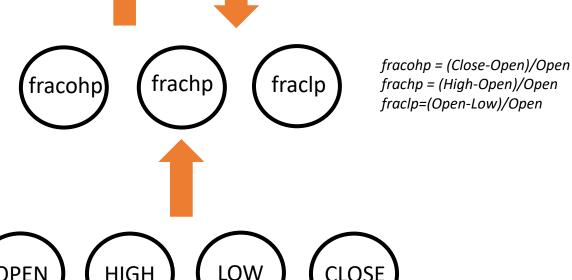
- Day trading strategy (to trade on the OPEN)
- If Predicted Close > Actual Open: Buy SPY at OPEN price and Sell at CLOSE
- If Predicted Close < Actual Open: Sell SPY at OPEN price and Buy at CLOSE

Using Augmented Price Features



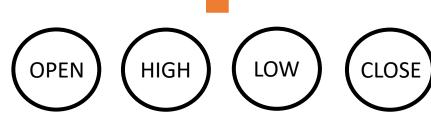
Observable Inputs:

Augmented Price Features

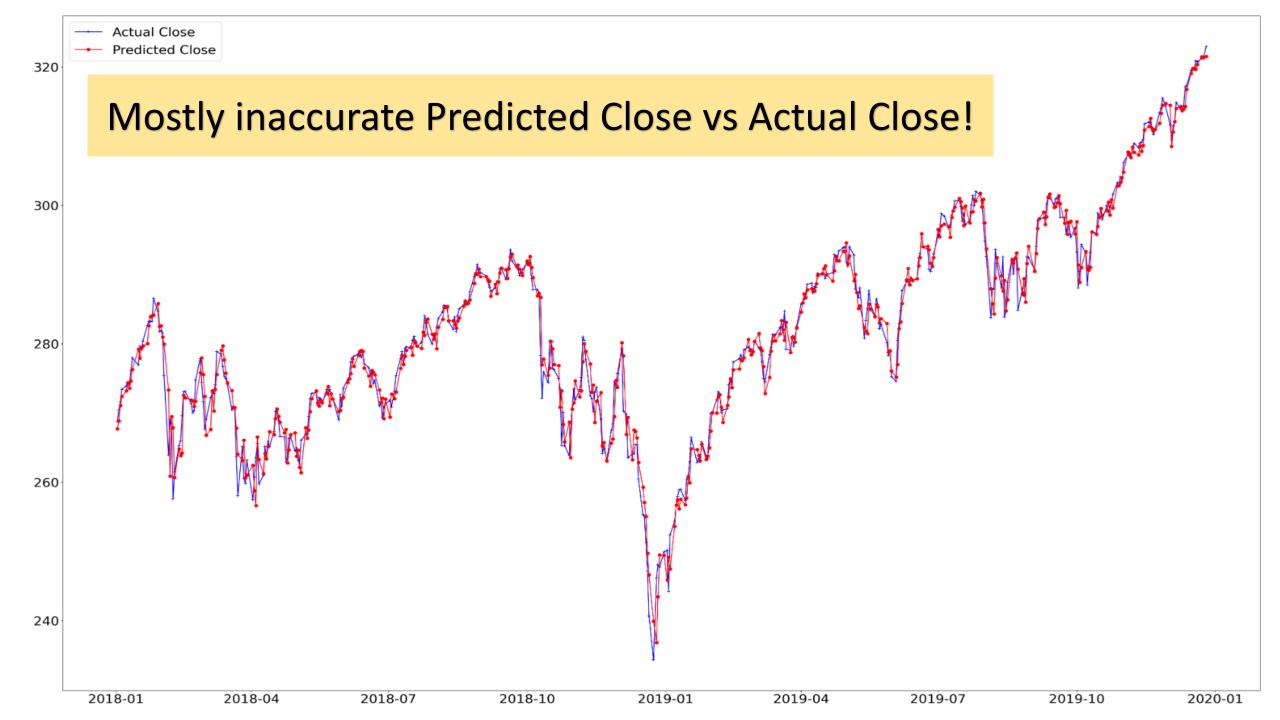


Observable Output:

Prices

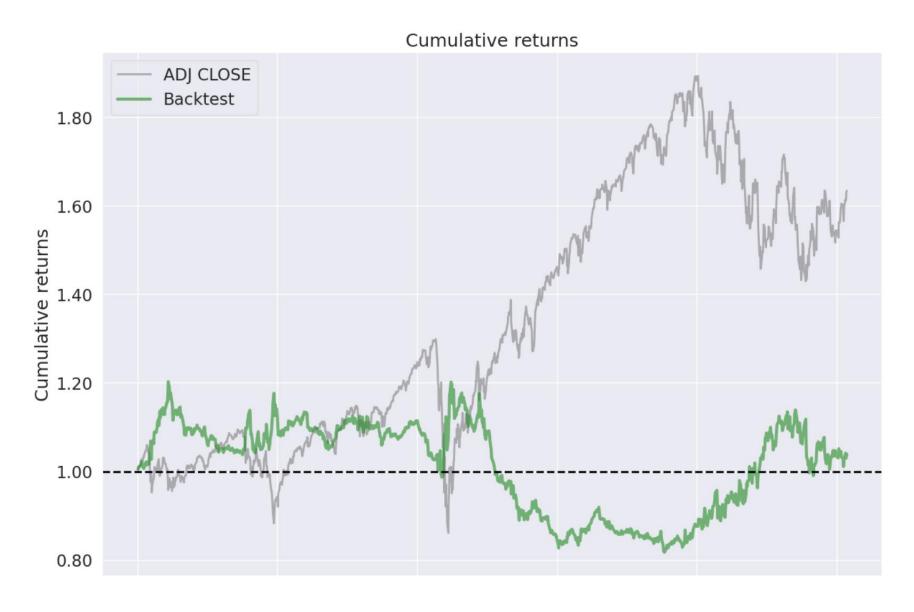






Strategy Backtest during 2018-2023/1



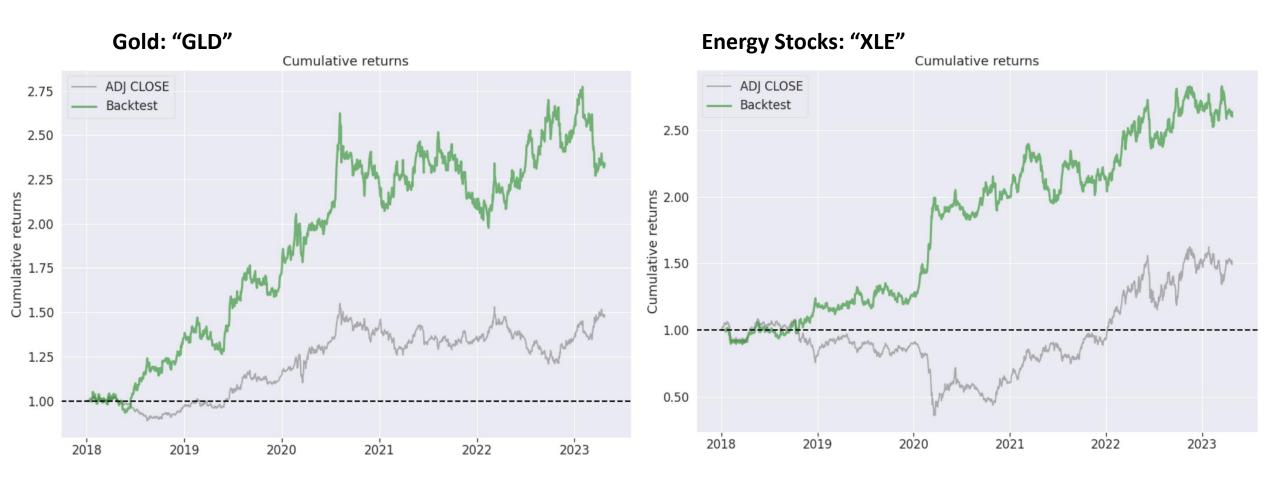


Conclusions & Recommendations

• Hidden Markov models proved to be reasonably capable of predicting current market regimes but underperformed considerably when trying to predict forward price returns.

• For the best alpha strategy, it is recommended to employ HMM models to **predict current market regime** just like Mr Partridge and not trade unnecessarily.

But market regime prediction strategy worked beautifully for some assets!



Limitations & Caveats

- Trading Costs such as short-financing costs, trading commissions, slippage and market impact costs have not been simulated in the backtesting.
- 2. Use of non-timeseries feature inputs may not be sufficient to capture underlying statistical patterns
- 3. More relevant features could be added to enhance model's accuracy

Model Uncertainty

- While HMM-clustering of SPY ETF market regimes enabled the model to outperform its benchmark up till Aug20, it failed to predict the market regime correctly following the global fiscal and monetary easing that fueled the unprecedented stock market recovery from Aug20-Dec21. From 2022, the model started to produce accurate regime classifications again.
- While the HMM regime-clustering model did not work as well for SPY ETF, it worked phenomenally well for more volatile assets such as oil and gold when a walk-forward training window was used to fit the HMM model.
- The choice of training window (static window, expanding window or walk-forward window) can result in vastly different backtesting profitability depending on the asset this risks explicit overfitting.
- The inherent mathematical randomness of the HMM model can also yield very different results resulting in overfitting. To reduce this variance, extra care must be exercised during the cross-validation process.
- In all assets backtested with similar models (not all shown here in this slides), there are varying lengths of time periods when the model underperformed and posed a drag on profitability. This underscores the fact that HMM regime-clustering model, like most quantitative strategies, is no Holy Grail to consistent profitability. Rigorous empirical analysis needs to be conducted to reduce overfitting variance and ensure the statistical significance of regime signals.
- As with all trading strategies, past performance is NOT an indicator of future performance (backtesting over longer periods of time easily proves this)
- 5. Due to the inconsistencies of quantitative strategies, one needs to be especially clear about the characteristics of an asset that allowed the strategy to work and re-train the model if such characteristics changes in the future.

What's Next

- "Best model" forecasts as of 26 April 2023:
- BULL market for SPY
- BULL market for GLD
- BULL market for XLE

Bonus Slides

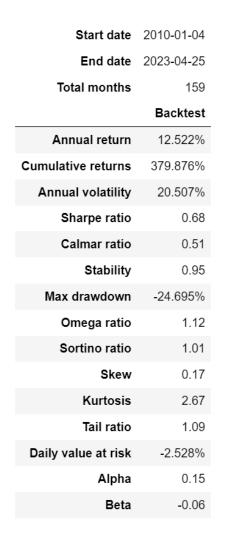
(not part of my presentation)

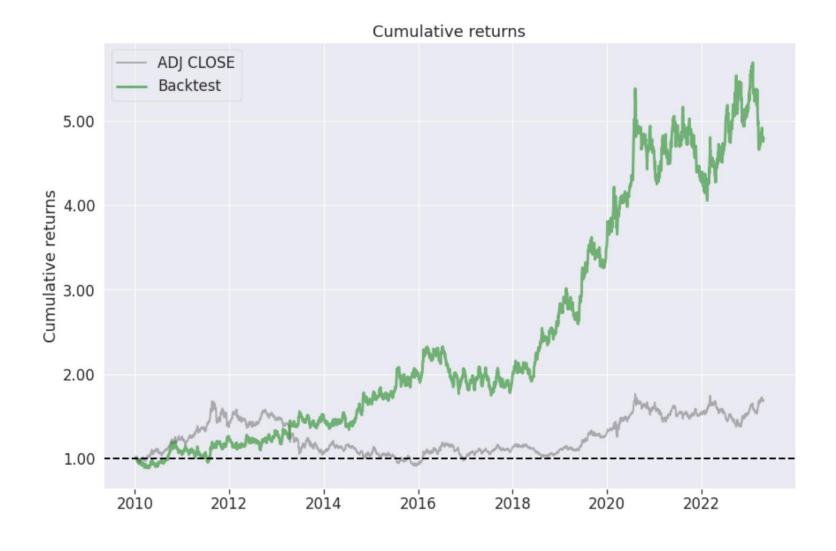
Maximum backtest periods possible with the data I can get.

Trained under a similar walk-forward fitting methodology and same trading strategy with 20% volatility-targeting.

For reference purposes only.

Even longer backtest for GLD





SLV





XLE (did not work until 2014-onwards)

Start date	2003-01-02
End date	2023-04-25
Total months	243
	Backtest
Annual return	11.748%
Cumulative returns	852.35%
Annual volatility	20.341%
Sharpe ratio	0.65
Calmar ratio	0.24
Stability	0.83
Max drawdown	- 48.743%
Omega ratio	1.11
Sortino ratio	0.96
Skew	0.18
Kurtosis	1.36
Tail ratio	1.09
Daily value at risk	-2.511%
Alpha	0.17
Beta	-0.16



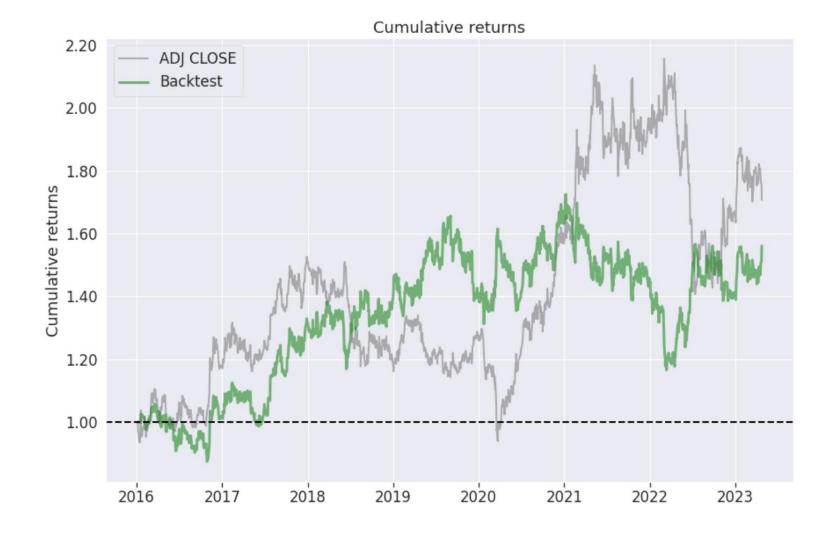
CPER

Start date 2016-01-04

End date 2023-04-25

Total months 87

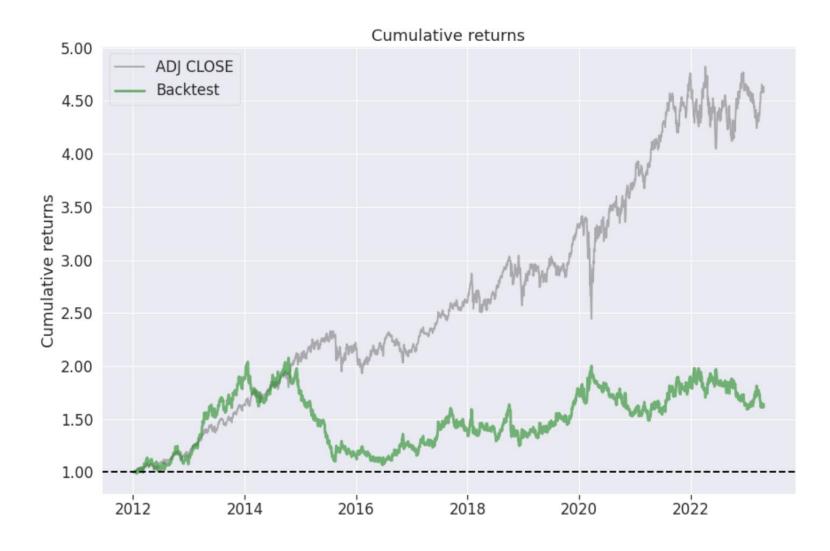
	Backtest
Annual return	6.29%
Cumulative returns	56.112%
Annual volatility	20.144%
Sharpe ratio	0.40
Calmar ratio	0.19
Stability	0.59
Max drawdown	-32.408%
Omega ratio	1.07
Sortino ratio	0.58
Skew	0.05
Kurtosis	1.32
Tail ratio	1.01
Daily value at risk	-2.506%
Alpha	0.11
Beta	-0.19



Some assets which didn't work

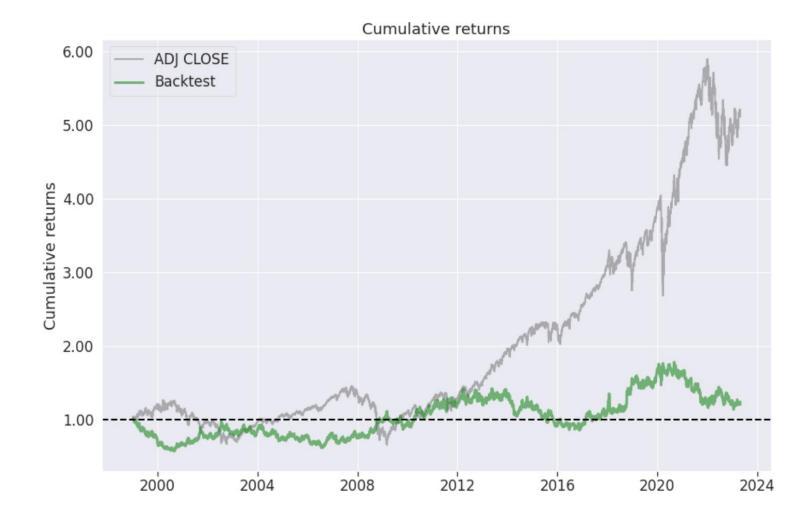
XLV

Start date	2012-01-03
End date	2023-04-25
Total months	135
	Backtest
Annual return	4.327%
Cumulative returns	61.343%
Annual volatility	20.526%
Sharpe ratio	0.31
Calmar ratio	0.09
Stability	0.25
Max drawdown	-48.64%
Omega ratio	1.05
Sortino ratio	0.45
Skew	0.02
Kurtosis	1.37
Tail ratio	1.02
Daily value at risk	-2.561%
Alpha	0.11
Beta	-0.28



SPY

Start date	1999-01-04
End date	2023-04-25
Total months	291
	Backtest
Annual return	0.759%
Cumulative returns	20.132%
Annual volatility	20.488%
Sharpe ratio	0.14
Calmar ratio	0.02
Stability	0.67
Max drawdown	-43.601%
Omega ratio	1.02
Sortino ratio	0.20
Skew	-0.07
Kurtosis	3.01
Tail ratio	1.01
Daily value at risk	-2.57%
Alpha	0.07
Beta	-0.41



VEU



