

Quantitative Strategy Proposal

New England Investment Consulting Group

designed for our quantitative strategies pipeline Phase 1: Idea
Generation and Defense.

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01

Strategy Overview: Cointegrated Pairs Trading

Trading

- ❖ Combine pre-selection tests, cointegration tests, technical indicator analysis, and portfolio optimization
- ❖ Focus on mid-large cap equities in the US (S&P 500)
- ❖ Identify pairs with mean reverting behavior
- ❖ Rely on z-score, moving averages as technical indicators

- ❖ Pairs follow mean reverting behavior
 - Long-term linear relationship

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Strategy Description

Trading

[1] Determining an investment universe

- ❖ Focus on longevity and stable long-term growth
- ❖ S&P 500
- ❖ Fundamental screening
 - Mid/Large Cap Companies
 - Stable YoY EPS Growth

Trading [2] Pre-Select Pairs Candidates

- ❖ [Brunetti & DeLuca \(2023\)](#): 7 pre-selection measures
 - Sum of Squared Deviations
 - Price Ratio
 - Correlation of Log-Price Time Series
 - etc.

Trading

[3] Cointegration Tests

- ❖ Run Engel-Granger/Johansen Tests

- [Brunetti & DeLuca \(2023\)](#)

```
from statsmodels.tsa.stattools import coint
from statsmodels.tsa.vector_ar.vecm import coint_johansen
```

Trading

[4] High-Variance Screening

- ❖ [Krauss \(2015\)](#): Pairs should exhibit
 - (1) high spread variance
 - (2) high mean reverting tendency
- ❖ Rank pairs based on these criteria

Trading

[5] Generate Buy/Sell Signals

- ❖ Compute ratio of prices of top 15 pairs from previous step
 - $P\{\text{Security A}\}/P\{\text{Security B}\}$
- ❖ Z-score of current ratio value w.r.t. previous 50 day window (medium.com)
 - When z-score is high (>1), Security A is relatively expensive (short) and Security B is relatively cheap (buy) → exit positions when reverted to mean (i.e. $-1 < z < 1$)
 - Further investigation to determine optimal windows
- ❖ Opportunity integrate further signals and machine learning
 - Bollinger Bands with pairs

03

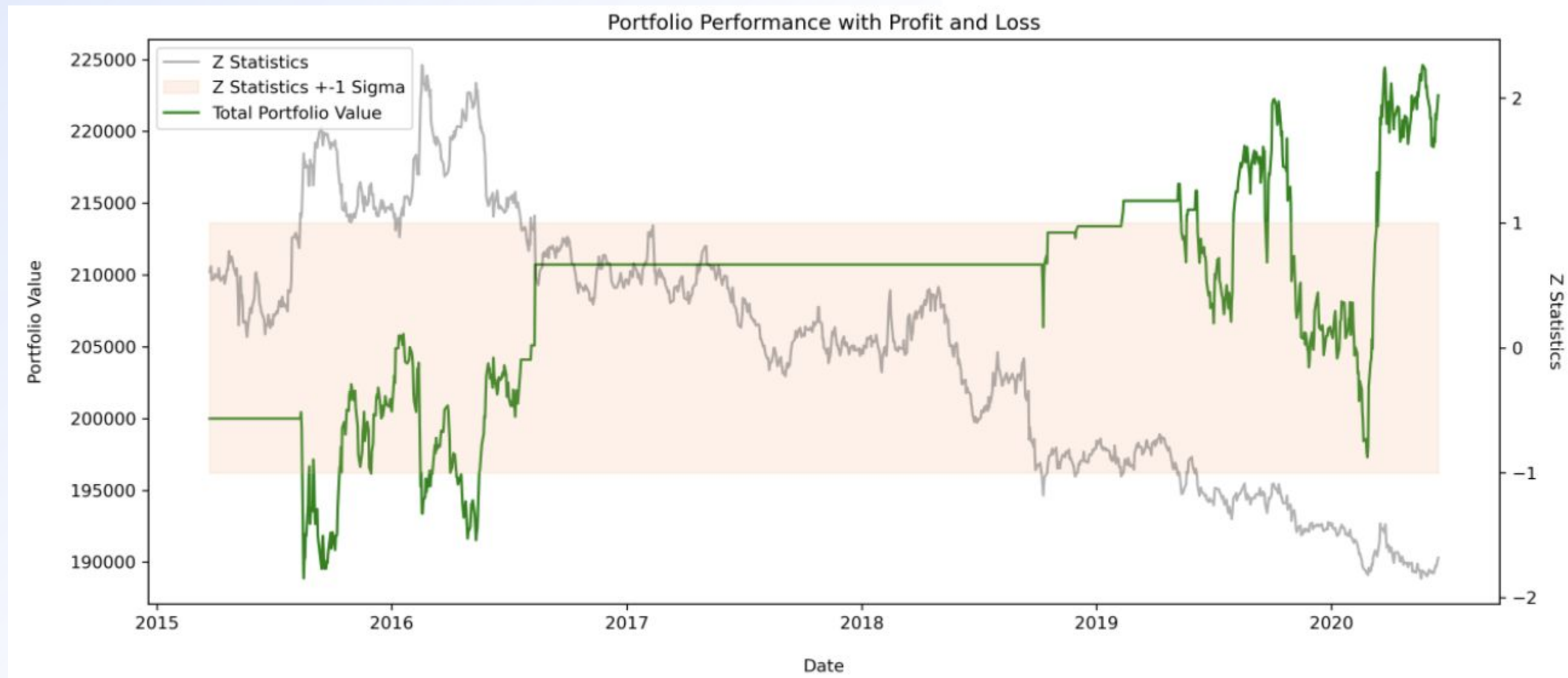
Strategy Rationale

- ❖ Thorough research exists
- ❖ "Statistical arbitrage with Cointegration" - Machine Learning for Trading
- ❖ yfinance for data
- ❖ statsmodels for cointegration tests
- ❖ Market-Neutral strategy - can yield profits in both bull and bear markets

- ❖ Thorough pre-selection measures for pairs
- ❖ Emphasis on high variance pairs
- ❖ Dynamically optimize allocations in portfolio
- ❖ Initial fundamental screening to weed out potentially unstable companies

❖ [Sabir Jana \(2020\)](#)

➤ 16.5% CAGR



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Literature Review

- ★ <https://numpy.org/doc/stable/reference/>
- ★ <https://www.investopedia.com/terms/p/pairstrade.asp>
- ★ <https://www.statsmodels.org/dev/generated/statsmodels.tsa.stattools.coint.html>
- ★ <https://seaborn.pydata.org/generated/seaborn.heatmap.html>
- ★ <https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html>
- ★ <https://www.statisticshowto.com/cointegration/>
- ★ <https://www.linkedin.com/pulse/momentum-reversion-poor-mans-trading-strategies-adam-darmanin-vhrif>

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Proposed Timeline

Dates	Objective
Jan 30 - Feb 6	Begin writing Python code and have a functioning code for the screening process, data collection, and pre-selection measures.
Feb 6 - Feb 13	Write code for cointegration testing and technical indicators.
Feb 13 - Feb 20	Backtest
Feb 20 - Feb 27	Continue backtesting and refining model parameters