

Financial Econometrics Presentation

# Analyzing Stocks

Exploring the world of investments and  
financial decision making

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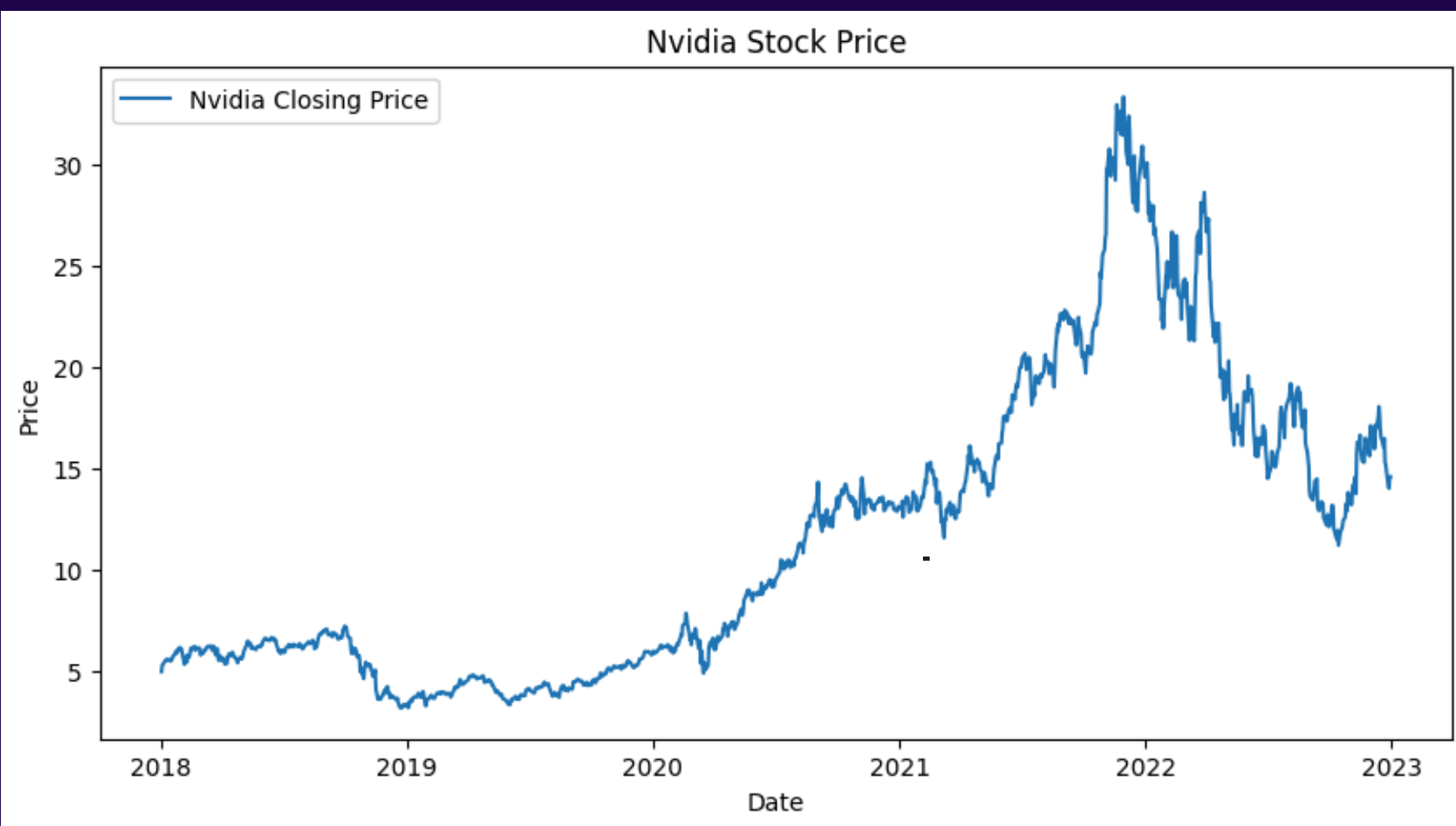
**TEAM SPADES**

# Contents

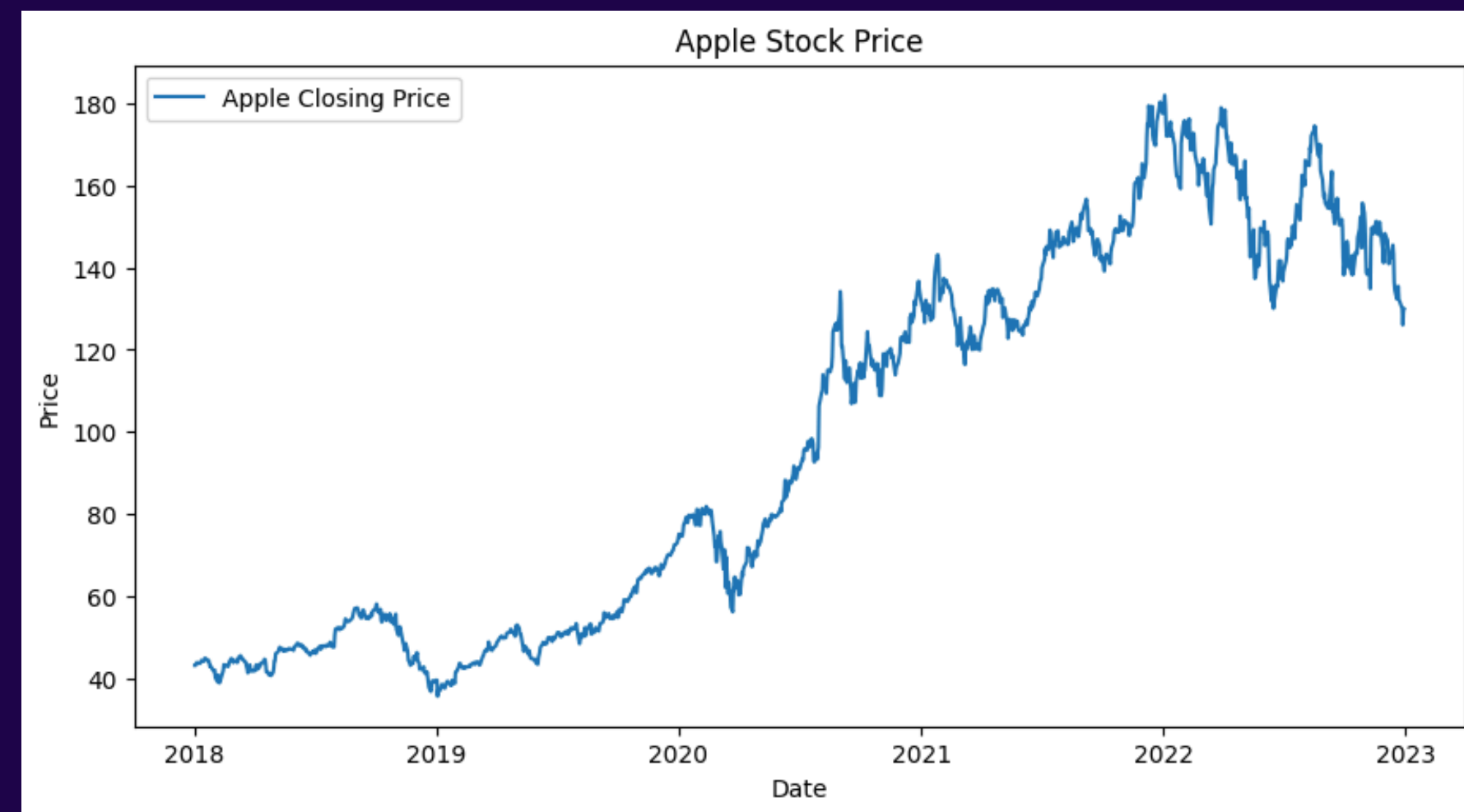
- Overview of Stocks We Use
- MA
- RSI
- Bollinger Bands
- Cointegration
- Mean Reversion

# Overview of Stocks Used

## NVDA



## AAPL



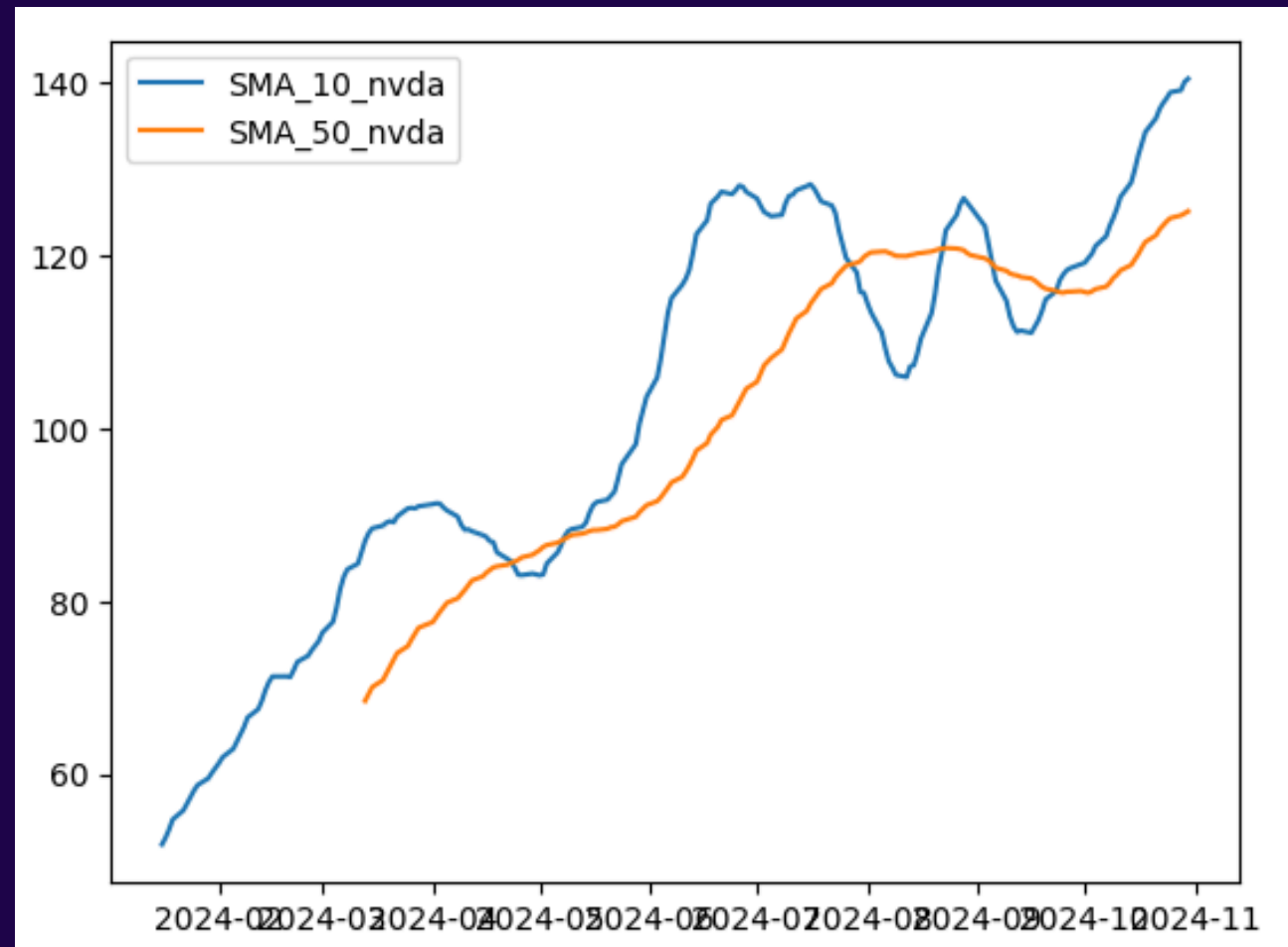
# MA STRATEGY

## NVDA

### EXAMPLE CODE

```
nvda['sma_50_nvda']=nvda['Adj Close'].rolling(window=50).mean()  
nvda['sma_10_nvda']=nvda['Adj Close'].rolling(window=10).mean()  
import matplotlib.pyplot as plt  
plt.plot(nvda['sma_10_nvda'], label='SMA_10_nvda')  
plt.plot(nvda['sma_50_nvda'], label='SMA_50_nvda')  
plt.legend()  
plt.show()
```

### VISUALIZATION OF NVDA MA10/50

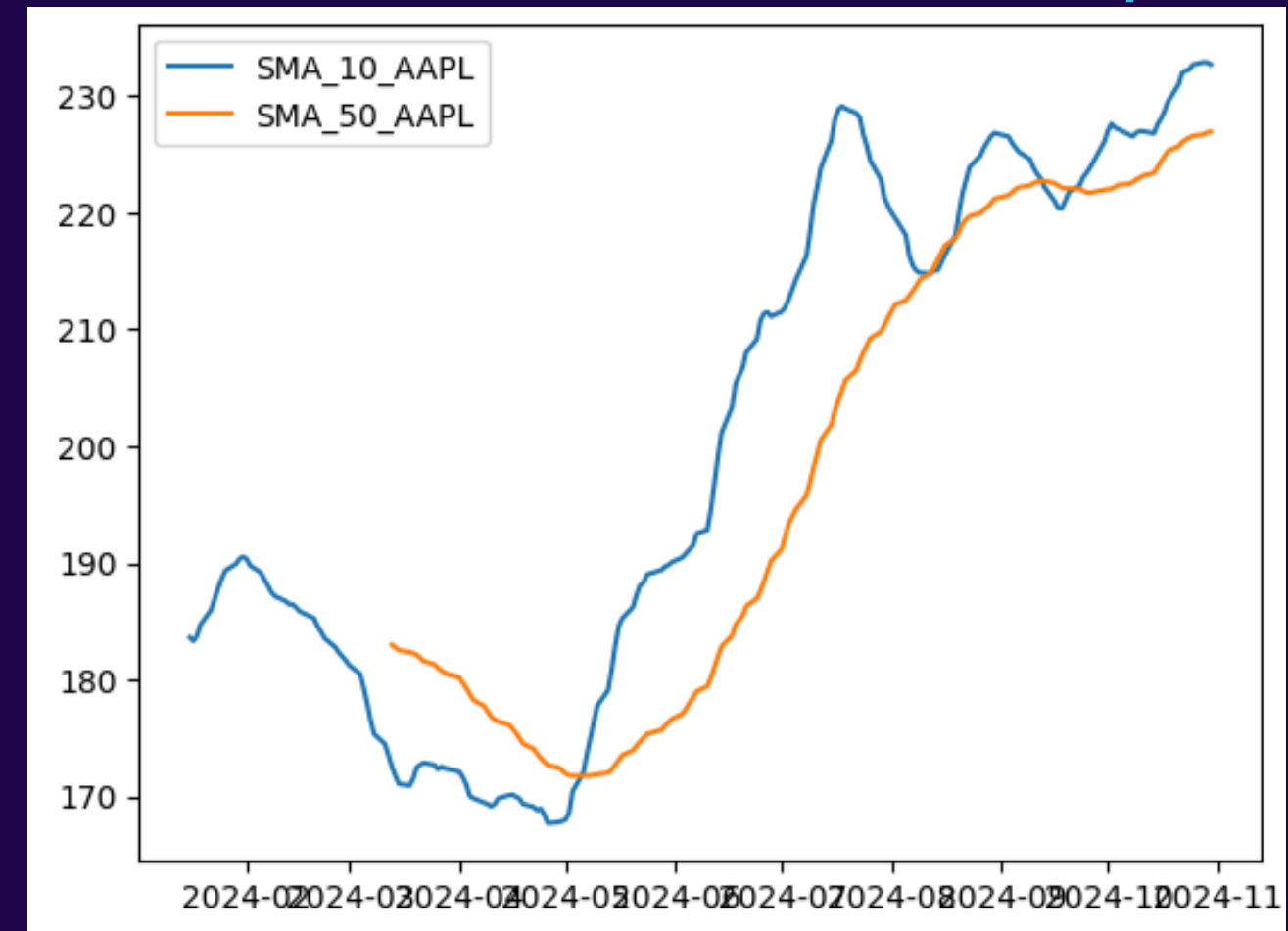


## AAPL

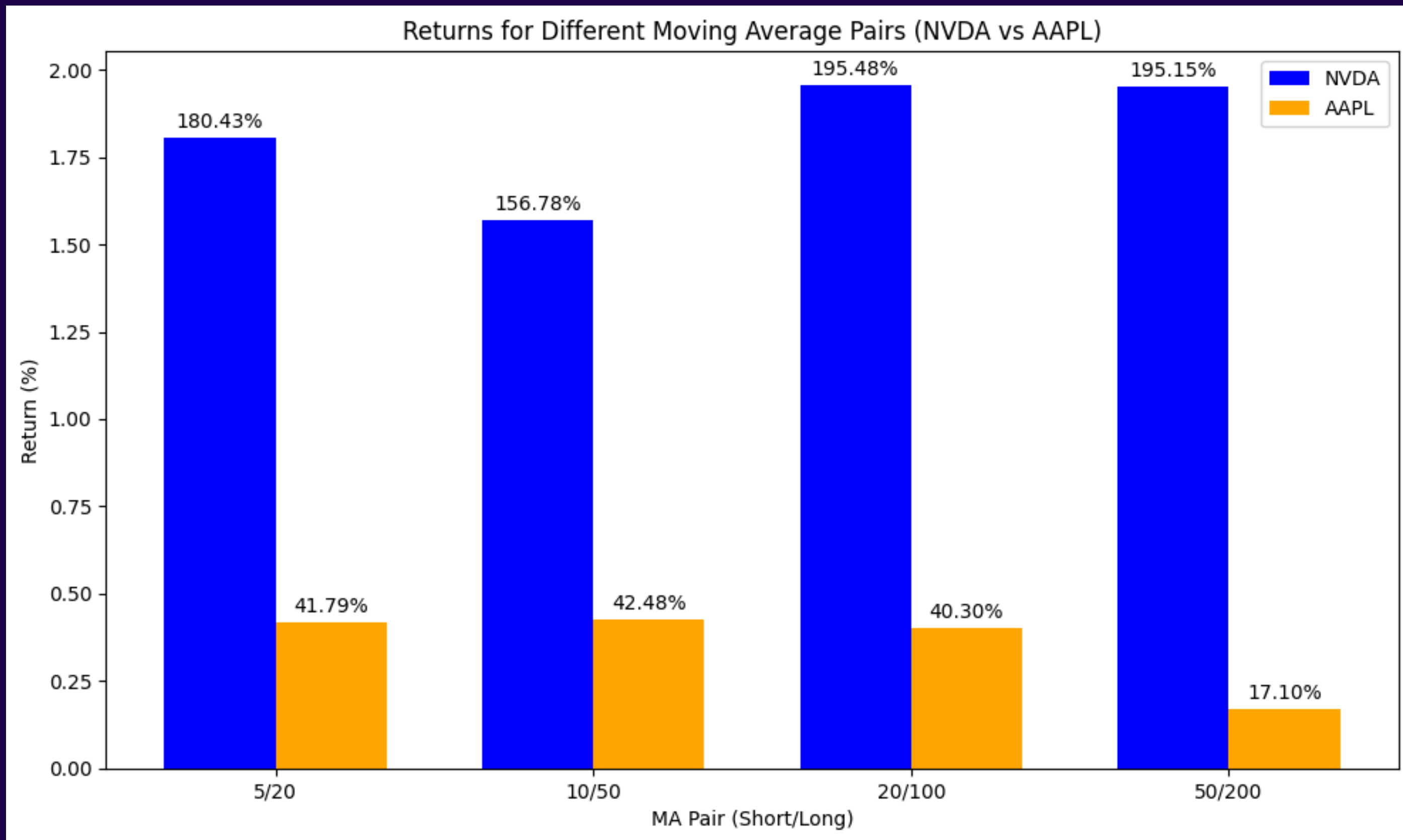
### EXAMPLE CODE

```
aapl = yf.download('AAPL', start='2024-1-1', end='2024-10-31')  
aapl['sma_50_AAPL']=aapl['Adj Close'].rolling(window=50).mean()  
aapl['sma_10_AAPL']=aapl['Adj Close'].rolling(window=10).mean()  
import matplotlib.pyplot as plt  
plt.plot(aapl['sma_10_AAPL'], label='SMA_10_AAPL')  
plt.plot(aapl['sma_50_AAPL'], label='SMA_50_AAPL')  
plt.legend()  
plt.show()
```

### VISUALIZATION OF AAPL MA10/50



# MA Optimazation



# RSI STRATEGY

## OUR CODE

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

def RSI(bbca, period=14):
    delta = bbca["Adj Close"].diff(1)
    delta = delta.dropna()
    up = delta.copy()
    down = delta.copy()
    up[up < 0] = 0
    down[down > 0] = 0

    # Calculate the EWMA
    roll_up1 = up.ewm(com=period - 1, adjust=False).mean()
    roll_down1 = down.ewm(com=period - 1, adjust=False).mean().abs()

    # Calculate the RS
    RS = roll_up1 / roll_down1

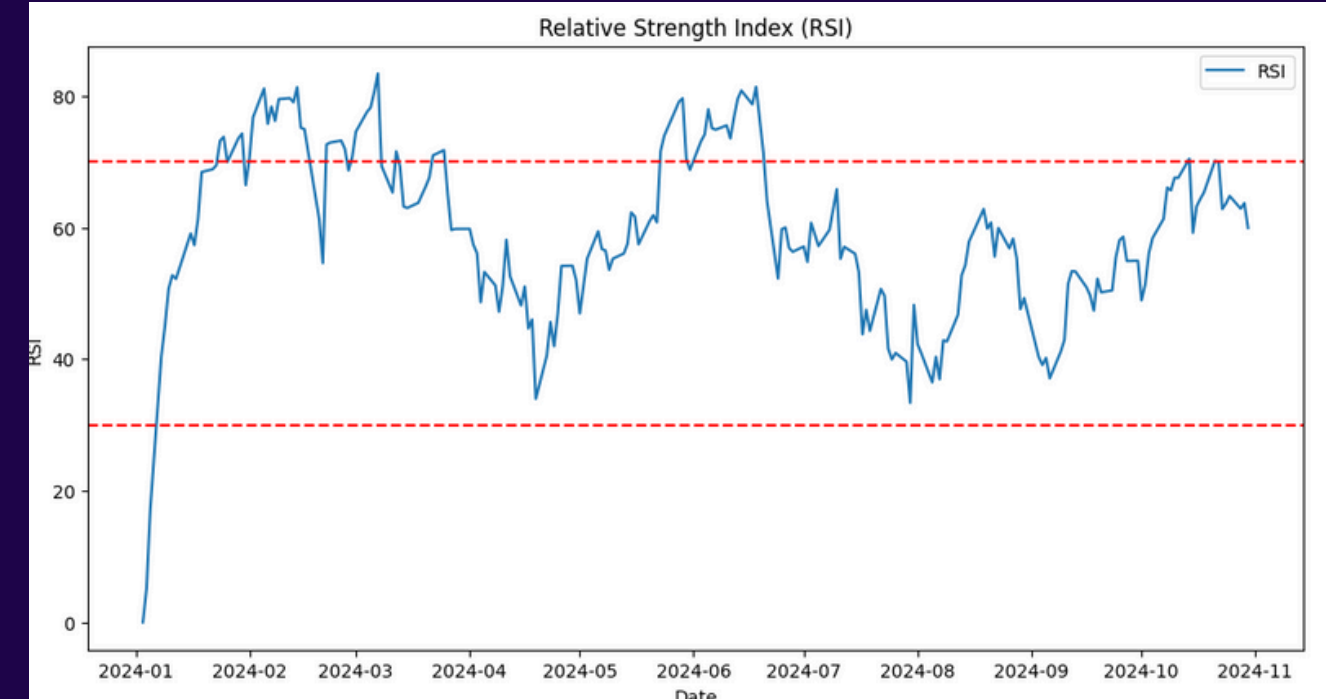
    # Calculate the RSI
    RSI = 100.0 - (100.0 / (1.0 + RS))

    bbca['RSI'] = RSI
    return bbca

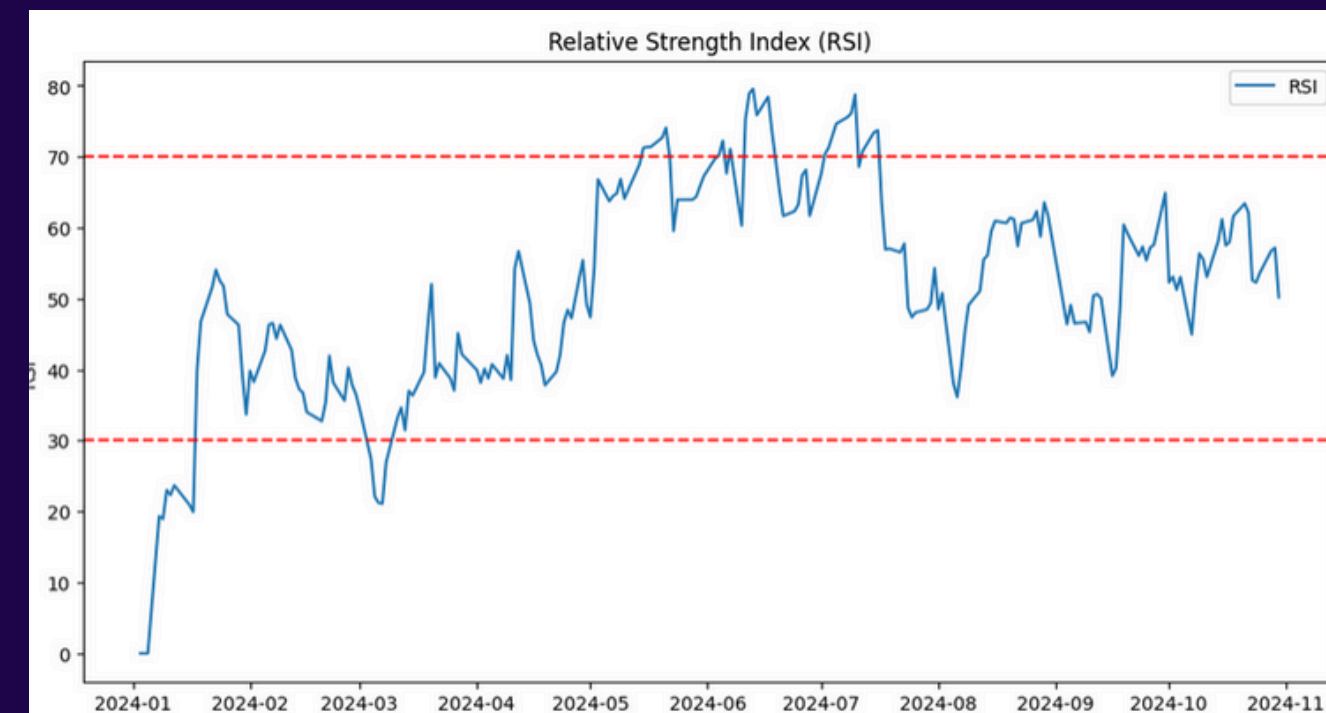
# Example usage:
# Assuming you have a DataFrame 'df' with 'Adj Close' column
bbca = RSI(bbca, period=14)

# Plot the RSI
plt.figure(figsize=(12, 6))
plt.plot(bbca.index, bbca['RSI'], label='RSI')
plt.axhline(30, color='r', linestyle='--')
plt.axhline(70, color='r', linestyle='--')
plt.xlabel('Date')
plt.ylabel('RSI')
plt.title('Relative Strength Index (RSI)')
plt.legend()
plt.show()
```

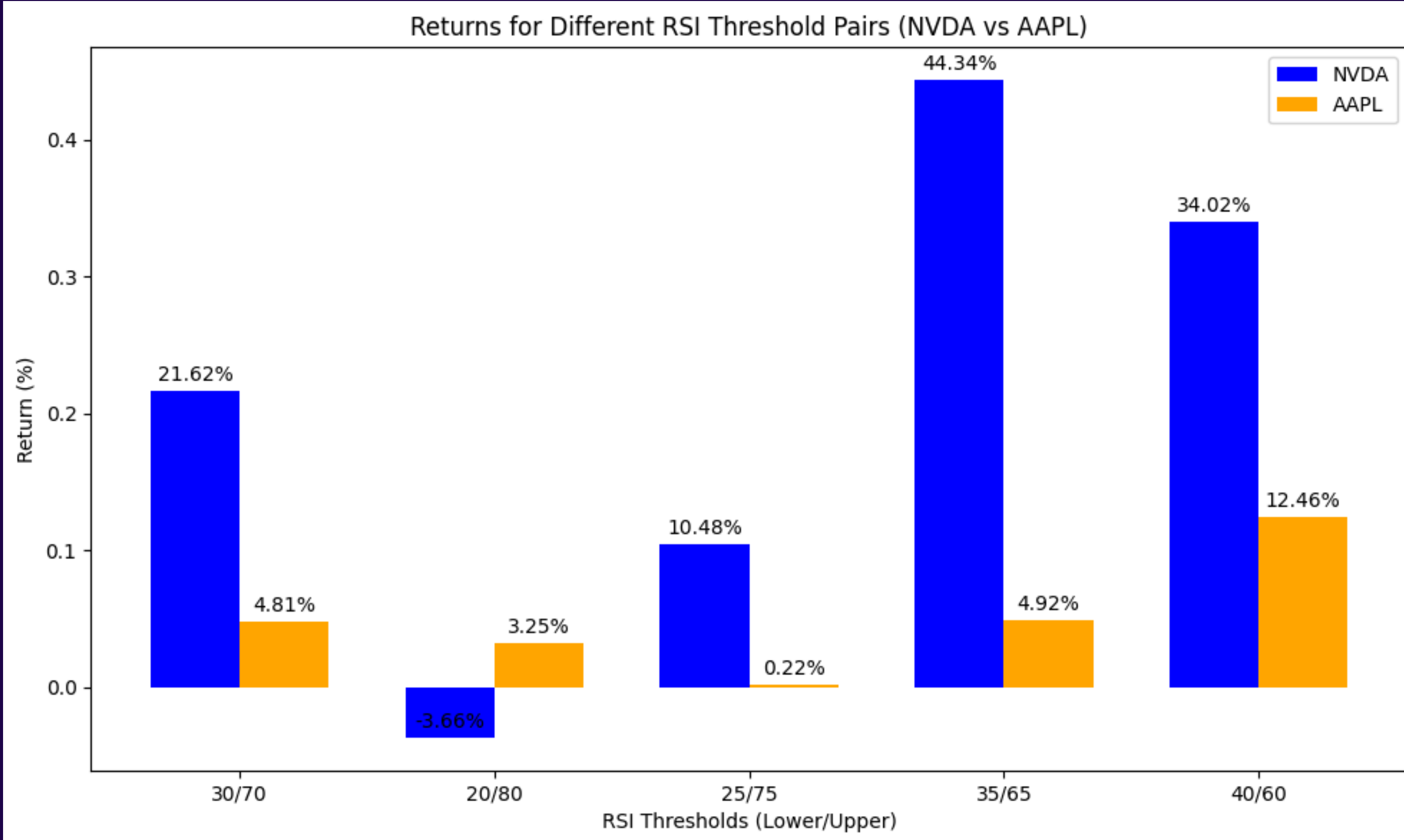
## NVDA



## AAPL



# RSI Optimization





# BOLLINGER BANDS STRATEGY

## OUR CODE

```
import yfinance as yf
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

# Fetch the data
ticker1 = "NVDA"
ticker2 = "AAPL"
start_date = "2024-01-01"
end_date = "2024-10-31"

data1 = yf.download(ticker1, start=start_date, end=end_date)
data2 = yf.download(ticker2, start=start_date, end=end_date)

# Function to calculate Bollinger Bands
def bollinger_bands(df, window=20, std_dev=2):
    rolling_mean = df['Close'].rolling(window).mean()
    rolling_std = df['Close'].rolling(window).std()

    df['Bollinger_Mid'] = rolling_mean
    df['Bollinger_Upper'] = rolling_mean + (rolling_std * std_dev)
    df['Bollinger_Lower'] = rolling_mean - (rolling_std * std_dev)

    return df

# Calculate Bollinger Bands for NVDA and AAPL
data1 = bollinger_bands(data1)
data2 = bollinger_bands(data2)

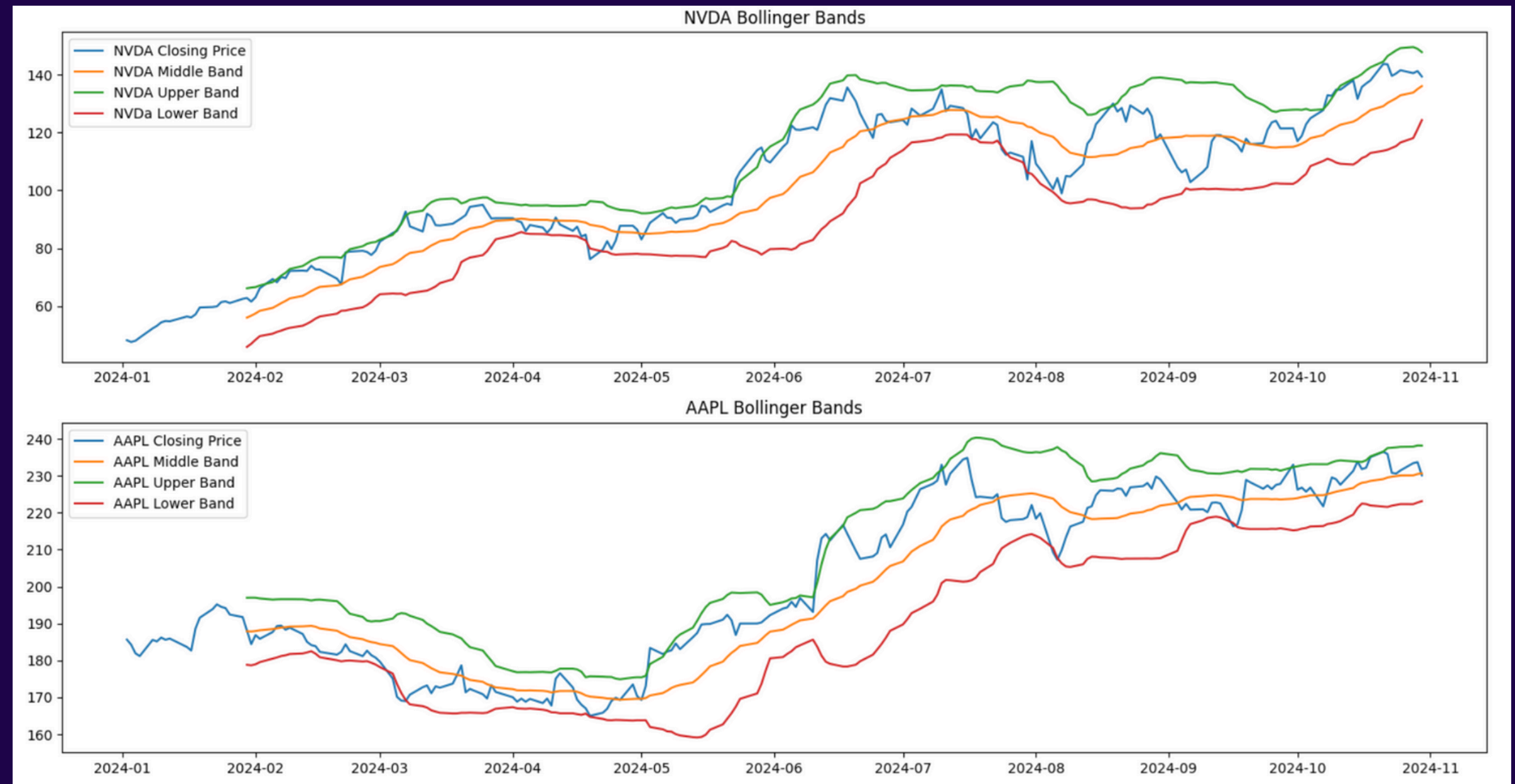
# Visualize the Bollinger Bands
plt.figure(figsize=(15, 8))

# Plot NVDA
plt.subplot(2, 1, 1)
plt.plot(data1.index, data1['Close'], label='NVDA Closing Price')
plt.plot(data1.index, data1['Bollinger_Mid'], label='NVDA Middle Band')
plt.plot(data1.index, data1['Bollinger_Upper'], label='NVDA Upper Band')
plt.plot(data1.index, data1['Bollinger_Lower'], label='NVDA Lower Band')
plt.title('NVDA Bollinger Bands')
plt.legend()

# Plot AAPL
plt.subplot(2, 1, 2)
plt.plot(data2.index, data2['Close'], label='AAPL Closing Price')
plt.plot(data2.index, data2['Bollinger_Mid'], label='AAPL Middle Band')
plt.plot(data2.index, data2['Bollinger_Upper'], label='AAPL Upper Band')
plt.plot(data2.index, data2['Bollinger_Lower'], label='AAPL Lower Band')
plt.title('AAPL Bollinger Bands')
plt.legend()

plt.tight_layout()
plt.show()
```

## OUR CHART

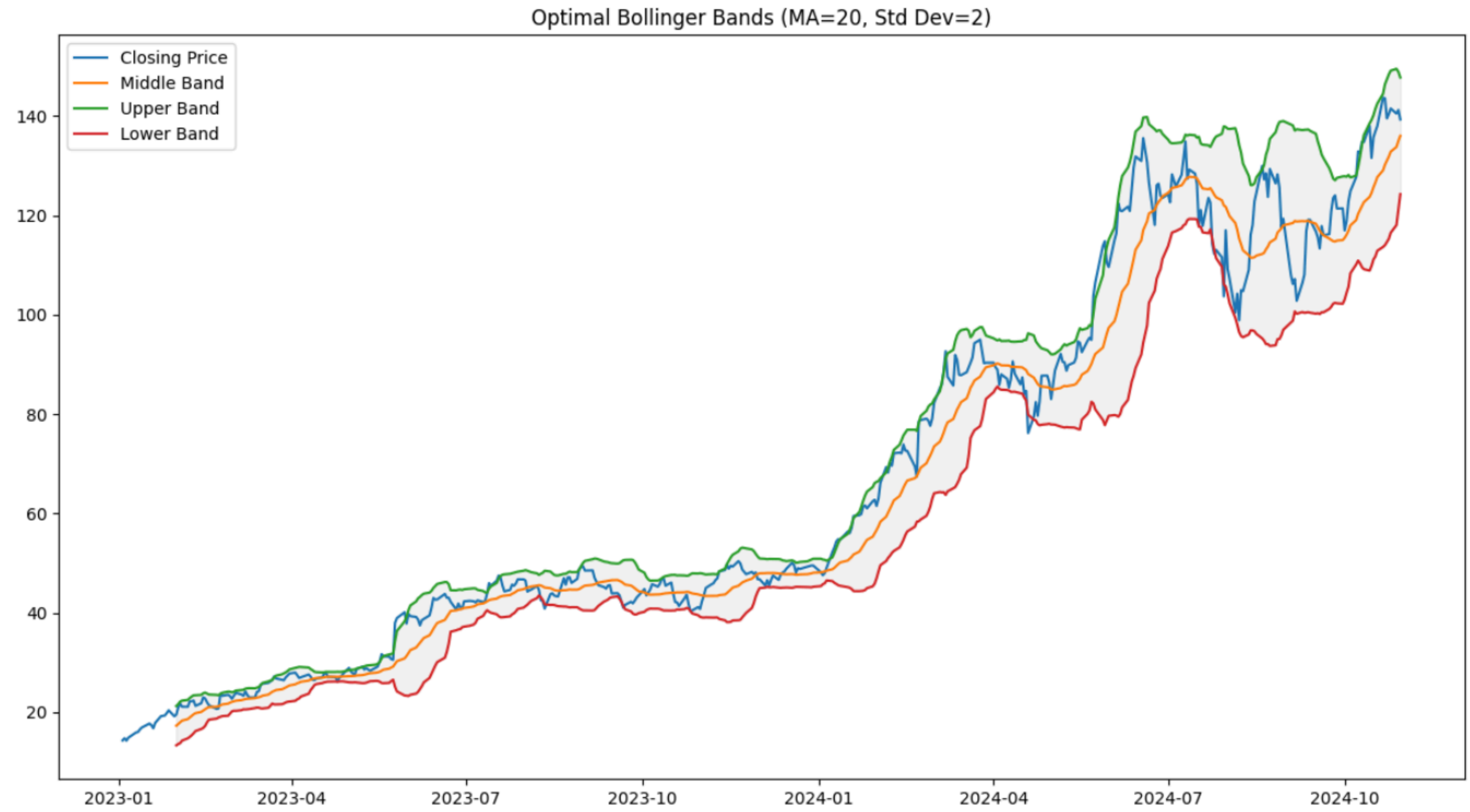




# Bollinger Bands Return

Best Parameters: MA Window = 20, Std Dev = 2

Best Strategy Return: 10.59%



# COINTEGRATION STRATEGY

## OUR CODE

```
import yfinance as yf
import pandas as pd
import numpy as np
import statsmodels.api as sm
import matplotlib.pyplot as plt

import yfinance as yf
bbri = yf.download('NVDA', start='2024-1-1', end='2024-10-31')
bbca = yf.download('AAPL', start='2024-1-1', end='2024-10-31')

[*****100%*****] 1 of 1 complete
[*****100%*****] 1 of 1 complete

# Check for cointegration
result = sm.tsa.stattools.coint(bbri['Adj Close'], bbca['Adj Close'])
print(result)

(np.float64(-2.350848491683655), np.float64(0.34847303289756365), 0.9999999999999999)

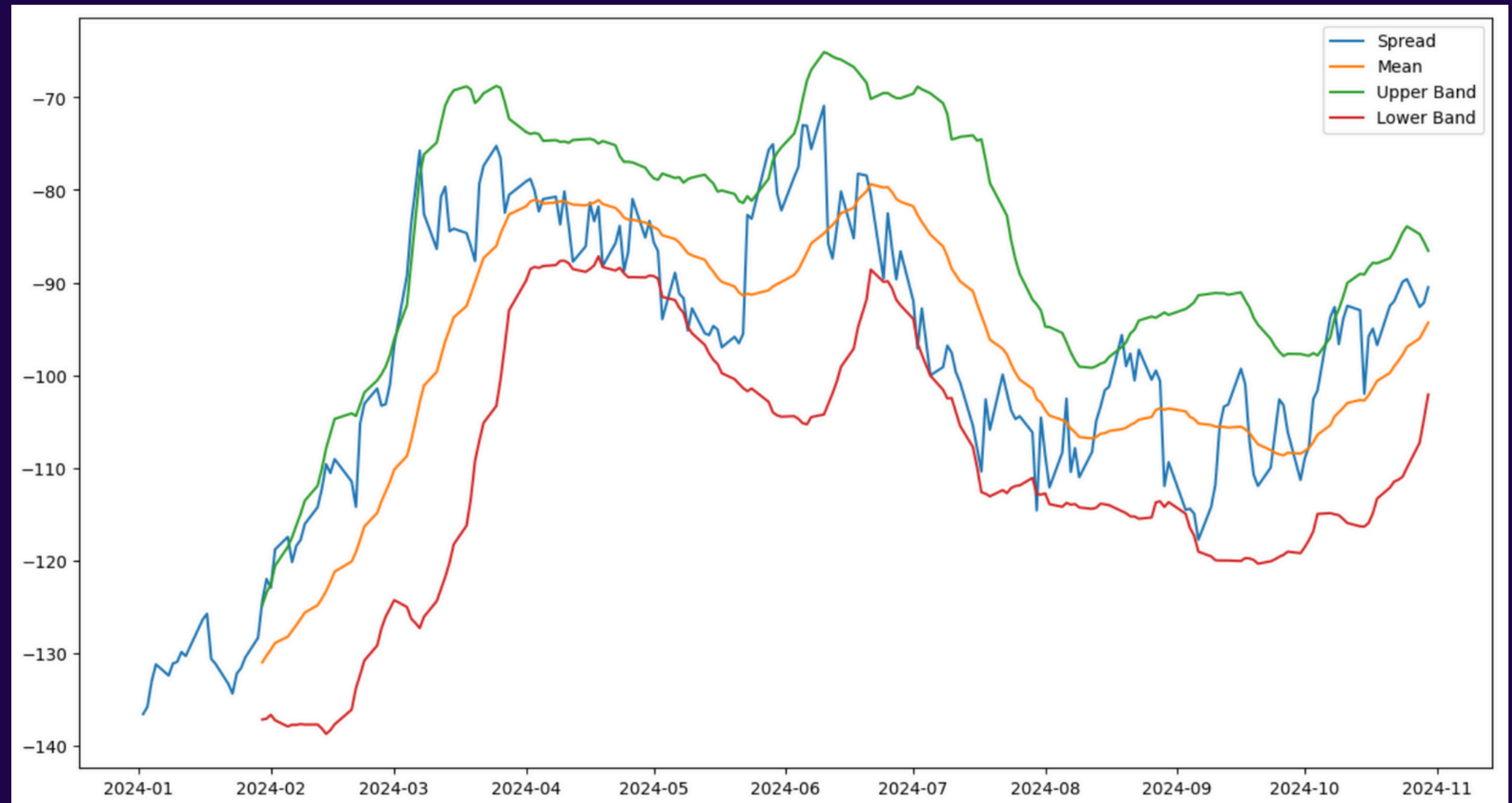
# Calculate the spread
bbri['spread'] = bbri['Adj Close'] - bbca['Adj Close']

# Calculate Z-score
bbri['z_score'] = (bbri['spread'] - bbri['spread'].rolling(window=20).mean()) / bbri['spread'].rolling(window=20).std()

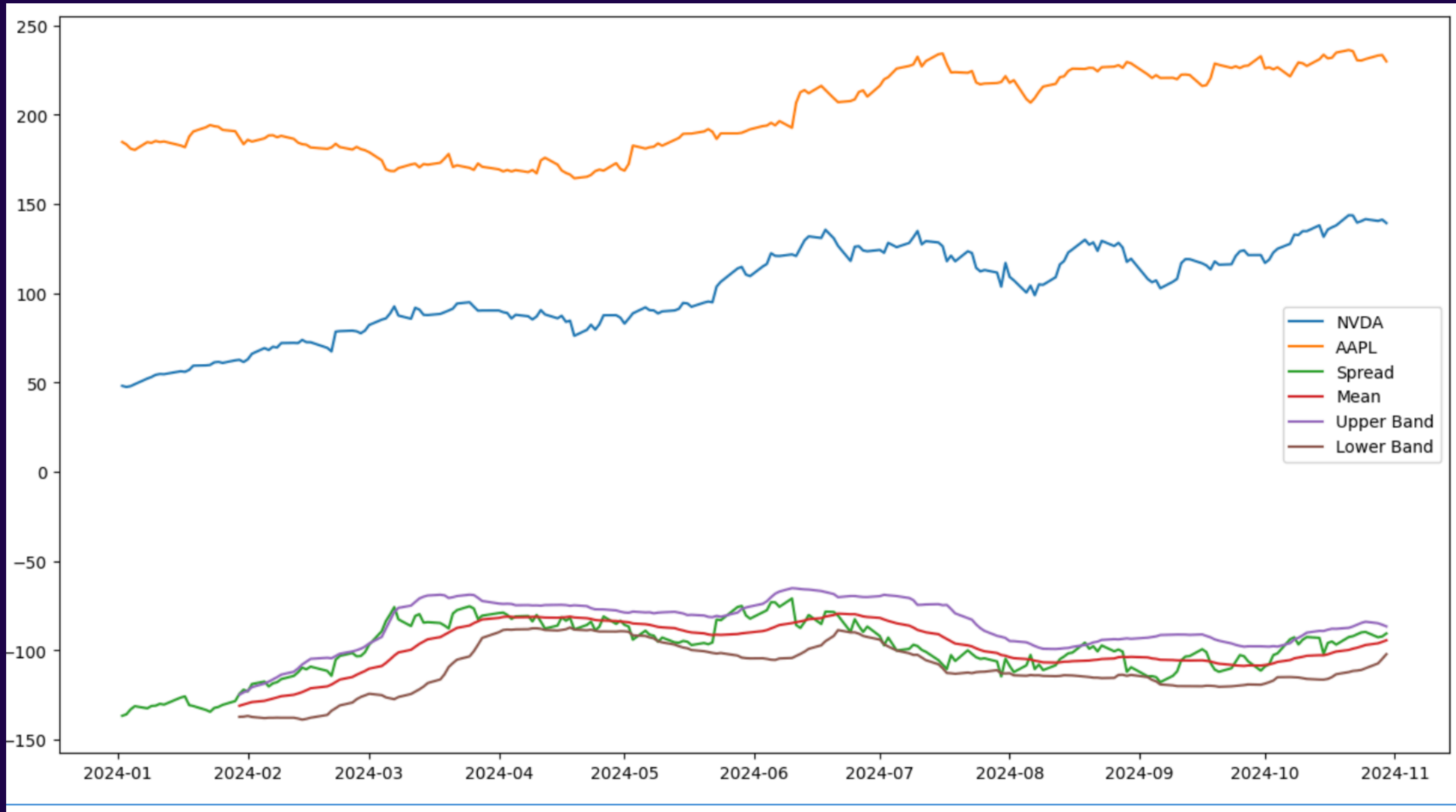
# Define trading signals
bbri['position'] = np.where(bbri['z_score'] > 2, -1, np.where(bbri['z_score'] < -2, 1, 0))

plt.figure(figsize=(15, 8))
plt.plot(bbri.index, bbri['spread'], label='Spread')
plt.plot(bbri.index, bbri['spread'].rolling(window=20).mean(), label='Mean')
plt.plot(bbri.index, bbri['spread'].rolling(window=20).mean() + 2*bbri['spread'].rolling(window=20).std(), label='Upper Band')
plt.plot(bbri.index, bbri['spread'].rolling(window=20).mean() - 2*bbri['spread'].rolling(window=20).std(), label='Lower Band')
plt.legend()
plt.show()
```

## OUR CHART



# COINTEGRATION STRATEGY cont.

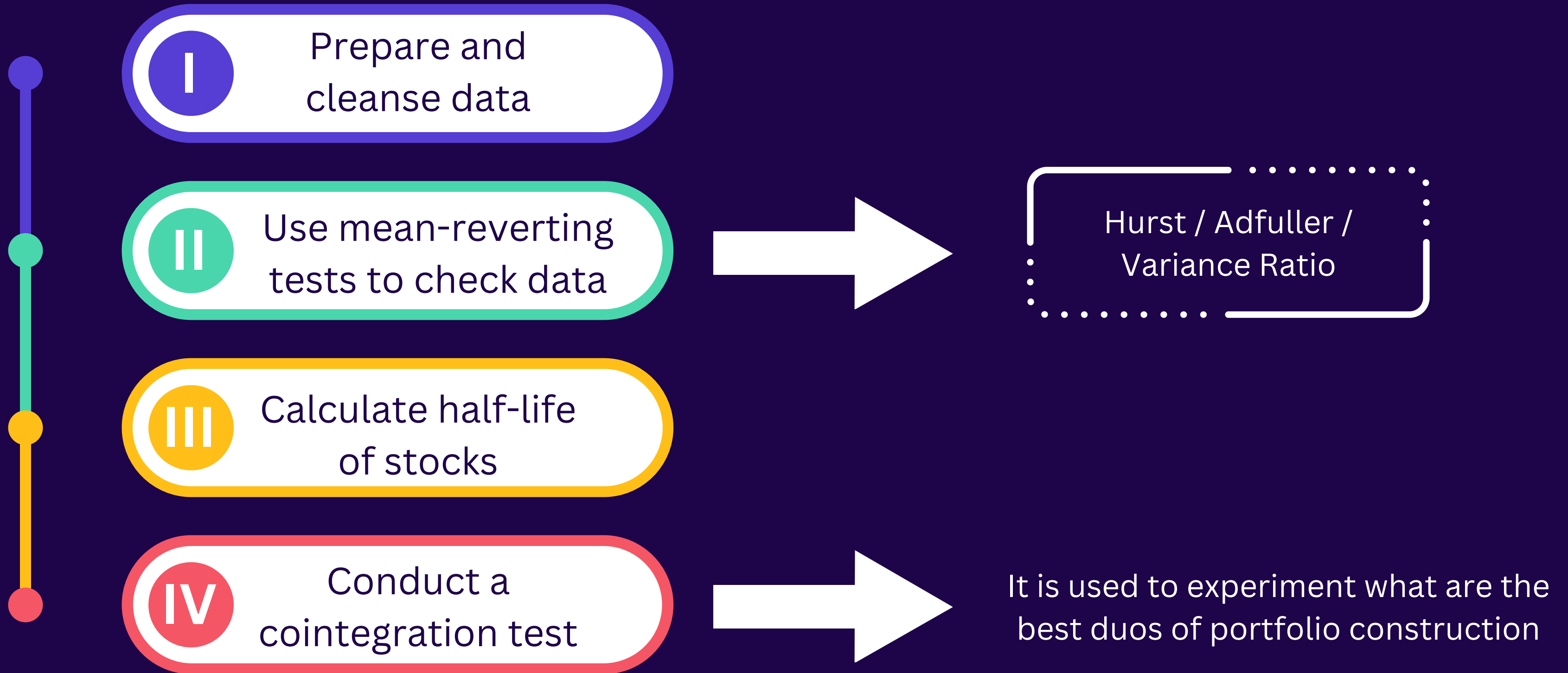


# Returns From Trading Strategies in 2 Years

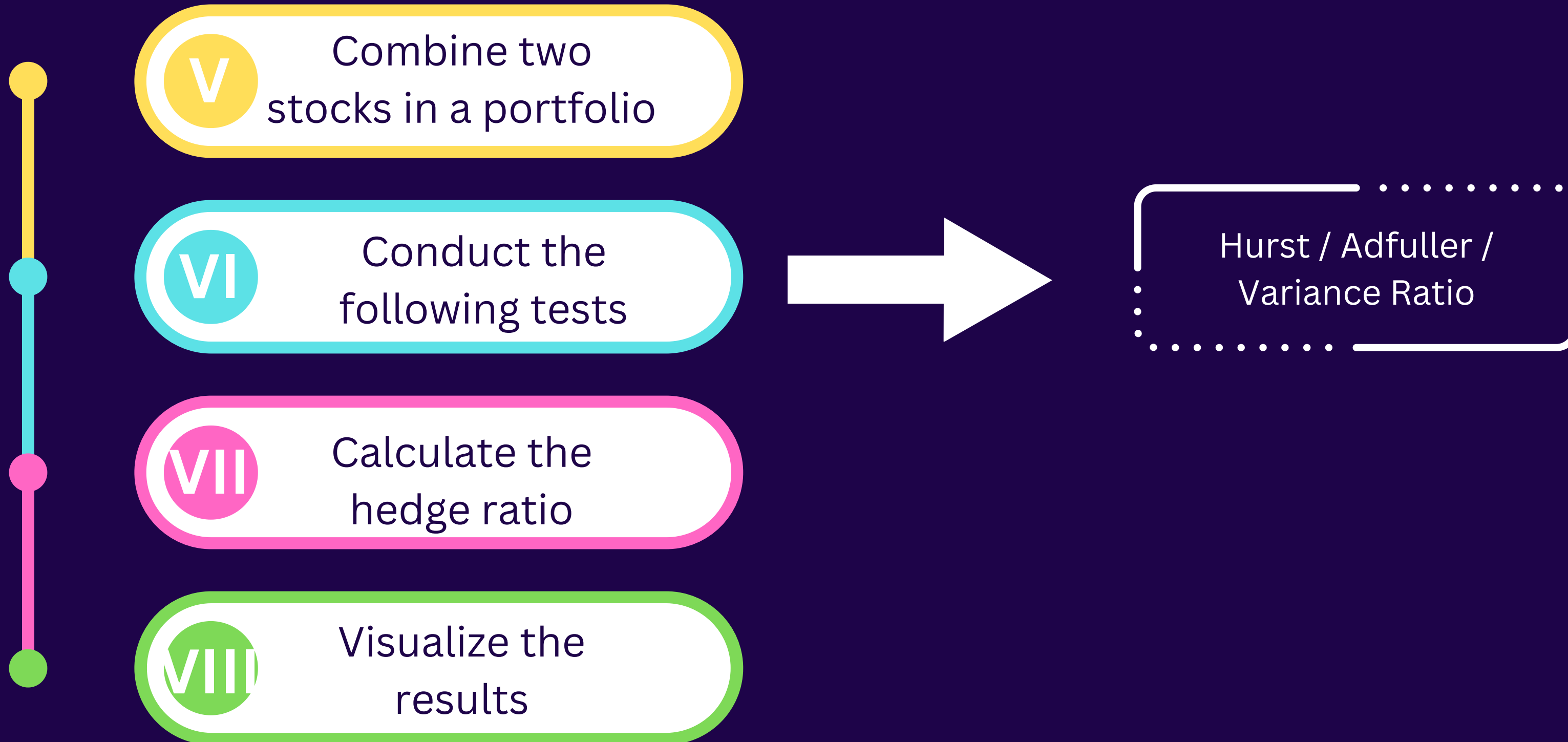
Test	NVIDIA	Apple
MA (20,100)	195%	40%
RSI (40,60)	34%	12,4%
Bollinger (20,2)	10,59% (joint)	
Cointegration		

# Mean Reversion Strategy

# Mean Reversion Strategy



# Mean Reversion Strategy





# Mean Reversion Strategy



IX

Incorporate Z-score index and stop-loss

X

Visualize the results

XI

Compare the performances

XII

Optimize the results

Conditions:

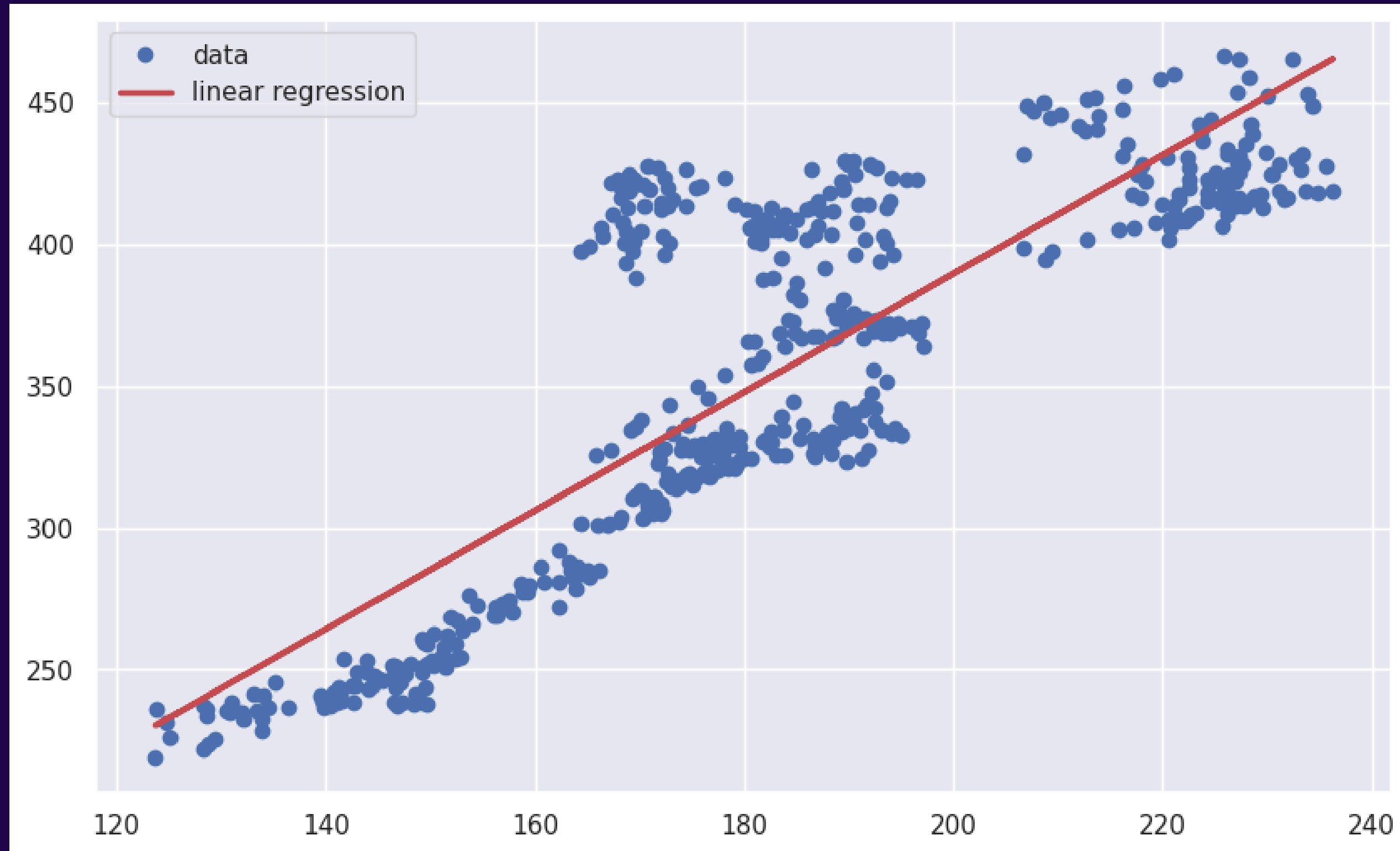
- If price-loss percentage < stop-loss: close the position if loss percentage > stop-loss threshold
- If Zscore < -threshold & no long-position => buy
- If Zscore > threshold & no short-position => sellll

Maximize Equity + Minimize Maximum Drawdown + Minimize Volatility

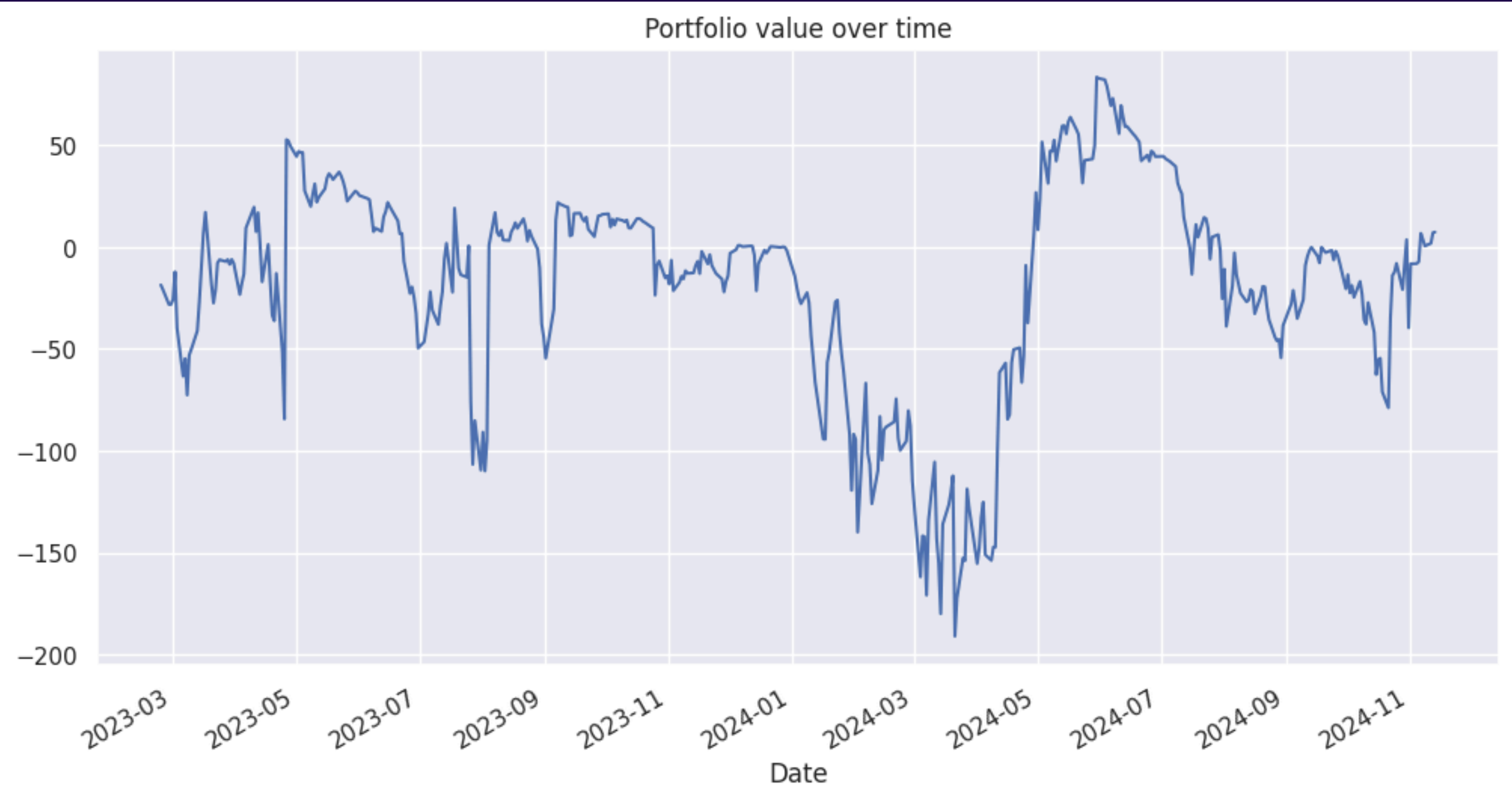
# Experimentation of Mean-Reverting Indicators

Test	NVIDIA	Apple	Microsoft
Hurst Exponent	0.718513	0.929772	0.784275
Variance Ratio	0.184508	0.181548	0.678936
Half-life	-788	181	153

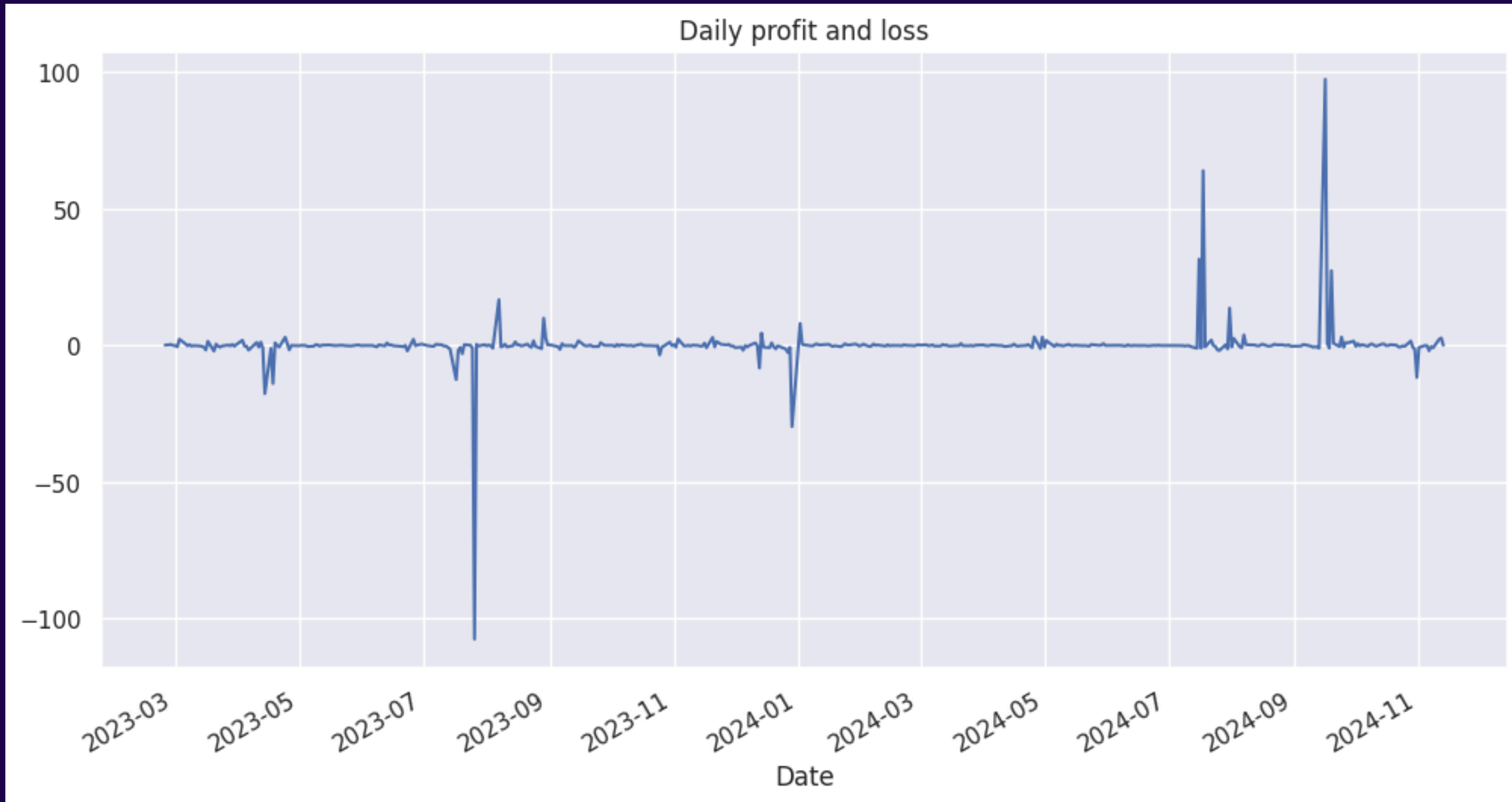
# Cointegration of Two stocks & Portfolio Construction



# Portfolio Value



# Daily profit & loss



# Backtesting

Given cash = 10,000  
interval = 2 years

Start	2022-11-14 00:00:00+00:00
End	2024-11-13 00:00:00+00:00
Duration	730 days 00:00:00
Exposure Time [%]	4.572565
Equity Final [\$]	80538.86331
Equity Peak [\$]	83574.661864
Return [%]	705.388633
Buy & Hold Return [%]	-76.250713
Return (Ann.) [%]	184.382816
Volatility (Ann.) [%]	2904.461708

# Appendrix



**Google Colab**

 [google.com](https://colab.google.com)



**arch**

ARCH for Python

 PyPI

The Bokeh logo, featuring the word "bokeh" in a stylized font with a colorful, abstract background.

**Bokeh**

Bokeh is a Python-based visualization library, capable of building plots from simple charts to interactive dashboards.


 [bokeh.org](https://bokeh.org)

edgetrader/**mean-reversion-strategy**

Mean Reversion Trading Strategy



 1  
Contributor

 0  
Issues

 26  
Stars

 9  
Forks



**mean-reversion-strategy/notebook/mean-reversion-strategy.ipynb at...**

Mean Reversion Trading Strategy. Contribute to edgetrader/mean-reversion-strategy development by creating an account on GitHub.

 GitHub



# Summary

Presentation

Thank you

Spades