

FinalCode

May 13, 2025

1 Mounting a drive

```
[1]: from google.colab import drive
     drive.mount('/content/drive')
```

Mounted at /content/drive

```
[2]: !pip install newsapi-python
```

Collecting newsapi-python

Downloading newsapi_python-0.2.7-py2.py3-none-any.whl.metadata (1.2 kB)

Requirement already satisfied: requests<3.0.0 in /usr/local/lib/python3.11/dist-packages (from newsapi-python) (2.32.3)

Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests<3.0.0->newsapi-python) (3.4.2)

Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests<3.0.0->newsapi-python) (3.10)

Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests<3.0.0->newsapi-python) (2.4.0)

Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests<3.0.0->newsapi-python) (2025.4.26)

Downloading newsapi_python-0.2.7-py2.py3-none-any.whl (7.9 kB)

Installing collected packages: newsapi-python

Successfully installed newsapi-python-0.2.7

2 Libraries

```
[3]: import pandas as pd
     import numpy as np
     import datetime
     import yfinance as yf
     from newsapi import NewsApiClient
     from transformers import pipeline
     from collections import Counter
```

```
import matplotlib.pyplot as plt
```

3 loading a Fine-tuned FinBERT model

```
[4]: model_path = "/content/drive/MyDrive/my-finbert-finetuned"
finbert = pipeline("sentiment-analysis", model=model_path)
```

Device set to use cuda:0

4 Fetching News Headlines using NewsAPI

```
[5]: newsapi = NewsApiClient(api_key='09458a27b33349439c94c2c7a302cfd0') #
    ↪ '51802ca249bf426da64b9b1b1bdae59b' backup key

def get_news_for_date(date='2025-05-05', query='stock market',
    ↪ max_articles=100):
    articles = newsapi.get_everything(
        q=query,
        language='en',
        sort_by='publishedAt',
        from_param=date,
        to=date,
        page_size=100
    )['articles']

    news_df = pd.DataFrame([
        'date': a['publishedAt'][:10],
        'headline': a['title']
    ] for a in articles)

    return news_df.head(max_articles)

news_df = get_news_for_date()
print(news_df.head())
```

	date	headline
0	2025-05-05	Rupee strengthens at 84.25 amid falling oil pr...
1	2025-05-05	Companies Like Compugates Holdings Berhad (KLS...
2	2025-05-05	Can CE Technology Berhad's (KLSE:CETECH) Decen...
3	2025-05-05	Barbie Toymaker Mattel Warns Tariffs May Force...
4	2025-05-05	Asia markets live: Stocks set to trade mixed

5 Ticker Dictionary

```
[6]: ticker_dict = {
    "Apple": "AAPL", "Microsoft": "MSFT", "Amazon": "AMZN",
    "Tesla": "TSLA", "Meta": "META", "Nvidia": "NVDA",
    "Google": "GOOGL", "Alphabet": "GOOGL", "Netflix": "NFLX", "Advanced Micro_
↳Devices Inc": "AMD", "S&P 500": "SPY"
}
```

- In this dictionary as much as ticker can be added for further process and identify more tickers in the News headlines

6 Ticker Extraction and Sentiment analysing

```
[7]: def extract_ticker(headline, ticker_dict):
    for company, ticker in ticker_dict.items():
        if company.lower() in headline.lower() or ticker.lower() in headline.
↳lower():
            return ticker, company
    return "SPY", "S&P 500"
```

```
[8]: def analyze_sentiment(text):
    result = finbert(text)[0]
    return result["label"].lower(), result["score"]
```

7 RSI Indicator

```
[9]: def calculate_rsi(data, period=14):
    delta = data['Close'].diff()
    gain = delta.clip(lower=0)
    loss = -delta.clip(upper=0)
    avg_gain = gain.rolling(period).mean()
    avg_loss = loss.rolling(period).mean()
    rs = avg_gain / avg_loss
    rsi = 100 - (100 / (1 + rs))
    return rsi.iloc[-1] if not rsi.empty else np.nan

def rsi_signal(rsi_value):
    if isinstance(rsi_value, pd.Series):
        rsi_value = rsi_value.iloc[0]

    if np.isnan(rsi_value):
        return 0
    if rsi_value < 30:
        return 1
```

```

elif rsi_value > 70:
    return -1
else:
    return 0

```

8 MACD Indicator

```

[10]: def calculate_macd(data):
    ema12 = data['Close'].ewm(span=12).mean()
    ema26 = data['Close'].ewm(span=26).mean()
    macd_line = ema12 - ema26
    signal_line = macd_line.ewm(span=9).mean()

    macd = macd_line.iloc[-1] if not macd_line.empty else np.nan
    signal = signal_line.iloc[-1] if not signal_line.empty else np.nan

    if isinstance(macd, pd.Series):
        macd = macd.iloc[0]
    if isinstance(signal, pd.Series):
        signal = signal.iloc[0]

    return macd, signal

def macd_signal(macd_value, signal_value):
    if isinstance(macd_value, pd.Series):
        macd_value = macd_value.iloc[0]
    if isinstance(signal_value, pd.Series):
        signal_value = signal_value.iloc[0]

    if np.isnan(macd_value) or np.isnan(signal_value):
        return 0

    if macd_value > signal_value:
        return 1
    elif macd_value < signal_value:
        return -1
    else:
        return 0

```

9 Mapping

```

[11]: decision_map = {"buy": 1, "hold": 0, "sell": -1}
    sentiment_map = {"positive": 1, "neutral": 0, "negative": -1}

```

10 Main Block for Calling function and final dataframe building

```
[12]: rows = []

for _, row in news_df.iterrows():
    date = row["date"]
    headline = row["headline"]

    # Extraction of company and its ticker
    ticker, company = extract_ticker(headline, ticker_dict)

    # sentiment and its score
    sentiment, sent_score = analyze_sentiment(headline)
    sentiment_action = sentiment_map.get(sentiment, 0)

    news_date = pd.to_datetime(date)
    start_date = (news_date - pd.Timedelta(days=60)).strftime('%Y-%m-%d')
    end_date = (news_date + pd.Timedelta(days=1)).strftime('%Y-%m-%d')

    price_data = yf.download(ticker, start=start_date, end=end_date,
↪progress=False)
    price_data.index = pd.to_datetime(price_data.index)

    # Ensuring we have OHLC data and news day are same
    if news_date not in price_data.index:
        print(f"No OHLC data found for {ticker} on {date}")
    else:
        same_day_ohlc = price_data.loc[news_date]

    if price_data.empty:
        continue

    # indicators call
    rsi_val = calculate_rsi(price_data)
    macd_val, macd_sig = calculate_macd(price_data)

    rsi_action = rsi_signal(rsi_val)
    macd_action = macd_signal(macd_val, macd_sig)

    # decision making by taking average of all scores
    final_score = np.mean([sentiment_action, rsi_action, macd_action])

    if final_score > 0.33:
        final_dec = "BUY"
    elif final_score < -0.33:
        final_dec = "SELL"
    else:
```

```

        final_dec = "HOLD"

    signal_components = {
        "Sentiment": sentiment_action,
        "RSI": rsi_action,
        "MACD": macd_action
    }

    # Find the strongest contributor (abs value)
    max_component = max(signal_components, key=lambda k: abs(
↪signal_components[k]))
    direction = "BUY" if signal_components[max_component] == 1 else "SELL" if
↪signal_components[max_component] == -1 else "HOLD"
    reason = f"{max_component} strongly suggested {direction}"

    rows.append({
        "date": date,
        "company": company,
        "ticker": ticker,
        "headline": headline,
        "sentiment": sentiment,
        "sentiment_score": round(sent_score, 4),
        "RSI": round(rsi_val, 2),
        "RSI_signal": rsi_action,
        "MACD": round(macd_val, 4),
        "MACD_signal": round(macd_sig, 4),
        "MACD_decision": macd_action,
        "final_score": round(final_score, 2),
        "final_decision": final_dec,
        "reason": reason
    })

# Final output DataFrame
final_df = pd.DataFrame(rows)
final_df.head()

```

YF.download() has changed argument auto_adjust default to True

You seem to be using the pipelines sequentially on GPU. In order to maximize efficiency please use a dataset

```

[12]:
      date  company ticker \
0  2025-05-05  S&P 500   SPY
1  2025-05-05  S&P 500   SPY
2  2025-05-05  S&P 500   SPY
3  2025-05-05  S&P 500   SPY
4  2025-05-05  S&P 500   SPY

```

	headline	sentiment	\
0	Rupee strengthens at 84.25 amid falling oil pr...	positive	
1	Companies Like Compugates Holdings Berhad (KLS...	positive	
2	Can CE Technology Berhad's (KLSE:CETECH) Decen...	positive	
3	Barbie Toymaker Mattel Warns Tariffs May Force...	negative	
4	Asia markets live: Stocks set to trade mixed	negative	

	sentiment_score		RSI	\
0	0.8046	Ticker		
	SPY	64.72		
	Name:	2025-05-05 00:00:00,...		
1	0.9996	Ticker		
	SPY	64.72		
	Name:	2025-05-05 00:00:00,...		
2	0.9577	Ticker		
	SPY	64.72		
	Name:	2025-05-05 00:00:00,...		
3	0.9516	Ticker		
	SPY	64.72		
	Name:	2025-05-05 00:00:00,...		
4	0.4354	Ticker		
	SPY	64.72		
	Name:	2025-05-05 00:00:00,...		

	RSI_signal	MACD	MACD_signal	MACD_decision	final_score	final_decision	\
0	0	4.2768	0.0634	1	0.67	BUY	
1	0	4.2768	0.0634	1	0.67	BUY	
2	0	4.2768	0.0634	1	0.67	BUY	
3	0	4.2768	0.0634	1	0.00	HOLD	
4	0	4.2768	0.0634	1	0.00	HOLD	

	reason
0	Sentiment strongly suggested BUY
1	Sentiment strongly suggested BUY
2	Sentiment strongly suggested BUY
3	Sentiment strongly suggested SELL
4	Sentiment strongly suggested SELL

11 Analysis of Ticker wise News

```
[13]: ticker = "AMZN"
      filtered_rows = final_df[final_df["ticker"] == ticker]
      print(filtered_rows)
```

Empty DataFrame

```
Columns: [date, company, ticker, headline, sentiment, sentiment_score, RSI,
RSI_signal, MACD, MACD_signal, MACD_decision, final_score, final_decision,
reason]
Index: []
```

12 Counting number of News per ticker

```
[14]: final_df["ticker"].value_counts()
```

```
[14]: ticker
      SPY      93
      NVDA       1
      AAPL       1
      TSLA       1
      NFLX       1
      Name: count, dtype: int64
```

13 Combining multiple news of same ticker and averaging into single decision

```
[15]: def aggregate_ticker_signals(final_df):
      aggregated_rows = []

      grouped = final_df.groupby(['ticker', 'date'])

      for (ticker, date), group in grouped:
          decisions = group['final_decision'].values
          sentiment_scores = group['sentiment_score'].values
          decision_counts = Counter(decisions)
          if len(decision_counts) == 1:
              final_decision = list(decision_counts.keys())[0]
          else:
              avg_score = np.mean(sentiment_scores)
              if avg_score > 0.33:
                  final_decision = 'BUY'
              elif avg_score < -0.33:
                  final_decision = 'SELL'
              else:
                  final_decision = 'HOLD'

          aggregated_rows.append({
              'ticker': ticker,
              'date': date,
              'final_decision': final_decision,
          })
```



```

    aggregated_df = pd.DataFrame(aggregated_rows)
    return aggregated_df

aggregated_df = aggregate_ticker_signals(final_df)

print("Aggregated Results:")
print(aggregated_df)

```

```

Aggregated Results:
   ticker      date final_decision
0  AAPL  2025-05-05             BUY
1  NFLX  2025-05-05             HOLD
2  NVDA  2025-05-05             BUY
3   SPY  2025-05-05             BUY
4  TSLA  2025-05-05             HOLD

```

14 Plotting Stock Price Chart with Aggregated BUY/SELL/HOLD Signals

```

[16]: # checking next available trading day if it is a weekend or not
def get_next_trading_day(date):
    date = pd.to_datetime(date)
    if date.weekday() == 5:
        date = date + pd.Timedelta(days=2)
    elif date.weekday() == 6:
        date = date + pd.Timedelta(days=1)
    return date

def plot_signals(aggregated_df, ticker):
    aggregated_df["date"] = pd.to_datetime(aggregated_df["date"])
    ticker_decisions = aggregated_df[aggregated_df["ticker"] == ticker].copy()
    if ticker_decisions.empty:
        print(f"No aggregated decisions found for {ticker}")
        return

    ticker_decisions["action_date"] = ticker_decisions["date"] + pd.
    ↪Timedelta(days=1)

    ticker_decisions["adjusted_action_date"] = ticker_decisions["action_date"].
    ↪apply(get_next_trading_day)

    first_news_date = ticker_decisions["date"].min()

    # plot range: 10 days before trading date to 10 days after
    plot_start = (first_news_date - pd.Timedelta(days=10)).strftime('%Y-%m-%d')

```

```

plot_end = (first_news_date + pd.Timedelta(days=10)).strftime('%Y-%m-%d')

price_data = yf.download(ticker, start=plot_start, end=plot_end,
↪progress=False)
if price_data.empty:
    print(f"No price data available for {ticker}.")
    return

price_data.index = pd.to_datetime(price_data.index)

ticker_decisions =
↪ticker_decisions[ticker_decisions["adjusted_action_date"].isin(price_data.
↪index)]

buy_dates = ticker_decisions[ticker_decisions["final_decision"] ==
↪"BUY"]["adjusted_action_date"]
sell_dates = ticker_decisions[ticker_decisions["final_decision"] ==
↪"SELL"]["adjusted_action_date"]
hold_dates = ticker_decisions[ticker_decisions["final_decision"] ==
↪"HOLD"]["adjusted_action_date"]

plt.figure(figsize=(12, 6))
plt.plot(price_data.index, price_data["Close"], label="Close Price",
↪linewidth=2)

# BUY signal - green
plt.scatter(buy_dates, price_data.loc[buy_dates]["Close"], color="green",
↪marker="^", s=100, label="BUY")

# SELL signal - red
plt.scatter(sell_dates, price_data.loc[sell_dates]["Close"], color="red",
↪marker="v", s=100, label="SELL")

# HOLD - yellow dot
plt.scatter(hold_dates, price_data.loc[hold_dates]["Close"], color="gold",
↪marker="o", s=70, label="HOLD")

plt.title(f"{ticker} Price Chart with Aggregated Trade Signals\n")
plt.xlabel("Date")
plt.ylabel("Price")
plt.xticks(rotation=45)
plt.grid(True)
plt.legend()
plt.tight_layout()
plt.show()

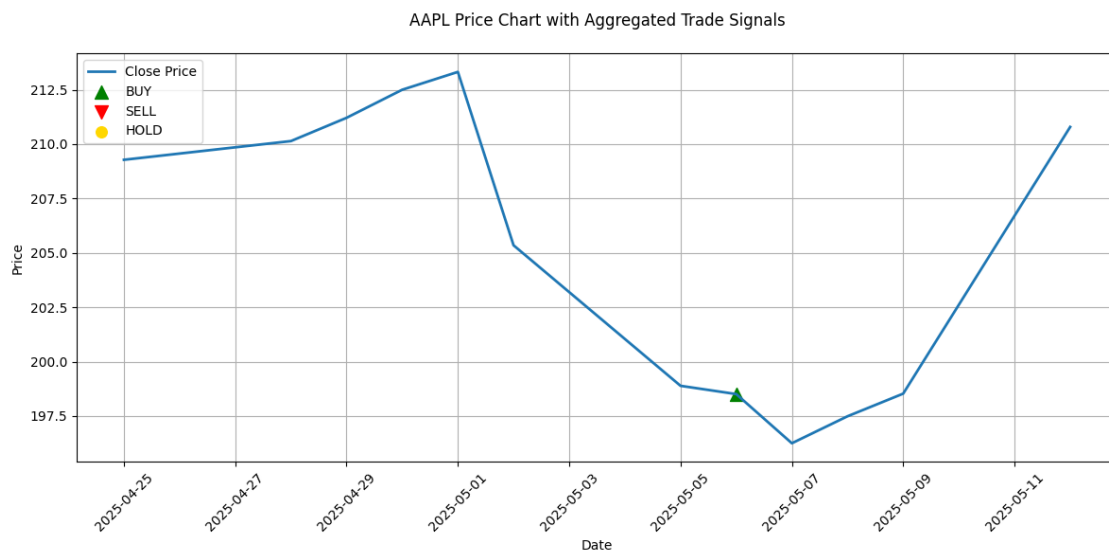
```

- Graph of each Ticker in Aggregated Dataframe and mapping decision to next day help investor

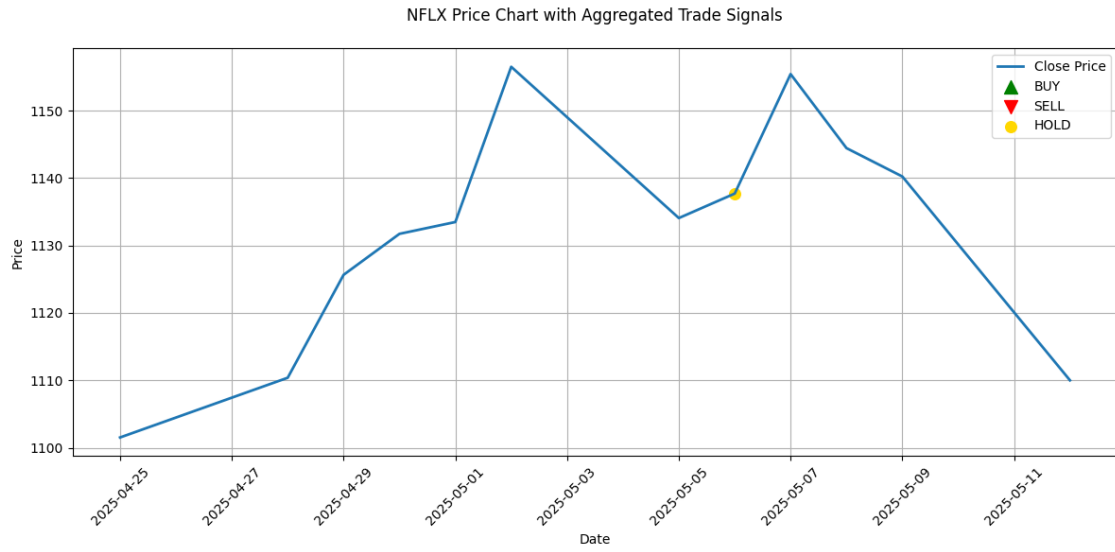
to visualize direction of market on trading day

```
[17]: def plot_all_tickers(aggregated_df):  
    tickers = aggregated_df["ticker"].unique()  
  
    for ticker in tickers:  
        print(f"Plotting signal for {ticker}...")  
        plot_signals(aggregated_df, ticker)  
  
plot_all_tickers(aggregated_df)
```

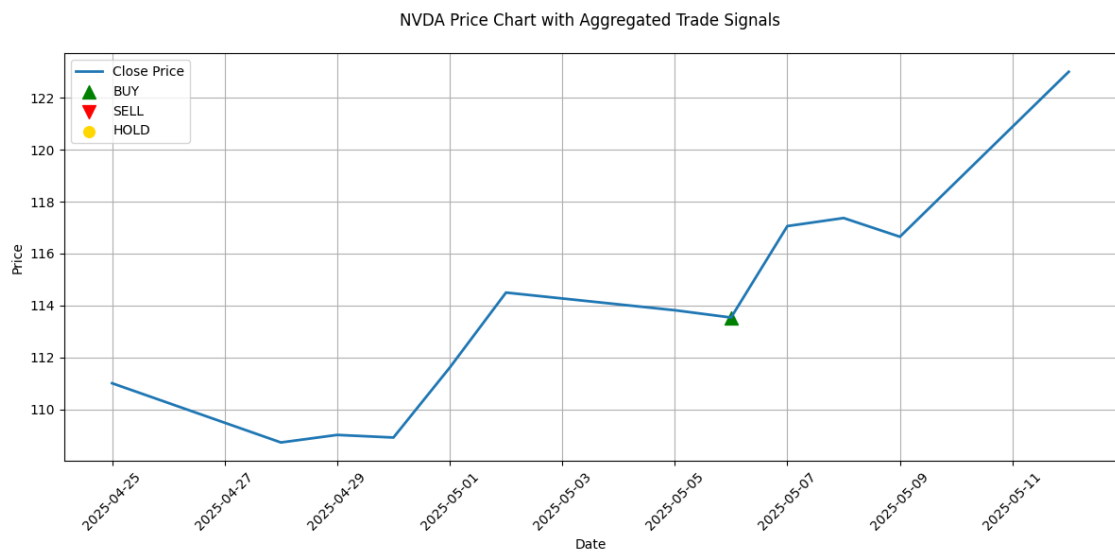
Plotting signal for AAPL...



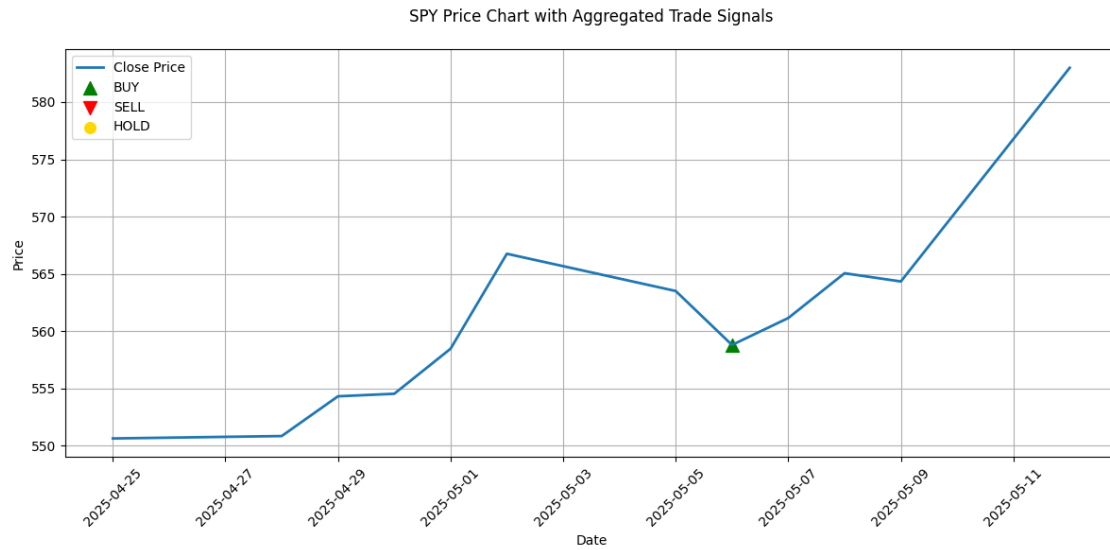
Plotting signal for NFLX...



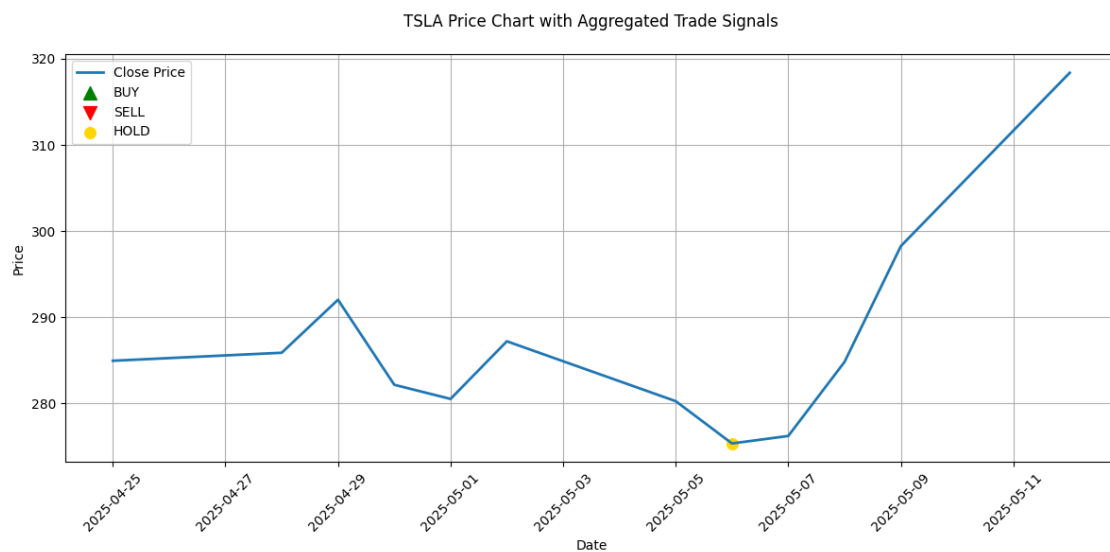
Plotting signal for NVDA...



Plotting signal for SPY...



Plotting signal for TSLA...



15 Backtesting

```
[18]: import datetime
import yfinance as yf
import pandas as pd

# Virtual Portfolio
```

```

initial_capital = 100000.0
portfolio_value = initial_capital
cash_balance = initial_capital
stock_holdings = 0
buy_date = None
buy_price = None
buy_ticker = None

trade_log = []

def is_weekend(date):
    return date.weekday() >= 5

def get_next_valid_date(date):
    while is_weekend(date):
        date += datetime.timedelta(days=1)
    return date.strftime("%Y-%m-%d")

def get_stock_data(ticker, target_date, search_range=5):
    target = pd.to_datetime(target_date)

    for offset in range(-search_range, search_range + 1):
        check_date = target + pd.Timedelta(days=offset)
        if is_weekend(check_date):
            continue

        start_date = check_date.strftime("%Y-%m-%d")
        end_date = (check_date + pd.Timedelta(days=1)).strftime("%Y-%m-%d")

        try:
            price_data = yf.download(ticker, start=start_date, end=end_date,
↳progress=False, interval="1d")
            if not price_data.empty:
                open_price = price_data['Open'].iloc[0]
                close_price = price_data['Close'].iloc[0]
                return open_price, close_price
        except Exception as e:
            print(f"Error fetching data for {ticker} on {check_date.date()}:
↳{e}")

        print(f"No price data found for {ticker} within ±{search_range} days of
↳{target_date}")
        return None, None

# backtesting loop

```

```

for i, row in aggregated_df.iterrows():
    trade_date = row['date']
    ticker = row['ticker']
    action = row['final_decision']

    if isinstance(trade_date, pd.Timestamp):
        trade_date = trade_date.strftime("%Y-%m-%d")

    next_date = datetime.datetime.strptime(trade_date, "%Y-%m-%d") + datetime.
    ⇨timedelta(days=1)
    next_valid_date = get_next_valid_date(next_date)

    open_price, close_price = get_stock_data(ticker, next_valid_date,
    ⇨search_range=5)

    if open_price is None or close_price is None:
        continue

    daily_profit = 0.0
    if action == "HOLD":
        pass
    elif action == "BUY" and stock_holdings == 0:
        buy_date = next_valid_date
        buy_price = open_price
        buy_ticker = ticker
        shares = cash_balance / buy_price
        if isinstance(shares, pd.Series):
            shares = shares.iloc[0]
        shares = float(shares)

        stock_holdings = shares

        stock_holdings = shares
        cash_balance -= shares * buy_price
    elif action == "SELL" and float(stock_holdings) > 0 and buy_date is not
    ⇨None:
        hold_days = (datetime.datetime.strptime(next_valid_date, "%Y-%m-%d") -
                     datetime.datetime.strptime(buy_date, "%Y-%m-%d")).days
        if hold_days >= 2:
            value_from_sale = stock_holdings * close_price
            cash_balance += value_from_sale
            daily_profit = value_from_sale - (stock_holdings * buy_price)
            portfolio_value = cash_balance

            trade_log.append({
                "date": next_valid_date,
                "ticker": ticker,

```

```

        "action": "SELL",
        "open": round(open_price, 2),
        "close": round(close_price, 2),
        "daily_profit": round(daily_profit, 2),
        "portfolio_value": round(portfolio_value, 2),
        "cash_balance": round(cash_balance, 2),
        "stock_holdings": 0
    })

    stock_holdings = 0
    buy_date = buy_price = buy_ticker = None

if isinstance(stock_holdings, pd.Series):
    stock_holdings = stock_holdings.iloc[0]
stock_holdings = float(stock_holdings)

if stock_holdings > 0 and buy_date is not None:
    if isinstance(cash_balance, pd.Series):
        cash_balance = cash_balance.iloc[0]
    if isinstance(close_price, pd.Series):
        close_price = close_price.iloc[0]
    if isinstance(stock_holdings, pd.Series):
        stock_holdings = stock_holdings.iloc[0]

portfolio_value = float(cash_balance) + float(stock_holdings) * \
    float(close_price)

trade_log.append({
    "date": next_valid_date,
    "ticker": buy_ticker,
    "action": "HOLD",
    "open": round(open_price, 2),
    "close": round(close_price, 2),
    "daily_profit": round(daily_profit, 2),
    "portfolio_value": round(portfolio_value, 2),
    "cash_balance": round(cash_balance, 2),
    "stock_holdings": round(stock_holdings, 2)
})

trade_log_df = pd.DataFrame(trade_log)
trade_log_df.head()

```

```

[18]:      date ticker action \
0  2025-05-06  AAPL  HOLD

```

```

open  close  daily_profit \

```



```

0 Ticker
TSLA      280.01
Name: 2025-05-01 00:00:00... 280.52      0.0

    portfolio_value  cash_balance  stock_holdings
0          134168.73           0.0          478.29

```

16 Summary

```

[19]: # profit/loss
profit = portfolio_value - initial_capital
profit_percent = (profit / initial_capital) * 100

# summary
print("Backtest Summary")
print(f" Starting Capital:${initial_capital:,.2f}")
print(f" Final Portfolio Value:${portfolio_value:,.2f}")
print(f" Profit/Loss:${profit:,.2f} ({profit_percent:+.2f}%)")
print(f" Final Cash Balance:${cash_balance:,.2f}")
print(f" Final Stock Holdings:{stock_holdings:.2f} shares")
print(f" Total Trades Executed:{len(trade_log_df)}")
print(f" Last Trade Date:{trade_log_df['date'].iloc[-1] if not trade_log_df.
    ↪empty else 'N/A'}")
print(f" Last Action Taken:{trade_log_df['action'].iloc[-1] if not trade_log_df.
    ↪empty else 'N/A'}")

```

```

Backtest Summary
Starting Capital:$100,000.00
Final Portfolio Value:$134,168.73
Profit/Loss:$34,168.73 (+34.17%)
Final Cash Balance:$0.00
Final Stock Holdings:478.29 shares
Total Trades Executed:1
Last Trade Date:2025-05-06
Last Action Taken:HOLD

```