

SimpleMovingStrategy (1)

August 5, 2024

0.1 Simple Moving Average Strategy Backtesting

Objective The objective of this project is to implement and backtest a basic trading strategy based on simple moving averages using historical price data

0.1.1 Background

Moving averages of a stock is the average price of a security over a duration and is used to help smooth out price data by creating a constantly updated average price. ##### Moving average Crossover For this project, I will be using the Moving average Crossover strategy. This strategy is typically used by traders to identify changes in trend direction in financial markets and involves the use of two moving averages with different periods, where one is typically a shorter-term moving average and the other is a longer-term moving average. “golden cross”: buy signal. This occurs when the shorter-term price average is moving upwards “faster” than the long-term average. “death cross”: sell signal. short-term price is falling quicker than the average long-term price. ##### Summary Overall, the Moving Average Crossover strategy aims to capture trends in asset prices by identifying key points of trend reversal. This is a simple strategy that can be used when complemented with proper risk management and additional analysis techniques.

```
[5]: import numpy as np
import pandas as pd
```

```
[6]: df = pd.read_csv('MSFT (1).csv')
df.head(5)
```

```
[6]:      Date      Open      High      Low      Close  Adj Close  \
0  2019-05-16  126.750000  129.380005  126.459999  128.929993  122.959991
1  2019-05-17  128.309998  130.460007  127.919998  128.070007  122.139816
2  2019-05-20  126.519997  127.589996  125.760002  126.220001  120.375465
3  2019-05-21  127.430000  127.529999  126.580002  126.900002  121.023979
4  2019-05-22  126.620003  128.240005  126.519997  127.669998  121.758331

      Volume
0  30112200
1  25770500
2  23706900
3  15293300
4  15396500
```

```
[7]: #calculating short term 50 day moving avergae for shorter terms and 200 day for
      ↪longer term
short_window = 20
df['Short_MA'] = df['Close'].rolling(window = short_window, min_periods = 1).
      ↪mean()

long_window = 50
df['Long_MA'] = df['Close'].rolling(window = long_window, min_periods = 1).
      ↪mean()
df.tail()
```

```
[7]:
```

	Date	Open	High	Low	Close	Adj Close	\
1254	2024-05-09	410.570007	412.720001	409.100006	412.320007	411.577637	
1255	2024-05-10	412.940002	415.380005	411.799988	414.739990	413.993256	
1256	2024-05-13	418.010010	418.350006	410.820007	413.720001	412.975098	
1257	2024-05-14	412.019989	417.489990	411.549988	416.559998	415.809998	
1258	2024-05-15	417.899994	423.809998	417.269989	423.079987	423.079987	

	Volume	Short_MA	Long_MA
1254	14689700	406.253000	413.728201
1255	13402300	405.895000	413.750200
1256	15440200	405.898999	413.714600
1257	15109300	405.998000	413.747400
1258	22217500	406.559999	414.156000

Now we will generate trade signals. the golden cross will be generated when the short-term SMA crosses above the long-term SMA, and the death cross will be generated when a short-term SMA crosses below the long-term SMA. Note that a golden cross occurs when (1) in the previous period when Short_Ma was below Long_Ma (2) in the current period, Short_Ma is > Long_Ma

```
[8]: #generating a solumn with zeros
df['Signal'] = 0

#golden cross
df.loc[(df['Short_MA'] > df['Long_MA']) & (df['Short_MA'].shift(1) <=
      ↪df['Long_MA'].shift(1)), 'Signal'] = 1

#death cross
df.loc[(df['Short_MA'] < df['Long_MA']) & (df['Short_MA'].shift(1) >=
      ↪df['Long_MA'].shift(1)), 'Signal'] = -1
df
```

```
[8]:
```

	Date	Open	High	Low	Close	Adj Close	\
0	2019-05-16	126.750000	129.380005	126.459999	128.929993	122.959991	
1	2019-05-17	128.309998	130.460007	127.919998	128.070007	122.139816	
2	2019-05-20	126.519997	127.589996	125.760002	126.220001	120.375465	
3	2019-05-21	127.430000	127.529999	126.580002	126.900002	121.023979	

4	2019-05-22	126.620003	128.240005	126.519997	127.669998	121.758331
...
1254	2024-05-09	410.570007	412.720001	409.100006	412.320007	411.577637
1255	2024-05-10	412.940002	415.380005	411.799988	414.739990	413.993256
1256	2024-05-13	418.010010	418.350006	410.820007	413.720001	412.975098
1257	2024-05-14	412.019989	417.489990	411.549988	416.559998	415.809998
1258	2024-05-15	417.899994	423.809998	417.269989	423.079987	423.079987

	Volume	Short_MA	Long_MA	Signal
0	30112200	128.929993	128.929993	0
1	25770500	128.500000	128.500000	0
2	23706900	127.740000	127.740000	0
3	15293300	127.530001	127.530001	0
4	15396500	127.558000	127.558000	0
...
1254	14689700	406.253000	413.728201	0
1255	13402300	405.895000	413.750200	0
1256	15440200	405.898999	413.714600	0
1257	15109300	405.998000	413.747400	0
1258	22217500	406.559999	414.156000	0

[1259 rows x 10 columns]

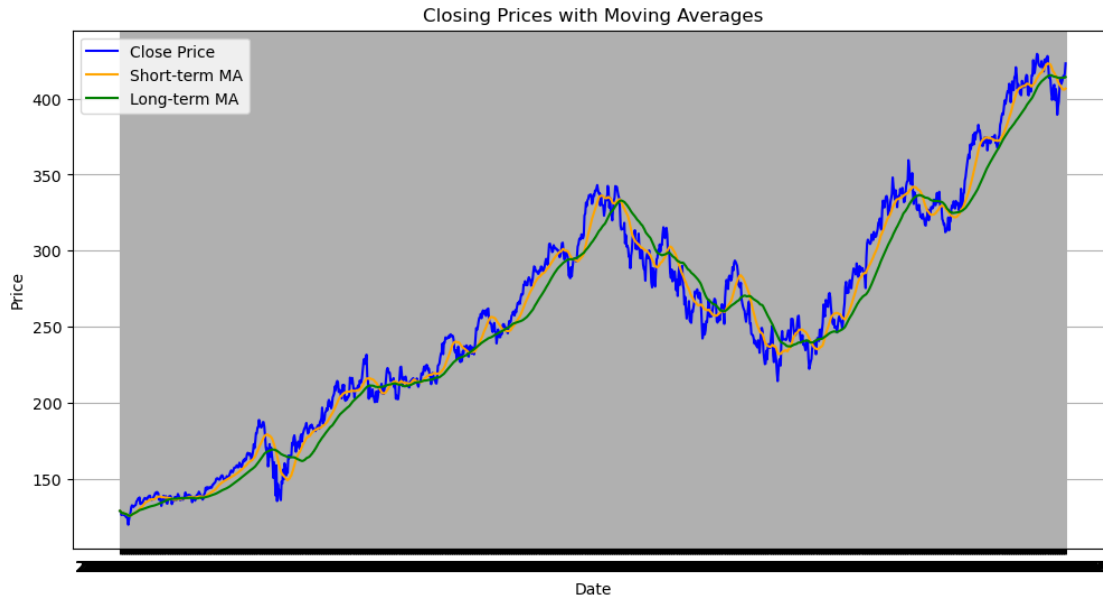
```
[9]: import matplotlib.pyplot as plt
```

```
[10]: plt.figure(figsize=(12, 6))
plt.plot(df['Date'], df['Close'], label='Close Price', color='blue')

# Plotting short-term moving average
plt.plot(df['Date'], df['Short_MA'], label='Short-term MA', color='orange')

# Plotting long-term moving average
plt.plot(df['Date'], df['Long_MA'], label='Long-term MA', color='green')

# Adding labels and title
plt.xlabel('Date')
plt.ylabel('Price')
plt.title('Closing Prices with Moving Averages')
plt.legend()
plt.grid(True)
plt.show()
```



We will now backtest and stimulate the strategy.

```
[11]: #intializing trading variables
portfolio_value = 10000
position = 0 # 0 = no position , 1 = long, -1 = short
entry_price = 0 #storing entry price for the current positon
trade_log = []
for index, row in df.iterrows():
    #cross signal
    if row['Signal'] == 1 and position == 0:
        position = 1 #enter long position
        entry_price = row['Close']
        print(f" buy at {row['Date']} for ${entry_price}")
    #death signal
    elif row['Signal'] == -1 and position == 1:
        position = 0 #exit long position
        exit_price = row['Close']
        trade_profit = (exit_price - entry_price) #calculating trade profit/
        ↪ loss
        portfolio_value += trade_profit
        trade_log.append({'entry_date': row['Date'], 'entry_price': entry_price,
                        'exit_date': row['Date'], 'exit_price': exit_price,
                        'position': 1})
        print(f" sell at {row['Date']} for ${exit_price}. Profit,
        ↪ ${trade_profit}")

final_portfolio_val = portfolio_value
```

```
print(f" Final Portfolio Value: ${final_portfolio_val}")
```

```
buy at 2019-06-18 for $135.160004
sell at 2019-08-23 for $133.389999. Profit $-1.7700049999999976
buy at 2019-09-23 for $139.139999
sell at 2020-03-13 for $158.830002. Profit $19.690003000000002
buy at 2020-04-22 for $173.520004
sell at 2020-09-28 for $209.440002. Profit $35.919997999999999
buy at 2020-10-21 for $214.800003
sell at 2020-12-14 for $214.199997. Profit $-0.60000600000000076
buy at 2020-12-18 for $218.589996
sell at 2021-03-23 for $237.580002. Profit $18.990005999999994
buy at 2021-04-07 for $249.899994
sell at 2021-05-26 for $251.490005. Profit $1.5900110000000004
buy at 2021-06-18 for $259.429993
sell at 2021-10-06 for $293.109985. Profit $33.679991999999997
buy at 2021-10-27 for $323.170013
sell at 2022-01-06 for $313.880005. Profit $-9.290008
buy at 2022-04-04 for $314.970001
sell at 2022-04-26 for $270.220001. Profit $-44.75
buy at 2022-07-21 for $264.839996
sell at 2022-07-25 for $258.829987. Profit $-6.0100089999999968
buy at 2022-07-28 for $276.410004
sell at 2022-09-13 for $251.990005. Profit $-24.419999000000002
buy at 2022-11-30 for $255.139999
sell at 2023-01-10 for $228.850006. Profit $-26.289992999999998
buy at 2023-02-07 for $267.559998
sell at 2023-08-14 for $324.040009. Profit $56.480010999999999
buy at 2023-10-25 for $340.670013
sell at 2024-04-29 for $402.25. Profit $61.579987000000002
Final Portfolio Value: $10114.799987999997
```

We will now evaluate performance using metrics such as total return and average return per trade.

```
[12]: #total return
total_return = ((final_portfolio_val- 10000)/10000)* 100
total_return
```

```
[12]: 1.14799987999999704
```

```
[13]: #average return per trade
num_trade = len(trade_log)
avg_return_per_trade = (total_return /num_trade)*100
avg_return_per_trade
```

```
[13]: 8.199999142856932
```

Overall, both metrics indicate that the trading strategy produced positive results during the back-test period. However, it's essential to consider other performance metrics and conduct further analysis to assess the strategy's robustness, risk-adjusted return, and potential for future performance.