## Collecting ta Downloading ta-0.11.0.tar.gz (25 kB) Preparing metadata (setup.py): started Preparing metadata (setup.py): finished with status 'done' Requirement already satisfied: numpy in c:\users\aditya kudva\anaconda3\lib\site-packages (from ta) (1.24.3)Requirement already satisfied: pandas in c:\users\aditya kudva\anaconda3\lib\site-packages (from ta) (1.5.3)Requirement already satisfied: python-dateutil>=2.8.1 in c:\users\aditya kudva\anaconda3\lib\site-pack ages (from pandas->ta) (2.8.2) Requirement already satisfied: pytz>=2020.1 in c:\users\aditya kudva\anaconda3\lib\site-packages (from pandas->ta) (2022.7) Requirement already satisfied: six>=1.5 in c:\users\aditya kudva\anaconda3\lib\site-packages (from pyt hon-dateutil>=2.8.1->pandas->ta) (1.16.0) Building wheels for collected packages: ta Building wheel for ta (setup.py): started Building wheel for ta (setup.py): finished with status 'done' Created wheel for ta: filename=ta-0.11.0-py3-none-any.whl size=29422 sha256=0a3d6b67f776f2ea472a7ff4 6b666b280eaa1914cfa6b6c4b7cdad55531d19e8 Stored in directory: c:\users\aditya kudva\appdata\local\pip\cache\wheels\a1\d7\29\7781cc5eb9a3659d0 32d7d15bdd0f49d07d2b24fec29f44bc4 Successfully built ta Installing collected packages: ta Successfully installed ta-0.11.0 In [9]: import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns from sklearn.model selection import train test split from sklearn.ensemble import RandomForestClassifier from sklearn.metrics import accuracy\_score, classification\_report, confusion\_matrix import ta import warnings warnings.filterwarnings('ignore') In [5]: | df = pd.read\_csv('AAPL.csv', parse\_dates=['Date'], index\_col='Date') df Out[5]: Open Price High Price Low Price Close Price Adj Close Price Volume Date 2014-05-27 87.982857 89 408569 87 947144 89 375717 80.948952 87216500 2014-05-28 89.111427 78870400 89.431427 89.975716 89.144287 80.739334 2014-05-29 89 692856 90 981430 89 681427 90 768570 82 210480 94118500 2014-05-30 91.139999 92.024284 89.842857 90.428574 81.902557 141005200 **2014-06-02** 90.565712 90.690002 88.928574 89.807144 81.339699 92337700 **2020-05-18** 313.170013 316.500000 310.320007 314.959991 314.959991 33843100 **2020-05-19** 315.029999 318.519989 313.010010 313.140015 313.140015 25432400 **2020-05-20** 316.679993 319.519989 316.519989 319.230011 319.230011 27876200

316.850006

318.890015

25672200

20430600

1510 rows × 6 columns

**2020-05-21** 318.660004 320.890015 315.869995 316.850006

**2020-05-22** 315.769989 319.230011 315.350006 318.890015

In [2]: !pip install ta

```
In [10]: print(df.head())
                     Open Price High Price Low Price Close Price Adj Close Price \
         Date
         2014-05-27
                     87.982857
                                  89.408569 87.947144
                                                          89.375717
                                                                           80.948952
                                                                           80.739334
         2014-05-28
                                  89.975716 89.111427
                      89,431427
                                                          89.144287
         2014-05-29
                      89.692856
                                  90.981430 89.681427
                                                          90.768570
                                                                           82.210480
         2014-05-30
                      91.139999
                                  92.024284 89.842857
                                                          90.428574
                                                                           81,902557
         2014-06-02
                      90.565712
                                  90.690002 88.928574
                                                          89.807144
                                                                           81.339699
                        Volume
         Date
         2014-05-27
                     87216500
         2014-05-28
                    78870400
         2014-05-29
                     94118500
         2014-05-30 141005200
         2014-06-02 92337700
In [11]: print(df.columns)
         Index(['Open Price', 'High Price', 'Low Price', 'Close Price',
                'Adj Close Price', 'Volume'],
               dtype='object')
In [6]: df.isnull().sum()
Out[6]: Open Price
         High Price
                            0
         Low Price
                            0
         Close Price
                            0
         Adj Close Price
                            0
         Volume
                            0
         dtype: int64
In [7]: | df.fillna(method='ffill', inplace=True)
In [15]: df['SMA_50'] = ta.trend.sma_indicator(df['Close Price'], window=50)
         df['SMA_200'] = ta.trend.sma_indicator(df['Close Price'], window=200)
         df['RSI'] = ta.momentum.rsi(df['Close Price'], window=14)
         df['MACD'] = ta.trend.macd_diff(df['Close Price'])
In [16]: | df['Future Close'] = df['Close Price'].shift(-1)
         df['Target'] = (df['Future Close'] > df['Close Price']).astype(int)
In [17]: | df.dropna(inplace=True)
In [18]: X = df[['SMA_50', 'SMA_200', 'RSI', 'MACD']]
         y = df['Target']
```

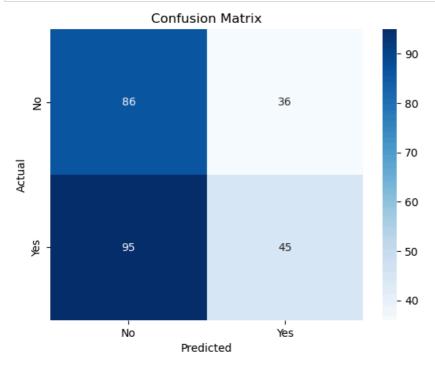
```
In [19]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, shuffle=False)
         X_train, X_test, y_train, y_test
Out[19]: (
                       SMA_50
                                  SMA 200
                                                 RSI
                                                         MACD
          Date
          2015-03-11 118.9348 105.731307 43.638948 -1.284230
          2015-03-12 119.1456
                               105.906679 49.410402 -1.247728
          2015-03-13 119.3670 106.078907 47.377260 -1.237522
          2015-03-16 119.6584 106.249814 50.823398 -1.099838
          2015-03-17 120.0126 106.432871 55.632029 -0.840376
          2019-05-02 190.4058 192.164050
                                           67.636408 -0.201529
          2019-05-03 191.2196 192.265550 70.432564 0.003276
          2019-05-06 191.9298 192.355950 63.053939 -0.123348
          2019-05-07 192.5024 192.410850 52.813487 -0.600758
          2019-05-08 193.0738 192.468150 52.872143 -0.910430
          [1048 rows x 4 columns],
                          SMA_50
                                    SMA_200
                                                   RSI
                                                           MACD
          Date
          2019-05-09 193.590800 192.513700 49.276357 -1.238334
          2019-05-10 194.071400 192.534600 44.038720 -1.644446
          2019-05-13 194.286400 192.489100 32.131856 -2.579975
          2019-05-14 194.542600 192.461350 36.849120 -2.873436
          2019-05-15 194.850400 192.461050 40.285105 -2.785926
          2020-05-15 273.114000 263.209399 61.940977 1.025197
          2020-05-18 273.632599
                                 263.742049 65.697653 0.996037
          2020-05-19 274.572000
                                 264.287649 63.990111 0.722853
          2020-05-20 275.249800 264.917100 67.073985
                                                        0.814189
          2020-05-21 276.078200 265.516350 64.740544 0.586820
          [262 rows x 4 columns],
          Date
          2015-03-11
          2015-03-12
                       0
          2015-03-13
          2015-03-16
                       1
          2015-03-17
                       1
          2019-05-02
                       1
          2019-05-03
          2019-05-06
                       0
          2019-05-07
                       1
          2019-05-08
                       а
          Name: Target, Length: 1048, dtype: int32,
          Date
          2019-05-09
                       0
          2019-05-10
                       0
          2019-05-13
                       1
          2019-05-14
          2019-05-15
                       0
          2020-05-15
                       1
          2020-05-18
                       0
          2020-05-19
                       1
          2020-05-20
                       0
          2020-05-21
                        1
          Name: Target, Length: 262, dtype: int32)
In [20]: model = RandomForestClassifier(n_estimators=100, random_state=42)
         model.fit(X_train, y_train)
Out[20]:
                  RandomForestClassifier
```

RandomForestClassifier(random\_state=42)

```
In [21]: y_pred = model.predict(X_test)
print(f'Accuracy: {accuracy_score(y_test, y_pred):.2f}')
print('Classification Report:')
print(classification_report(y_test, y_pred))
```

```
Accuracy: 0.50
Classification Report:
              precision
                           recall f1-score
                                               support
           0
                   0.48
                              0.70
                                        0.57
                                                   122
           1
                   0.56
                              0.32
                                        0.41
                                                   140
    accuracy
                                        0.50
                                                   262
                   0.52
                              0.51
                                        0.49
                                                   262
   macro avg
weighted avg
                   0.52
                              0.50
                                        0.48
                                                    262
```

```
In [22]: cm = confusion_matrix(y_test, y_pred)
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['No', 'Yes'], yticklabels=['No', 'Yes']
    plt.xlabel('Predicted')
    plt.ylabel('Actual')
    plt.title('Confusion Matrix')
    plt.show()
```



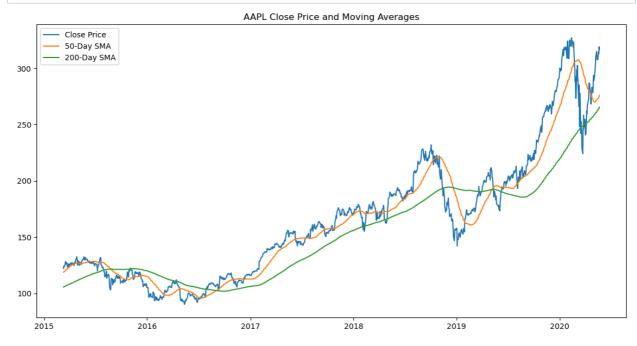
```
In [23]:
    test_df = df.iloc[len(X_train):].copy()
    test_df['Predicted Signal'] = y_pred
    test_df['Strategy Returns'] = test_df['Close Price'].pct_change() * test_df['Predicted Signal'].shift(1)
```

```
In [24]: test_df['Cumulative Market Returns'] = (1 + test_df['Close Price'].pct_change()).cumprod()
test_df['Cumulative Strategy Returns'] = (1 + test_df['Strategy Returns']).cumprod()
```

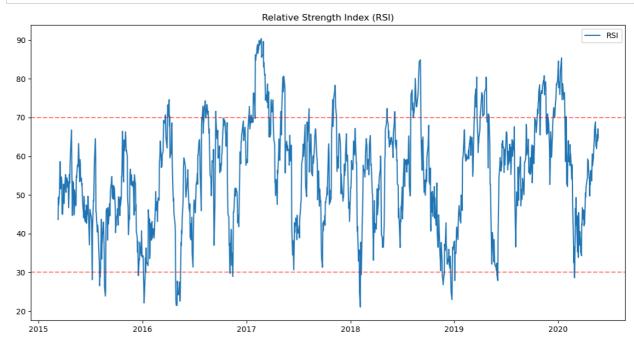
```
In [25]: plt.figure(figsize=(14, 7))
    plt.plot(test_df['Cumulative Market Returns'], label='Market Returns')
    plt.plot(test_df['Cumulative Strategy Returns'], label='Strategy Returns')
    plt.legend()
    plt.title('Market vs Strategy Returns')
    plt.show()
```



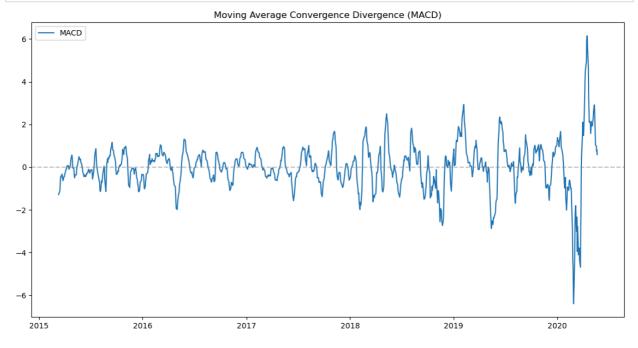
```
In [26]: plt.figure(figsize=(14, 7))
    plt.plot(df['Close Price'], label='Close Price')
    plt.plot(df['SMA_50'], label='50-Day SMA')
    plt.plot(df['SMA_200'], label='200-Day SMA')
    plt.legend()
    plt.title('AAPL Close Price and Moving Averages')
    plt.show()
```



```
In [27]: plt.figure(figsize=(14, 7))
    plt.plot(df['RSI'], label='RSI')
    plt.axhline(30, linestyle='--', alpha=0.5, color='red')
    plt.axhline(70, linestyle='--', alpha=0.5, color='red')
    plt.title('Relative Strength Index (RSI)')
    plt.legend()
    plt.show()
```



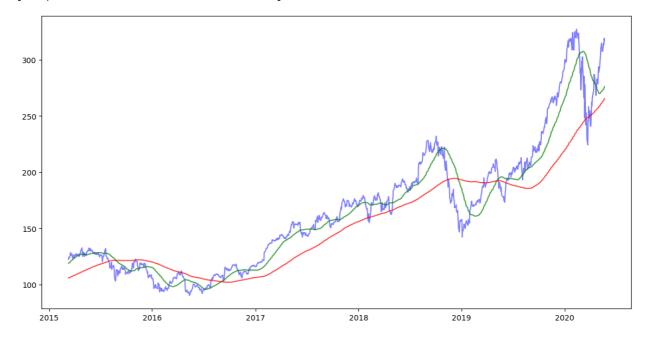
```
In [28]: plt.figure(figsize=(14, 7))
    plt.plot(df['MACD'], label='MACD')
    plt.axhline(0, linestyle='--', alpha=0.5, color='grey')
    plt.title('Moving Average Convergence Divergence (MACD)')
    plt.legend()
    plt.show()
```



```
In [29]: import matplotlib.dates as mdates import matplotlib.ticker as mticker from matplotlib.dates import DateFormatter
```

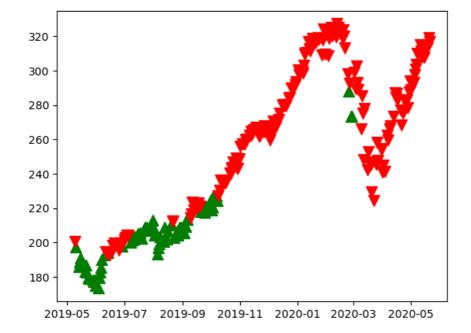
```
In [30]: plt.figure(figsize=(14, 7))
    plt.plot(df.index, df['Close Price'], label='Close Price', color='blue', alpha=0.5)
    plt.plot(df.index, df['SMA_50'], label='50-Day SMA', color='green', alpha=0.7)
    plt.plot(df.index, df['SMA_200'], label='200-Day SMA', color='red', alpha=0.7)
```

Out[30]: [<matplotlib.lines.Line2D at 0x1df39976a50>]



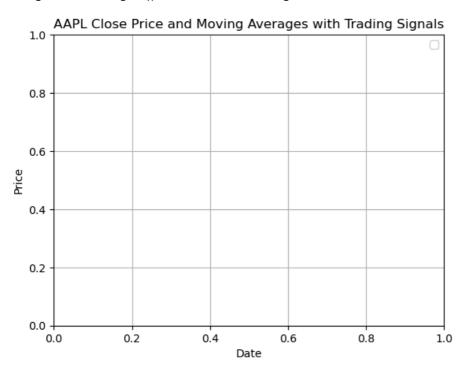
```
In [31]:
buy_signals = test_df[test_df['Predicted Signal'] == 1]
sell_signals = test_df[test_df['Predicted Signal'] == 0]
plt.scatter(buy_signals.index, df.loc[buy_signals.index]['Close Price'], marker='^', color='green', alp
plt.scatter(sell_signals.index, df.loc[sell_signals.index]['Close Price'], marker='v', color='red', alp
```

Out[31]: <matplotlib.collections.PathCollection at 0x1df39a04290>



```
In [32]: plt.legend()
    plt.title('AAPL Close Price and Moving Averages with Trading Signals')
    plt.xlabel('Date')
    plt.ylabel('Price')
    plt.grid(True)
    plt.show()
```

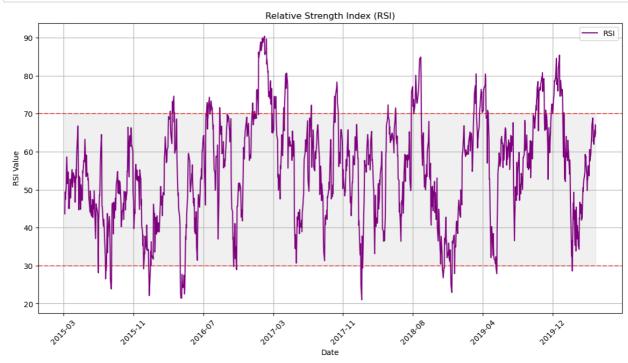
No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.



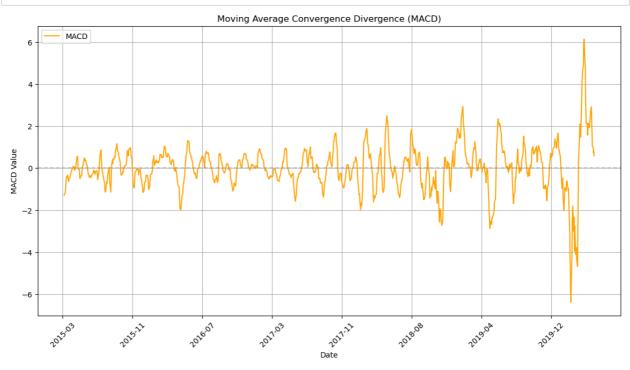
```
In [33]:
    fig, ax = plt.subplots(figsize=(14, 7))
        ax.plot(test_df.index, test_df['Cumulative Market Returns'], label='Market Returns', color='blue')
        ax.plot(test_df.index, test_df['Cumulative Strategy Returns'], label='Strategy Returns', color='green')
        ax.set_title('Cumulative Returns: Market vs Strategy')
        ax.set_xlabel('Date')
        ax.set_ylabel('Cumulative Returns')
        ax.legend()
        ax.grid(True)
        ax.xaxis.set_major_locator(mticker.MaxNLocator(10))
        ax.xaxis.set_major_formatter(DateFormatter('%Y-%m'))
        plt.xticks(rotation=45)
        plt.show()
```



```
In [34]:
    fig, ax = plt.subplots(figsize=(14, 7))
    ax.plot(df.index, df['RSI'], label='RSI', color='purple')
    ax.axhline(30, linestyle='--', alpha=0.5, color='red')
    ax.axhline(70, linestyle='--', alpha=0.5, color='red')
    ax.fill_between(df.index, y1=30, y2=70, alpha=0.1, color='grey')
    ax.set_title('Relative Strength Index (RSI)')
    ax.set_xlabel('Date')
    ax.set_ylabel('RSI Value')
    ax.legend()
    ax.grid(True)
    ax.xaxis.set_major_locator(mticker.MaxNLocator(10))
    ax.xaxis.set_major_formatter(DateFormatter('%Y-%m'))
    plt.xticks(rotation=45)
    plt.show()
```



```
In [35]: fig, ax = plt.subplots(figsize=(14, 7))
    ax.plot(df.index, df['MACD'], label='MACD', color='orange')
    ax.axhline(0, linestyle='--', alpha=0.5, color='grey')
    ax.set_title('Moving Average Convergence Divergence (MACD)')
    ax.set_xlabel('Date')
    ax.set_ylabel('MACD Value')
    ax.legend()
    ax.grid(True)
    ax.xaxis.set_major_locator(mticker.MaxNLocator(10))
    ax.xaxis.set_major_formatter(DateFormatter('%Y-%m'))
    plt.xticks(rotation=45)
    plt.show()
```



## In [37]: !pip install mpf

## Collecting mpf

ERROR: pip's dependency resolver does not currently take into account all the packages that are inst alled. This behaviour is the source of the following dependency conflicts.

tables 3.8.0 requires blosc2~=2.0.0, which is not installed.

tables 3.8.0 requires cython>=0.29.21, which is not installed.

conda 23.7.2 requires ruamel-yaml<0.18,>=0.11.14, but you have ruamel-yaml 0.18.6 which is incompatible.

python-lsp-black 1.2.1 requires black>=22.3.0, but you have black 0.0 which is incompatible.

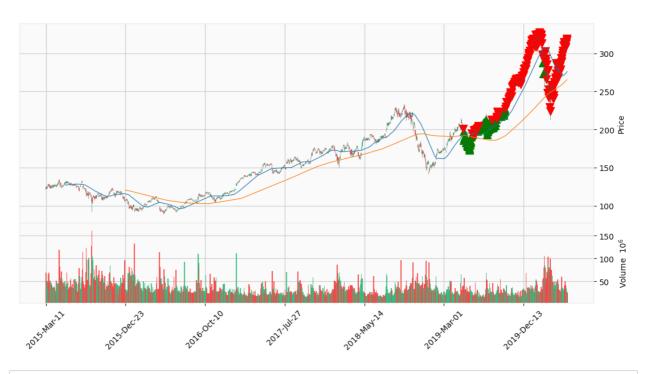
Obtaining dependency information for mpf from https://files.pythonhosted.org/packages/c6/16/a77a69 be0090883f490e74303f966369176f940331a8ac5904e8977f3510/mpf-0.57.1-py3-none-any.whl.metadata (https://files.pythonhosted.org/packages/c6/16/a77a69be0090883f490e74303f966369176f940331a8ac5904e8977f3510/mpf-0.57.1-py3-none-any.whl.metadata)

Downloading mpf-0.57.1-py3-none-any.whl.metadata (5.8 kB) Collecting asciimatics==1.15.0 (from mpf)

Obtaining dependency information for asciimatics==1.15.0 from https://files.pythonhosted.org/packages/35/bf/9cad857b630c840738003eb24c1adb63490a1024ec40a9dcc3a753300c38/asciimatics-1.15.0-py3-none-any.whl.metadata (https://files.pythonhosted.org/packages/35/bf/9cad857b630c840738003eb24c1adb63490a1

```
In [39]: !pip install mplfinance
         Collecting mplfinance
          Obtaining dependency information for mplfinance from https://files.pythonhosted.org/packages/d7/d9/3
         1c436ea7673c21a5bf3fc747bc7f63377582dfe845c3004d3e46f9deee0/mplfinance-0.12.10b0-py3-none-any.whl.meta
         data (https://files.pythonhosted.org/packages/d7/d9/31c436ea7673c21a5bf3fc747bc7f63377582dfe845c3004d3
         e46f9deee0/mplfinance-0.12.10b0-py3-none-any.whl.metadata)
          Downloading mplfinance-0.12.10b0-py3-none-any.whl.metadata (19 kB)
         Requirement already satisfied: matplotlib in c:\users\aditya kudva\anaconda3\lib\site-packages (from m
         plfinance) (3.7.1)
         Requirement already satisfied: pandas in c:\users\aditya kudva\anaconda3\lib\site-packages (from mplfi
         nance) (1.5.3)
         Requirement already satisfied: contourpy>=1.0.1 in c:\users\aditya kudva\anaconda3\lib\site-packages
         (from matplotlib->mplfinance) (1.0.5)
         Requirement already satisfied: cycler>=0.10 in c:\users\aditya kudva\anaconda3\lib\site-packages (from
         matplotlib->mplfinance) (0.11.0)
         Requirement already satisfied: fonttools>=4.22.0 in c:\users\aditya kudva\anaconda3\lib\site-packages
         (from matplotlib->mplfinance) (4.25.0)
         Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\aditya kudva\anaconda3\lib\site-packages
         (from matplotlib->mplfinance) (1.4.4)
         Requirement already satisfied: numpy>=1.20 in c:\users\aditya kudva\anaconda3\lib\site-packages (from
         matplotlib->mplfinance) (1.24.3)
         Requirement already satisfied: packaging>=20.0 in c:\users\aditya kudva\anaconda3\lib\site-packages (f
         rom matplotlib->mplfinance) (23.2)
         Requirement already satisfied: pillow>=6.2.0 in c:\users\aditya kudva\anaconda3\lib\site-packages (fro
         m matplotlib->mplfinance) (9.5.0)
         Requirement already satisfied: pyparsing>=2.3.1 in c:\users\aditya kudva\anaconda3\lib\site-packages
         (from matplotlib->mplfinance) (3.0.9)
         Requirement already satisfied: python-dateutil>=2.7 in c:\users\aditya kudva\anaconda3\lib\site-packag
         es (from matplotlib->mplfinance) (2.8.2)
         Requirement already satisfied: pytz>=2020.1 in c:\users\aditya kudva\anaconda3\lib\site-packages (from
         pandas->mplfinance) (2022.7)
         Requirement already satisfied: six>=1.5 in c:\users\aditya kudva\anaconda3\lib\site-packages (from pyt
         hon-dateutil>=2.7->matplotlib->mplfinance) (1.16.0)
         Downloading mplfinance-0.12.10b0-py3-none-any.whl (75 kB)
           ----- 0.0/75.0 kB ? eta -:--:--
            ----- 71.7/75.0 kB 2.0 MB/s eta 0:00:01
            ----- 75.0/75.0 kB 1.0 MB/s eta 0:00:00
         Installing collected packages: mplfinance
         Successfully installed mplfinance-0.12.10b0
In [71]: import mplfinance as mpf
         import pandas as pd
In [72]: df_candle = df[['Open Price', 'High Price', 'Low Price', 'Close Price', 'Volume']].copy()
         df_candle.columns = ['Open', 'High', 'Low', 'Close', 'Volume']
In [73]: buy_signals = test_df[test_df['Predicted Signal'] == 1].index
         sell_signals = test_df[test_df['Predicted Signal'] == 0].index
In [74]: buy_prices = df.loc[buy_signals, 'Close Price']
         sell_prices = df.loc[sell_signals, 'Close Price']
In [75]: buy_prices = buy_prices.reindex(df_candle.index, fill_value=pd.NA)
         sell_prices = sell_prices.reindex(df_candle.index, fill_value=pd.NA)
In [65]: print(f"df_candle columns: {df_candle.columns}")
         print(f"df_candle index length: {len(df_candle.index)}")
         df_candle columns: Index(['Open', 'High', 'Low', 'Close', 'Volume'], dtype='object')
         df_candle index length: 1310
```

## **AAPL Candlestick Chart with SMA and Trading Signals**



In [ ]: