

# Trend indicator MAs

FINANCIAL TRADING IN PYTHON



**Chelsea Yang**

Data Science Instructor

# What are technical indicators?

- Mathematical calculations based on historical market data
- Assume the market is efficient and the price has incorporated all public information
- Help traders to gain insight into past price patterns

# Types of indicators

- **Trend indicators:** measure the direction or strength of a trend
  - Example: Moving averages (MA), Average Directional Movement Index (ADX)
- **Momentum indicators:** measure the velocity of price movement
  - Example: Relative Strength Index (RSI)
- **Volatility indicators:** measure the magnitude of price deviations
  - Example: Bollinger Bands

# The TA-Lib package

## TA-Lib : Technical Analysis Library

- Includes 150+ technical indicator implementations

```
import talib
```

# Moving average indicators

- SMA: Simple Moving average
- EMA: Exponential Moving Average
- Characteristics:
  - Move with the price
  - Smooth out the data to better indicate the price direction

# Simple Moving Average (SMA)

$$SMA = (P_1 + P_2 + \dots + P_n) / n$$

```
# Calculate two SMAs
stock_data['SMA_short'] = talib.SMA(stock_data['Close'], timeperiod=10)
stock_data['SMA_long'] = talib.SMA(stock_data['Close'], timeperiod=50)
# Print the last five rows
print(stock_data.tail())
```

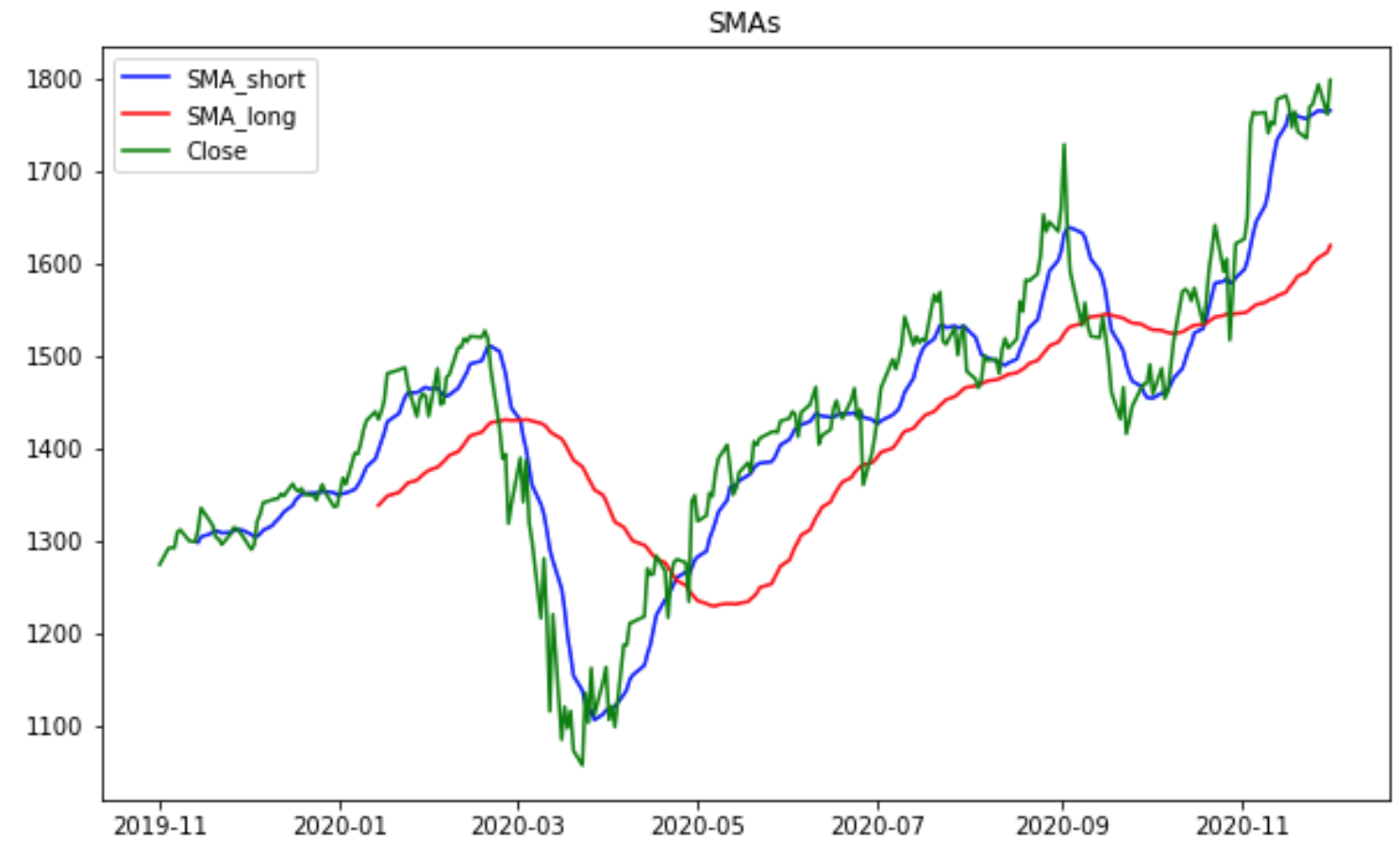
	Close	SMA_short	SMA_long
Date			
2020-11-24	1768.88	1758.77	1594.74
2020-11-25	1771.43	1760.65	1599.75
2020-11-27	1793.19	1764.98	1605.70
2020-11-30	1760.74	1763.35	1611.71
2020-12-01	1798.10	1765.02	1619.05

# Plotting the SMA

```
import matplotlib.pyplot as plt

# Plot SMA with the price
plt.plot(stock_data['SMA_short'],
         label='SMA_short')
plt.plot(stock_data['SMA_long'],
         label='SMA_long')
plt.plot(stock_data['Close'],
         label='Close')

# Customize and show the plot
plt.legend()
plt.title('SMAs')
plt.show()
```



# Exponential Moving Average (EMA)

$$EMA_n = P_n \times multiplier + \text{previous EMA} \times (1 - multiplier)$$

$$multiplier = 2/(n + 1)$$

```
# Calculate two EMAs
stock_data['EMA_short'] = talib.EMA(stock_data['Close'], timeperiod=10)
stock_data['EMA_long'] = talib.EMA(stock_data['Close'], timeperiod=50)
# Print the last five rows
print(stock_data.tail())
```

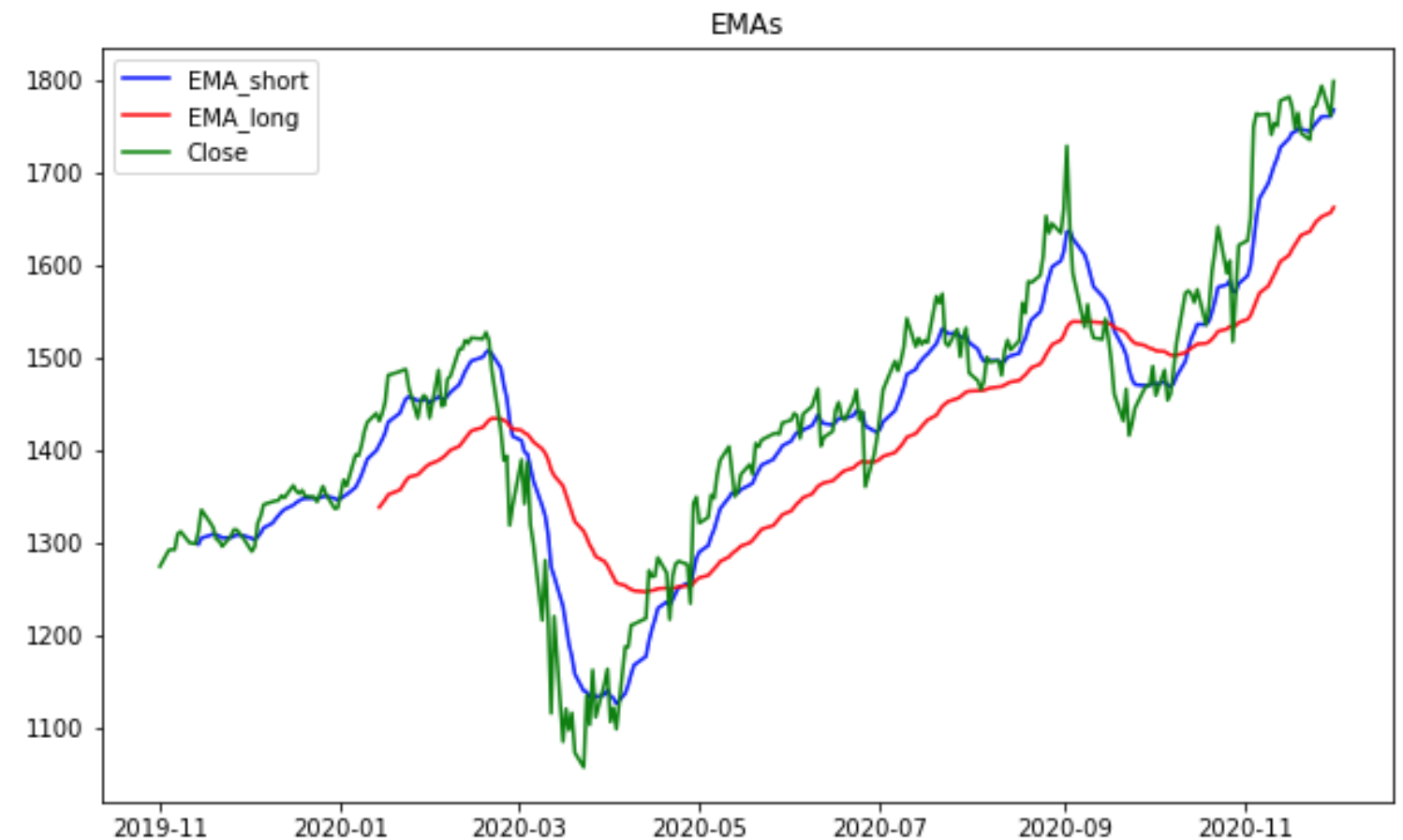
	Close	EMA_short	EMA_long
Date			
2020-11-24	1768.88	1748.65	1640.98
2020-11-25	1771.43	1752.79	1646.09
2020-11-27	1793.19	1760.13	1651.86
2020-11-30	1760.74	1760.24	1656.13
2020-12-01	1798.10	1767.13	1661.70



# Plotting the EMA

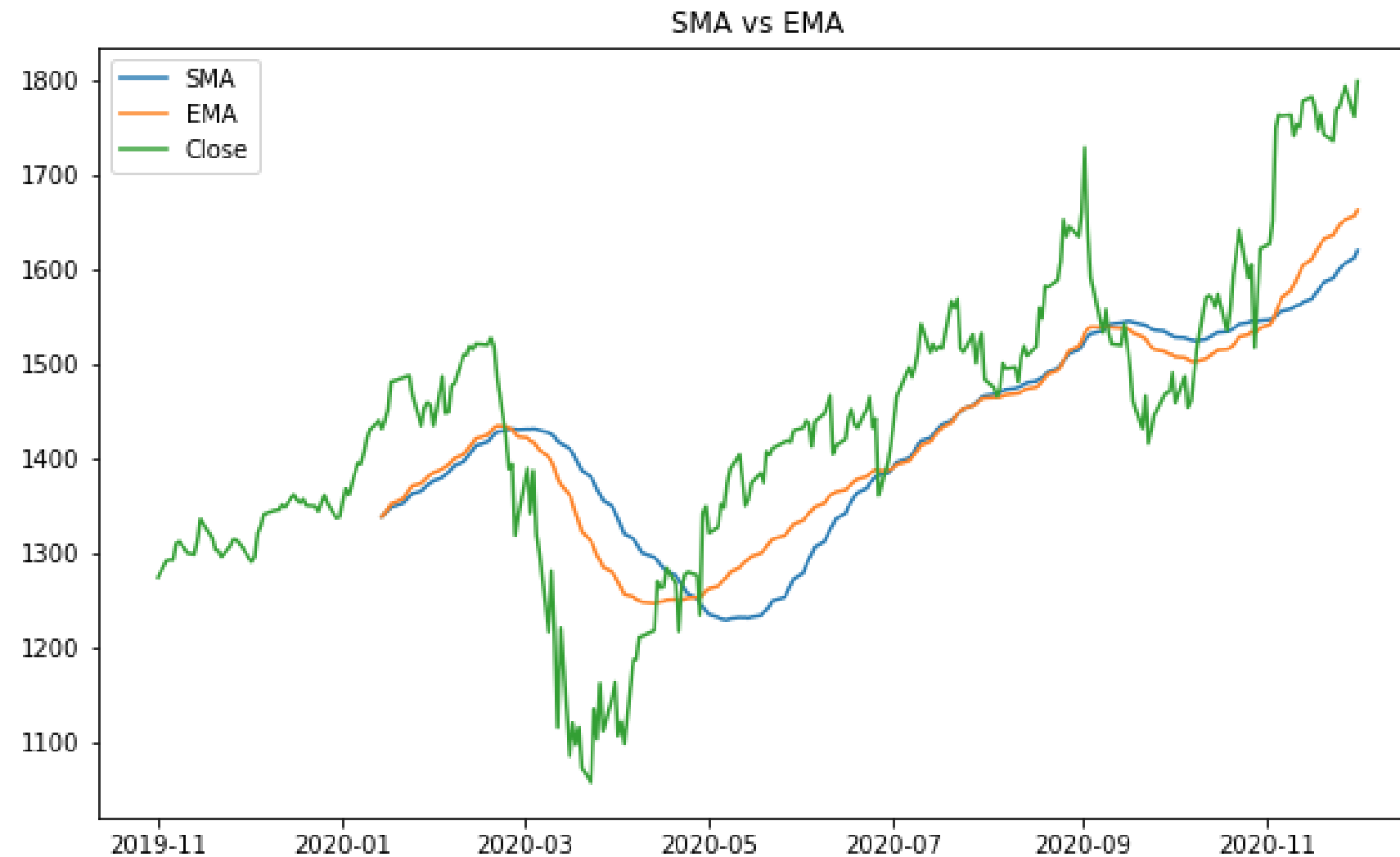
```
import matplotlib.pyplot as plt
# Plot EMA with the price
plt.plot(stock_data['EMA_short'],
         label='EMA_short')
plt.plot(stock_data['EMA_long'],
         label='EMA_long')
plt.plot(stock_data['Close'],
         label='Close')

# Customize and show the plot
plt.legend()
plt.title('EMAs')
plt.show()
```



# SMA vs. EMA

EMA is more sensitive to the most recent price movement



**Let's practice!**  
FINANCIAL TRADING IN PYTHON

# Strength indicator: ADX

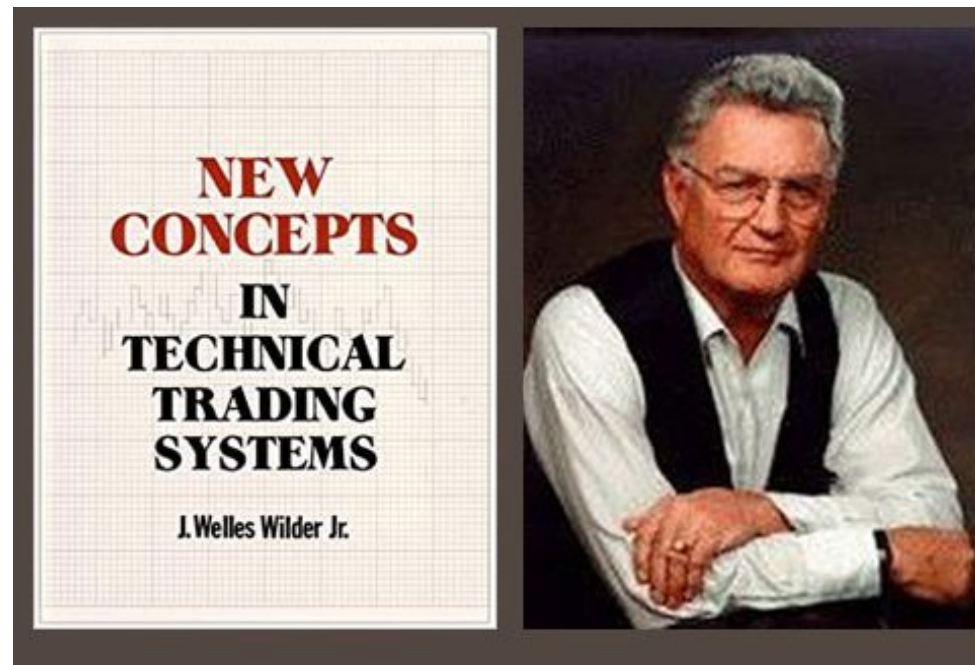
FINANCIAL TRADING IN PYTHON



**Chelsea Yang**  
Data Science Instructor

# What is ADX?

- Stands for "Average Directional Movement Index"
- Developed by J. Welles Wilder
  - "New Concepts in Technical Systems" (1987)
- Measures the strength of a trend
  - Oscillates between 0 and 100
  - $ADX \leq 25$ : no trend
  - $ADX > 25$ : trending market
  - $ADX > 50$ : strong trending market



# How is ADX calculated?

- Derived from the smoothed averages of the difference between +DI and -DI
  - +DI (Plus Directional Index): quantify the presence of an uptrend
  - -DI (Minus Directional Index): quantify the presence of a downtrend
- Calculation input:
  - high, low, and close prices of each period

# Implementing ADX in Python

```
# Calculate ADX
stock_data['ADX'] = talib.ADX(stock_data['High'], stock_data['Low'], stock_data['Close'],
                             timeperiod=14)

# Print the last five rows
print(stock_data.tail())
```

	Open	High	Low	Close	ADX
Date					
2020-11-24	540.40	559.99	526.20	555.38	21.16
2020-11-25	550.06	574.00	545.37	574.00	23.92
2020-11-27	581.16	598.78	578.45	585.76	26.82
2020-11-30	602.21	607.80	554.51	567.60	28.25
2020-12-01	597.59	597.85	572.05	584.76	29.58

# Plotting ADX

```
import matplotlib.pyplot as plt

# Create subplots
fig, (ax1, ax2) = plt.subplots(2)

# Plot ADX with the price
ax1.set_ylabel('Price')
ax1.plot(stock_data['Close'])
ax2.set_ylabel('ADX')
ax2.plot(stock_data['ADX'])

ax1.set_title('Price and ADX')
plt.show()
```





**Let's practice!**  
FINANCIAL TRADING IN PYTHON

# Momentum indicator: RSI

FINANCIAL TRADING IN PYTHON



**Chelsea Yang**

Data Science Instructor

# What is RSI?

- Stands for "Relative Strength Index"
- Developed by J. Welles Wilder
  - "New Concepts in Technical Systems" (1987)
- Measures the momentum of a trend
  - Oscillates between 0 and 100
  - $RSI > 70$ : Overbought
  - $RSI < 30$ : Oversold

# How is RSI calculated?

$$RSI = 100 - 100 / (1 + RS)$$

Where:

- RS: relative strength
- RS = average of upward price changes / average of downward price changes

# Implementing RSI in Python

```
# Calculate RSI
stock_data['RSI'] = talib.RSI(stock_data['Close'], timeperiod=14)
# Print the last five rows
print(stock_data.tail())
```

	Open	High	Low	Close	RSI
Date					
2020-11-24	1730.50	1771.60	1727.69	1768.88	62.78
2020-11-25	1772.89	1778.54	1756.54	1771.43	63.10
2020-11-27	1773.09	1804.00	1772.44	1793.19	65.81
2020-11-30	1781.18	1788.06	1755.00	1760.74	58.87
2020-12-01	1774.37	1824.83	1769.37	1798.10	63.63

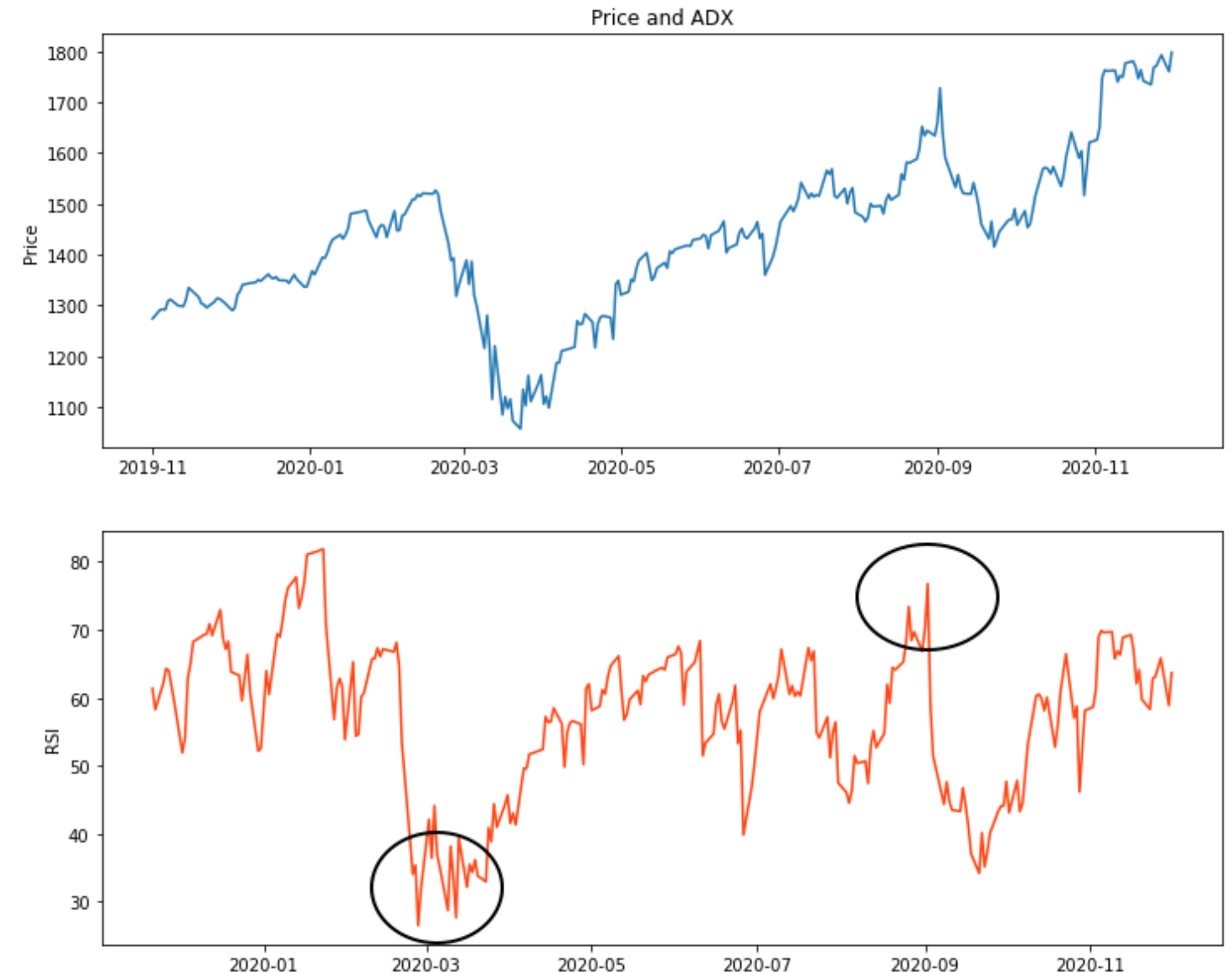
# Plotting RSI

```
import matplotlib.pyplot as plt

# Create subplots
fig, (ax1, ax2) = plt.subplots(2)

# Plot RSI with the price
ax1.set_ylabel('Price')
ax1.plot(stock_data['Close'])
ax2.set_ylabel('RSI')
ax2.plot(stock_data['RSI'])

ax1.set_title('Price and RSI')
plt.show()
```



**Let's practice!**  
FINANCIAL TRADING IN PYTHON

# Volatility indicator: Bollinger Bands

FINANCIAL TRADING IN PYTHON



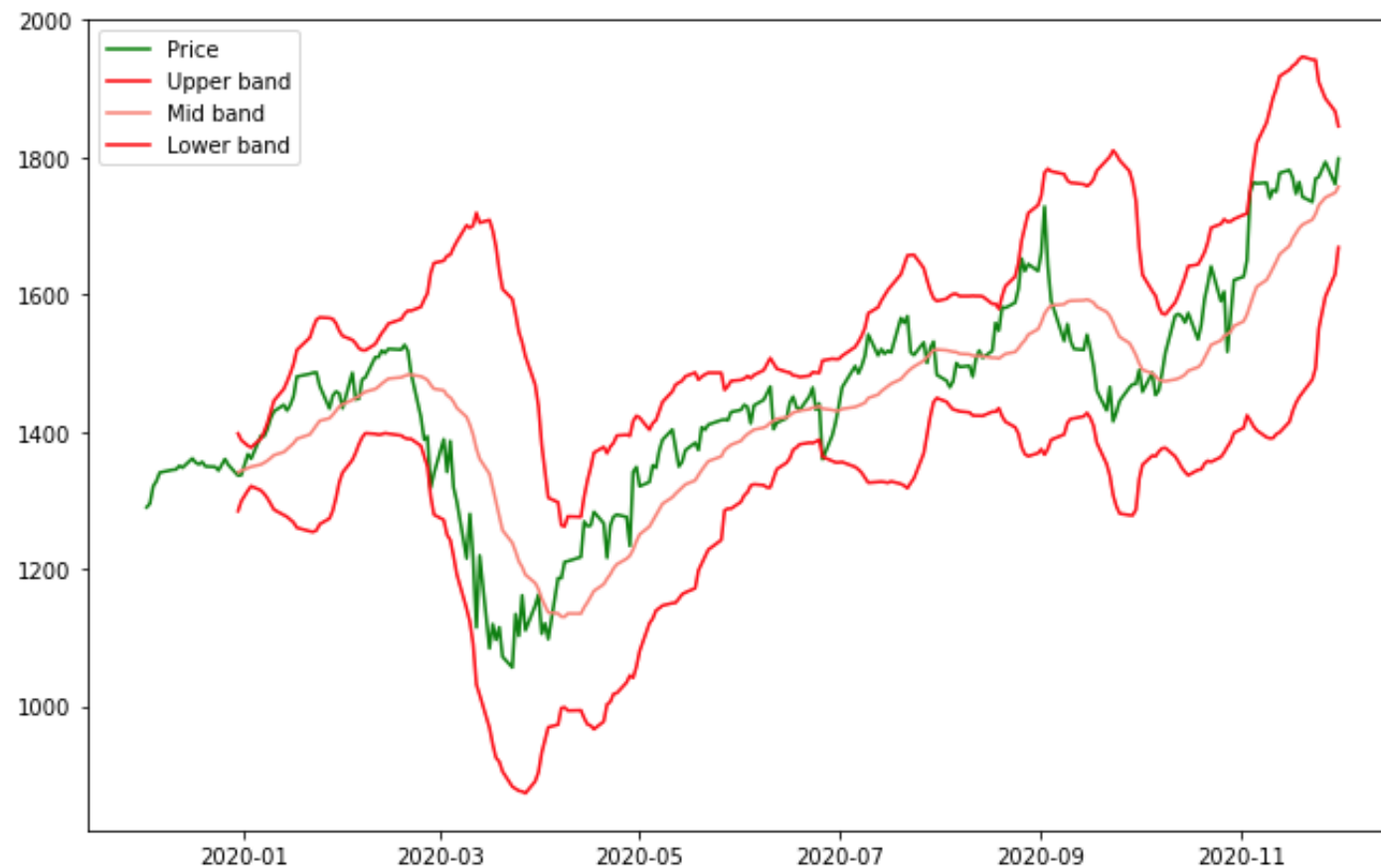
**Chelsea Yang**

Data Science Instructor



# What are Bollinger Bands?

- Developed by John Bollinger
  - "Bollinger on Bollinger Bands"



- Measure price volatility
- Composed of three lines:
  - Middle band: n-period simple moving average
  - Upper band: k-standard deviations above the middle band
  - Lower band: k-standard deviations below the middle band

# Bollinger Bands implications

- The wider the bands, the more volatile the asset prices.
- Measure whether a price is too high or too low on a relative basis:
  - Relatively high: price close to the upper band
  - Relatively low: price close to the lower band

# Implementing Bollinger Bands in Python

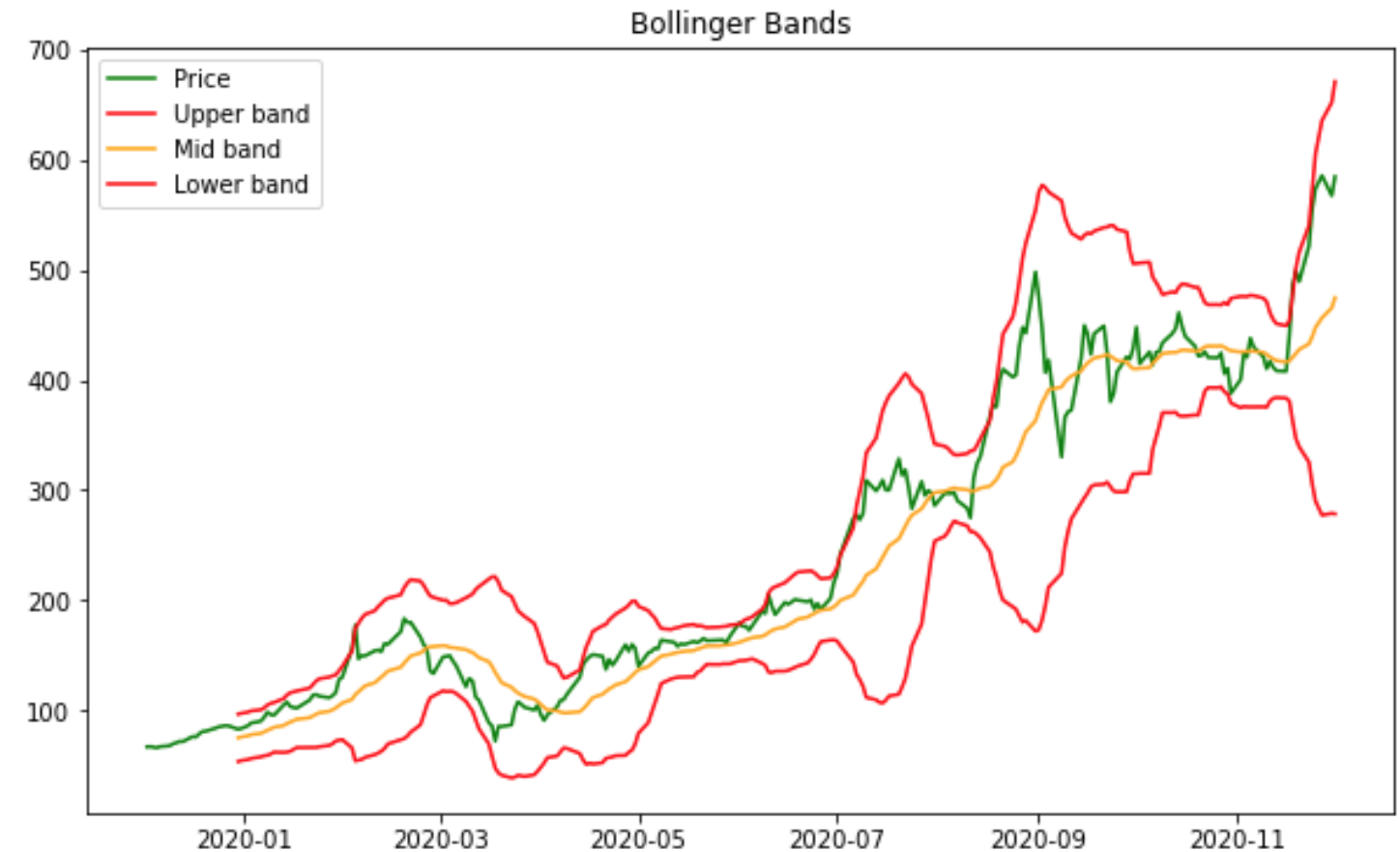
```
# Define the Bollinger Bands
upper, mid, lower = talib.BBANDS(stock_data['Close'],
                                nbdevup=2,
                                nbdevdn=2,
                                timeperiod=20)
```

# Plotting Bollinger Bands

```
import matplotlib.pyplot as plt

# Plot the Bollinger Bands
plt.plot(stock_data['Close'], label='Price')
plt.plot(upper, label="Upper band")
plt.plot(mid, label='Middle band')
plt.plot(lower, label='Lower band')

# Customize and show the plot
plt.title('Bollinger Bands')
plt.legend()
plt.show()
```



**Let's practice!**  
FINANCIAL TRADING IN PYTHON