

Python初級數據分析員證書

(六) 數據分析及可視化專案

# 13. 數據分析專案

## Demo3 - TA library

# Review

- Statistics
- Hypothesis testing
- Algebra
- Linear regression
- Propositional logic
- Python
- R
- SQL
- Pandas, NumPy, SciPy
- Data Visualization, Matplotlib, Seaborn, Plotly
- Dashboard Visualization, Business Intelligence
- Storytelling



# 13. 數據分析專案 Data Analysis Project – Demo3

## Chapter Summary

- Leading and Lagging Indicators
- TA-Lib introduction and installation
- STOCH RSI
- Bollinger Band
- MACD
- ADOSC
- ATR & TR

## Recap

In last chapter, we learnt simple backtest for multiple Moving Average combination. [Google Colab](#) is a great choice for long running computing on cloud.

In this chapter, we will discover more [Technical Analysis](#) method with library. TA is [not](#) just used in finance and stock market analysis, it is statistic tools to analyse [time series](#) problems. We often use stock data because its variety and convenience of source.

# Disclaimer

All content on from our course is for informational and educational purposes of a general nature only, and does not address any circumstances of any particular individual or entity. Do not construe any such information or material as investment, financial, professional or any other advice.



# Leading Indicator and Lagging Indicator

**Leading** and **lagging indicators** are two types of measurements used when assessing performance in a **business or economy**.

**Leading indicators** are considered to point toward **future events**.

- **Heads-up** for economists and investors who hope to anticipate **trends**.
- **Bond yields** are thought to be a good leading indicator of the stock market because bond traders anticipate and speculate about trends in the economy.

**Lagging indicators** are seen as **confirming** a pattern that is in progress.

- Be known **after the event**, but that doesn't make them useless.
- They can **clarify and confirm** a pattern that is occurring over time.
- The unemployment rate is one of the most reliable lagging indicators.

# Leading Indicator and Lagging Indicator

## Popular leading indicators in finance

- The relative strength index (RSI)
- The stochastic oscillator
- Williams %R
- On-balance volume (OBV)

## Popular lagging indicators in finance

- Moving averages
- The MACD indicator
- Bollinger bands

Relying solely on either could have negative effects on a strategy, which is why many traders will aim to find a **balance of the two**.

# TA library

There are two commonly used TA library that trader would use:

- **Pandas-ta**
  - Provide more than 130+ indicator. Easier to use.
  - Pandas TA - <https://github.com/twopirllc/pandas-ta>
- **ta-lib**
  - Fast library and available in other language like C, Java, Perl
  - Since its backend is written in C, the speed is faster
  - TA Lib - <https://github.com/TA-Lib/ta-lib-python>

## Ta-lib

**TA-Lib** is widely used by trading software developers requiring to perform technical analysis of financial market data.

- Includes 150+ indicators such as ADX, MACD, RSI, Stochastic, Bollinger Bands, etc.
- Candlestick pattern recognition
- Open-source API for C/C++, Java, Perl, Python and 100% Managed .NET
- Although it might be a little tricky while installation, it definitely a great tool afterwards.

# Ta-lib

In this chapter, we focus using ta-lib. First let's install the ta library.

<https://github.com/TA-Lib/ta-lib-python>

Add two lines of scripts in `.zshrc`

Then `source .zshrc`

```
156 # TA-Lib
157 export TA_INCLUDE_PATH="$(brew --prefix ta-lib)/include"
158 export TA_LIBRARY_PATH="$(brew --prefix ta-lib)/lib"
159
```

PROBLEMS    OUTPUT    DEBUG CONSOLE    TERMINAL    COMMENTS



```
1 pip install ta-lib
```

Collecting ta-lib

```
Using cached TA-Lib-0.4.26.tar.gz (272 kB)
Installing build dependencies ... done
Getting requirements to build wheel ... done
Installing backend dependencies ... done
Preparing metadata (pyproject.toml) ... done
```

## 13. 數據分析專案 Data Analysis Project – Demo3

```
1 import pandas as pd  
2 import numpy as np  
3 import talib  
4 import math  
5 import yfinance as yf  
6 import plotly.express as px  
7 import plotly.graph_objects as go  
8 from plotly.subplots import make_subplots
```

```
1 df_nvda = yf.download('NVDA', start='2022-04-01',  
2                         end='2023-01-01')  
3 df_nvda
```

[\*\*\*\*\*100%\*\*\*\*\*] 1 of 1 completed

	Open	High	Low	Close	Adj Close	Volume
Date						
2022-04-01	273.750000	274.959991	262.670013	267.119995	266.870575	51723500
2022-04-04	267.279999	275.579987	266.130005	273.600006	273.344543	39712000

# STOCHASTIC RSI

Stochastic RSI (StochRSI) is an indicator used in technical analysis that ranges between zero and one (or zero and 100 on some charting platforms) and is created by applying the Stochastic oscillator formula to a set of relative strength index (RSI) values rather than to standard price data.

StochRSI reading above 0.8 is considered overbought, while a reading below 0.2 is considered oversold. On the zero to 100 scale, above 80 is overbought, and below 20 is oversold.

Lowest RSI = Lowest RSI reading over last 14 periods (or chosen lookback period);  
Highest RSI = Highest RSI reading over last 14 period (or lookback period).

# STOCHASTIC RSI

**The Formulas For the Stochastic RSI (StochRSI) are:**

$$\text{StochRSI} = \frac{RSI - \min [RSI]}{\max [RSI] - \min [RSI]}$$

**where:**

$RSI$  = Current RSI reading

$\min [RSI]$  = Lowest RSI reading over the last 14 periods  
(or your chosen lookback interval)

$\max [RSI]$  = Highest RSI reading over the last 14 periods  
(or your chosen lookback interval)

# STOCHASTIC & RSI

Using talib.STOCHRSI to prepare k and d line. Function setting as follow:

```
k, d = talib.STOCH(df_nvda['High'], df_nvda['Low'], df_nvda['Close'],  
fastk_period, slowk_period, slowk_matype, slowd_period, slowd_matype)
```

K line is the fast line.

```
1 rsi = talib.RSI(df_nvda['Close'], timeperiod=14)  
2 k, d = talib.STOCH(df_nvda['High'], df_nvda['Low'], df_nvda['Close'], 14)
```

# STOCHASTIC & RSI

```
1 fig = make_subplots(rows=3, cols=1, shared_xaxes=True,
2                     vertical_spacing=0.01, row_heights=[0.7, 0.15, 0.15])
3 fig.add_trace(go.Candlestick(x=df_nvda.index,
4                             open=df_nvda['Open'], high=df_nvda['High'],
5                             low=df_nvda['Low'], close=df_nvda['Close'],
6                             showlegend=True, name='Close Price'), row=1, col=1)
7 fig.add_trace(go.Scatter(x=df_nvda.index,
8                          y=k, name='Stoch - k',
9                          line=dict(color='red', width=1),
10                         ), row=2, col=1)
11 fig.add_trace(go.Scatter(x=df_nvda.index,
12                          y=d, name='Stoch - d',
13                          line=dict(color='blue', width=1),
14                         ), row=2, col=1)
15 fig.add_trace(go.Scatter(x=df_nvda.index, y=rsi, name='RSI',
16                          marker=dict(color ='purple')), row=3, col=1)
17 fig.update_layout(title="NVIDIA Share Price (Close) US$",
18                    xaxis_rangeslider_visible=False)
19 fig.show()
```

# STOCHASTIC & RSI

NVIDIA Share Price (Close) US\$



# Bollinger Bands

Bollinger Bands are envelopes plotted at a standard deviation level above and below a simple moving average of the price. Because the distance of the bands is based on standard deviation, they adjust to volatility swings in the underlying price. Bollinger Bands use 2 parameters, Period and Standard Deviations.

Typical values used:

Short term: 10 day moving average, bands at 1.5 standard deviations. (1.5 times the standard dev. +/- the SMA)

Medium term: 20 day moving average, bands at 2 standard deviations.

Long term: 50 day moving average, bands at 2.5 standard deviations.

# Bollinger Bands

```
1 from talib import MA_Type  
2  
3 upper, middle, lower = talib.BBANDS(df_nvda['Close'], matype=MA_Type.T3)
```

Note (for your ref):

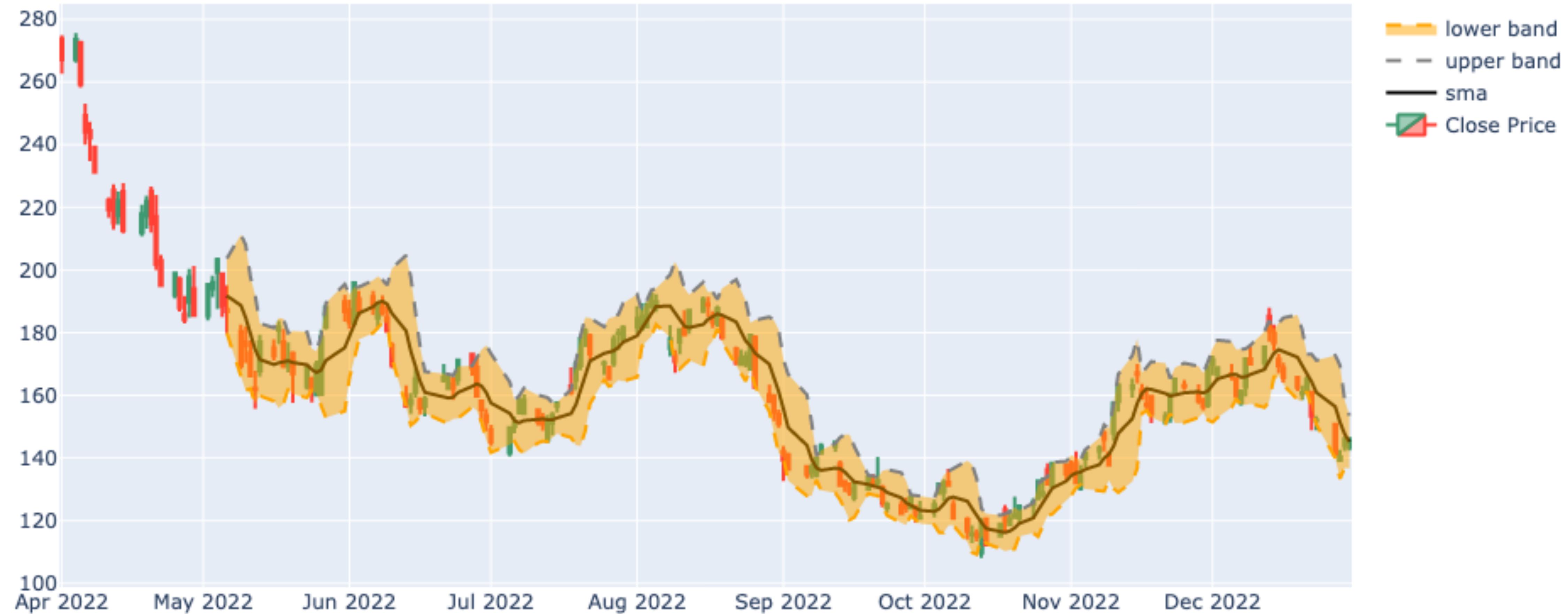
MA\_Type.T3 = talib.MA(closed, timeperiod=10, matype=3)

# Bollinger Bands

```
1 fig=go.Figure(data=[go.Candlestick(x=df_nvda.index,
2                                 open=df_nvda['Open'], high=df_nvda['High'],
3                                 low=df_nvda['Low'], close=df_nvda['Close'],
4                                 showlegend=True, name='Close Price') ] )
5
6 # Moving Average
7 fig.add_trace(go.Scatter(x = df_nvda.index, y = middle,
8                         line_color = 'black', name = 'sma'))
9
10 # Upper Bound
11 fig.add_trace(go.Scatter(x = df_nvda.index, y = upper,
12                         line_color = 'gray', line = {'dash': 'dash'},
13                         name = 'upper band', opacity = 0.5))
14
15 # Lower Bound fill in between with parameter 'fill': 'tonexty'
16 fig.add_trace(go.Scatter(x = df_nvda.index, y = lower,
17                         line_color = 'orange', line = {'dash': 'dash'},
18                         fill = 'tonexty', name = 'lower band', opacity = 0.4))
19
20 fig.update_layout(title="NVDIA Share Price (Close) US$",
21                   xaxis_rangeslider_visible=False)
```

# Bollinger Bands

NVIDIA Share Price (Close) US\$



# MACD

Moving average convergence divergence (**MACD**) is one of the most popular technical indicators in trading. The MACD is appreciated by traders worldwide for its simplicity and flexibility, as it can be used as a trend or momentum indicator and signal opportunities to enter and exit positions.

# MACD

There are three different elements involved with the histogram, which is mapped out around a baseline:

- The **MACD line** (subtracting a long-term exponential moving average (EMA) from a shorter-term EMA)
- The **signal line** (subtracting the two EMAs and creating a 9-day moving average)
- The **histogram** (subtracting the MACD line from the signal line)

## TA lib setting

```
Macd, macdsignal, macdhist = talib.MACD(close, fastperiod=12, slowperiod=26, signalperiod=9)
```

# MACD

```
1 # Colorize the histogram values
2 colors = np.where(macdhist < 0, '#008000', '#FF00FF')
3
4 fig = make_subplots(rows=2, cols=1, shared_xaxes=True,
5                      vertical_spacing=0.01, row_heights=[0.8, 0.2])
6 fig.add_trace(go.Candlestick(x=df_nvda.index,
7                             open=df_nvda['Open'], high=df_nvda['High'],
8                             low=df_nvda['Low'], close=df_nvda['Close'],
9                             showlegend=True, name='Close Price'), row=1, col=1)
10 # MACD
11 fig.append_trace(
12     go.Scatter(x=df_nvda.index, y=macd,
13                 line=dict(color='#ff9900', width=2), name='macd',
14                 legendgroup='2', ), row=2, col=1)
15
16 # MACD Signal
17 fig.append_trace(
18     go.Scatter(x=df_nvda.index, y=macdsignal,
19                 line=dict(color='#000000', width=2),
20                 legendgroup='2', name='signal'), row=2, col=1)
21
22 # histogram
23 fig.append_trace(
24     go.Bar(x=df_nvda.index, y=macdhist, name='histogram',
25             marker_color=colors,), row=2, col=1)
26
27 fig.update_layout(title="NVIDIA Share Price (Close) US$",
28                    xaxis_rangeslider_visible=False)
29 fig.show()
```

# MACD

NVIDIA Share Price (Close) US\$



# ADOSC

Accumulation/Distribution Oscillator (ADOSC) is a volume indicator. The ADOSC is a cumulative indicator that uses **volume** and **price** to assess whether a stock is being accumulated or distributed.

## The Accumulation/Distribution Indicator (A/D) Formula

$$MFM = \frac{(Close - Low) - (High - Close)}{High - Low}$$

**where:**

MFM = Money Flow Multiplier

Close = Closing price

Low = Low price for the period

High = High price for the period

$$\text{Money Flow Volume} = MFM \times \text{Period Volume}$$

$$A/D = \text{Previous A/D} + CMFV$$

**where:**

CMFV = Current period money flow volume

# ADOSC

## TA Lib setting

```
real = talib.ADOSC(high, low, close, volume, fastperiod=3, slowperiod=10)
```

```
1 adosc = talib.ADOSC(df_nvda['High'], df_nvda['Low'], df_nvda['Close'],
2                      df_nvda['Volume'], fastperiod=3, slowperiod=10)
```

```
1 fig = make_subplots(rows=2, cols=1, shared_xaxes=True,
2                      vertical_spacing=0.01, row_heights=[0.7, 0.3])
3 fig.add_trace(go.Candlestick(x=df_nvda.index,
4                               open=df_nvda['Open'], high=df_nvda['High'],
5                               low=df_nvda['Low'], close=df_nvda['Close'],
6                               showlegend=True, name='Close Price'), row=1, col=1)
7 fig.add_trace(go.Scatter(x=df_nvda.index, y=adosc, name='ADOSC',
8                           marker=dict(color='purple')), row=2, col=1)
9 fig.update_layout(title="NVDIA Share Price (Close) US$",
10                    xaxis_rangeslider_visible=False)
11 fig.show()
```

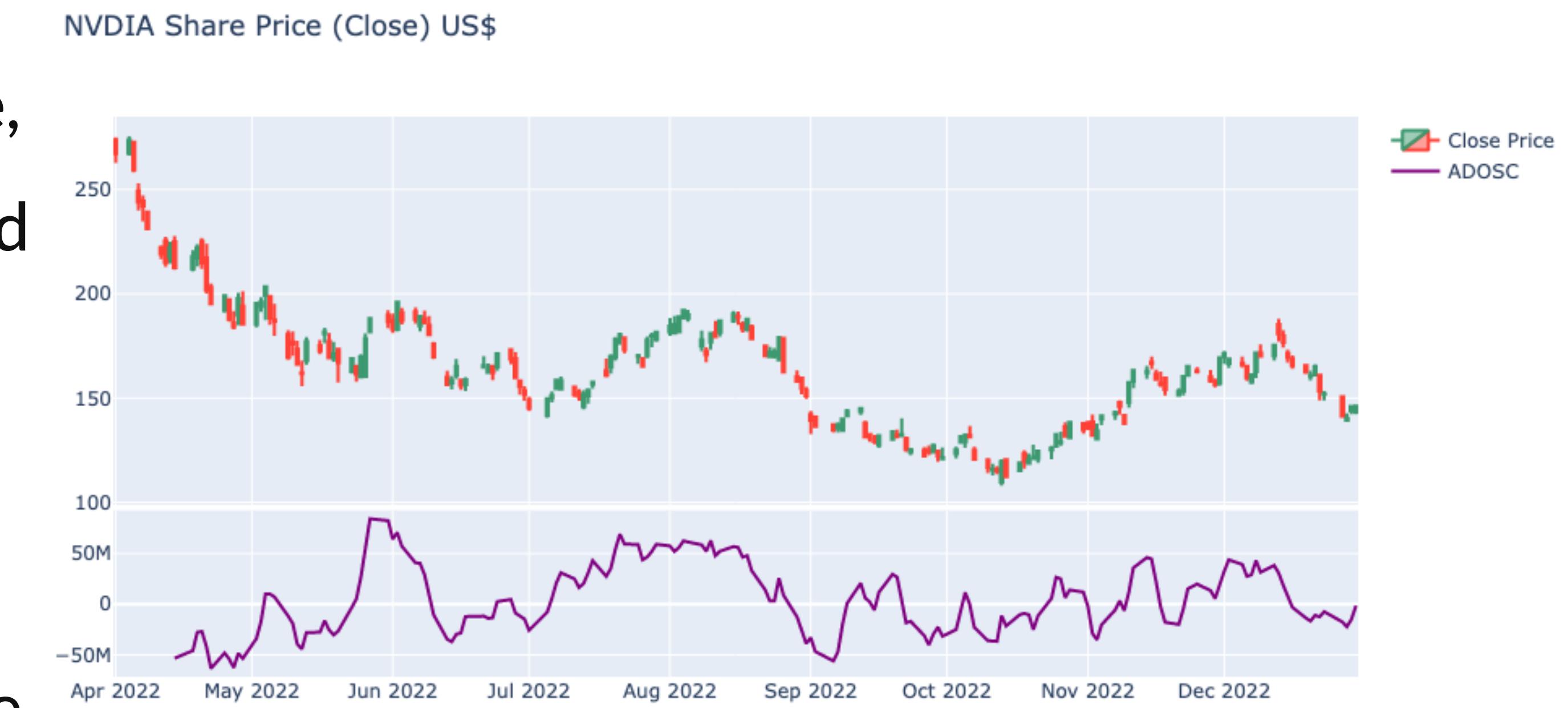
# ADOSC

NVIDIA Share Price (Close) US\$



# ADOSC

- The accumulation/distribution (ADOSC) indicates **supply** and **demand** of an asset or security by looking at where the price closed within the period's range and then multiplying that by volume.
- The ADOSC indicator is cumulative, meaning one period's value is added or subtracted from the last.
- In general, a **rising ADOSC** helps **confirm a rising price trend**, while a **falling ADOSC** helps confirm a price **downtrend**.



# ADOSC – Further Consideration

You may check its holder of of the stock on morningstars or yfinance. And calculate the approximately market shares and the non-institute share portion. That makes hint for you to estimate the minimum and maximum ADOSC.

## Major Holders

### Breakdown

4.18%	% of Shares Held by All Insider
68.08%	% of Shares Held by Institutions
71.06%	% of Float Held by Institutions
4,015	Number of Institutions Holding Shares

<https://finance.yahoo.com/quote/NVDA/holders?p=NVDA>

# ATR

The **average true range (ATR)** is a technical analysis indicator introduced by market technician J. Welles Wilder Jr. in his book *New Concepts in Technical Trading Systems* that **measures market volatility** by decomposing the entire range of an asset price for that period.

# ATR & TR

- ATR is typically derived from the 14-day simple moving average of a series of true range indicators.
- The ATR was initially developed for use in commodities markets but has since been applied to all types of securities.
- ATR shows investors the average range prices swing for an investment over a specified period.
- True Range is the daily based swing range.

# ATR

TA lib setting

`ATR = talib.ATR(high, low, close, timeperiod=14)`

`TR = talib.TRANGE(high, low, close)`

## The Average True Range (ATR) Formula

The formula to calculate ATR for an investment with a previous ATR calculation is :

$$\frac{\text{Previous ATR}(n - 1) + \text{TR}}{n}$$

**where:**

$n$  = Number of periods

TR = True range

# ATR & TR

```
1 atr = talib.ATR(df_nvda['High'], df_nvda['Low'], df_nvda['Close'], timeperiod=14)
2 tr = talib.TRANGE(df_nvda['High'], df_nvda['Low'], df_nvda['Close'])
```

```
1 fig = make_subplots(rows=2, cols=1, shared_xaxes=True,
2                         vertical_spacing=0.01, row_heights=[0.7, 0.3])
3 fig.add_trace(go.Candlestick(x=df_nvda.index,
4                             open=df_nvda['Open'], high=df_nvda['High'],
5                             low=df_nvda['Low'], close=df_nvda['Close'],
6                             showlegend=True, name='Close Price'),row=1,col=1 )
7 fig.add_trace(go.Scatter(x=df_nvda.index, y=atr, name='ATR',
8                           marker=dict(color ='orange' )), row=2, col=1)
9 fig.add_trace(go.Scatter(x=df_nvda.index, y=tr, name='TR',
10                           marker=dict(color ='purple' )), row=2, col=1)
11 fig.update_layout(title="NVIDIA Share Price (Close) US$",
12                     xaxis_rangeslider_visible=False)
13 fig.show()
```

# ATR & TR

NVIDIA Share Price (Close) US\$



# ATR & TR

NVDIA Share Price (Close) US\$



Close Price  
ATR  
TR

For case TR greater than ATR, that probably means the current trend might change soon.

If you are an option trader, you should pay more attention to the asset volatility.

# Chapter Wrapping

TA-Lib is a C backend library support Python and other main stream core language. We only demonstrated 6 out of 130+ TA with plotly graphs. Statistical TA data analysis will be discussed in later chapters.

# Academic Resources

TA Lib - <https://github.com/TA-Lib/ta-lib-python>

TA Lib Documentation - [https://ta-lib.github.io/ta-lib-python/doc\\_index.html](https://ta-lib.github.io/ta-lib-python/doc_index.html)

Pandas TA - <https://github.com/twopirllc/pandas-ta>

QuantConnect - <https://www.quantconnect.com/>

Google Colab - <https://colab.research.google.com/>

Google Scholar - <https://scholar.google.com/>

Research Gate - <https://www.researchgate.net/>

JSTOR - <https://www.jstor.org/>

Research SPJ - <https://spj.science.org/>

