Time series analysis means analyzing and finding patterns in a time series dataset. A timeseries dataset is a sequence of data collected over an interval of time. Stock price data, monthly sales data, daily rainfall data, hourly website traffic data are some examples of timeseries data that you will get to solve business problems as a data scientist

I will be using the plotly library in Python here as it is easy to analyze data with plotly because of less code and interactive results. I will recommend using a Jupyter Notebook or Google Colaboratory for Time series analysis instead of using a code editor or an IDE like VS Code or PyCharm.

```
In [1]: # Importing required libraries
        import pandas as pd
        import plotly.express as px
         import plotly.graph_objects as go
        import yfinance as yf
        from datetime import date, timedelta
        # Getting today's date and formatting it
        today=date.today()
        end_date = today.strftime("%Y-%m-%d")
        # Getting the date 720 days ago from today and formatting it
        d1 = date.today() - timedelta(days=720)
        start_date = d1.strftime("%Y-%m-%d")
         # Fetching historical stock market data for AAPL (Apple Inc.)
        data = yf.download('AAPL',
                               start=start date,
                               end=end date,
                               progress=False)
```

In [2]: data.head(8)

Out[2]:

	Open	High	Low	Close	Adj Close	Volume
Date						
2022-04-04	174.570007	178.490005	174.440002	178.440002	176.423874	76468400
2022-04-05	177.500000	178.300003	174.419998	175.059998	173.082062	73401800
2022-04-06	172.360001	173.630005	170.130005	171.830002	169.888565	89058800
2022-04-07	171.160004	173.360001	169.850006	172.139999	170.195053	77594700
2022-04-08	171.779999	171.779999	169.199997	170.089996	168.168213	76575500
2022-04-11	168.710007	169.029999	165.500000	165.750000	163.877258	72246700
2022-04-12	168.020004	169.869995	166.639999	167.660004	165.765686	79265200
2022-04-13	167.389999	171.039993	166.770004	170.399994	168.474701	70618900

```
In [3]: data.tail(8)
```

Date						
2024-03-13	172.770004	173.190002	170.759995	171.130005	171.130005	52488700
2024-03-14	172.910004	174.309998	172.050003	173.000000	173.000000	72913500
2024-03-15	171.169998	172.619995	170.289993	172.619995	172.619995	121664700
2024-03-18	175.570007	177.710007	173.520004	173.720001	173.720001	75604200
2024-03-19	174.339996	176.610001	173.029999	176.080002	176.080002	55215200
2024-03-20	175.720001	178.669998	175.089996	178.669998	178.669998	53423100
2024-03-21	177.050003	177.490005	170.839996	171.369995	171.369995	106181300
2024-03-22	171.860001	173.050003	170.059998	172.279999	172.279999	70618124

Low

Close

Adj Close

Volume

```
In [4]: data.info()
```

Out[3]:

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 495 entries, 2022-04-04 to 2024-03-22
Data columns (total 6 columns):
```

#	Column	Non-	Null Count	Dtype
0	Open	495	non-null	float64
1	High	495	non-null	float64
2	Low	495	non-null	float64
3	Close	495	non-null	float64
4	Adj Close	495	non-null	float64
5	Volume	495	non-null	int64
dtynes: float64(5)			int64(1)	

dtypes: float64(5), int64(1) memory usage: 27.1 KB

Open

High

The reason we only have 494 values instead of the expected 720 days worth of data could be due to several factors:

- Trading Days vs. Calendar Days: Stock markets are not open on weekends and holidays. Therefore, if you're looking for 720 calendar days, the number of trading days within that period might be less.
- Missing Data: Sometimes, there are missing data points for various reasons such as holidays, data errors, or the unavailability of trading data.

In [5]: data.describe()

	Open	High	Low	Close	Adj Close	Volume
count	495.000000	495.000000	495.000000	495.000000	495.000000	4.950000e+02
mean	165.119213	166.881899	163.492808	165.266061	164.413930	6.953084e+07
std	18.827272	18.503530	19.051000	18.784394	19.079842	2.408641e+07
min	126.010002	127.769997	124.169998	125.019997	124.166634	2.404830e+07
25%	148.720001	150.760002	147.139999	149.075005	147.929649	5.234125e+07
50%	167.320007	168.880005	165.649994	167.229996	165.765198	6.508660e+07
75%	180.949997	182.384995	179.004997	180.849998	180.245224	8.030145e+07
max	198.020004	199.619995	197.000000	198.110001	197.857529	1.826020e+08

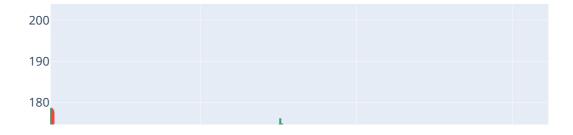
Out[5]:

Time Series Analysis (Line Plot)

```
200
190
180
```

```
xaxis_rangeslider_visible = False)
figure.show()
```

Time Series Analysis (Candlestick Chart)



Time Series Analysis (Bar Plot)



Time Series Analysis (Custom Date Range)

```
200190180
```

```
In [10]: # Interactive Candlestick Chart with Buttons and Slider: Visualizing open, high, lo
         figure = go.Figure(data = [go.Candlestick(x = data.index,
                                                  open = data["Open"],
                                                  high = data["High"],
                                                  low = data["Low"],
                                                  close = data["Close"])])
         figure.update_layout(title = "Time Series Analysis (Candlestick Chart with Buttons
         figure.update_xaxes(
             rangeslider_visible = True,
             rangeselector = dict(
                 buttons = list([
                     dict(count = 1, label = "1m", step = "month", stepmode = "backward"),
                     dict(count = 6, label = "6m", step = "month", stepmode = "backward"),
                     dict(count = 1, label = "YTD", step = "year", stepmode = "todate"),
                     dict(count = 1, label = "1y", step = "year", stepmode = "backward"),
                     dict(step = "all")
                 ])
         figure.show()
```

Time Series Analysis (Candlestick Chart with Buttons and Slider)

