

# Advanced Python Module :

Project: Trading Bot-BTC :

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## Part I

### Abstract :

Cryptocurrencies are known for being incredibly volatile, with prices fluctuating dramatically even in the space of minutes. Investors also have the opportunity to take part in cryptocurrency trading around the world and at any hour of the day. Combined, these factors limit the effectiveness of human cryptocurrency trading in several ways.

First, investors in many cases cannot react quickly enough to changes in price to achieve the optimal trades that are theoretically available to them. Slowdowns in exchanges and transaction times further exacerbate this problem. Second, investors can simply not dedicate as much time to the cryptocurrency markets as necessary to always achieve the best trades. Doing so would require round-the-clock monitoring of cryptocurrency exchanges all over the globe.

Fortunately for many investors, there are solutions to these issues. One of the primary solutions is bots. Crypto trading bots are automated software that helps you to buy and sell cryptocurrencies at the correct time. The main goal of these software is to increase revenue and reduce losses and risks. These applications enable you to manage all crypto exchange account in one place. Many such programs allow you to trade for Ethereum, Litecoin, Bitcoin (BTC), and more with ease.



## Part II

### Introduction :

Crypto bots analyze data, predict risk and buy and sell assets as per their calculations. They watch the market and trade when certain market conditions are met. Most crypto trading bots work by connecting directly to a cryptocurrency exchange. After the bot successfully connects to an exchange, it starts watching the market and waiting for certain events or changes in prices. Once it detects an event, it will either send you a signal to take action or make a decision based on the rules you've defined and take the required action (buy/sell) itself. Some bots can even take into account historical data to make more accurate decisions.



# 1 Advantages of Bots in trading :

-Bots are used by traders to take advantage of the cryptocurrency markets that trade 24/7 all over the world.

-The advantage bots have over investors is they can react quicker.

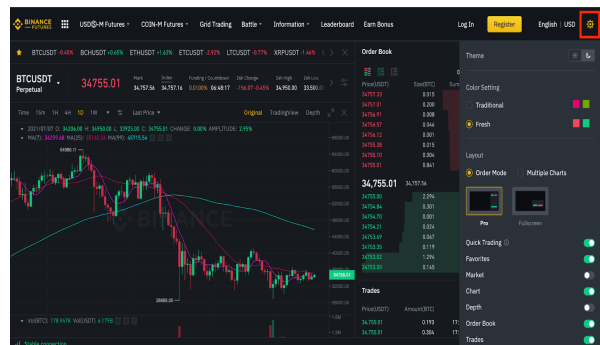
-Meanwhile, most investors also don't have the time to dedicate to always get the best trade—something that bots can do.

-One key type of bot is the arbitrage bot, which looks to take advantage of price discrepancies across exchanges.

## 2 Binance :

Binance is a cryptocurrency exchange which is the largest exchange in the world in terms of daily trading volume of cryptocurrencies. It was founded in 2017 and is registered in the Cayman Islands.

Binance was founded by Changpeng Zhao, a developer who had previously created high frequency trading software. Binance was initially based in China, but later moved its headquarters out of China following the Chinese government's increasing regulation of cryptocurrency.



## Part III Project :

### 3 Install packages :

#### 3.1 What Is Technical Analysis TA ?

Technical analysis is a trading discipline employed to evaluate investments and identify trading opportunities by analyzing statistical trends gathered from trading activity, such as price movement and volume.

```
1 pip install ta

Requirement already satisfied: ta in e:\anaconda\lib\site-packages (0.9.0)
Requirement already satisfied: numpy in e:\anaconda\lib\site-packages (from ta) (1.20.1)
Requirement already satisfied: pandas in e:\anaconda\lib\site-packages (from ta) (1.2.4)
Requirement already satisfied: python-dateutil>=2.7.3 in e:\anaconda\lib\site-packages (from pandas->ta) (2.8.1)
Requirement already satisfied: pytz>=2017.3 in e:\anaconda\lib\site-packages (from pandas->ta) (2021.1)
Requirement already satisfied: six>=1.5 in e:\anaconda\lib\site-packages (from python-dateutil>=2.7.3->pandas->ta) (1.15.0)
Note: you may need to restart the kernel to use updated packages.
```

## 3.2 What is Python-Binance ?

The Binance API is a method that allows you to connect to the Binance servers via Python or several other programming languages. With it, you can automate your trading. More specifically, Binance has a RESTful API that uses HTTP requests to send and receive data.

```
1 pip install python-binance

Collecting python-binance
  Using cached python_binance-1.0.15-py2.py3-none-any.whl (63 kB)
Collecting websockets==9.1
  Using cached websockets-9.1-cp38-cp38-win_amd64.whl (90 kB)
Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-packages (from python-binance) (1.15.0)
Collecting aiohttp
  Using cached aiohttp-3.8.1-cp38-cp38-win_amd64.whl (555 kB)
Collecting dateparser
  Using cached dateparser-1.1.0-py2.py3-none-any.whl (288 kB)
Requirement already satisfied: ujson in c:\programdata\anaconda3\lib\site-packages (from python-binance) (4.0.2)
Requirement already satisfied: requests in c:\programdata\anaconda3\lib\site-packages (from python-binance) (2.25.1)
Collecting aiosignal>=1.1.2
  Using cached aiosignal-1.2.0-py3-none-any.whl (8.2 kB)
Collecting charset-normalizer<3.0,>=2.0
  Using cached charset-normalizer-2.0.10-py3-none-any.whl (39 kB)
Collecting yarl<2.0,>=1.0
  Using cached yarl-1.7.2-cp38-cp38-win_amd64.whl (122 kB)
Collecting multidict<7.0,>=4.5
  Using cached multidict-6.0.2-cp38-cp38-win_amd64.whl (28 kB)
Collecting frozenlist>=1.1.1
  Using cached frozenlist-1.3.0-cp38-cp38-win_amd64.whl (33 kB)
Collecting async-timeout<5.0,>=4.0.0a3
  Using cached async_timeout-4.0.2-py3-none-any.whl (5.8 kB)
Requirement already satisfied: attrs>=17.3.0 in c:\programdata\anaconda3\lib\site-packages (from aiohttp->python-binance) (20.3.0)
Requirement already satisfied: idna>=2.0 in c:\programdata\anaconda3\lib\site-packages (from yarl<2.0,>=1.0->aiohttp->python-binance) (2.10)
Requirement already satisfied: python-dateutil in c:\programdata\anaconda3\lib\site-packages (from dateparser->python-binance) (2.8.1)
Collecting tzlocal
  Using cached tzlocal-4.1-py3-none-any.whl (19 kB)
Requirement already satisfied: regex!=2019.02.19,!=2021.8.27 in c:\programdata\anaconda3\lib\site-packages (from dateparser->python-binance) (2021.4.4)
Requirement already satisfied: pytz in c:\programdata\anaconda3\lib\site-packages (from dateparser->python-binance) (2021.1)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\programdata\anaconda3\lib\site-packages (from requests->python-binance) (1.26.4)
Requirement already satisfied: certifi>=2017.4.17 in c:\programdata\anaconda3\lib\site-packages (from requests->python-binance) (2020.12.5)
```

## 4 Libraries used :

Pandas, Numpy, ta, Seaborn, Matplotlib, ...



```

1 import pandas as pd
2 from binance.client import Client
3 import ta
4 import matplotlib.pyplot as plt
5 import seaborn as sns
6 import numpy as np
7
8 import os
9 print(os.listdir(r"C:\Users\Oussama\Desktop\python"))

```

['.ipynb\_checkpoints', 'archive (1).zip', 'athlete\_events.csv', 'BTC-USD.csv', 'BTC1.csv', 'c.csv', 'covid-19-datasets-1200x900.jpg', 'Covid\_19\_Dataset\_Morocco.csv', 'data\_Bitcoin.csv', 'Data\_F.csv', 'data\_MAR\_2.csv', 'df10.csv', 'df9.csv', 'df\_BT.csv', 'new3.csv', 'new\_cases\_analysis.ipynb', 'noc\_regions.csv', 'Olympics.ipynb', 'Olympics\_Boujbair.ipynb', 'our\_data.csv', 'projet\_SC.ipynb', 'serie\_temp1.ipynb', 'serie\_temp - Copie (2).ipynb', 'serie\_temp - Copie.ipynb', 'serie\_temp.ipynb', 'Untitled.ipynb', 'Untitled1.ipynb', 'Untitled2.ipynb', 'Untitled3.ipynb', 'Untitled4.ipynb', 'Untitled5.ipynb', 'Untitled6.ipynb', 'Untitled7.ipynb', 'VosQuestions.ipynb']

## 5 Dataset :

### 5.1 Download Data

Download and retrieve our dataset from the binance platform, this dataset is made up of a table of 12 columns: Time, open, close, high, low, volume .... since 01/01/2018 until the hour code execution.

```

1 recup_data = Client().get_historical_klines("BTCUSD", Client.KLINE_INTERVAL_1HOUR,"01 January 2018")
2 BTC_data = pd.DataFrame(recup_data, columns = ['Time','Open','High','Low','Close','Volume','close_time','quote_av','trades',

```

|       | Time          | Open           | High           | Low            | Close          | Volume        | close_time    | quote_av          | trades | tb_base_a    |
|-------|---------------|----------------|----------------|----------------|----------------|---------------|---------------|-------------------|--------|--------------|
| 0     | 1514768400000 | 13715.65000000 | 13715.65000000 | 13400.01000000 | 13529.01000000 | 443.35619900  | 1514768399999 | 5993909.83603800  | 5228   | 228.52192100 |
| 1     | 1514768400000 | 13528.99000000 | 13595.89000000 | 13155.38000000 | 13203.08000000 | 383.69700600  | 1514771999999 | 5154521.55513544  | 4534   | 180.84040300 |
| 2     | 1514772000000 | 13203.00000000 | 13418.43000000 | 13200.00000000 | 13330.18000000 | 429.06457200  | 1514775599999 | 5710192.01852959  | 4887   | 192.23793500 |
| 3     | 1514775600000 | 13330.26000000 | 13611.27000000 | 13290.00000000 | 13410.03000000 | 420.08703000  | 1514779199999 | 5657448.45730419  | 4789   | 137.91840700 |
| 4     | 1514779200000 | 13434.98000000 | 13623.29000000 | 13322.15000000 | 13801.01000000 | 340.80732900  | 1514782799999 | 4588046.99615741  | 4563   | 172.95763500 |
| ...   | ...           | ...            | ...            | ...            | ...            | ...           | ...           | ...               | ...    | ...          |
| 35632 | 1643468400000 | 37620.04000000 | 37735.67000000 | 37268.44000000 | 37566.82000000 | 1881.67209000 | 1643471999999 | 70837969.49429300 | 51754  | 889.31896000 |
| 35633 | 1643472000000 | 37567.98000000 | 37603.57000000 | 37317.37000000 | 37538.19000000 | 1102.74218000 | 1643475599999 | 41323390.93888780 | 32444  | 550.48690000 |
| 35634 | 1643475600000 | 37538.19000000 | 37720.77000000 | 37532.15000000 | 37646.31000000 | 784.47184000  | 1643479199999 | 29522144.27770180 | 27730  | 349.29032000 |
| 35635 | 1643479200000 | 37646.31000000 | 37904.19000000 | 37558.56000000 | 37806.23000000 | 772.06693000  | 1643482799999 | 29138395.49170340 | 31389  | 428.81978000 |
| 35636 | 1643482800000 | 37806.24000000 | 37940.19000000 | 37781.49000000 | 37891.83000000 | 305.89610000  | 1643486399999 | 11578326.31702000 | 12413  | 168.24119000 |

35637 rows x 12 columns

### 5.2 Organize the Data for analysis:

Clean Dataset, Remove some columns :

```

1 del BTC_data['ignore']
2 del BTC_data['close_time']
3 del BTC_data['tb_quote_av']
4 del BTC_data['tb_base_av']
5 del BTC_data['trades']
6 del BTC_data['quote_av']

1 BTC_data.sample(5)

```

|       | Time          | Open           | High           | Low            | Close          | Volume        |
|-------|---------------|----------------|----------------|----------------|----------------|---------------|
| 12404 | 1559896400000 | 7754.79000000  | 7799.99000000  | 7734.44000000  | 7759.70000000  | 1140.08916300 |
| 25497 | 1606921200000 | 18910.21000000 | 18989.00000000 | 18728.38000000 | 18891.57000000 | 3966.23942200 |
| 1899  | 1521720000000 | 8645.00000000  | 8753.54000000  | 8632.74000000  | 8715.00000000  | 1770.70372400 |
| 9317  | 1548525600000 | 3568.97000000  | 3572.74000000  | 3565.15000000  | 3569.42000000  | 460.47190200  |
| 6238  | 1537405200000 | 8421.50000000  | 8422.00000000  | 8400.12000000  | 8407.12000000  | 1159.04455100 |

Numiric columns Close, High, Low, Open :

```

1 BTC_data['Close'] = pd.to_numeric(BTC_data['Close'])
2 BTC_data['High'] = pd.to_numeric(BTC_data['High'])
3 BTC_data['Low'] = pd.to_numeric(BTC_data['Low'])
4 BTC_data['Open'] = pd.to_numeric(BTC_data['Open'])
5

1 BTC_data.head()

```

|   | Time          | Open     | High     | Low      | Close    | Volume       |
|---|---------------|----------|----------|----------|----------|--------------|
| 0 | 1514764800000 | 13715.65 | 13715.65 | 13400.01 | 13529.01 | 443.35619900 |
| 1 | 1514768400000 | 13528.99 | 13595.89 | 13155.38 | 13203.06 | 383.69700800 |
| 2 | 1514772000000 | 13203.00 | 13418.43 | 13200.00 | 13330.18 | 429.06457200 |
| 3 | 1514775600000 | 13330.26 | 13611.27 | 13290.00 | 13410.03 | 420.08703000 |
| 4 | 1514779200000 | 13434.98 | 13623.29 | 13322.15 | 13601.01 | 340.80732900 |

```

1 BTC_data.to_csv(r'C:\Users\Oussama\Desktop\python\BTC1.csv', index = False) # save BTC_data in a csv file

```

Convert Time and define it as index :



```

1 Data_A= pd.read_csv(r'C:\Users\Oussama\Desktop\python\BTC1.csv')
2 Data_A = Data_A.set_index(Data_A['Time'])
3 Data_A.index = pd.to_datetime(Data_A.index, unit='ms')
4 del Data_A['Time']
5 Data_A

```

|                     | Open     | High     | Low      | Close    | Volume      |
|---------------------|----------|----------|----------|----------|-------------|
| Time                |          |          |          |          |             |
| 2018-01-01 00:00:00 | 13715.65 | 13715.65 | 13400.01 | 13529.01 | 443.356199  |
| 2018-01-01 01:00:00 | 13528.99 | 13595.89 | 13155.38 | 13203.06 | 383.697006  |
| 2018-01-01 02:00:00 | 13203.00 | 13418.43 | 13200.00 | 13330.18 | 429.064572  |
| 2018-01-01 03:00:00 | 13330.26 | 13611.27 | 13290.00 | 13410.03 | 420.087030  |
| 2018-01-01 04:00:00 | 13434.98 | 13623.29 | 13322.15 | 13601.01 | 340.807329  |
| ...                 | ...      | ...      | ...      | ...      | ...         |
| 2022-01-29 15:00:00 | 37620.04 | 37735.67 | 37268.44 | 37566.82 | 1881.672090 |
| 2022-01-29 16:00:00 | 37567.98 | 37603.57 | 37317.37 | 37538.19 | 1102.742180 |
| 2022-01-29 17:00:00 | 37538.19 | 37720.77 | 37532.15 | 37646.31 | 784.471840  |
| 2022-01-29 18:00:00 | 37646.31 | 37904.19 | 37556.56 | 37806.23 | 772.066930  |
| 2022-01-29 19:00:00 | 37806.24 | 37940.19 | 37781.49 | 37891.83 | 305.896100  |

35637 rows × 5 columns

```
1 Data_A.describe()
```

|       | Open         | High         | Low          | Close        | Volume       |
|-------|--------------|--------------|--------------|--------------|--------------|
| count | 35637.000000 | 35637.000000 | 35637.000000 | 35637.000000 | 35637.000000 |
| mean  | 18792.468892 | 18903.449612 | 18673.027058 | 18793.145426 | 2275.837916  |
| std   | 17869.060310 | 17976.505457 | 17754.698380 | 17869.330804 | 2182.479426  |
| min   | 3172.620000  | 3184.750000  | 3156.260000  | 3172.050000  | 0.000000     |
| 25%   | 7134.490000  | 7165.000000  | 7096.000000  | 7134.690000  | 1064.036896  |
| 50%   | 9528.230000  | 9571.040000  | 9483.780000  | 9527.750000  | 1660.792103  |
| 75%   | 32596.490000 | 32893.930000 | 32342.910000 | 32604.720000 | 2704.863840  |
| max   | 68635.120000 | 69000.000000 | 68451.190000 | 68633.690000 | 47255.762685 |

```
1 Data_A.info()
```

```

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 35637 entries, 2018-01-01 00:00:00 to 2022-01-29 19:00:00
Data columns (total 5 columns):
#   Column  Non-Null Count  Dtype
---  -
0   Open    35637 non-null     float64
1   High    35637 non-null     float64
2   Low     35637 non-null     float64
3   Close   35637 non-null     float64
4   Volume  35637 non-null     float64
dtypes: float64(5)
memory usage: 1.6 MB

```

## 6 Training data, Testing data :

we divide the data into two parts: Training data 80% and Testing data 20%.

```
1 df_T_T = Data_A.loc[:,['Close']]
2 df_T_T
```

|                     | Close    |
|---------------------|----------|
| Time                |          |
| 2018-01-01 00:00:00 | 13529.01 |
| 2018-01-01 01:00:00 | 13203.06 |
| 2018-01-01 02:00:00 | 13330.18 |
| 2018-01-01 03:00:00 | 13410.03 |
| 2018-01-01 04:00:00 | 13601.01 |
| ...                 | ...      |
| 2022-01-29 15:00:00 | 37566.82 |
| 2022-01-29 16:00:00 | 37538.19 |
| 2022-01-29 17:00:00 | 37646.31 |
| 2022-01-29 18:00:00 | 37806.23 |
| 2022-01-29 19:00:00 | 37891.83 |

35637 rows × 1 columns

```
1 test=df_T_T.drop(train.index)
2 test
```

|                     | Close    |
|---------------------|----------|
| Time                |          |
| 2018-01-01 03:00:00 | 13410.03 |
| 2018-01-01 04:00:00 | 13601.01 |
| 2018-01-01 05:00:00 | 13558.99 |
| 2018-01-01 15:00:00 | 13247.00 |
| 2018-01-01 17:00:00 | 13022.00 |
| ...                 | ...      |
| 2022-01-27 07:00:00 | 36173.87 |
| 2022-01-27 08:00:00 | 36450.00 |
| 2022-01-27 09:00:00 | 36525.85 |
| 2022-01-27 17:00:00 | 36282.03 |
| 2022-01-27 20:00:00 | 35583.50 |

7119 rows × 1 columns

```
1 train=df_T_T.sample(frac=0.8,random_state=200)
2 train
```

|                     | Close    |
|---------------------|----------|
| Time                |          |
| 2021-11-11 07:00:00 | 65228.40 |
| 2018-05-10 08:00:00 | 9354.08  |
| 2020-05-06 09:00:00 | 9082.59  |
| 2019-05-16 04:00:00 | 7978.83  |
| 2020-04-01 14:00:00 | 6251.13  |
| ...                 | ...      |
| 2019-04-13 07:00:00 | 5083.45  |
| 2020-02-13 18:00:00 | 10208.44 |
| 2021-02-03 05:00:00 | 36480.12 |
| 2021-07-13 04:00:00 | 33118.22 |
| 2020-08-17 00:00:00 | 11845.31 |

28510 rows × 1 columns

## 7 Data visualization :

We analyze Close price of BTC over time from 01/01/2018 to 29/01/2022.  
Let's use matplotlib to visualise the series.

```
1 fig,ax = plt.subplots(figsize=(17,6))
2 rolling_avg = 1
3 ax.plot(Data_A.index,Data_A['Close'].rolling(window=rolling_avg).mean(),color='blue',label='Close')
4
5 ax.figure.legend()
6 sns.set()
7 sns.set_style("darkgrid")
8 ax.set_xlabel('Time')
9 ax.set_ylabel('BTC_Close')
```

Text(0, 0.5, 'BTC\_Close')



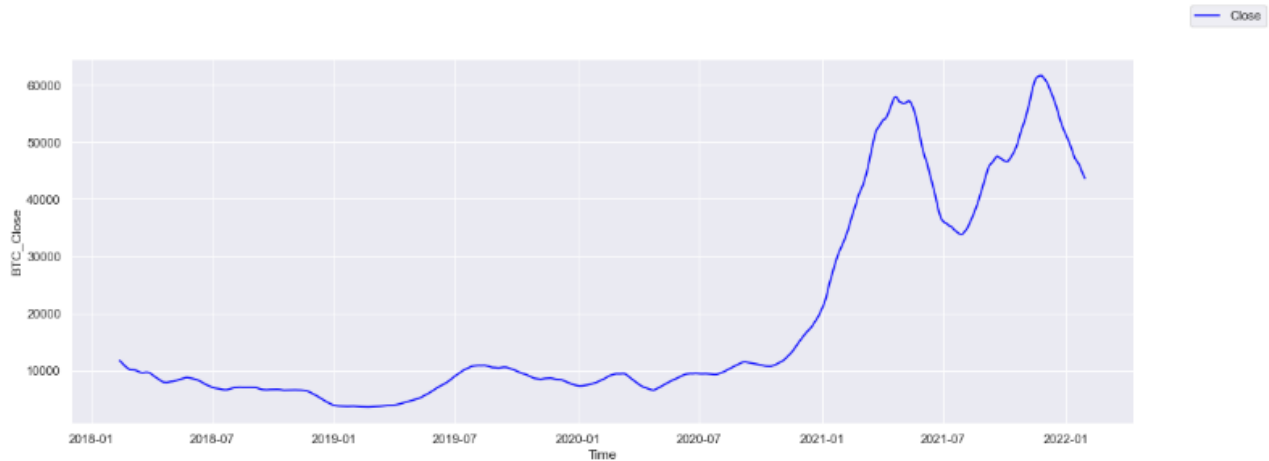


```

1 fig,ax = plt.subplots(figsize=(17,6))
2 rolling_avg = 1000
3 ax.plot(Data_A.index,Data_A['Close'].rolling(window=rolling_avg).mean(),color='blue',label='Close')
4
5 ax.figure.legend()
6 sns.set()
7 sns.set_style("darkgrid")
8 ax.set_xlabel('Time')
9 ax.set_ylabel('BTC_Close')

```

Text(0, 0.5, 'BTC\_Close')

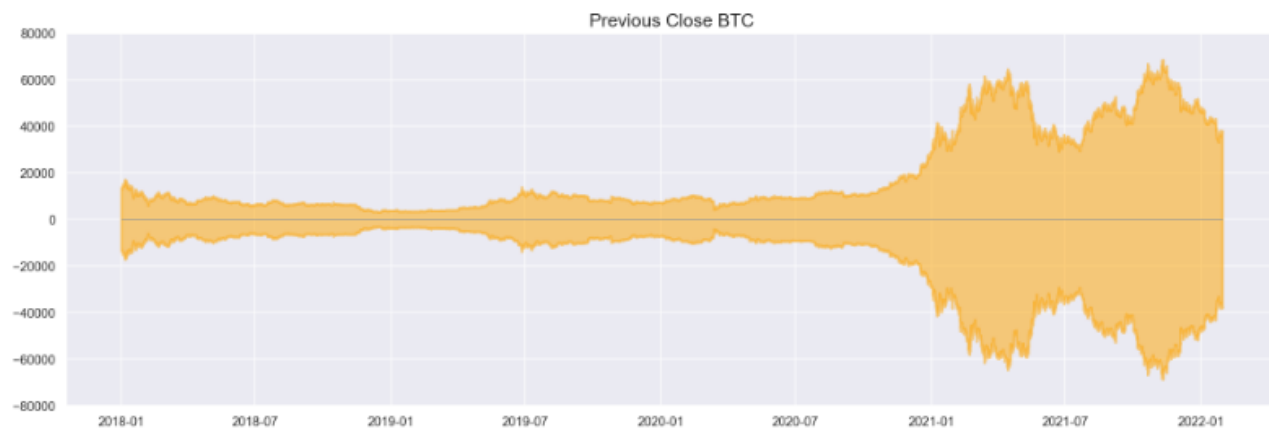


Since all values are positive, you can show this on both sides of the Y axis to emphasize the growth.

```

1 x = Data_A.index
2 y1 = Data_A['Close'].values
3
4 fig, ax = plt.subplots(1, 1, figsize=(19,6), dpi= 120)
5 plt.fill_between(x, y1=y1, y2=-y1, alpha=0.5, linewidth=2, color='orange')
6 plt.ylim(-80000, 80000)
7 plt.title('Previous Close BTC', fontsize=16)
8 plt.hlines(y=0, xmin=np.min(x), xmax=np.max(x), linewidth=.5)
9 plt.show()

```



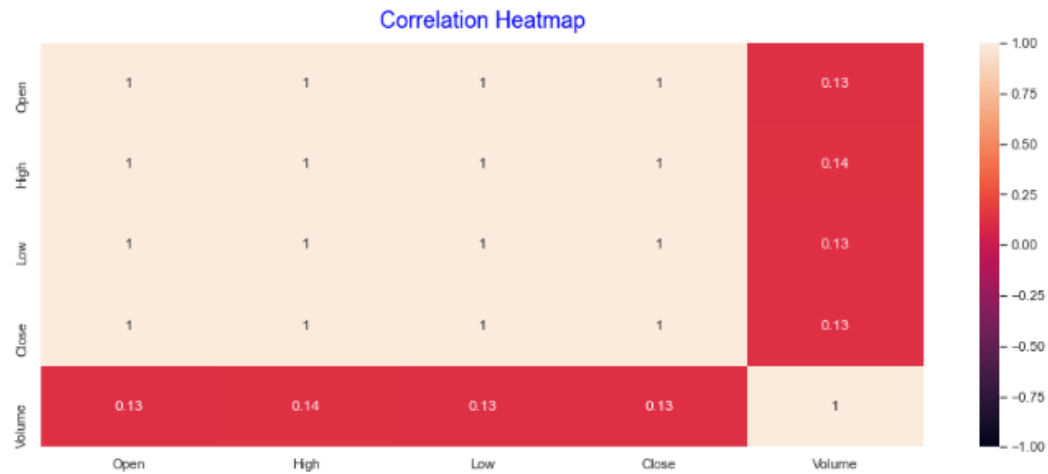
A correlation heatmap is a heatmap that shows a 2D correlation matrix between two discrete dimensions, using colored cells to represent data from usually a monochromatic scale. The values of the first dimension appear as the rows of the table while of the second dimension as a column.

```
1 Data_A.corr()
```

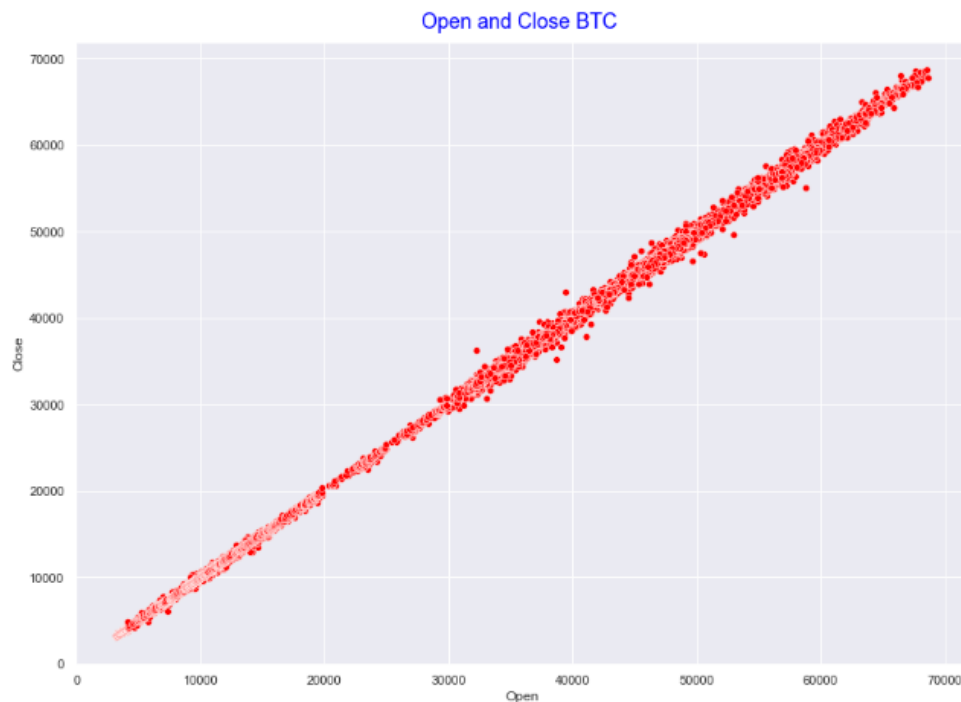
|        | Open     | High     | Low      | Close    | Volume   |
|--------|----------|----------|----------|----------|----------|
| Open   | 1.000000 | 0.999980 | 0.999938 | 0.999928 | 0.132344 |
| High   | 0.999980 | 1.000000 | 0.999907 | 0.999980 | 0.135888 |
| Low    | 0.999938 | 0.999907 | 1.000000 | 0.999950 | 0.127203 |
| Close  | 0.999928 | 0.999980 | 0.999950 | 1.000000 | 0.131835 |
| Volume | 0.132344 | 0.135888 | 0.127203 | 0.131835 | 1.000000 |

```
1 plt.figure(figsize=(16, 6))
2 heatmap = sns.heatmap(Data_A.corr(), vmin=-1, vmax=1, annot=True)
3 heatmap.set_title('Correlation Heatmap', fontdict={'fontsize':18}, pad=12, color='blue')
```

Text(0.5, 1.0, 'Correlation Heatmap')



```
1 plt.figure(figsize=(14, 10))
2 ax = sns.scatterplot(x="Open", y="Close", data=Data_A, color='red')
3 plt.title('Open and Close BTC ', fontdict={'fontsize':18}, pad=12, color='blue');
```



## 8 Trading bot strategy :

Many traders use moving averages to place their trades.

The issue with this, as for almost all indicators, is if you want to trade on a short time frame it is very likely that you will miss the exact right time to execute the trade.

In our situation we will use Moving Average 200 and 600.

```

1 Data_A['SMA200'] = ta.trend.sma_indicator(Data_A['Close'],200)
2 Data_A['SMA600'] = ta.trend.sma_indicator(Data_A['Close'],600)
3 Data_A

```

|                     | Open     | High     | Low      | Close    | Volume      | SMA200      | SMA600       |
|---------------------|----------|----------|----------|----------|-------------|-------------|--------------|
| Time                |          |          |          |          |             |             |              |
| 2018-01-01 00:00:00 | 13715.65 | 13715.65 | 13400.01 | 13529.01 | 443.356199  | NaN         | NaN          |
| 2018-01-01 01:00:00 | 13528.99 | 13595.89 | 13155.38 | 13203.06 | 383.697006  | NaN         | NaN          |
| 2018-01-01 02:00:00 | 13203.00 | 13418.43 | 13200.00 | 13330.18 | 429.064572  | NaN         | NaN          |
| 2018-01-01 03:00:00 | 13330.26 | 13611.27 | 13290.00 | 13410.03 | 420.087030  | NaN         | NaN          |
| 2018-01-01 04:00:00 | 13434.98 | 13623.29 | 13322.15 | 13601.01 | 340.807329  | NaN         | NaN          |
| ...                 | ...      | ...      | ...      | ...      | ...         | ...         | ...          |
| 2022-01-29 15:00:00 | 37620.04 | 37735.67 | 37268.44 | 37566.82 | 1881.672090 | 36475.93270 | 40596.781033 |
| 2022-01-29 16:00:00 | 37567.98 | 37803.57 | 37317.37 | 37538.19 | 1102.742180 | 36467.83550 | 40581.379233 |
| 2022-01-29 17:00:00 | 37538.19 | 37720.77 | 37532.15 | 37646.31 | 784.471840  | 36461.26820 | 40566.531400 |
| 2022-01-29 18:00:00 | 37646.31 | 37904.19 | 37556.56 | 37806.23 | 772.068930  | 36455.46975 | 40553.291783 |
| 2022-01-29 19:00:00 | 37806.24 | 37940.19 | 37781.49 | 37891.83 | 305.896100  | 36450.43800 | 40539.690900 |

35637 rows x 7 columns

Visualize SMA200 and SMA600 graphics :

```

1 fig,ax = plt.subplots(figsize=(17,6))
2 rolling_avg = 1
3 ax.plot(Data_A.index,Data_A['Close'].rolling(window=rolling_avg).mean(),color='blue',label='Close')
4 ax.plot(Data_A.index,Data_A['SMA200'].rolling(window=rolling_avg).mean(),color='red',label='Moving_Average_200')
5 ax.plot(Data_A.index,Data_A['SMA600'].rolling(window=rolling_avg).mean(),color='green',label='Moving_Average_600')
6
7 ax.figure.legend()
8 sns.set()
9 sns.set_style("darkgrid")
10 ax.set_xlabel('Time')
11 ax.set_ylabel('BTC_Close')

```

Text(0, 0.5, 'BTC\_Close')



Define the trading bot strategy :

We start with 100000 USDT and 0 BTC.

Firstly, we iterate through our table line by line, secondly we test if the graph of SMA200 is above of SMA600 graph and we have a balance usdt, then we buy BTC.  
If the graph of SMA600 is above of SMA200 graph and we have a balance BTC, then we sell BTC.

```

1 from termcolor import colored
2 # starting balance
3 usdt = 100000 # we start with 100000 usdt
4 btc = 0 # we start with 0 btc
5 li = Data_A.first_valid_index() # recover the closing price
6
7 for index, row in Data_A.iterrows(): # iterate through our table line by line
8     # Buy BTC
9     if Data_A['SMA200'][li] > Data_A['SMA600'][li] and usdt > 0:
10         btc = usdt/Data_A['Close'][index]
11         usdt = 0
12         print (colored("Buy BTC at", 'red'),Data_A['Close'][index], '$ the', index)
13     # Sell BTC
14     if Data_A['SMA200'][li] < Data_A['SMA600'][li] and btc > 0:
15         usdt = btc * Data_A['Close'][index]
16         btc = 0
17         print (colored("Sell BTC at", 'blue'),Data_A['Close'][index], '$ the', index)
18     li = index

```

```

Buy BTC at 10940.0 $ the 2018-02-19 17:00:00
Sell BTC at 8704.0 $ the 2018-03-11 07:00:00
Buy BTC at 8018.98 $ the 2018-04-16 21:00:00
Sell BTC at 8665.97 $ the 2018-05-14 00:00:00
Buy BTC at 6584.36 $ the 2018-07-07 10:00:00
Sell BTC at 7087.99 $ the 2018-08-07 11:00:00
Buy BTC at 6902.74 $ the 2018-08-28 04:00:00
Sell BTC at 6355.0 $ the 2018-09-11 03:00:00
Buy BTC at 6669.36 $ the 2018-09-27 19:00:00
Sell BTC at 6309.02 $ the 2018-10-12 22:00:00
Buy BTC at 6573.58 $ the 2018-10-20 10:00:00
Sell BTC at 6491.85 $ the 2018-10-28 10:00:00
Buy BTC at 4125.4 $ the 2018-12-24 04:00:00
Sell BTC at 3588.24 $ the 2019-01-13 02:00:00
Buy BTC at 3626.58 $ the 2019-02-12 04:00:00
Sell BTC at 7660.98 $ the 2019-06-09 18:00:00
Buy BTC at 9105.21 $ the 2019-06-17 06:00:00
Sell BTC at 9431.52 $ the 2019-07-17 04:00:00
Buy BTC at 11692.8 $ the 2019-08-06 01:00:00
Sell BTC at 10886.0 $ the 2019-08-20 01:00:00
Buy BTC at 10541.4 $ the 2019-09-08 03:00:00
Sell BTC at 10028.87 $ the 2019-09-22 23:00:00
Buy BTC at 9660.0 $ the 2019-10-27 17:00:00
Sell BTC at 8564.64 $ the 2019-11-15 05:00:00
Buy BTC at 7206.3 $ the 2019-12-27 00:00:00
Sell BTC at 9601.09 $ the 2020-02-24 19:00:00
Buy BTC at 6368.68 $ the 2020-04-01 03:00:00
Sell BTC at 8859.62 $ the 2020-05-27 04:00:00
Buy BTC at 10115.56 $ the 2020-06-02 11:00:00
Sell BTC at 9403.01 $ the 2020-06-18 13:00:00
Buy BTC at 9290.23 $ the 2020-07-13 18:00:00
Sell BTC at 9117.69 $ the 2020-07-17 10:00:00
Buy BTC at 9164.57 $ the 2020-07-18 16:00:00

```

```

Buy BTC at 9164.57 $ the 2020-07-18 16:00:00
Sell BTC at 11400.12 $ the 2020-08-27 01:00:00
Buy BTC at 10417.22 $ the 2020-09-21 23:00:00
Sell BTC at 10571.36 $ the 2020-10-06 19:00:00
Buy BTC at 11363.33 $ the 2020-10-10 18:00:00
Sell BTC at 32801.15 $ the 2021-01-24 07:00:00
Buy BTC at 37366.02 $ the 2021-02-05 03:00:00
Sell BTC at 47789.87 $ the 2021-03-02 22:00:00
Buy BTC at 54632.79 $ the 2021-03-10 08:00:00
Sell BTC at 57854.32 $ the 2021-03-29 11:00:00
Buy BTC at 58670.64 $ the 2021-04-02 20:00:00
Sell BTC at 54744.24 $ the 2021-04-22 14:00:00
Buy BTC at 57584.01 $ the 2021-05-07 19:00:00
Sell BTC at 47964.69 $ the 2021-05-15 22:00:00
Buy BTC at 39952.7 $ the 2021-06-16 09:00:00
Sell BTC at 33982.26 $ the 2021-06-23 07:00:00
Buy BTC at 38242.96 $ the 2021-07-27 14:00:00
Sell BTC at 45150.33 $ the 2021-09-12 06:00:00
Buy BTC at 49476.51 $ the 2021-10-05 07:00:00
Sell BTC at 56989.77 $ the 2021-11-19 09:00:00
Buy BTC at 51450.57 $ the 2021-12-27 15:00:00
Sell BTC at 46999.98 $ the 2022-01-03 02:00:00

```

Using the strategy we win 715353.249818794 USDT, if we start with 100000 USDT

```

1 Final_usdt = usdt + btc * Data_A['Close'].iloc[-1] # calculate the sum of usdt after purchases and sales
2 print("Final usdt", Final_usdt, 'USDT')

```

Final usdt 815353.249818794 USDT

If we have bought BTC with 100000 usdt since 01/01/2018 how much we will earn if we sell them with the current price of BTC.(280078.3649357935 USDT)

```

1 # if we have bought BTC with 100000 usdt since 01/01/2018 how much we will earn if we sell them with the current price of B
2 print("Buy and hold result", (100000 / Data_A['Close'].iloc[0]) * Data_A['Close'].iloc[-1], 'USDT')

```

Buy and hold result 280078.3649357935 USDT

Here is the GitHub repo link:

[https://github.com/Boujbair/Trading\\_Bot\\_Python](https://github.com/Boujbair/Trading_Bot_Python)

# END