

```
In [44]: # !pip3 install yfinance
```

```
In [ ]: # Assignment 1 for MFE407
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```

## Assignment 1

The goal of this assignment is to write a simple back-testing algorithm. It is not a test for you to come up with new strategies. I have laid out all the instructions. ChatGPT is the best friend in coding. P.S. if this is too easy, happy to give you more difficult tasks:-)

```
In [1]: import pandas as pd
import numpy as np
import yfinance as yf
import requests
import matplotlib.pyplot as plt
from random import sample
```

```
/Users/ashutosh/Library/Python/3.9/lib/python/site-packages/urllib3/
__init__.py:34: NotOpenSSLWarning: urllib3 v2 only supports OpenSSL
1.1.1+, currently the 'ssl' module is compiled with 'LibreSSL 2.8.
3'. See: https://github.com/urllib3/urllib3/issues/3020 (https://github.com/urllib3/urllib3/issues/3020)
warnings.warn(
```

**Read in NYSE.txt into a DataFrame, which includes stock tickers and names**

```
In [2]: #read the stock tickers and names into a DataFrame
df = pd.read_csv('~Downloads/NYSE.txt', sep='\t', header=0) ## please
#Create a list contains all tickers: iterate through stock list and ap
all_tickers_list = df.Symbol.to_list()
```

**Using one stock as an example to construct buy-sell strategy**

Step 1: Download data and calculate necessary summary statistics

```
In [3]: #download stock data for the FIRST stock and place in DataFrame (using
first_stock_data = yf.download(all_tickers_list[0])
print(type(first_stock_data))
```

```
[*****100%*****] 1 of 1 completed
<class 'pandas.core.frame.DataFrame'>
```

```
In [4]: #create column to hold our 90 day rolling standard deviation
# print(first_stock_data.head(5))
first_stock_data['90_SD'] = first_stock_data['Adj Close'].rolling(90).
# print(first_stock_data.head(-10))
```

```
In [5]: #create a column to hold our 20 day moving average
first_stock_data['20_MA'] = first_stock_data['Adj Close'].rolling(20).
```

Step 2: Create "BUY" signal according to two conditions

```
In [6]: #BUY Condition 1: create a column which holds a TRUE value if the gap
first_stock_data['Cond_1'] = np.where((first_stock_data['Low'].shift(
print("Condition 1 was satisfied {} times.".format(first_stock_data['C
```

Condition 1 was satisfied 28 times.

```
In [7]: #BUY Condition 2: create a column which holds a TRUE value if the open
first_stock_data['Cond_2'] = np.where(first_stock_data['Open'] > first
print("Condition 2 was satisfied {} times.".format(first_stock_data['C
```

Condition 2 was satisfied 5522 times.

```
In [8]: # "BUY" signal: create a column that holds a TRUE value if both buy cri
first_stock_data['signal'] = np.where((first_stock_data['Cond_1'] ==
print("Buy signal {} times.".format(first_stock_data['signal'].sum()))
```

Buy signal 15 times.

Step 3: Create "SELL" signal according to two conditions

```
In [9]: #SELL Condition 1: create a column which holds a TRUE value if the gap
first_stock_data['Sell_Cond_1'] = np.where((first_stock_data['Open']
print("Sell condition 1 was satisfied {} times.".format(first_stock_da
```

Sell condition 1 was satisfied 32 times.

```
In [10]: #SELL Condition 2: create a column which holds a TRUE value if the open
first_stock_data['Sell_Cond_2'] = np.where(first_stock_data['Open'] <
print("Sell condition 2 was satisfied {} times.".format(first_stock_da
```

Sell condition 2 was satisfied 539 times.

```
In [11]: # "SELL" signal: create a column that holds a TRUE value if both sell c
first_stock_data['sell_signal'] = np.where((first_stock_data['Sell_Co
print("Sell signal {} times.".format(first_stock_data['sell_signal'].s
```

Sell signal 0 times.

Step 4: Show the results of Trading Algo

```
In [12]: #calculate daily % return series for stock
first_stock_data['daily_returns'] = first_stock_data['Adj Close'].pct_
```

**A Challenge: create an indicator which equals to 1 if you hold one share and to -1 if you sell one share.**

As a example, if at day 10, the algo tells you to buy a share, and at day 15, the algo tells you to sell a share, then you do not hold any share from day 1 to day 10, and hold one share from day 11 to day 15, and short-sell a share from day 16. How can you construct such an indicator in the dataset?

Step 1: Calculate shares of holding for each day

```
In [13]: # Create a column contains your holding of shares according to "BUY" and "SELL" signals
# first_stock_data['Share'] = np.where(first_stock_data["signal"]=="BUY", 1, 0)
buy_indices = first_stock_data[first_stock_data['signal']=='BUY'].index
sell_indices = first_stock_data[first_stock_data['sell_signal']=='SELL'].index
indicator = []
sig_to_append = 0
for idx in first_stock_data.index:
    if idx in buy_indices:
        sig_to_append = 1
    elif idx in sell_indices:
        sig_to_append = -1
    else:
        pass
    indicator.append(sig_to_append)
first_stock_data['Share'] = indicator
```

```
In [14]: # Verify. The new column is "Share" (verification is an important step)
first_stock_data[['daily_returns', 'signal', 'sell_signal', 'Share']]
```

```
Out[14]:
```

	daily_returns	signal	sell_signal	Share
2016-08-25	0.011992	False	False	1
2016-08-26	0.001270	False	False	1
2016-08-29	0.006340	False	False	1
2016-08-30	-0.010080	False	False	1

Date				
2016-08-25	0.011992	False	False	1
2016-08-26	0.001270	False	False	1
2016-08-29	0.006340	False	False	1
2016-08-30	-0.010080	False	False	1

```
In [15]: print(first_stock_data["Share"].value_counts())
```

```
Share
1    4828
0    1252
Name: count, dtype: int64
```

Step 2: Calculate strategic returns according to your holdings of shares and daily stock returns

```
In [16]: #create a strategy return series by using the daily stock returns multiplied by the share holding
first_stock_data['Rets'] = first_stock_data['daily_returns'] * first_stock_data['Share']
```

```
In [17]: # Verify again
first_stock_data[['daily_returns', 'signal', 'sell_signal', 'Share', 'Rets', 'Date']]
```

Out[17]:

	daily_returns	signal	sell_signal	Share	Rets
Date					
2016-08-25	0.011992	False	False	1	0.011992
2016-08-26	0.001270	False	False	1	0.001270
2016-08-29	0.006340	False	False	1	0.006340
2016-08-30	-0.010080	False	False	1	-0.010080

Good Job! Apply the strategy to all stocks in stocks\_list

```
In [18]: #create empty list to hold our return series DataFrame for each stock
frames = []
# lets randomly select 200 stocks from the list
sample_stocks = sample(all_tickers_list, 200)
for stock in sample_stocks:
    try:
        ### Copy Paste Previous Code For One Stock ###
        first_stock_data = yf.download(stock)
        first_stock_data['90_SD'] = first_stock_data['Adj Close'].rolling(90).std()
        first_stock_data['20_MA'] = first_stock_data['Adj Close'].rolling(20).mean()
        first_stock_data['Cond_1'] = np.where((first_stock_data['Low'] < first_stock_data['20_MA']) && (first_stock_data['90_SD'] > 0.02), 1, 0)
        first_stock_data['Cond_2'] = np.where((first_stock_data['Open'] > first_stock_data['High']) && (first_stock_data['90_SD'] > 0.02), 1, 0)
        first_stock_data['signal'] = np.where((first_stock_data['Cond_1'] == 1 && first_stock_data['Cond_2'] == 0), 1, 0)
        first_stock_data['Sell_Cond_1'] = np.where((first_stock_data['Cond_1'] == 0 && first_stock_data['Cond_2'] == 1), 1, 0)
        first_stock_data['Sell_Cond_2'] = np.where((first_stock_data['Cond_1'] == 1 && first_stock_data['Cond_2'] == 1), 1, 0)
        first_stock_data['sell_signal'] = np.where((first_stock_data['Sell_Cond_1'] == 1 && first_stock_data['Sell_Cond_2'] == 0), 1, 0)
        first_stock_data['daily_returns'] = first_stock_data['Adj Close'].pct_change()
        buy_indices = first_stock_data[first_stock_data['signal'] == True]
        sell_indices = first_stock_data[first_stock_data['sell_signal'] == True]
        indicator = []
        sig_to_append = 0
        for idx in first_stock_data.index:
            if idx in buy_indices:
                sig_to_append = 1
            elif idx in sell_indices:
                sig_to_append = -1
            else:
                pass
            indicator.append(sig_to_append)
        first_stock_data['Share'] = indicator
        first_stock_data['Rets'] = first_stock_data['daily_returns'] * first_stock_data['Share']
        #append the strategy return series to our list
        frames.append(first_stock_data['Rets'])
    except:
        pass
```

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

1 Failed download:

```
['VR-A']: Exception('%ticker%: No timezone found, symbol may be delisted')
```

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

1 Failed download:

```
['HFC']: Exception('%ticker%: No timezone found, symbol may be delisted')
```

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

1 Failed download:

```
['OAK-A']: Exception('%ticker%: No timezone found, symbol may be delisted')
```

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

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

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

## Plot cummulative returns of strategy

Step 1: Calculate cumulative returns

```
In [19]: #concatenate the individual DataFrames held in our list- and do it also
new_df = pd.concat(frames, ignore_index=False, axis=1, keys=sample_stocks)
```

```
/var/folders/44/r2pt84y14r968g9_vxmvr1xh0000gn/T/ipykernel_64955/3577510563.py:2: FutureWarning: The behavior of pd.concat with len(keys) != len(objs) is deprecated. In a future version this will raise instead of truncating to the smaller of the two sequences
  new_df = pd.concat(frames, ignore_index=False, axis=1, keys=sample_stocks)
```

```
In [20]: #create a column to hold the sum of all the individual daily strategy
new_df['daily_returns_sum'] = new_df.sum(axis=1, numeric_only=True)
```

```
In [22]: #fill 'NaNs' with zeros to allow our "count" function below to work properly
print(new_df.daily_returns_sum.sum())
```

```
297.00562770480803
```

```
In [23]: #create a column that hold the count of the number of stocks that were added
#we minus one from it so that we dont count the "Total" column we added
new_df['count'] = new_df.count(axis=1)
```

```
In [24]: print(new_df.head())
```

	BGH	VR-A	HFC	OAK-A	NKE	IHD	DDR-K	UMH	ASH	CP
... NGVC \										
Date										
...										
1962-01-02	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
... NaN										
1962-01-03	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
... NaN										
1962-01-04	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
... NaN										
1962-01-05	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
... NaN										
1962-01-08	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
... NaN										

	VSH	CHH	BDC-B	SBH	PSA-Y	DNOW	GHY	daily_returns_sum
m count								
Date								
1962-01-02	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.
0 1								
1962-01-03	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.
0 2								
1962-01-04	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.
0 2								
1962-01-05	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.
0 2								
1962-01-08	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.
0 2								

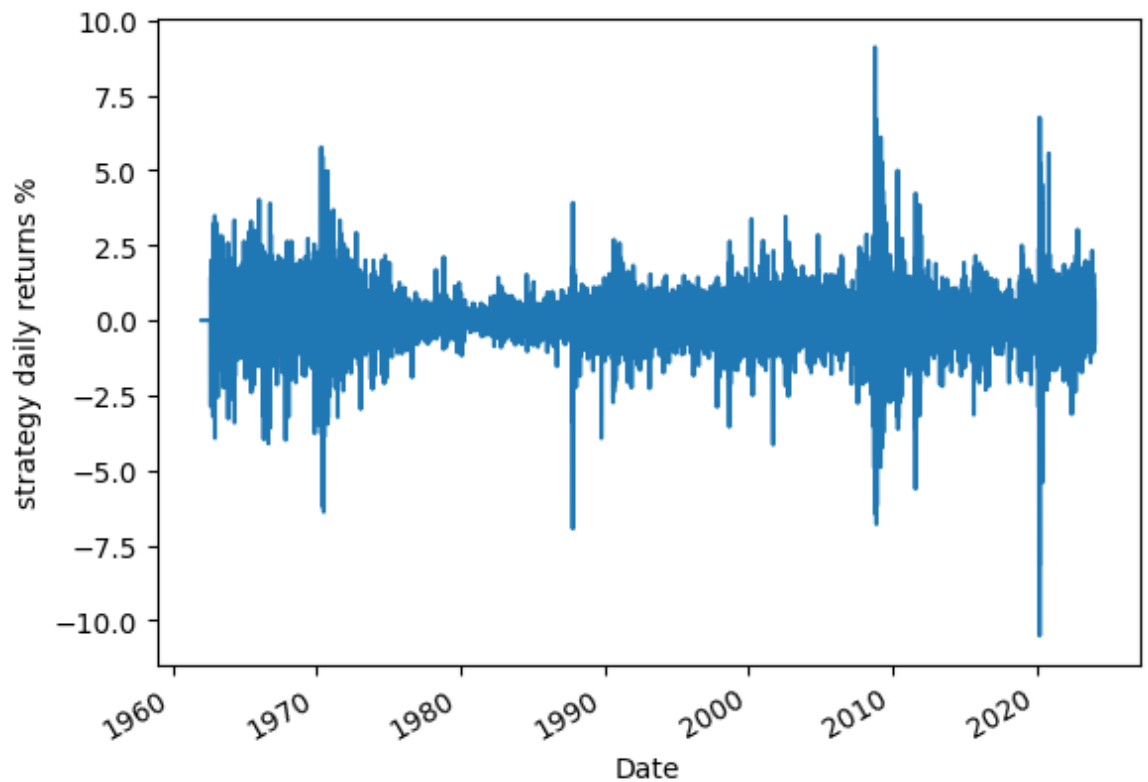
```
[5 rows x 122 columns]
```

```
In [26]: #create a column that divides the "total" strategy return each day by  
new_df['daily_equally_wt_returns'] = new_df['daily_returns_sum'] / new
```

Step 2: Plot

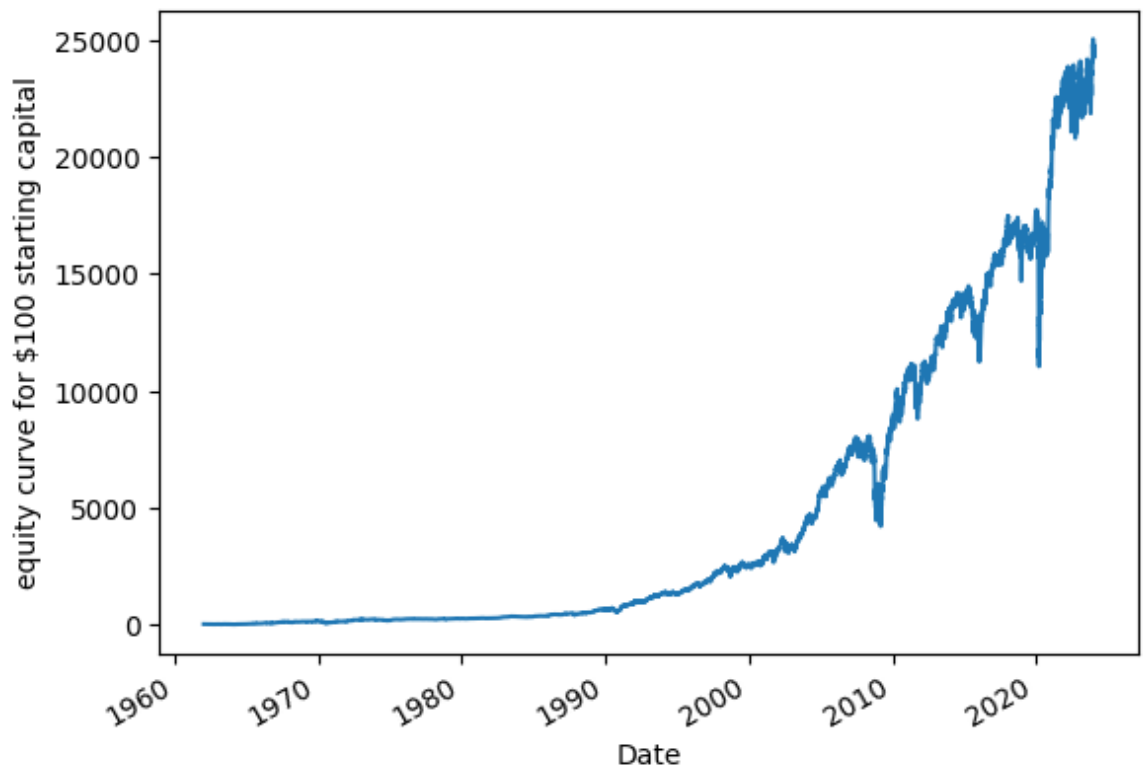
```
In [35]: #plot the strategy returns  
(new_df['daily_equally_wt_returns'] * 100).plot(ylabel="strategy daily
```

```
Out[35]: <Axes: xlabel='Date', ylabel='strategy daily returns %'>
```



```
In [42]: new_df['daily_cum_returns'] = (1 + new_df['daily_equally_wt_returns'])
         (new_df['daily_cum_returns'] * 100).plot(ylabel="equity curve for $100
```

```
Out[42]: <Axes: xlabel='Date', ylabel='equity curve for $100 starting capita
l'>
```



Step 3: Calculate annual returns and sharpe ratio for your strategy

```
In [38]: # Annual Return
annual_returns = new_df['daily_equally_wt_returns'].mean() * 255
print("Annual returns of the strategy are {:.2f} %".format(annual_retu
```

Annual returns of the strategy are 9.81 %

```
In [39]: # Sharpe Ratio (risk free element excluded for simplicity)
def sharpe(returns):
    ann = returns.mean() * 255
    std_dev_ann = returns.std() * np.sqrt(255)
    return ann / std_dev_ann

strat_sharpe = sharpe(new_df['daily_equally_wt_returns'])
print("Strategy sharpe ratio is {:.2f}".format(strat_sharpe))
```

Strategy sharpe ratio is 0.77

**Congraduation!!! You have written the first backtesting code yourself!!!**