

## Loading Libraries and importing files

```
In [50]: # Importing files and Loading Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib.patches as mpatches
import matplotlib.dates as mdates
import os
from matplotlib.pyplot import figure
import warnings
warnings.filterwarnings('ignore')
```

```
In [51]: # Function to read all CSV files from directory and return them in one data
         # frame
def read_dir(dir):
    files = pd.concat((pd.read_csv(dir + fname).assign(File = fname) for fn
ame in os.listdir(dir)))
    files["File"] = [file.replace(".csv", "") for file in files["File"]]
    return(files)

# Utilized code from: Topic 9 Lab questions
```

```
In [52]: # Loading in stocks
stocks = read_dir('/Users/siam/Desktop/Coding Assessment/Given_Files/Stock
s/')
```

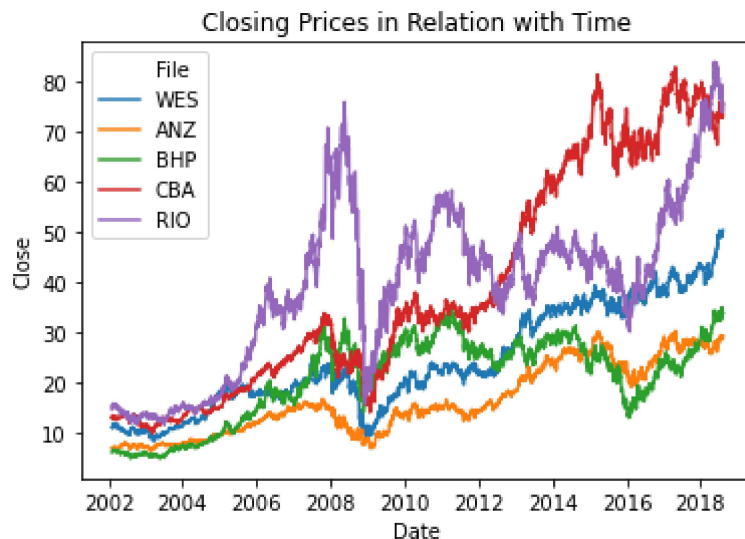
## Task 1

```
In [53]: # Converting 'Date' to date-time object
stocks.Date = pd.to_datetime(stocks.Date)
```

```
In [55]: # Creating line plot that showcases the relationship between closing prices
          and time
          filtered_stocks = stocks[stocks.File.isin(["ANZ", "BHP", "CBA", "RIO", "WES"])]

          sns.lineplot(x = "Date", y = "Close", hue = "File", data = filtered_stocks)
          plt.title("Closing Prices in Relation with Time")
          plt.show()

          # Code utilized from: Lab Questions Topic 9 Task 5
```



```
In [56]: # Finding correlation of closing stock prices
          close_corr = stocks[['Date', 'Close', 'File']]

          close_corr_pivot = close_corr.pivot('Date', 'File', 'Close').reset_index()

          close_corr = close_corr_pivot.corr(method='pearson')

          close_corr.head(10)

          # Code utilized from: https://www.interviewqs.com/blog/py_stock_correlation
```

Out[56]:

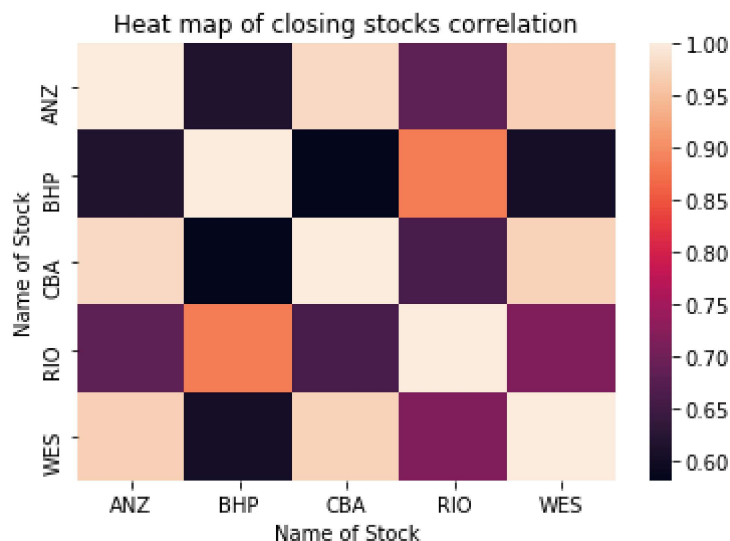
File	ANZ	BHP	CBA	RIO	WES
File					
ANZ	1.000000	0.614569	0.980258	0.683818	0.969762
BHP	0.614569	1.000000	0.581494	0.884080	0.604126
CBA	0.980258	0.581494	1.000000	0.660790	0.973428
RIO	0.683818	0.884080	0.660790	1.000000	0.716878
WES	0.969762	0.604126	0.973428	0.716878	1.000000

The table above is a representation of the correlation between the stocks, ANZ, BHP, CBA, RIO and WES. The stock correlation identifies the relationship between these stocks and the price movements. It is evident that the column 'ANZ' and the row 'ANZ' represent a positive 1 which is a rare occurrence of perfect positive correlation, including the columns and rows, 'BHP', 'CBA', 'RIO', and 'WES'.

The remaining data in the table above are between +1 and -1. Therefore, the investor should choose assets that have low correlation with each other in order to reduce the risk within the portfolio.

```
In [57]: # Heat map of 'stocks.corr'
sns.heatmap(close_corr)
plt.title("Heat map of closing stocks correlation")
plt.xlabel("Name of Stock")
plt.ylabel("Name of Stock")
plt.show()

# Utilized code from: https://seaborn.pydata.org/generated/seaborn.heatmap.html
```



The heatmap graph above representing the closing correlation between the stocks illustrates that the rare occurrences of the perfect positive correlations is a diagonal line from the top towards the bottom. Furthermore, the graph showcases the remaining data points of the correlation of the stocks.

## Task 2

### Moving Averages

```
In [58]: # Finding moving average of 5 and 20
filtered_stocks['MAV5'] = filtered_stocks['Open'].rolling(window=5).mean()
filtered_stocks['MAV20'] = filtered_stocks['Open'].rolling(window=20).mean()

# Code utilized from: https://www.learndatasci.com/tutorials/python-finance-part-3-moving-average-trading-strategy/
```

## Last 5 values of each stock

```
In [59]: # Sorting each stock into a seperate dataframe
wes_t5 = filtered_stocks[filtered_stocks['File'].str.contains("WES")]
anz_t5 = filtered_stocks[filtered_stocks['File'].str.contains("ANZ")]
bhp_t5 = filtered_stocks[filtered_stocks['File'].str.contains("BHP")]
cba_t5 = filtered_stocks[filtered_stocks['File'].str.contains("CBA")]
rio_t5 = filtered_stocks[filtered_stocks['File'].str.contains("RIO")]
```

```
In [60]: # For WES
wes_t5[['MAV5', 'MAV20']].tail(5)
```

Out[60]:

	MAV5	MAV20
4224	49.764	49.4980
4225	49.864	49.5005
4226	49.972	49.5400
4227	50.022	49.5730
4228	50.062	49.6090

```
In [61]: # For ANZ
anz_t5[['MAV5', 'MAV20']].tail(5)
```

Out[61]:

	MAV5	MAV20
4231	29.156	29.1310
4232	29.074	29.1175
4233	29.054	29.1450
4234	29.002	29.1555
4235	29.050	29.1595

```
In [62]: # For BHP
bhp_t5[['MAV5', 'MAV20']].tail(5)
```

Out[62]:

	MAV5	MAV20
4235	34.450	33.7615
4236	34.232	33.7335
4237	34.062	33.7350
4238	33.968	33.7730
4239	33.992	33.7970

```
In [63]: # For CBA
cba_t5[['MAV5', 'MAV20']].tail(5)
```

```
Out[63]:
```

	MAV5	MAV20
4229	74.122	74.9180
4230	73.802	74.7830
4231	73.762	74.7795
4232	73.998	74.8135
4233	74.342	74.8155

```
In [64]: # For RIO
rio_t5[['MAV5', 'MAV20']].tail(5)
```

```
Out[64]:
```

	MAV5	MAV20
4234	77.04448	77.705415
4235	76.19038	77.447830
4236	75.29754	77.250765
4237	74.87042	77.162865
4238	75.02888	77.074650

## Graphing each stock by the opening price and moving averages

```
In [65]: # Placing data into a new dataframe that accomodates the date 2008 to 2010
filtered_stocks_date = filtered_stocks

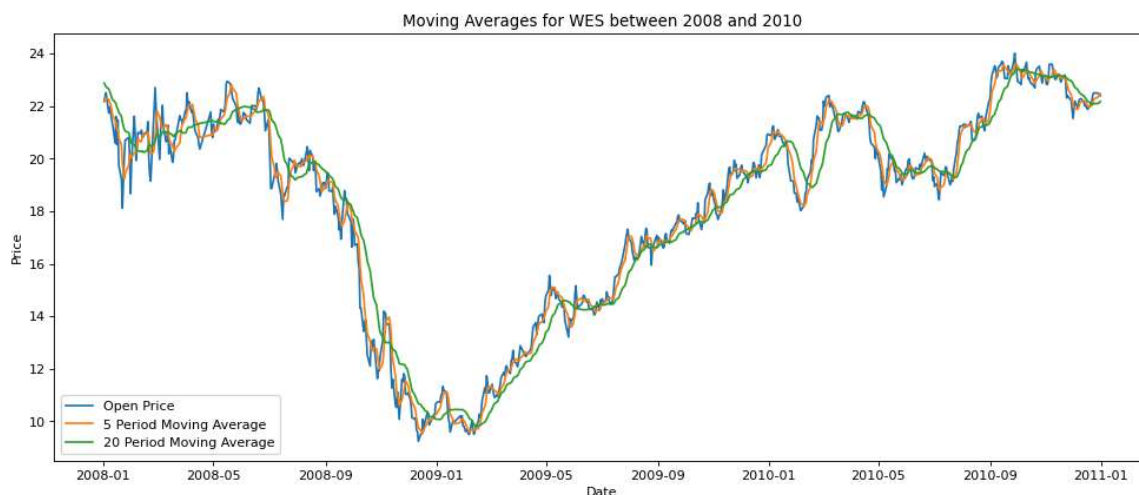
filtered_stocks_date.Date = pd.to_datetime(filtered_stocks_date.Date)
filtered_stocks_date.set_index('Date', inplace=True, drop=True)
filtered_stocks_date.index = pd.to_datetime(filtered_stocks_date.index)

filtered_stocks_date = filtered_stocks_date.loc['2008-01-01': '2010-12-31']

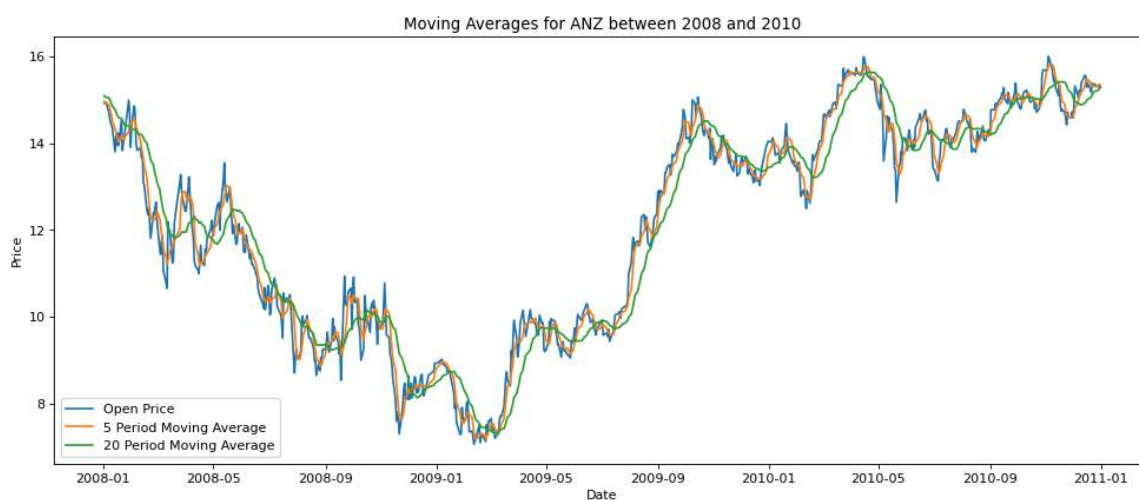
# Code utilized from: https://www.youtube.com/watch?v=\_LWjaAiKaf8
```

```
In [66]: # Selecting each stock seperately
wes = filtered_stocks_date[filtered_stocks_date['File'].str.contains("WE
S")]
anz = filtered_stocks_date[filtered_stocks_date['File'].str.contains("AN
Z")]
bhp = filtered_stocks_date[filtered_stocks_date['File'].str.contains("BH
P")]
cba = filtered_stocks_date[filtered_stocks_date['File'].str.contains("CB
A")]
rio = filtered_stocks_date[filtered_stocks_date['File'].str.contains("RI
O")]
```

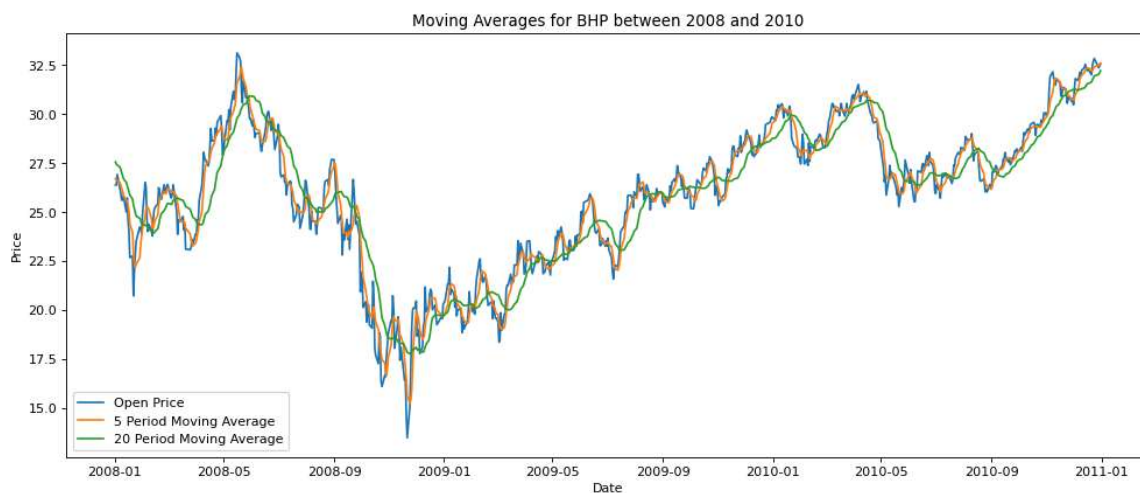
```
In [67]: # WES opening price and moving averages graph
figure(num=None, figsize=(15, 6), dpi=80, facecolor='w', edgecolor='k')
plt.plot(wes['Open'], label='Open Price')
plt.plot(wes['MAV5'], label='5 Period Moving Average')
plt.plot(wes['MAV20'], label='20 Period Moving Average')
plt.title('Moving Averages for WES between 2008 and 2010')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend()
plt.show()
```



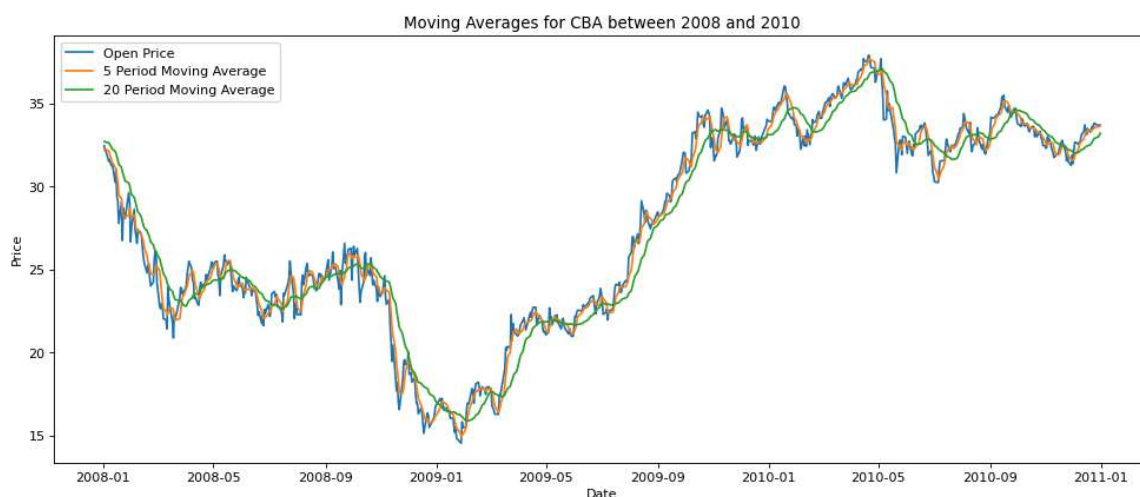
```
In [68]: # ANZ opening price and moving averages graph
figure(num=None, figsize=(15, 6), dpi=80, facecolor='w', edgecolor='k')
plt.plot(anz['Open'], label='Open Price')
plt.plot(anz['MAV5'], label='5 Period Moving Average')
plt.plot(anz['MAV20'], label='20 Period Moving Average')
plt.title('Moving Averages for ANZ between 2008 and 2010')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend()
plt.show()
```



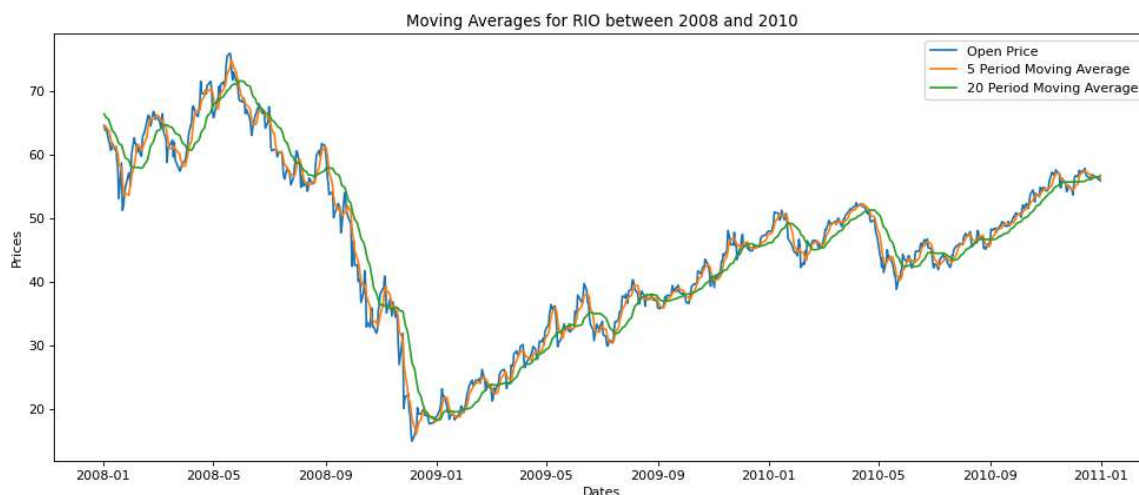
```
In [69]: # BHP opening price and moving averages graph
figure(num=None, figsize=(15, 6), dpi=80, facecolor='w', edgecolor='k')
plt.plot(bhp['Open'], label='Open Price')
plt.plot(bhp['MAV5'], label='5 Period Moving Average')
plt.plot(bhp['MAV20'], label='20 Period Moving Average')
plt.title('Moving Averages for BHP between 2008 and 2010')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend()
plt.show()
```



```
In [70]: # CBA opening price and moving averages graph
figure(num=None, figsize=(15, 6), dpi=80, facecolor='w', edgecolor='k')
plt.plot(cba['Open'], label='Open Price')
plt.plot(cba['MAV5'], label='5 Period Moving Average')
plt.plot(cba['MAV20'], label='20 Period Moving Average')
plt.title('Moving Averages for CBA between 2008 and 2010')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend()
plt.show()
```



```
In [71]: # RIO opening price and moving averages graph
figure(num=None, figsize=(15, 6), dpi=80, facecolor='w', edgecolor='k')
plt.plot(rio['Open'], label='Open Price')
plt.plot(rio['MAV5'], label='5 Period Moving Average')
plt.plot(rio['MAV20'], label='20 Period Moving Average')
plt.title('Moving Averages for RIO between 2008 and 2010')
plt.xlabel('Dates')
plt.ylabel('Prices')
plt.legend()
plt.show()
```



## Strategy Implementation

```
In [72]: # Adding all the stocks in one dataframe
wes_strat = filtered_stocks[filtered_stocks['File'].str.contains("WES")]
anz_strat = filtered_stocks[filtered_stocks['File'].str.contains("ANZ")]
bhp_strat = filtered_stocks[filtered_stocks['File'].str.contains("BHP")]
cba_strat = filtered_stocks[filtered_stocks['File'].str.contains("CBA")]
rio_strat = filtered_stocks[filtered_stocks['File'].str.contains("RIO")]

stocks_strat = pd.concat([wes_strat, anz_strat, bhp_strat, cba_strat, rio_s
trат])

# Code utilized from: https://stackoverflow.com/questions/39301465/appending-one-data-frame-into-another
```

```
In [73]: # Selecting only the opening price, closing price, MAV5 and MAV20
stocks_strat = stocks_strat.drop(['High', 'Low', 'Volume', 'Member', 'Unnam
ed: 0'], axis = 1)

# Code utilized from: https://www.geeksforgeeks.org/python-delete-rows-columns-from-dataframe-using-pandas-drop/
```



```

In [74]: # Recording buy and sell transaction of the strategy
def strategy(stocks_strat):
    buy_price = []
    sell_price = []
    flag = -1

    for i in range(len(stocks_strat)):
        if stocks_strat['MAV5'][i] < stocks_strat['MAV20'][i]:
            if flag != 1:
                buy_price.append(stocks_strat['Open'][i])
                sell_price.append(np.nan)
            else:
                buy_price.append(np.nan)
                sell_price.append(np.nan)
        elif stocks_strat['MAV5'][i] > stocks_strat['MAV20'][i]:
            if flag != 0:
                buy_price.append(np.nan)
                sell_price.append(stocks_strat['Open'][i])
            else:
                buy_price.append(np.nan)
                sell_price.append(np.nan)
        else:
            buy_price.append(np.nan)
            sell_price.append(np.nan)

    return (buy_price, sell_price)

# Code utilized fromL https://www.youtube.com/watch?v=SEQbb8w7VTw

```

### Task 3

```

In [75]: # Storing function data into variable
results = strategy(stocks_strat)
buy_price = results[0]
sell_price = results[1]

del buy_price[-1]
del sell_price[-1]

stocks_strat['Buy_Price'] = buy_price
stocks_strat['Sell_Price'] = sell_price

```

```
In [76]: # Adding final columns to the dataframe stocks_strat
stocks_strat.rename(columns={'File': 'Stock'}, inplace=True)

stocks_strat['Quantity'] = 1000
stocks_strat.loc[:19, "Quantity"] = np.nan

stocks_strat['Profit'] = stocks_strat['Sell_Price'].sum() - stocks_strat['Buy_Price'].sum()
stocks_strat.loc[:19, "Profit"] = np.nan
print(stocks_strat)
```

	Stock	Open	Close	MAV5	MAV20	Buy_Price \
Date						
2002-01-31	WES	10.9880	11.1336	NaN	NaN	NaN
2002-02-01	WES	11.1517	11.0786	NaN	NaN	NaN
2002-02-04	WES	11.1554	11.1325	NaN	NaN	NaN
2002-02-05	WES	11.1041	11.0972	NaN	NaN	NaN
2002-02-06	WES	11.0972	10.9407	11.09928	NaN	NaN
...	...	...	...	...	...	...
2018-08-06	RIO	75.1465	74.5848	77.04448	77.705415	75.1465
2018-08-07	RIO	74.2556	73.9844	76.19038	77.447830	74.2556
2018-08-08	RIO	74.6623	74.7494	75.29754	77.250765	74.6623
2018-08-09	RIO	75.1800	75.9400	74.87042	77.162865	75.1800
2018-08-10	RIO	75.9000	75.6600	75.02888	77.074650	75.9000

	Sell_Price	Quantity	Profit
Date			
2002-01-31	NaN	NaN	NaN
2002-02-01	NaN	NaN	NaN
2002-02-04	NaN	NaN	NaN
2002-02-05	NaN	NaN	NaN
2002-02-06	NaN	NaN	NaN
...	...	...	...
2018-08-06	NaN	1000.0	87541.7446
2018-08-07	NaN	1000.0	87541.7446
2018-08-08	NaN	1000.0	87541.7446
2018-08-09	NaN	1000.0	87541.7446
2018-08-10	NaN	1000.0	87541.7446

[21178 rows x 9 columns]

## Overview of stocks\_strat

```
In [77]: # Stocks_strat describe
stocks_described = stocks_strat.describe().loc[['mean', '50%', 'std', 'max', 'min']]
print(stocks_described)
```

	Open	Close	MAV5	MAV20	Buy_Price	Sell_Price	\
mean	28.017879	28.009648	28.015015	28.003588	27.511335	28.433033	
50%	23.930700	23.932250	23.925950	23.961470	23.534400	24.354350	
std	17.084839	17.082790	17.066134	17.004874	16.878374	17.232445	
max	84.152400	84.007200	82.699860	81.582780	82.825700	84.152400	
min	4.922500	4.900100	4.988300	5.113300	4.922500	5.173700	

	Quantity	Profit
mean	1000.0	8.754174e+04
50%	1000.0	8.754174e+04
std	0.0	4.039707e-08
max	1000.0	8.754174e+04
min	1000.0	8.754174e+04

## Calculating profit of each stock

```
In [78]: # WES trading profit
wes_profit = stocks_strat[stocks_strat['Stock'].str.contains("WES")]
wes_profit['Profit'] = wes_profit['Sell_Price'].sum() - wes_profit['Buy_Price'].sum()
print('Total profit from WES: ', wes_profit['Profit'].iloc[-1])

# ANZ trading profit
anz_profit = stocks_strat[stocks_strat['Stock'].str.contains("ANZ")]
anz_profit['Profit'] = anz_profit['Sell_Price'].sum() - anz_profit['Buy_Price'].sum()
print('Total profit from ANZ: ', anz_profit['Profit'].iloc[-1])

# BHP trading profit
bhp_profit = stocks_strat[stocks_strat['Stock'].str.contains("BHP")]
bhp_profit['Profit'] = bhp_profit['Sell_Price'].sum() - bhp_profit['Buy_Price'].sum()
print('Total profit from BHP: ', bhp_profit['Profit'].iloc[-1])

# CBA trading profit
cba_profit = stocks_strat[stocks_strat['Stock'].str.contains("WES")]
cba_profit['Profit'] = cba_profit['Sell_Price'].sum() - cba_profit['Buy_Price'].sum()
print('Total profit from CBA: ', cba_profit['Profit'].iloc[-1])

# RIO trading profit
rio_profit = stocks_strat[stocks_strat['Stock'].str.contains("WES")]
rio_profit['Profit'] = rio_profit['Sell_Price'].sum() - rio_profit['Buy_Price'].sum()
print('Total profit from RIO: ', rio_profit['Profit'].iloc[-1])
```

```
Total profit from WES: 19213.982799999998
Total profit from ANZ: 11377.8253
Total profit from BHP: 8001.5336000000025
Total profit from CBA: 19213.982799999998
Total profit from RIO: 19213.982799999998
```

```
In [79]: # Number of trades
buy_trades = stocks_strat['Buy_Price'].shape[0] - stocks_strat['Buy_Pric
e'].isnull().sum()

sell_trades = stocks_strat['Sell_Price'].shape[0] - stocks_strat['Sell_Pric
e'].isnull().sum()

total_trades = sell_trades - buy_trades

print('Total number of trades: ', total_trades)
```

Total number of trades: 2781

### Comments on the strategy

The strategy involves purchasing stock whenever the 5 moving average of a stock crosses over the 20 moving average and selling when it crosses back over would generate a profit of 87541 dollars based on the historical prices of the stocks, WES, ANZ, CBA, BHP and RIO.

It is evident that the 3 stocks (WES, CBA and RIO) returned the same amount of profit of 19213.98 dollars, while the second best performing stock was ANZ with a return of 11377.825 dollars. The worst performing stock was BHP which generated a return of 8001.53 dollars.

This strategy could be used in the future if the trader studies the market and invests into the highest performing stocks as the strategy did generate a profit.

However, the strategy could be improved by following Yahoo Finance's suggested strategy of having a crossover of 13 days and a 48.5 days. This strategy utilizes the 'Golden Cross' where returns are at a 4.9 percent gain.

Reference: <https://finance.yahoo.com/news/study-determines-best-moving-average-195042216.html>  
(<https://finance.yahoo.com/news/study-determines-best-moving-average-195042216.html>)