

```
In [~]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline

In [~]: import os
os.chdir('/Users/divine/Documents/MBS /PGE 2 /S2/Finance appliquée avec Python /Rapport/2021+2022')

In [~]: df = pd.read_excel('NDAQ 2021.xlsx')

In [~]: print(df)

      Date      META      NASDAQ
0  2021-01-04  268.654968  42.049213
1  2021-01-05  270.682800  42.036465
2  2021-01-06  263.030914  43.129025
3  2021-01-07  268.455170  43.574966
4  2021-01-08  267.286407  44.215218
...
498 2022-12-23  117.914894  60.059288
499 2022-12-27  116.756119  59.765259
500 2022-12-28  115.497459  59.490841
501 2022-12-29  120.132538  60.637527
502 2022-12-30  120.212448  60.127888

[503 rows x 3 columns]

In [~]: plt.plot(df['Date'], df['META'])
Out[7]: <matplotlib.lines.Line2D at 0x12d707cd0>

350
300
250
200
150
100

2021-012021-042021-072021-102022-012022-042022-072022-102023-01

In [~]: plt.plot(df['Date'], df['NASDAQ'])
Out[8]: <matplotlib.lines.Line2D at 0x12db57690>

65
60
55
50
45

2021-012021-042021-072021-102022-012022-042022-072022-102023-01

In [~]: meta = df['META']
nasdaq = df['NASDAQ']

In [~]: meta_daily_returns = meta.pct_change()

In [~]: nasdaq_daily_returns = nasdaq.pct_change()

In [~]: meta_daily_returns.mean()*252
Out[12]: -0.27203999231524006

In [~]: nasdaq_daily_returns.mean()*252
Out[13]: 0.2082185274696086

In [~]: meta_daily_returns.std()*252**0.5
Out[14]: 0.5008556318801559

In [~]: nasdaq_daily_returns.std()*252**0.5
Out[15]: 0.23939344332419366

In [~]: clean_meta_daily_returns = meta_daily_returns.dropna(axis=0)
clean_nasdaq_daily_returns = nasdaq_daily_returns.dropna(axis=0)

In [~]: data = np.array([clean_meta_daily_returns,clean_nasdaq_daily_returns])

In [~]: covMatrix = np.cov(data,bias=True)
print(covMatrix)
[[0.00099348 0.00020938]
 [0.00020938 0.0002696]]

In [~]: beta = covMatrix[0][1] / covMatrix[1][1]
print(beta)
0.9225097415184129

In [~]: import os
os.chdir('/Users/divine/Documents/MBS /PGE 2 /S2/Finance appliquée avec Python /Rapport/2021+2022')

In [~]: data = pd.read_excel('META2021.xlsx')

In [~]: data.head()
Out[22]:
      Date      META      NVIDIA      ALPHABET INC.      MICROSOFT      BAIDU
0  2021-01-04  268.654968  130.833954      86.412003      211.605286      176.399994
1  2021-01-05  270.682800  133.739792      87.045998      211.809448      174.800003
2  2021-01-06  263.030914  125.855415      86.764503      206.317368      171.800003
3  2021-01-07  268.455170  133.133682      89.362503      212.188538      170.000000
4  2021-01-08  267.286407  132.462723      90.360497      213.481369      191.000000

In [~]: stocks = data.drop(columns=["Date"])
stocks.head()
Out[23]:
      META      NVIDIA      ALPHABET INC.      MICROSOFT      BAIDU
0  268.654968  130.833954      86.412003      211.605286      176.399994
1  270.682800  133.739792      87.045998      211.809448      174.800003
2  263.030914  125.855415      86.764503      206.317368      171.800003
3  268.455170  133.133682      89.362503      212.188538      170.000000
4  267.286407  132.462723      90.360497      213.481369      191.000000

In [~]: stock_normed = stocks/stocks.iloc[0]

In [~]: stock_normed.plot()
Out[25]: <Axes: >

2.5
2.0
1.5
1.0
0.5
0

0      100      200      300      400      500
META
NVIDIA
ALPHABET INC.
MICROSOFT
BAIDU

In [~]: log_ret = np.log(stocks/stocks.shift(1))

In [~]: log_ret.mean() * 252
Out[27]:
META      -0.403686
NVIDIA      0.055292
ALPHABET INC.  0.013288
MICROSOFT      0.057473
BAIDU      -0.250959
dtype: float64

In [~]: log_ret.describe().transpose()
Out[28]:
      count      mean      std      min      25%      50%      75%      max
META      502.0      -0.001602  0.032776      -0.306391      -0.013486      -0.000310      0.014581      0.162064
NVIDIA      502.0      0.000219  0.034583      -0.099518      -0.020369      0.001376      0.020590      0.133913
ALPHABET INC.  502.0      0.000053  0.020306      -0.101313      -0.009954      0.000720      0.010154      0.074606
MICROSOFT      502.0      0.000228  0.018335      -0.080295      -0.009206      -0.000084      0.011290      0.079059
BAIDU      502.0      -0.000996  0.039350      -0.165775      -0.022895      0.000000      0.018669      0.333690

In [~]: log_ret.cov()
Out[29]:
      META      NVIDIA      ALPHABET INC.      MICROSOFT      BAIDU
META      0.001074      0.000611      0.000425      0.000352      0.000051
NVIDIA      0.000611      0.001196      0.000479      0.000462      0.000027
ALPHABET INC.  0.000425      0.000479      0.000412      0.000302      0.000021
MICROSOFT      0.000352      0.000462      0.000302      0.000336      -0.000002
BAIDU      0.000051      0.000027      0.000021      -0.000002      0.001548

In [~]: log_ret.corr()
print(log_ret.corr())
META      NVIDIA      ALPHABET INC.      MICROSOFT      BAIDU
META      1.000000      0.538802      0.638188      0.585666      0.039530
NVIDIA      0.538802      1.000000      0.681950      0.728488      0.020084
ALPHABET INC.  0.638188      0.681950      1.000000      0.811912      0.026548
MICROSOFT      0.585666      0.728488      0.811912      1.000000      -0.003158
BAIDU      0.039530      0.020084      0.026548      -0.003158      1.000000

In [~]: num_ports = 15000
all_weights = np.zeros((num_ports,len(stocks.columns)))
ret_arr = np.zeros(num_ports)
vol_arr = np.zeros(num_ports)
sharpe_arr = np.zeros(num_ports)
nb_assets = stocks.shape[1]

In [~]: for ind in range(num_ports):
weights = np.array(np.random.random(nb_assets))
weights = weights / np.sum(weights)
all_weights[ind,:] = weights
ret_arr[ind] = np.sum(log_ret.mean() * weights) *252
vol_arr[ind] = np.sqrt(np.dot(weights.T, np.dot(log_ret.cov() * 252, weights)))
sharpe_arr[ind] = ret_arr[ind]/vol_arr[ind]

In [~]: sharpe_arr.max()
optimal = sharpe_arr.argmax()
print(optimal)
4636

In [~]: optimal_pf = all_weights[optimal,: ]
o_w = optimal_pf.tolist()
weights_sim = dict(zip(stocks, o_w))

In [~]: optimal_weights = pd.DataFrame.from_dict(weights_sim, orient='index', columns=['Weight in %'])
optimal_weights = optimal_weights*100
optimal_weights = optimal_weights.round(decimals = 2)
optimal_weights = optimal_weights.sort_values(by=['Weight in %'], ascending = False)
optimal_weights
Out[35]:
      Weight in %
MICROSOFT      65.45
NVIDIA      19.06
ALPHABET INC.      13.01
BAIDU      2.38
META      0.10

In [~]: max_sr_ret = ret_arr[optimal]
max_sr_vol = vol_arr[optimal]
max_sr = sharpe_arr[optimal]

In [~]: print('La rentabilité du portefeuille optimal est',max_sr_ret*100,'%')
print('La volatilité du portefeuille optimal est',max_sr_vol*100,'%')
print('Le ratio de Sharpe du portefeuille optimal est',max_sr)
La rentabilité du portefeuille optimal est 4.349397649445135 %
La volatilité du portefeuille optimal est 31.22392546629376 %
Le ratio de Sharpe du portefeuille optimal est 0.13929695208055473

In [~]: plt.scatter(max_sr_vol,max_sr_ret,c='black',s=50,edgecolors='black')
plt.xlabel('Volatility')
plt.ylabel('Return')
Out[38]: Text(0, 0.5, 'Return')

0.045
0.044
0.043
0.042

0.300      0.305      0.310      0.315      0.320      0.325
Volatility

In [~]: plt.figure(figsize=(12,8))
plt.scatter(vol_arr,ret_arr,c=sharpe_arr,cmap='plasma')
plt.scatter(max_sr_vol,max_sr_ret,c='black',s=50,edgecolors='black')
plt.colorbar(label='Sharpe Ratio')
plt.xlabel('Volatility')
plt.ylabel('Return')
Out[39]: Text(0, 0.5, 'Return')

0.05
0.00
-0.05
-0.10
-0.15
-0.20
-0.25
-0.30
-0.35

0.30      0.35      0.40      0.45      0.50
Volatility

Sharpe Ratio
1.0
0.8
0.6
0.4
0.2
0.0
```