

Determine Beta for CAPM application

```
In [1]: 1 import numpy as np
        2 import pandas as pd
        3 import matplotlib.pyplot as plt
        4 %matplotlib inline
```

Load Dataset from resident file

```
In [2]: 1 security_data = pd.read_csv('^GSPC_MSFT Monthly V2 .csv', index_col='Date')
```

```
In [3]: 1 security_data
```

Out[3]:

	GSPC	MSFT
Date		
01/11/2015	2080.409912	49.264046
01/12/2015	2043.939941	50.627274
01/01/2016	1940.239990	50.271385
01/02/2016	1932.229980	46.429623
01/03/2016	2059.739990	50.760994
...
01/07/2020	3271.120117	203.981583
01/08/2020	3500.310059	224.398651
01/09/2020	3363.000000	209.780777
01/10/2020	3269.959961	201.941315
01/11/2020	3621.629883	213.511017

61 rows × 2 columns

Calculate the returns using the log function and based on differencing method. Check the data is complete for all dates in the series

```
In [4]: 1 sec_returns = np.log ( security_data / security_data.shift(1))
```

```
In [5]: 1 Ret =sec_returns*12
        2 Ret
```

Out[5]:

	^GSPC	MSFT
Date		
01/11/2015	NaN	NaN
01/12/2015	-0.212228	0.327551
01/01/2016	-0.624811	-0.084653
01/02/2016	-0.049643	-0.953980
01/03/2016	0.766860	1.070287
...
01/07/2020	0.643642	0.088124
01/08/2020	0.812629	1.144733
01/09/2020	-0.480217	-0.808333
01/10/2020	-0.336668	-0.457031
01/11/2020	1.225757	0.668536

61 rows × 2 columns

Determine the Covariance between the security and the market portoflio. Check date for accuracy and interpret results

```
In [6]: 1 Ret_secs = Ret.iloc[1:61] #Index 1 to 61
```

```
In [7]: 1 Ret_secs.head( )
```

Out[7]:

	^GSPC	MSFT
Date		
01/12/2015	-0.212228	0.327551
01/01/2016	-0.624811	-0.084653
01/02/2016	-0.049643	-0.953980
01/03/2016	0.766860	1.070287
01/04/2016	0.032349	-1.225041

```
In [8]: 1 cov=Ret_secs.cov( )
```

```
In [9]: 1 cov.round(4)
```

```
Out[9]:
```

	[^]GSPC	MSFT
[^]GSPC	0.2814	0.2248
MSFT	0.2248	0.3849

CoviM/Varm is the Beta. We apply the ILOC method to the derived covariance of returns. The preceding output is the nominator we require to calculate the beta value

```
In [10]: 1 #iloc locates the cell that corresponds to the covariance between MSFT a
2 cov_with_market = cov.iloc[1,0]
```

Note- ILOC is a cell location in a dataframe. Therefore, iloc[1,0] is the second row and first column in the dataframe.

```
In [11]: 1 cov_with_market.round(4)
```

```
Out[11]: 0.2248
```

We require the Variance of the market index to also calculate the BETA. Find this out next.

```
In [12]: 1 mkt_var= Ret_secs['^GSPC'].var()
```

```
In [13]: 1 mkt_var
```

```
Out[13]: 0.28138771172648686
```

Finally we calculate the Beta value of the stocks by using the formula discussed in class.

```
In [14]: 1 MSFT_beta= cov_with_market / mkt_var
```

```
In [15]: 1 MSFT_beta
```

```
Out[15]: 0.7987772298943374
```

Interpretation of results: MSFT Beta is Defensive security if beta value is less than 1. Aggressive if higher.

The CAPM Formula $r_i = r_f - (\text{beta}(i_m) \cdot (r_m - r_f))$

In the US, we get the 3mth T-bill rate of 0.09% (22nd Dec 2020)

```
In [16]: 1 RM = Ret_secs[ '^GSPC' ].mean()
          2 RM
```

```
Out[16]: 0.11087184421142973
```

```
In [17]: 1 ri= Ret_secs[ 'MSFT' ].mean()
          2 ri
```

```
Out[17]: 0.29329878165689066
```

```
In [18]: 1 rf= 0.0009
```

Determine the Expected return of the security (MSFT_er) using the CAPM

```
In [19]: 1 #Risk premium is Rm-Rf
          2 RP = RM - rf
          3 RP
```

```
Out[19]: 0.10997184421142973
```

```
In [20]: 1 MSFT_er= rf +MSFT_beta*(RP)
```

```
In [21]: 1 MSFT_er
```

```
Out[21]: 0.08874300508557746
```

```
In [22]: 1 #8.87% is predicted. Technique can be applied to estimate the return on
```

```
In [23]: 1 #EXERCISE 2 Apply the Sharpe Ratio in practice
          2 #Sharpe Ratio, S = (Ri-Rf) Vari
```

```
In [27]: 1 (Ret_secs [ 'MSFT' ].var())**0.5
```

```
Out[27]: 0.6204211352023237
```

```
In [29]: 1 Ret_secs [ 'MSFT' ].std()
```

```
Out[29]: 0.6204211352023237
```

```
In [30]: 1 Sharpe= (MSFT_er - rf) / Ret_secs [ 'MSFT' ].std()
```

```
In [31]: 1 Sharpe
```

```
Out[31]: 0.14158609386659796
```

```
In [ ]: 1 Sharpe Ratio is used to compare different stocks in portfolios.
```

