

```
In [15]: import pandas as pd  
import numpy as np
```

```
In [16]: stock_prices = pd.read_csv("sample_prices.csv")
```

```
In [17]: stock_prices.head()
```

Out[17]:

	BLUE	ORANGE
0	8.7000	10.6600
1	8.9055	11.0828
2	8.7113	10.7100
3	8.4346	11.5907
4	8.7254	12.1070

```
In [18]: returns = stock_prices.pct_change()
```

```
In [19]: returns
```

Out[19]:

	BLUE	ORANGE
0	NaN	NaN
1	0.023621	0.039662
2	-0.021807	-0.033638
3	-0.031763	0.082232
4	0.034477	0.044544
5	0.037786	-0.026381
6	-0.011452	-0.049187
7	0.032676	0.117008
8	-0.012581	0.067353
9	0.029581	0.078249
10	0.006151	-0.168261
11	0.012162	0.024041
12	0.021149	-0.055623

```
In [20]: # drop first row
returns = returns.dropna()
returns
```

```
Out[20]:
```

	BLUE	ORANGE
1	0.023621	0.039662
2	-0.021807	-0.033638
3	-0.031763	0.082232
4	0.034477	0.044544
5	0.037786	-0.026381
6	-0.011452	-0.049187
7	0.032676	0.117008
8	-0.012581	0.067353
9	0.029581	0.078249
10	0.006151	-0.168261
11	0.012162	0.024041
12	0.021149	-0.055623

```
In [21]: # standard deviation
returns.std()
```

```
Out[21]: BLUE      0.023977
ORANGE      0.079601
dtype: float64
```

```
In [28]: # How to calculate STD
deviations = returns - returns.mean()
sqr_deviations = deviations**2
variance = sqr_deviations.mean()
volatility = np.sqrt(variance)
volatility
```

```
Out[28]: BLUE      0.022957
ORANGE      0.076212
dtype: float64
```

```
In [29]: number_of_obs = returns.shape[0]
variance = sqr_deviations.sum()/(number_of_obs - 1)
volatility = variance**0.5
```

```
In [30]: # this will give you the std
volatility
```

```
Out[30]: BLUE      0.023977
ORANGE      0.079601
dtype: float64
```

```
In [31]: # the return Annualized option 1
returns.std()*np.sqrt(12)
```

```
Out[31]: BLUE      0.083060
ORANGE      0.275747
dtype: float64
```

```
In [32]: #option 2
returns.std()*(12**0.5)
```

Out[32]: BLUE 0.083060
ORANGE 0.275747
dtype: float64

```
In [36]: returns_portfolio = pd.read_csv("Portfolios_Formed_on_ME_monthly_EW.csv", header=
0, index_col=0,
parse_dates=True, na_values=-99.99)
```

```
In [37]: returns_portfolio.head()
```

Out[37]:

	<=	Lo	Med	Hi	Lo	Qnt	Qnt	Qnt	Hi	Lo	Dec	Dec	Dec	Dec	Dec	Dec
	0	30	40	30	20	2	3	4	20	10	2	3	4	5	6	7
192607	NaN	-0.43	1.52	2.68	-0.57	0.59	1.60	1.47	3.33	-1.45	0.29	-0.15	1.33	1.24	1.98	1.55
192608	NaN	3.90	3.04	2.09	3.84	3.59	3.71	1.61	2.33	5.12	2.59	4.03	3.15	2.72	4.72	1.60
192609	NaN	-1.08	-0.54	0.16	-0.48	-1.40	0.00	-0.50	-0.09	0.93	-1.87	-2.27	-0.53	0.07	-0.07	-1.64
192610	NaN	-3.32	-3.52	-3.06	-3.29	-4.10	-2.89	-3.36	-2.95	-4.84	-1.77	-3.36	-4.83	-2.98	-2.80	-3.45
192611	NaN	-0.46	3.82	3.09	-0.55	2.18	3.41	3.39	3.16	-0.78	-0.32	-0.29	4.65	3.24	3.57	3.82

```
In [38]: columns = ['Lo 10', 'Hi 10']
returns_portfolio_one = returns_portfolio[columns]
returns_portfolio_one.head()
```

Out[38]:

	Lo 10	Hi 10
192607	-1.45	3.29
192608	5.12	3.70
192609	0.93	0.67
192610	-4.84	-2.43
192611	-0.78	2.70

```
In [39]: # to work with the raw data
returns_portfolio_one = returns_portfolio_one/100
returns_portfolio_one.head()
```

Out[39]:

	Lo 10	Hi 10
192607	-0.0145	0.0329
192608	0.0512	0.0370
192609	0.0093	0.0067
192610	-0.0484	-0.0243
192611	-0.0078	0.0270

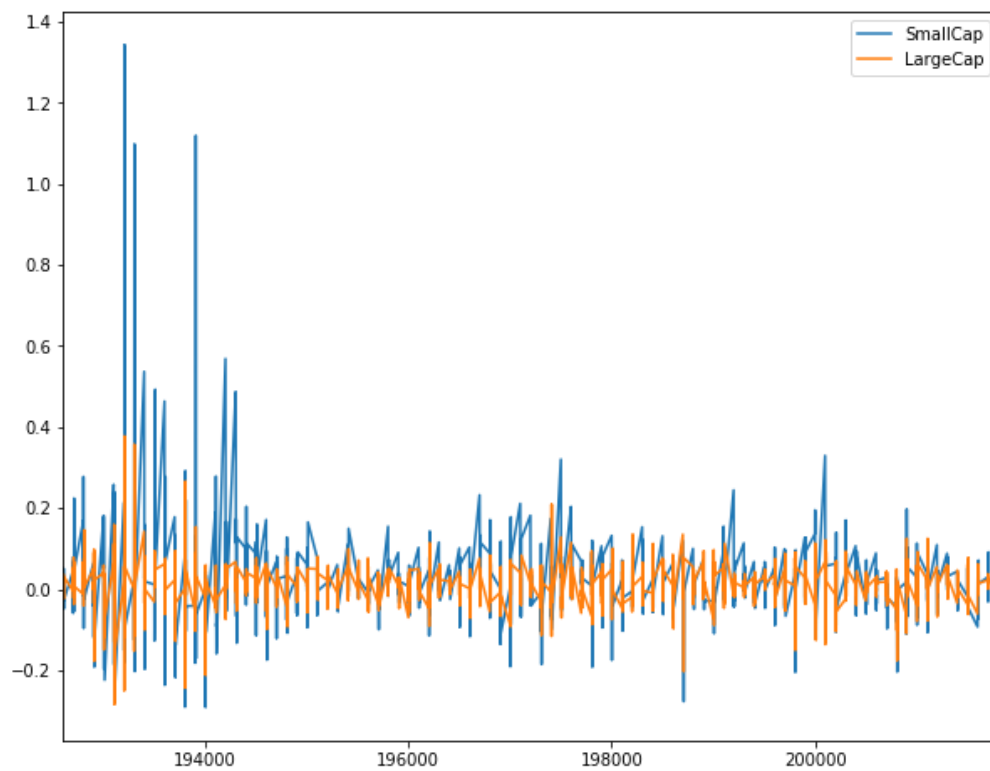
```
In [41]: #Change columns names
returns_portfolio_one.columns = ['SmallCap', 'LargeCap']
returns_portfolio_one.head()
```

Out[41]:

	SmallCap	LargeCap
192607	-0.0145	0.0329
192608	0.0512	0.0370
192609	0.0093	0.0067
192610	-0.0484	-0.0243
192611	-0.0078	0.0270

```
In [44]: returns_portfolio_one.plot(figsize=(10,8))
```

Out[44]: <matplotlib.axes._subplots.AxesSubplot at 0x11f7f2940>



```
In [45]: returns_portfolio_one.std()
```

Out[45]: SmallCap 0.106288
LargeCap 0.053900
dtype: float64

```
In [46]: annualized_vol = returns_portfolio_one.std()*np.sqrt(12)
annualized_vol
```

Out[46]: SmallCap 0.368193
LargeCap 0.186716
dtype: float64

In [47]: *# Calculate monthly returns*

```
# Find how many months in the data set
n_months = returns_portfolio_one.shape[0]

#Calculate the return per month
return_per_month = (returns_portfolio_one + 1).prod()**(1/n_months) - 1
return_per_month
```

Out[47]: SmallCap 0.012986
LargeCap 0.007423
dtype: float64

In [48]: annualized_vol

Out[48]: SmallCap 0.368193
LargeCap 0.186716
dtype: float64

In [49]: annualized_return = (return_per_month + 1)**12-1
annualized_return

Out[49]: SmallCap 0.167463
LargeCap 0.092810
dtype: float64

In [50]: *# Another way to calculate annualized returns*
annualized_return_simple_format = (returns_portfolio_one + 1).prod()**(12/n_months) - 1
annualized_return_simple_format

Out[50]: SmallCap 0.167463
LargeCap 0.092810
dtype: float64

In [51]: annualized_return / annualized_vol

Out[51]: SmallCap 0.454825
LargeCap 0.497063
dtype: float64

In [52]: risk_free_rate = 0.025
excess_return = annualized_return - risk_free_rate
sharpe_ratio = excess_return / annualized_vol
sharpe_ratio

Out[52]: SmallCap 0.386926
LargeCap 0.363170
dtype: float64

In []: