



PORTFOLIO OPTIMIZATION

MUKESH S BENGALURU
DUSHYANTH S NANDEESH
SUMIT GUPTA

CONTENTS

SHARPE RATIO

PORTFOLIO
ALLOCATION

PORTFOLIO
OPTIMIZATION

EFFICIENT
FRONTIER

PORTFOLIO

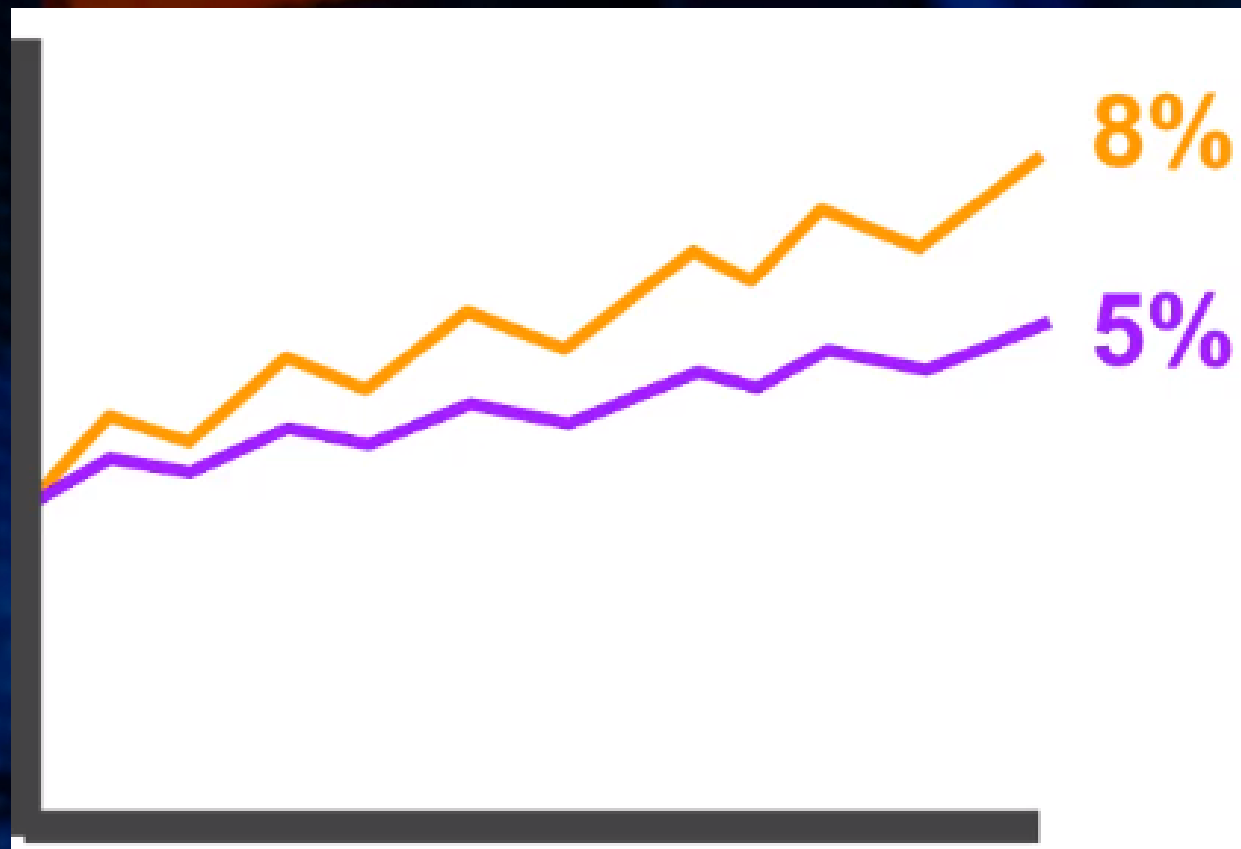
- PORTFOLIO IS JUST A SET OF ALLOCATIONS IN A VARIETY OF SECURITIES.
- FOR EXAMPLE,
 - 1) 20% in APPLE
 - 2) 30% in FACEBOOK
 - 3) 50% in GOOGLE
- THESE PERCENTAGES SHOULD ADD UP TO 100% OR IF DEFINED AS WEIGHTS THEY SHOULD ADD UP TO 1.

KEY STATISTICS FOR A PORTFOLIO

- **DAILY RETURNS** – THE PERCENT RETURNED FROM 1 DAY TO THE NEXT FOR A STOCK.
- **CUMULATIVE RETURN** – THE AMOUNT RETURNED AFTER AN ENTIRE TIME PERIOD.
- **AVERAGE DAILY RETURN** – MEAN OF DAILY RETURNS.
- **STANDARD DAILY RETURNS** – STANDARD DEVIATION OF DAILY RETURNS.

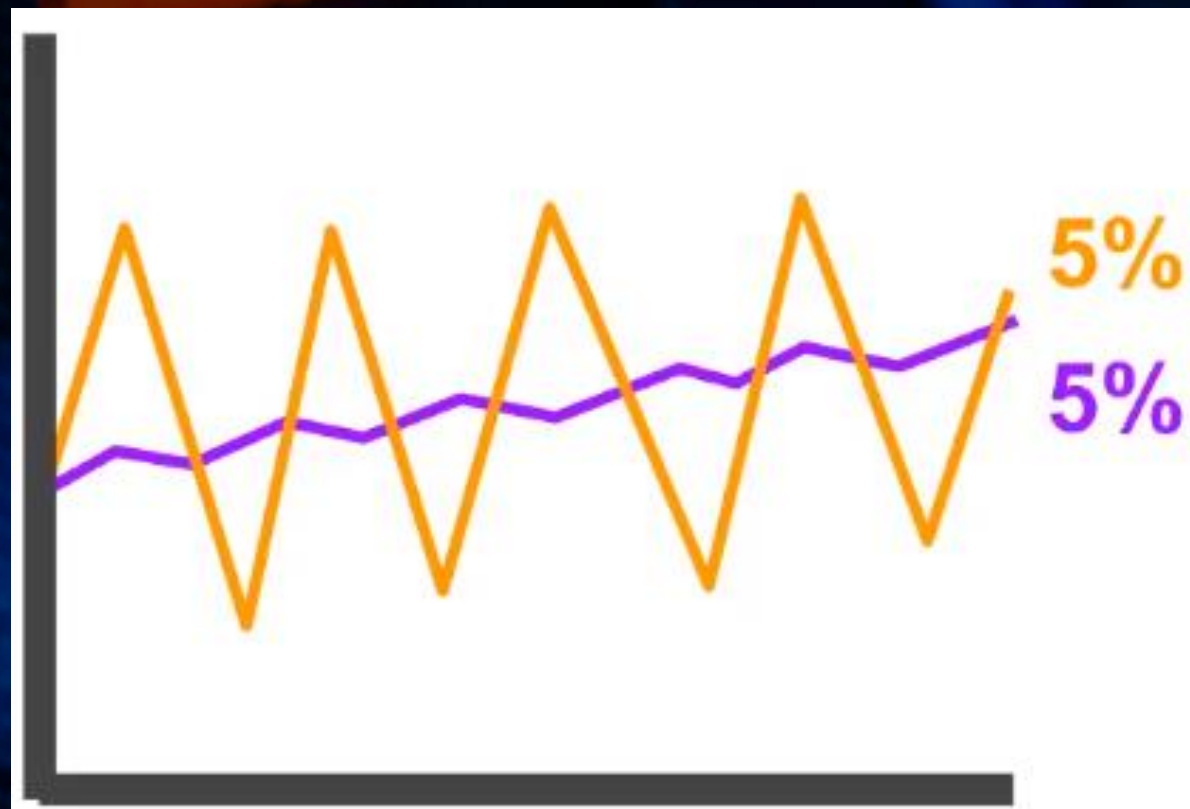
WHICH PORTFOLIO IS BETTER?

- CASE I



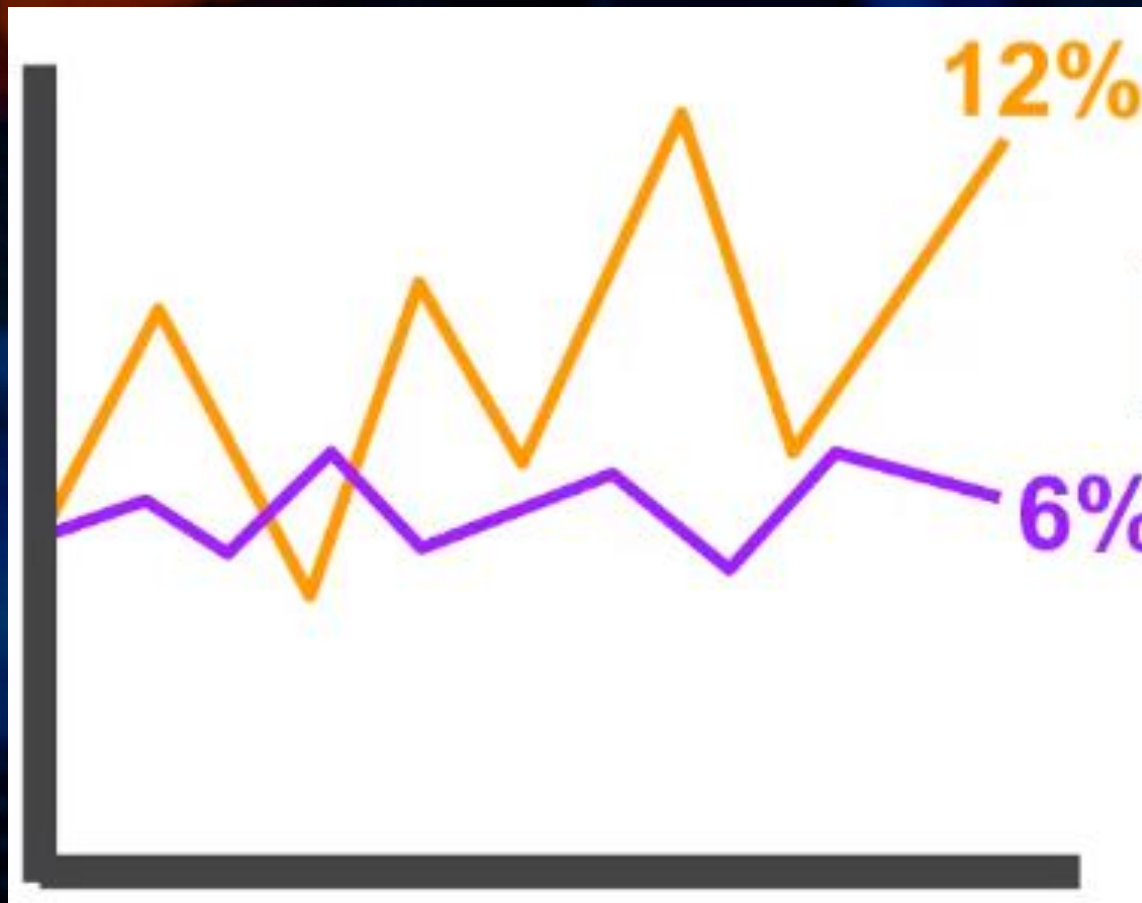
WHICH PORTFOLIO IS BETTER?

- CASE 2



WHICH PORTFOLIO IS BETTER?

- CASE 3



SHARPE RATIO

- SHARPE RATIO IS A MEASURE FOR CALCULATING RISK-ADJUSTED RETURN, AND THIS RATIO HAS BECOME THE INDUSTRY STANDARD FOR SUCH CALCULATIONS.
- IT WAS DEVELOPED BY NOBEL LAUREATE WILLIAM F. SHARPE.

Sharpe Ratio

$$\text{Sharpe Ratio} = \frac{(R_x - R_f)}{\text{StdDev } R_x}$$

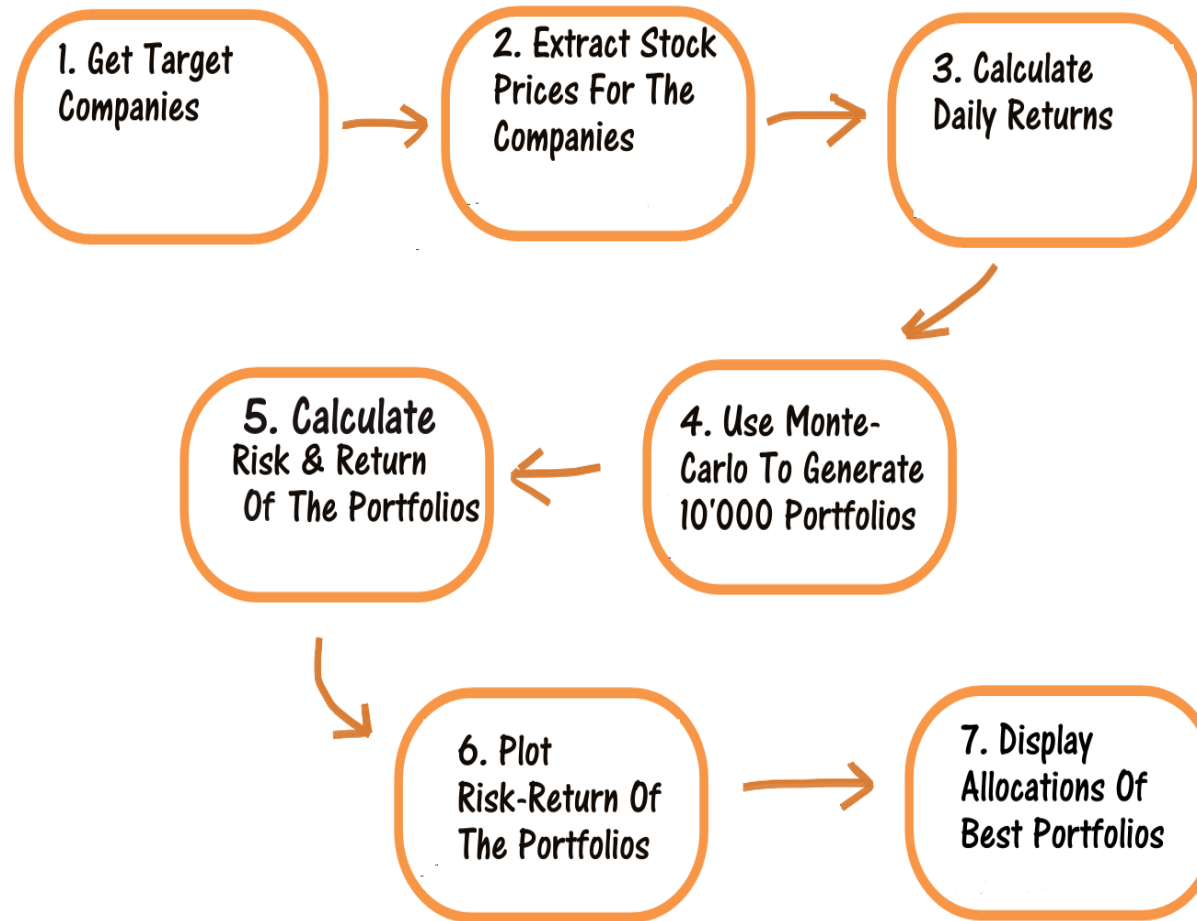
Where:

R_x = Expected portfolio return

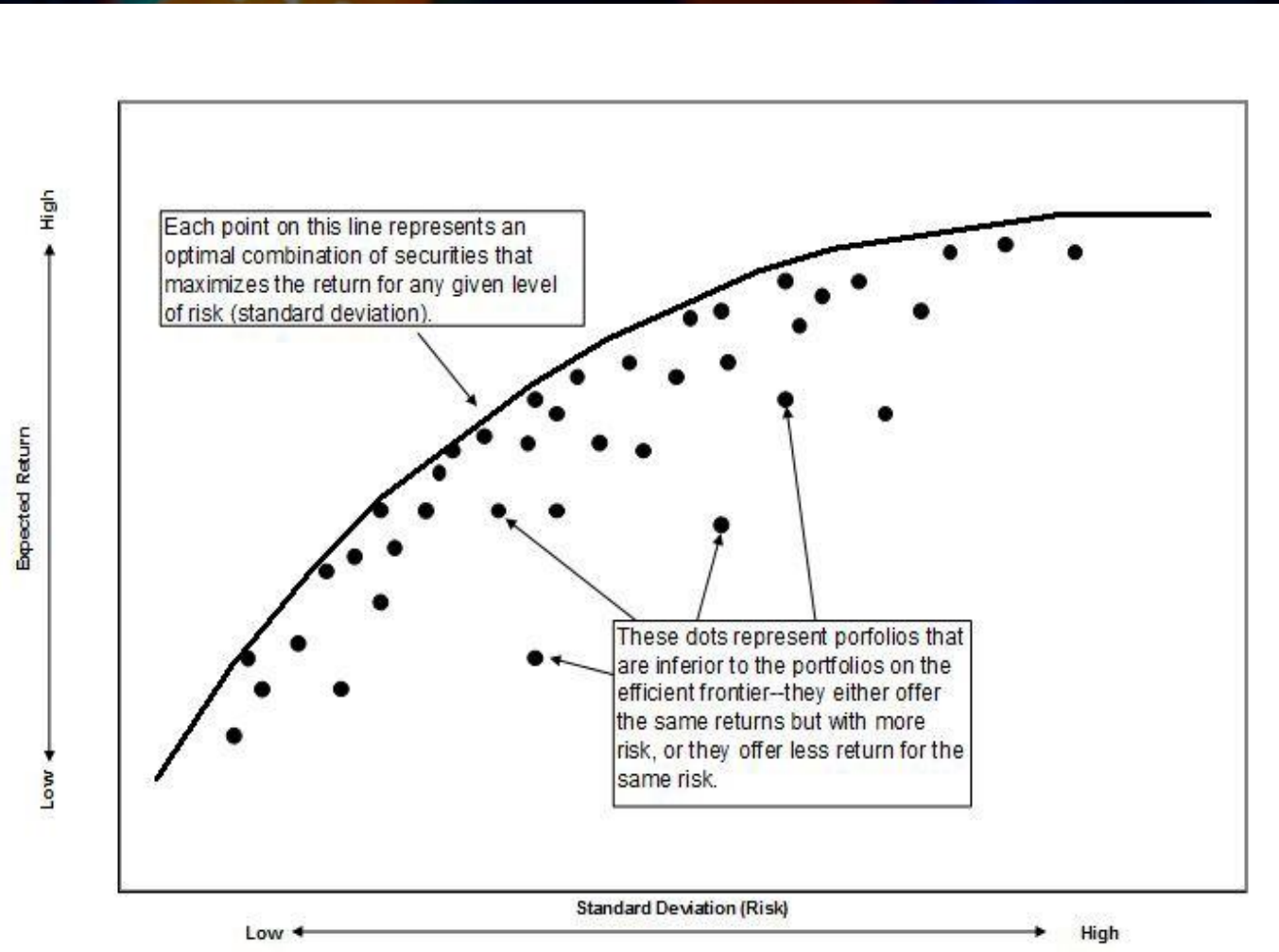
R_f = Risk free rate of return

StdDev R_x = Standard deviation of portfolio return / volatility

MONTE CARLO SIMULATION



MODERN PORTFOLIO THEORY





LIVE DEMO

DATA

```
# Dates for which stock data is collected
start = pd.to_datetime('2013-01-01')
end = pd.to_datetime('2019-01-01')
# Collecting stock data using QUANDL
aapl = quandl.get('WIKI/AAPL.11', start_date=start, end_date=end)
nike = quandl.get('WIKI/NKE.11', start_date=start, end_date=end)
intel = quandl.get('WIKI/INTC.11', start_date=start, end_date=end)
visa = quandl.get('WIKI/V.11', start_date=start, end_date=end)
msft = quandl.get('WIKI/MSFT.11', start_date=start, end_date=end)
hodp = quandl.get('WIKI/HD.11', start_date=start, end_date=end)
disc = quandl.get('WIKI/DIS.11', start_date=start, end_date=end)
ba = quandl.get('WIKI/BA.11', start_date=start, end_date=end)
pfizer = quandl.get('WIKI/PFE.11', start_date=start, end_date=end)
jnj = quandl.get('WIKI/JNJ.11', start_date=start, end_date=end)
```

DATA

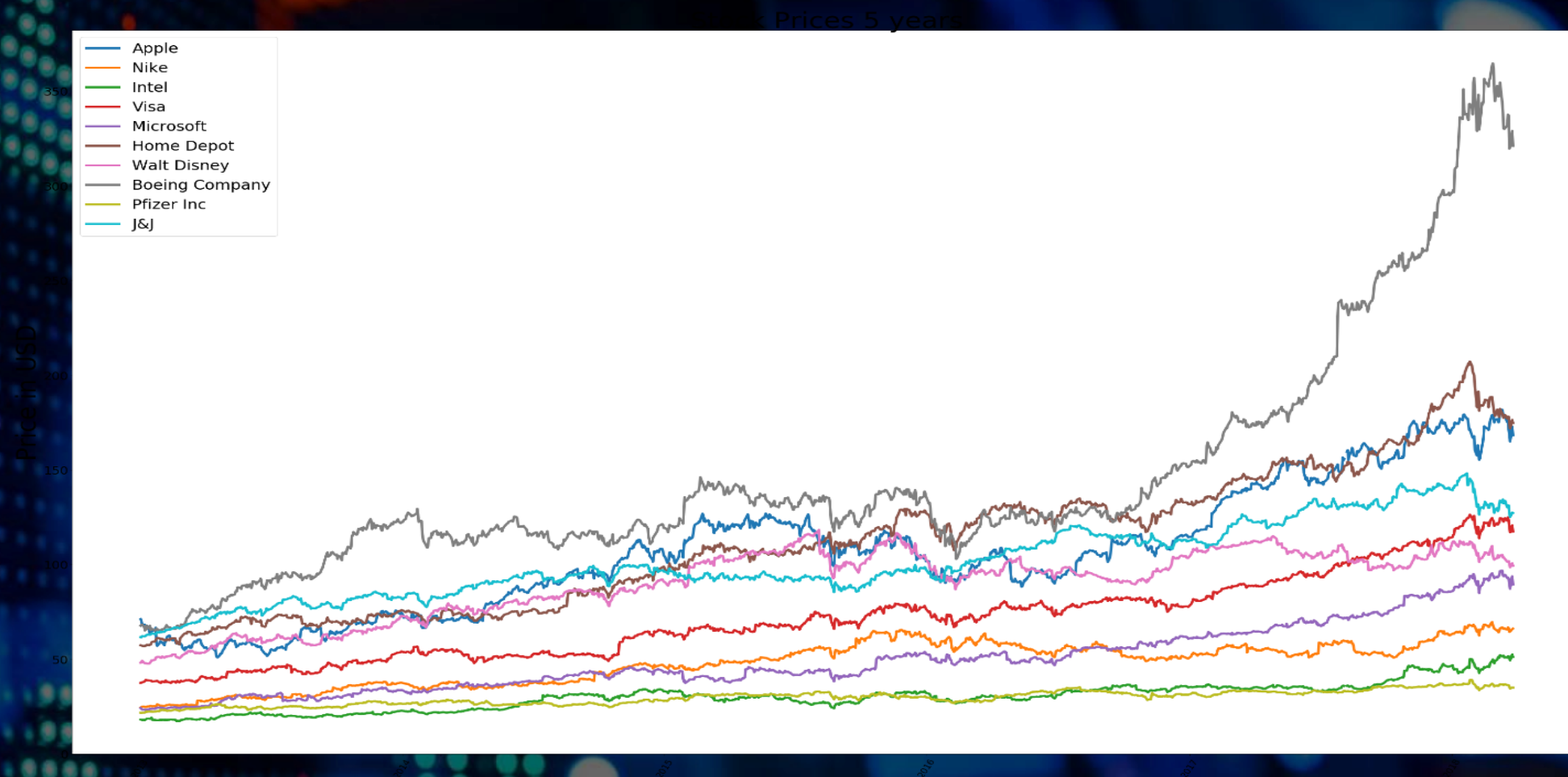
Date	Apple
2013-01-02	71.195748
2013-01-03	70.296565
2013-01-04	68.338996
2013-01-07	67.937002
2013-01-08	68.119845

DATA PROCESSING

```
PS C:\Users\sumit\OneDrive\Desktop\project> python .\portfolio.py
```

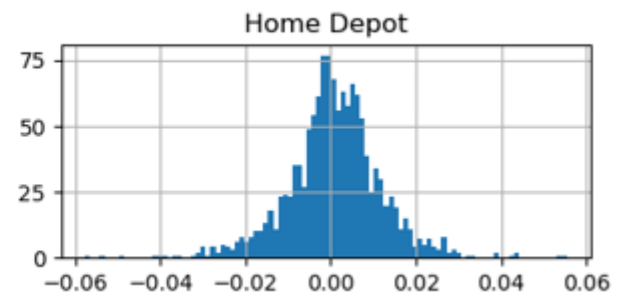
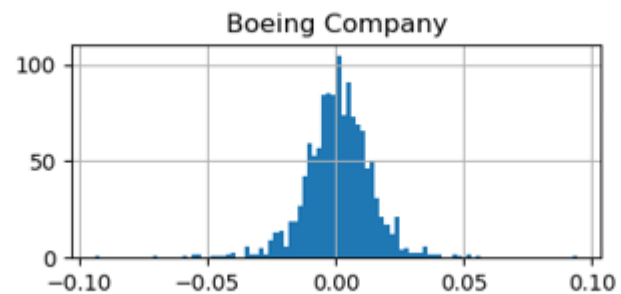
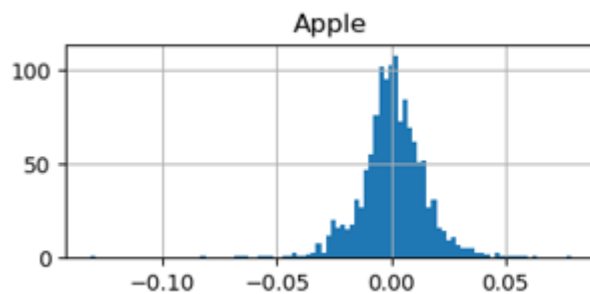
	Apple	Nike	Intel	Visa	Microsoft	Home Depot	Walt Disney	Boeing Company	Pfizer Inc	J&J
Date										
2013-01-02	71.195748	24.456742	18.101359	37.507891	24.194478	57.369885	48.169335	67.733862	21.715242	61.595109
2013-01-03	70.296565	24.706782	18.050560	37.536859	23.870367	57.207211	48.273026	68.085406	21.664955	61.508159
2013-01-04	68.338996	24.947387	17.915096	37.843430	23.423618	57.098761	49.196821	68.278756	21.757147	62.212451
2013-01-07	67.937002	24.985129	17.991295	38.113792	23.379820	56.792571	48.046790	66.907732	21.773909	62.082027
2013-01-08	68.119845	24.720935	17.855831	38.468642	23.257183	57.134911	47.848834	65.150009	21.807433	62.090722

GRAPH



LOG RETURNS

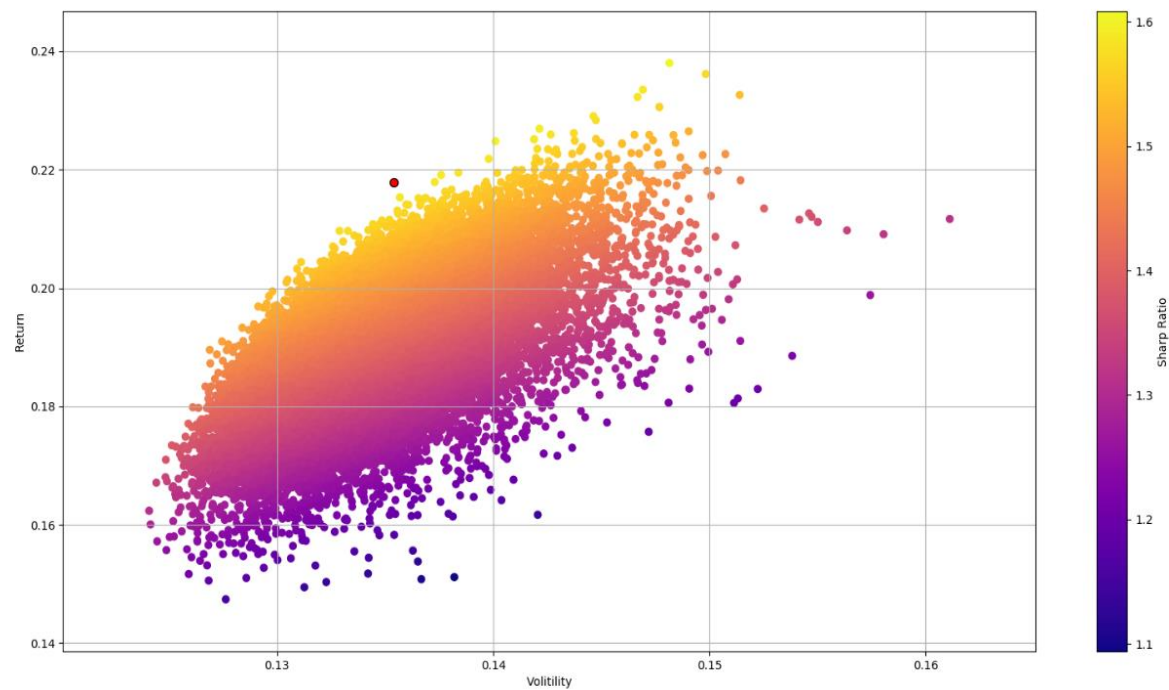
```
-----Daily return-----
      Apple      Nike      Intel      Visa      Microsoft      Home Depot      Walt Disney      Boeing Company      Pfizer Inc      J&J
Date
2013-01-02      NaN      NaN      NaN      NaN      NaN      NaN      NaN      NaN      NaN      NaN
2013-01-03 -0.012710  0.010172 -0.002810  0.000772 -0.013487 -0.002840  0.002150  0.005177 -0.002318 -0.001413
2013-01-04 -0.028242  0.009691 -0.007533  0.008134 -0.018893 -0.001898  0.018956  0.002836  0.004246  0.011385
2013-01-07 -0.005900  0.001512  0.004244  0.007119 -0.001872 -0.005377 -0.023654 -0.020284  0.000770 -0.002099
2013-01-08  0.002688 -0.010630 -0.007558  0.009267 -0.005259  0.006010 -0.004129 -0.026622  0.001538  0.000140
```



Monte Carlo Simulation

```
np.random.seed(1276)
# Finding optimum in 25000 repetitions
num_ports = 25000
all_weight = np.zeros((num_ports, len(stock.columns)))
ret_arr = np.zeros(num_ports)
vol_arr = np.zeros(num_ports)
sharp_arr = np.zeros(num_ports)
for i in range(num_ports):
    ... weight = np.array(np.random.random(10))
    ... weight = weight/np.sum(weight)
    ... #Save the weight
    ... all_weight[i, :] = weight
    ... # Expected Return
    ... ret_arr[i] = np.sum((log_ret.mean()* weight)*252)
    ... #Expected Volatility
    ... vol_arr[i] = np.sqrt(np.dot(weight, np.dot(log_ret.cov()*252, weight)))
    ... #Sharp Ratio
    ... sharp_arr[i] = ret_arr[i]/vol_arr[i]
```


MONTE CARLO SIMULATION



Monte Carlo Simulation

=====Random generated=====

Maximum Sharp Ratio(using random number generation) : 1.608817368819721

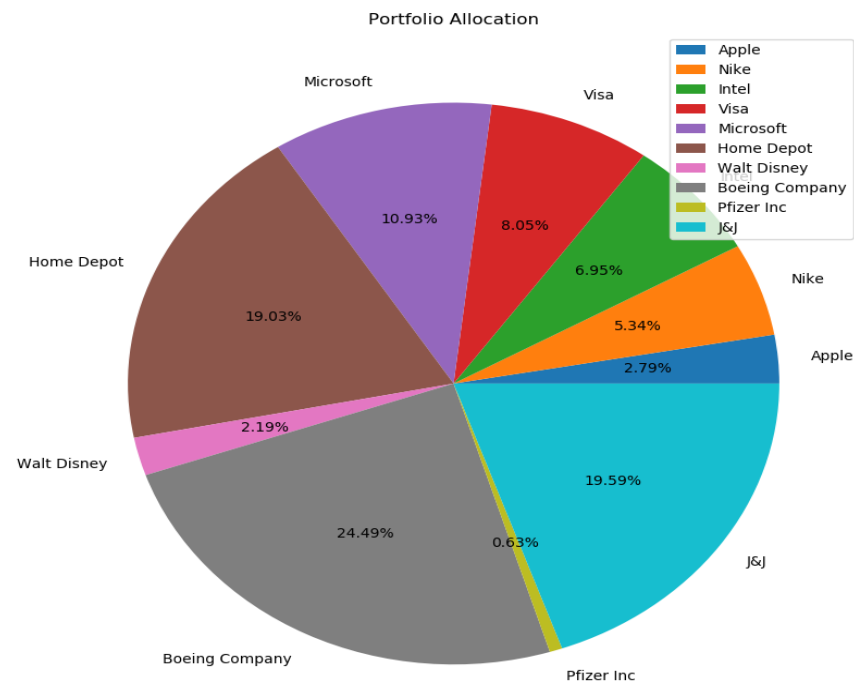
Maximum Sharpe Ratio Portfolio Allocation

	allocation
Apple	2.79
Nike	5.34
Intel	6.95
Visa	8.05
Microsoft	10.93
Home Depot	19.03
Walt Disney	2.19
Boeing Company	24.49
Pfizer Inc	0.63
J&J	19.59

Optimization using random ssampling graph generate...

=====

MONTE CARLO SIMULATION



MARKOWITS OPTIMIZATION

-----Mathematical Optimized Result set-----

fun: -1.6449962611572726

jac: array([3.19207013e-02, -4.72798944e-04, 4.73111868e-05, 5.46053052e-04,
-3.39001417e-04, 2.58386135e-05, 2.11826891e-01, 9.77218151e-05,
3.08924764e-01, -4.81456518e-05])

message: 'Optimization terminated successfully.'

nfev: 97

nit: 8

njev: 8

status: 0

success: True

x: array([2.15998543e-17, 4.54312129e-02, 1.50290440e-02, 7.87281837e-02,
1.77136337e-01, 2.57398502e-01, 6.03604739e-17, 3.54493409e-01,
6.63321610e-17, 7.17833118e-02])

MARKOWITS OPTIMIZATION

=====Mathematically Maximized=====

Mathematically Maximized Sharp ratio : 1.6449962611572726

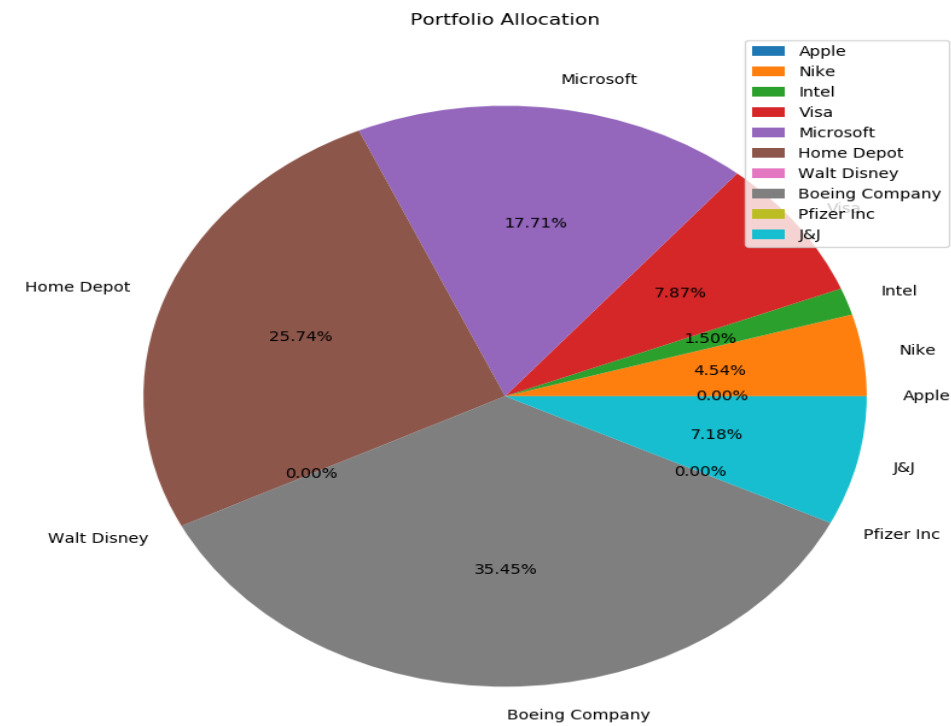
Maximum Sharpe Ratio Portfolio Allocation

	allocation
Apple	0.00
Nike	4.54
Intel	1.50
Visa	7.87
Microsoft	17.71
Home Depot	25.74
Walt Disney	0.00
Boeing Company	35.45
Pfizer Inc	0.00
J&J	7.18

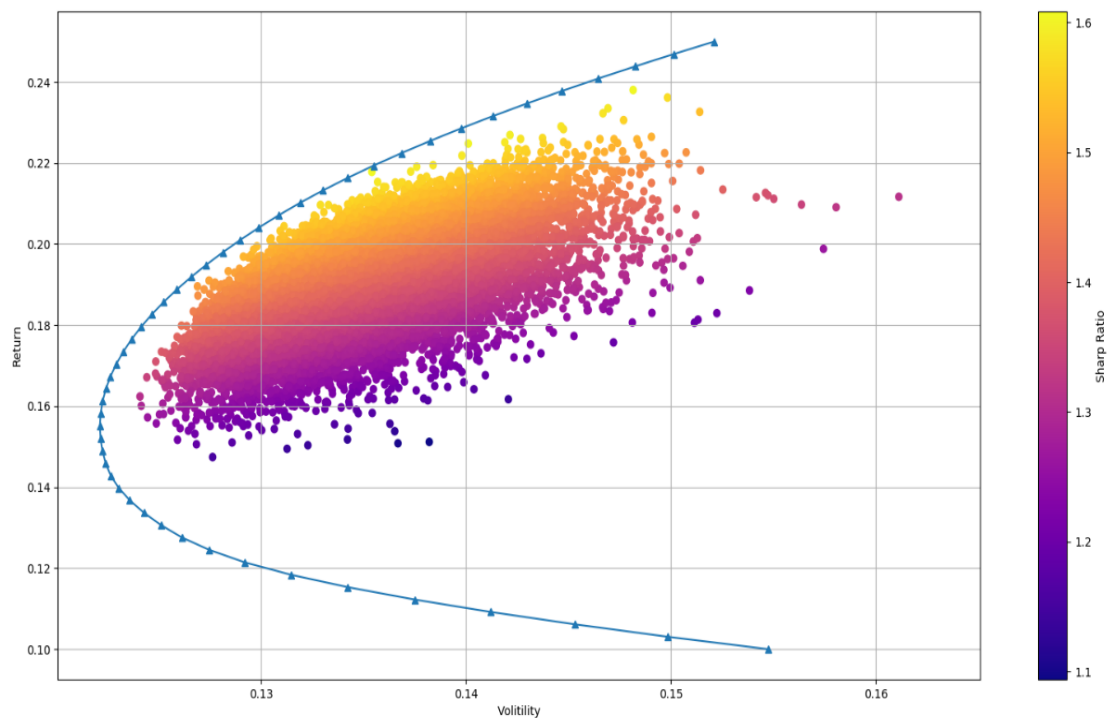
Mathematical Optimization and Efficient frontier graph generated...

=====

MARKOWITS OPTIMIZATION



ALL OPTIMAL PORTFOLIOS (EFFICIENT FRONTIER)





THANK YOU