```
In [18]:
           import pandas as pd
           import matplotlib.pyplot as plt
           from numpy import *
           from scipy import stats
           from numpy.linalg import multi dot
           #pip install cufflinks
           #import cufflinks as cf
           #cf.set config file(offline=True, dimensions=((1000,600)))
In [19]:
          # Nasdaq-listed stocklist
           symbols = ['NVDA', 'AAPL', 'JPM', 'AMZN', 'CAT', 'WMT', 'BA', 'JNJ', 'MCD', 'UNH',
           # Number of assets
          numofasset = len(symbols)
           # Number of portfolio for optimization
          numofportfolio = 5000
          #Retrieve Data
          df = pd.read excel('D:\\Derivatives Trading\\US equity data.xlsx', sheet name='Pric
                   NVDA
Out[19]:
                             AAPL
                                     JPM
                                                            WMT
                                                                                         UNH
                                                                                                  VZ
                                             AMZN
                                                      CAT
                                                                     BA
                                                                            JNJ
                                                                                  MCD
           Date
          2016-
                  23.3125
                           27.8650
                                    78.92
                                            38.1260
                                                     94.04
                                                            71.37 151.64 112.48
                                                                                120.68 157.59 50.960
          11-29
          2016-
                  23.0500
                           27.6300
                                    80.17
                                            37.5285
                                                     95.56
                                                            70.43 150.56 111.30 119.27 158.32 49.900
          11-30
          2016-
                                                            70.67 152.39 111.38 118.47 160.94 49.870
                  21.9100
                           27.3725
                                    81.79
                                            37.1825
                                                     96.24
          12-01
          2016-
                  22.1125
                           27.4750
                                    81.60
                                            37.0170
                                                     95.14
                                                            70.88 152.25 111.96 118.24 160.73 49.810
          12-02
          2016-
                  22.9700
                                            37.9680
                           27.2775
                                    83.26
                                                     94.45
                                                            69.94 152.16 111.94 119.29 157.63 49.750
          12-05
              •••
          2024-
                 682.2300 189.3000 175.10 169.1500 322.72 169.81 208.58 158.06 284.65 510.67 41.100
          02-06
          2024-
                 700.9900 189.4100 175.43 170.5300 323.59 169.38 211.92 157.98 287.33 519.39 40.420
          02-07
          2024-
                 696.4100 188.3200 174.80 169.8400
                                                   322.00 169.37
                                                                  209.22 156.40 291.93 520.09
          02-08
          2024-
                                                                  209.20 156.76
                 721.3300
                          188.8500
                                   175.01
                                          174.4500
                                                    317.16
                                                          169.28
                                                                               289.47
                                                                                        518.22 39.720
          02-09
          2024-
                 741.2500 187.7500 176.26 173.3150 320.80 169.90 209.71 157.67 290.01 516.47 39.975
          02-12
         1812 rows × 19 columns
In [20]: # Plot
```

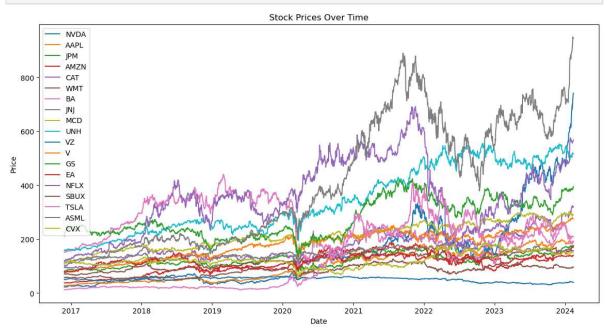
Assuming 'df' is a pandas DataFrame with datetime index and stock symbols as colu

file:///C:/Users/user/Downloads/US equity portfolio.html

and you want to plot the prices over time

```
# Plot all columns in the DataFrame
plt.figure(figsize=(14, 7)) # Set the figure size
for column in df.columns:
    plt.plot(df.index, df[column], label=column)

plt.title('Stock Prices Over Time')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend() # Add a legend to display the symbols
plt.show()
```



```
In [21]: # Calculate returns
  returns = df.pct_change().fillna(0)
  returns
```

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Out[21]: NVDA AAPL JPM AMZN CAT WMT BA JNJ

Date									
2016- 11-29	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000
2016- 11-30	-0.011260	-0.008434	0.015839	-0.015672	0.016163	-0.013171	-0.007122	-0.010491	-0.01
2016- 12-01	-0.049458	-0.009320	0.020207	-0.009220	0.007116	0.003408	0.012155	0.000719	-0.006
2016- 12-02	0.009242	0.003745	-0.002323	-0.004451	-0.011430	0.002972	-0.000919	0.005207	-0.00
2016- 12-05	0.038779	-0.007188	0.020343	0.025691	-0.007252	-0.013262	-0.000591	-0.000179	0.008
•••	•••	•••	***	•••	•••	•••	•••	•••	
2024- 02-06	-0.015995	0.008632	0.003438	-0.006811	0.004107	0.006818	0.009437	0.014506	-0.004
2024- 02-07	0.027498	0.000581	0.004005						
		0.000561	0.001885	0.008158	0.002696	-0.002532	0.016013	-0.000506	0.009
2024- 02-08	-0.006534	-0.005755	-0.003591	-0.004046	-0.004914	-0.002532 -0.000059	0.016013	-0.000506 -0.010001	0.009
	-0.006534 0.035784								

1812 rows × 19 columns

```
# Import optimization module from scipy
        import scipy.optimize as sco
        # Define portfolio stats function
        def portfolio_stats(weights):
            weights = array(weights)[:,newaxis]
            port_rets = weights.T @ array(returns.mean() * 252)[:,newaxis]
            port_vols = sqrt(multi_dot([weights.T, returns.cov() * 252, weights]))
            return array([port_rets, port_vols, port_rets/port_vols]).flatten()
        #Example 1 : Maximum sharpe ratio portfolio
        # Maximizing sharpe ratio
        def min_sharpe_ratio(weights):
            return -portfolio_stats(weights)[2]
        # Define initial weights
        initial_wts = numofasset * [1./numofasset]
        initial_wts
        # Each asset boundary ranges from 0 to 1 bounds
        bnds = tuple((0,1) for x in range(numofasset))
        # Specify constraints
        cons = (\{'type': 'eq', 'fun': lambda x: sum(x)-1\})
```

```
cons
          # Optimizing for maximum sharpe ratio
          opt_sharpe = sco.minimize(min_sharpe_ratio, initial_wts, method='SLSQP', bounds=bnd
          opt_sharpe
          bnds
Out[22]: ((0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1),
           (0, 1))
In [23]:
          # Portfolio weights
          list(zip(symbols, opt_sharpe['x']))
          [('NVDA', 0.273053733372647),
Out[23]:
           ('AAPL', 0.08671842689290042),
           ('JPM', 8.851964168858266e-17),
           ('AMZN', 7.061435335512556e-17),
           ('CAT', 0.06908207851383266),
           ('WMT', 0.22097735511694094),
           ('BA', 5.584220102641086e-17),
           ('JNJ', 1.7758401840368728e-17),
           ('MCD', 0.0325772972057475),
           ('UNH', 0.1523305674945298),
           ('VZ', 8.216394721800937e-17),
           ('V', 2.5704335403502673e-17),
           ('GS', 6.091325034838343e-17),
           ('EA', 1.323772414385636e-17),
           ('NFLX', 0.004014765956249008),
           ('SBUX', 0.0),
           ('TSLA', 0.12605456060649503),
           ('ASML', 0.03519121484065745),
           ('CVX', 1.2611413626254993e-16)]
In [25]: # Portfolio stats
          stats = ['Returns', 'Volatility', 'Sharpe Ratio']
          list(zip(stats, around(portfolio_stats(opt_sharpe['x']),4)))
          [('Returns', 0.3594), ('Volatility', 0.2614), ('Sharpe Ratio', 1.3747)]
Out[25]:
In [26]: #Example 2 : Minumum variance portfolio
          # Minimize the variance
          def min_variance(weights):
              return portfolio_stats(weights)[1]**2
          # Optimizing for minimum variance
          opt var = sco.minimize(min variance, initial wts, method='SLSQP', bounds=bnds, cons
          opt_var
```

```
message: Optimization terminated successfully
Out[26]:
           success: True
           status: 0
               fun: 0.020815086532275064
                 x: [ 0.000e+00 4.962e-18 ... 0.000e+00 2.231e-04]
              nit: 13
               jac: [ 5.735e-02 4.736e-02 ... 4.670e-02 4.193e-02]
              nfev: 261
              njev: 13
         # Portfolio weights
In [27]:
          list(zip(symbols, around(opt_var['x']*100,2)))
         [('NVDA', 0.0),
Out[27]:
           ('AAPL', 0.0),
           ('JPM', 0.0),
           ('AMZN', 3.11),
           ('CAT', 0.95),
           ('WMT', 18.36),
           ('BA', 0.0),
           ('JNJ', 24.53),
           ('MCD', 18.41),
           ('UNH', 0.0),
           ('VZ', 24.63),
           ('V', 0.0),
           ('GS', 0.0),
           ('EA', 9.99),
           ('NFLX', 0.0),
           ('SBUX', 0.0),
           ('TSLA', 0.0),
           ('ASML', 0.0),
           ('CVX', 0.02)]
In [28]: # Portfolio stats
          stats = ['Returns', 'Volatility', 'Sharpe Ratio']
          list(zip(stats, around(portfolio_stats(opt_var['x']),4)))
         [('Returns', 0.088), ('Volatility', 0.1443), ('Sharpe Ratio', 0.6102)]
Out[28]:
In [29]:
          #Example 3 : Efficient Frontier portfolio
          def min_volatility(weights):
              return portfolio_stats(weights)[1]
          # Efficient frontier params
          targetrets = linspace(0.30, 0.60, 100)
          tvols = []
          for tr in targetrets:
              ef_cons = ({'type': 'eq', 'fun': lambda x: portfolio_stats(x)[0] - tr},
                         {'type': 'eq', 'fun': lambda x: sum(x) - 1})
              opt_ef = sco.minimize(min_volatility, initial_wts, method='SLSQP', bounds=bnds,
              tvols.append(opt_ef['fun'])
          targetvols = array(tvols)
          # Dataframe for EF
          efport = pd.DataFrame({
              'targetrets' : around(100*targetrets[14:],2),
              'targetvols': around(100*targetvols[14:],2),
              'targetsharpe': around(targetrets[14:]/targetvols[14:],2)
          })
```

```
efport.head(5)
```

targetrets targetvols targetsharpe Out[29]: 0 34.24 24.93 1.37 1 34.55 25.14 1.37 2 34.85 25.36 1.37 3 35.15 25.57 1.37 4 35.45 25.79 1.37

```
# Plot efficient frontier portfolio
In [30]:
          #pip install plotly
          import plotly.express as px
          # Plot starts
          fig = px.scatter(
              efport, x='targetvols', y='targetrets', color='targetsharpe',
              labels={'targetrets': 'Expected Return', 'targetvols': 'Expected Volatility',
              title="Efficient Frontier Portfolio"
          ).update_traces(mode='markers', marker=dict(symbol='cross'))
          # Plot maximum sharpe portfolio
          fig.add_scatter(
              mode='markers',
              x=[100*portfolio_stats(opt_sharpe['x'])[1]],
              y=[100*portfolio_stats(opt_sharpe['x'])[0]],
              marker=dict(color='red', size=20, symbol='star'),
              name='Max Sharpe'
          ).update(layout_showlegend=True)
          # Plot minimum variance portfolio
          fig.add_scatter(
              mode='markers',
              x=[100*portfolio_stats(opt_var['x'])[1]],
             y=[100*portfolio_stats(opt_var['x'])[0]],
              marker=dict(color='green', size=20, symbol='star'),
              name='Min Variance'
          ).update(layout showlegend=True)
          # Show spikes
          fig.update_xaxes(showspikes=True)
          fig.update_yaxes(showspikes=True)
          fig.show()
```

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In []:	
In []:	