```
In [1]: import yfinance as yf
        import numpy as np
        import pandas as pd
         import matplotlib.pyplot as plt
In [2]: sp_df = yf.download('^GSPC', start='2000-01-01', end='2014-12-31')
        df = sp_df.loc[:, ['Adj Close']]
        df = df.rename(columns={'Adj Close': 'Price'})
        df.head(10)
        [********* 100%********* 1 of 1 completed
Out[2]:
                         Price
              Date
        2000-01-03 1455.219971
        2000-01-04 1399.420044
        2000-01-05 1402.109985
        2000-01-06 1403.449951
        2000-01-07 1441.469971
        2000-01-10 1457.599976
        2000-01-11 1438.560059
        2000-01-12 1432.250000
        2000-01-13 1449.680054
        2000-01-14 1465.150024
In [3]: def get_vol(data, frequency):
            function:calculate the annural volatility in an given frequency
            data:pandas dataframe or numpy array
            frequency:int, represent the sampling frequency in days
            data = np.array(data)
            fprices = data[::frequency]
            log_return = np.log((fprices/np.roll(fprices,1))[1:])
            length_ = log_return.shape[0]
            delta_t = frequency/252
            mu = np.sum(log_return/length_)
            vol = np.sum(np.square(log return-mu)/(length -1))
            ann_sigma = np.sqrt(vol/delta_t)
            return ann_sigma
In [4]:
        def calculate_exposure(df_, lookback=252):
            function:calculate the exposure of the given strategy
            df :dataframe with a column named Price
            lookback:int, days to lookback
```

```
df_['temp'] = 0
            daily_vol = get_vol(df_['Price'], frequency=1)
            weekly_vol = get_vol(df_['Price'], frequency=5)
            df_['temp'] = np.where(daily_vol > weekly_vol, 1,
                                        np.where(daily_vol < weekly_vol, -1, 0))</pre>
            df_.iloc[:lookback, df_.columns.get_loc('temp')] = 0
            df_['First Day'] = df_.index - pd.to_timedelta(df_.index.dayofweek, unit='d')
            df_['First Day Price'] = df_.groupby('First Day')['Price'].transform('first')
            exposures = ((1 / df ['Price']) - (1 / df ['First Day Price'])) * df ['temp']
             return exposures.to frame(name='Exposure')
In [5]:
        def strategy performance(cash, df, lookback=252):
            function:calculate the performance of the strategy
            df_ = calculate_exposure(df, lookback=lookback)
            df_['Daily P/L'] = cash * df_['Exposure'].shift(1) * (df['Price'] - df['Price'
            df_['Cumulative P/L'] = df_['Daily P/L'].cumsum()
            df_['Return'] = df_['Cumulative P/L'].pct_change()
            df_ = df_.replace(np.inf,0)
            df_['Cumulative Return'] = (1 + df_['Return']).cumprod() - 1
             return df_.drop(['Daily P/L','Return'], axis=1)
```

In [6]: performance = strategy_performance(10000, df, lookback=252)
 performance

Out[6]: Exposure Cumulative P/L Cumulative Return

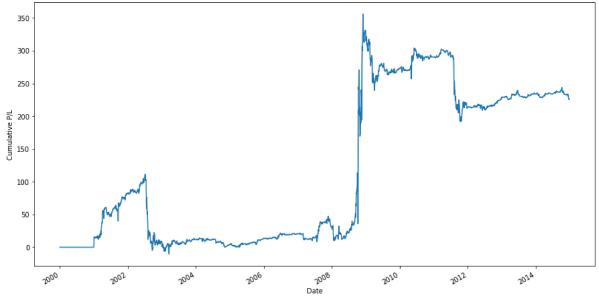
Date

2000-01-03	0.000000e+00	NaN	NaN
2000-01-04	0.000000e+00	0.000000	NaN
2000-01-05	0.000000e+00	0.000000	NaN
2000-01-06	0.000000e+00	0.000000	NaN
2000-01-07	0.000000e+00	0.000000	NaN
•••	•••	•••	
2014-12-23	-8.387220e-07	225.994552	41.748219
2014-12-24	-7.718130e-07	225.996985	41.748679
2014-12-26	-2.356274e-06	225.943806	41.738620
2014-12-29	0.000000e+00	225.901392	41.730598
2014-12-30	2.349895e-06	225.901392	41.730598

3772 rows × 3 columns

```
In [7]: fig = plt.figure(figsize=(15, 8))
```

```
performance['Cumulative P/L'].plot()
plt.ylabel("Cumulative P/L")
plt.show()
```



In []: