```
import pandas as pd
         import numpy as np
         import yfinance as yf
         import datetime as dt
         import matplotlib.pyplot as plt
         import seaborn as sns
         import random
        Defaulting to user installation because normal site-packages is not writeable
        Requirement already satisfied: yfinance in /opt/tljh/user/lib/python3.9/site-packages (0.1.66)
        Requirement already satisfied: pandas>=0.24 in /opt/tljh/user/lib/python3.9/site-packages (from yfinance) (1.3.4)
        Requirement already satisfied: requests>=2.20 in /opt/tljh/user/lib/python3.9/site-packages (from yfinance) (2.26.0)
        Requirement already satisfied: lxml>=4.5.1 in /opt/tljh/user/lib/python3.9/site-packages (from yfinance) (4.6.4)
        Requirement already satisfied: numpy>=1.15 in /opt/tljh/user/lib/python3.9/site-packages (from yfinance) (1.19.5)
        Requirement already satisfied: multitasking>=0.0.7 in /opt/tljh/user/lib/python3.9/site-packages (from yfinance) (0.0.9)
        Requirement already satisfied: python-dateutil>=2.7.3 in /opt/tljh/user/lib/python3.9/site-packages (from pandas>=0.24->yfinance) (2.8.2)
        Requirement already satisfied: pytz>=2017.3 in /opt/tljh/user/lib/python3.9/site-packages (from pandas>=0.24->yfinance) (2021.3)
        Requirement already satisfied: idna<4,>=2.5 in /opt/tljh/user/lib/python3.9/site-packages (from requests>=2.20->yfinance) (3.1)
        Requirement already satisfied: certifi>=2017.4.17 in /opt/tljh/user/lib/python3.9/site-packages (from requests>=2.20->yfinance) (2021.10.8)
        Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/tljh/user/lib/python3.9/site-packages (from requests>=2.20->yfinance) (1.26.7)
        Requirement already satisfied: charset-normalizer~=2.0.0 in /opt/tljh/user/lib/python3.9/site-packages (from requests>=2.20->yfinance) (2.0.8)
        Requirement already satisfied: six>=1.5 in /opt/tljh/user/lib/python3.9/site-packages (from python-dateutil>=2.7.3->pandas>=0.24->yfinance) (1.15.0)
In [2]: # Step 1 - Obtain tickers of SPY, (the chosen etf) and place them in a list
         state_street = list(pd.read_excel("https://www.ssga.com/us/en/intermediary/etfs/library-content/products/fund-data/etfs/us/holdings-daily-us-en-spy.xlsx", header=4).dropna().Ticker)
         blackrock = list(pd.read_csv("https://www.blackrock.com/us/individual/products/239726/ishares-core-sp-500-etf/1464253357814.ajax?fileType=csv&fileName=IVV_holdings&dataType=fund", header = 9).Ticker.dropna())
         tickers = list(set(state_street) & set(blackrock))
In [3]: # step 2 -> Obtain the data metrics we need being adj close + Volume + returns from yahoo finance
         data = yf.download(tickers, period='7y')[["Adj Close", "Volume"]]
         data = data.dropna(how="all", axis=1)
         close = data['Adj Close']
         volume = data['Volume']
         returns = data['Adj Close'].pct_change().resample("M").sum().shift(-1)
         Γ******** 500 of 500 completed
In [4]:
         # step 3 -> Calculating price momentum factor scores
         class Momentum_Factors():
           def __init__(self, stock_data, tickers):
             self.price = close
             self.volume = volume
             self.returns = returns
             self.tickers = tickers
             self.mom_factors_df = pd.DataFrame(index = self.tickers)
         # Calculates the 52 week price trend line
           def fifty_two_trend_line(self):
             self.fifty_two_trend_line_df = pd.DataFrame(index=self.tickers)
             for i in range(1, 21):
                   self.fifty_two_trend_line_df[i] = np.polyfit(range(len(self.price[-i - 252 : -i])), self.price[-i -252 : -i], 1)[0]
             return self.fifty_two_trend_line_df.mean(axis=1)
           def factors(self):
         # Gives the percent increase from 252 day low
                 self.roll_min = self.price.rolling(252).min()
                 self.mom_factors_df["pct_above"] = ((self.price - self.roll_min) / self.roll_min)[-20:].mean()
         # Gives the momentum factor of price oscillator
                 self.mom_factors_df["price_oscillator"] = ((self.price.rolling(20).mean() - self.price.rolling(260).mean())
                  /self.price.rolling(260).std())[-20:].mean()
         # Gives the 39 week return momentum factor
                 self.mom_factors_df["39_wk_ret"] = self.price.pct_change(189)[-20:].mean()
         # Gives the 51 week volume momentum factor
                 self.mom_factors_df["price_volume"] = (
                    (self.returns * self.volume).rolling(252).sum()[-20:].mean())
         # Places the 52 week trend into mom_factors_df
                 self.mom_factors_df["fifty_two_trend_line"] = self.fifty_two_trend_line()
                 return self.mom_factors_df
         ## Step 4 -> Aggregating the factor scores using z-score
           def aggregate(self):
               self.mom_factors_df = self.factors()
               self.scored = ((self.mom_factors_df - self.mom_factors_df.mean()) / self.mom_factors_df.std()).sum(axis=1)
               return self.scored
         ## step 5 -> Ranking the stocks
           def baskets(self):
               self.z_scored = self.aggregate()
               return self.z_scored.nlargest(10).index, self.z_scored.nsmallest(10).index
In [5]: # Step 6 -> 5 year backtest to validate the trading algorithm
         dates_range = pd.DataFrame(pd.date_range("2017-12-18", dt.datetime.today(), freq="M"))
         dates_range.rename(columns={0: "date"}, inplace=True)
         long_returns, short_returns = pd.DataFrame(), pd.DataFrame()
         for date in dates_range.date:
           MF = Momentum_Factors(data[data.index <= date], tickers)</pre>
           longs, short = MF.baskets()
           long_returns = long_returns.append(returns[longs].loc[date])
           short_returns = short_returns.append(returns[short].loc[date])
         long_returns.iloc[-1].dropna()
        ENPH -0.046497
        0XY
              -0.101937
              -0.101345
        VL0
              -0.108636
        EQT
              -0.124717
        XOM
              -0.060144
        CAH
              -0.009352
               0.001567
        MRO -0.110440
        SLB -0.034990
        Name: 2022-11-30 00:00:00, dtype: float64
In [7]: # Step 7 -> Plotting the selected stocks and comparing them to the S&P 500
         sum_returns = pd.DataFrame(long_returns.mean(axis=1) - short_returns.mean(axis=1)).shift()
         sum_returns = sum_returns[-60:]
         sum_returns["cum_ret"] = np.exp(np.log1p(sum_returns).cumsum())
         sum_returns["SPY"] = (
             yf.download("SPY", period="7y")["Adj Close"].pct_change().resample("M").sum()
         In [8]: # Plotting a Line Graph Comparing the Portfolio Returns vs the ETF SPY Returns
         plt.plot(sum_returns["cum_ret"],
                  label="SPY Returns",
                 color="g")
         plt.plot(sum_returns["SPY"],
                  label="Portfolio Returns",
                 color="b")
         plt.legend()
         plt.title('Portfolio Returns vs SPY Returns');
                    Portfolio Returns vs SPY Returns
             — SPY Returns
                Portfolio Returns
         1.2
         1.0
         0.8
         0.6
         0.4
         0.2
            2018
         # Plotting the Monthly Return Differential Between Our Portfolio Returns & Spy
         sum_returns['colors'] = (sum_returns[0] > 0)
         sum_returns[0].plot(kind='bar'
                           , figsize=(15,5)
                           , title='Monthly Returns Differential Between Portfolio and SPY'
                           , xlabel='Dates'
                           , ylabel='Returns'
                           , color=['purple' if i == True else 'blue' for i in sum_returns['colors']]);
                                             Monthly Returns Differential Between Portfolio and SPY
           0.2
           0.1
           -0.1
           -0.2
              In [10]:
         # Plotting the Long vs Short Returns
         fig = plt.subplots(figsize=(20,5))
         plt.title('Long vs Short Return Comparison');
         long_returns.mean(axis=1).plot(
             color="purple",
             label="Long Returns"
         short_returns.mean(axis=1).plot(
             color="blue",
             label="Short Returns"
         plt.legend()
         plt.show()
                                                                    Long vs Short Return Comparison
          0.4
                                                                                                                                      Long Returns
                                                                                                                                         Short Returns
          0.3
          0.2
          0.1
          0.0
         -0.1
         -0.2
         -0.3
             2018
                                         2019
                                                                    2020
                                                                                               2021
                                                                                                                           2022
In [11]:
         # Credit to professor John Droescher's code
 In [ ]:
 In [ ]:
 In [ ]:
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!pip install yfinance