```
1 !nin install vfinance
          Looking in indexes: <a href="https://gypi.org/simple">https://gypi.org/simple</a>, <a href="https://gypi.org/simple">https://gypi.org/simple</a>, <a href="https://gypi.org/simple/">https://gypi.org/simple</a>, <a href="https://gypi.org/simple/">https://g
          Requirement already satisfied: multitasking>=0.0.7 in /usr/local/lib/python3.8/dist-packages (from yfinance) (0.0.11) Requirement already satisfied: pytz>=2022.5 in /usr/local/lib/python3.8/dist-packages (from yfinance) (2022.7.1)
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Requirement already satisfied: requests>=2.26 in /usr/local/lib/python3.8/dist-packages (from yfinance) (2.28.2)
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Requirement already satisfied: six>=1.2 in /usr/local/lib/python3.8/dist-packages (from html5lib>=1.1->yfinance) (1.15.0)
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          Requirement already satisfied: webencomings in /osi/local/lib/python3.8/dist-packages (from pandas>=1.3.0->yfinance) (2.8.2)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.8/dist-packages (from requests>=2.26->yfinance) (3.0.1)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.8/dist-packages (from requests>=2.26->yfinance) (2022.12.7)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.8/dist-packages (from requests>=2.26->yfinance) (1.24.3)
          Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.8/dist-packages (from requests>=2.26-yyfinance) (2.10)
Requirement already satisfied: pycparser in /usr/local/lib/python3.8/dist-packages (from cffi>=1.12->cryptography>=3.3.2->yfinance) (2.21)
1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import yfinance as yf
1 price_history = yf.Ticker('TSLA').history(period='2y', # valid periods: 1d,5d,1mo,3mo,6mo,1y,2y,5y,10y,ytd,max
                                                                                               interval='1d', # valid intervals: 1m,2m,5m,15m,30m,60m,90m,1h,1d,5d,1wk,1mo,3mo
                                                                                               actions=False)
1 price history
                                                                      0pen
                                                                                                    High
                                                                                                                                    Low
                                                                                                                                                            Close
                                                                                                                                                                                     Volume
                                        Date
                    2021-03-01
                                                        230 036667 239 666672 228 350006 239 476669
                                                                                                                                                                                81408600
                00:00:00-05:00
                    2021-03-02
                                                        239 426666 240 369995 228 333328 228 813339 71196600
                00:00:00-05:00
                    2021-03-03
                                                        229.330002 233.566666 217.236664 217.733337 90624000
                00:00:00-05:00
                    2021-03-04
                                                        218.600006 222.816666 200.000000 207.146667 197758500
                00:00:00-05:00
                    2021-03-05
                                                        208.686661 209.279999 179.830002 199.316666 268189500
                00:00:00-05:00
                    2023-02-21
                                                        204.990005 209.710007 197.220001 197.369995 180018600
                00:00:00-05:00
1 def find volatility(ticker):
          data = yf.Ticker(ticker).history(period='3mo', interval='1d', actions=False)
          data['Log returns'] = np.log(data['Close']/data['Close'].shift())
          volatility = data['Log returns'].std()*252**.5
          return volatility
1 def find_beta_against_spy(ticker):
2   data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
          data['Log returns'] = np.log(data['Close']/data['Close'].shift())
          data_spy = yf.Ticker('SPY').history(period='12mo', interval='1d', actions=False)
data_spy['Log returns'] = np.log(data_spy['Close']/data_spy['Close'].shift())
          cov = data['Log returns'].cov(data_spy['Log returns'])
          var = data_spy['Log returns'].var()
         beta_spy = cov/var
return beta_spy
1 def find_beta_against_iwm(ticker):
          data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
          data['Log returns'] = np.log(data['Close']/data['Close'].shift())
```

```
1 def find_beta_against_iwm(ticker):
2   data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
3   data['log returns'] = np.log(data['close']/data['close'].shift())
4   data_iwm = yf.Ticker('iwm').history(period='12mo', interval='1d', actions=False)
5   data_iwm['log returns'] = np.log(data_iwm['close']/data_iwm['Close'].shift())
6   cov = data['log returns'].cov(data_iwm['log returns'])
7   var = data_iwm['log returns'].var()
8   beta_iwm = cov/var
9   return beta_iwm

1 def find_beta_against_dia(ticker):
2   data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
3   data['log returns'] = np.log(data_i'close']/data['close'].shift())
4   data_dia = yf.Ticker('dia').history(period='12mo', interval='1d', actions=False)
5   data_dia['log returns'] = np.log(data_dia['close']/data_dia['close'].shift())
6   cov = data_dia['log returns'] = np.log(data_dia['close']/data_dia['close'].shift())
7   var = data_dia['log returns'].cov(data_dia['log returns'])
8   beta_dia = cov/var
9   return beta_dia
```

```
1 def find average weekly drawdown(ticker):
    data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
    avg_weekly_drawdown = (data['Close'].min() - data['Close'].max())/data['Close'].max()
     return avg_weekly_drawdown
1 def find_maximum_weekly_drawdown(ticker):
    data = yf.Ticker(ticker).history(period='12mo', interval='1d', actions=False)
    avg_maximum_drawdown = (data['Close'].min() - data['Close'].max())/data['Close'].max()
    return avg_maximum_drawdown
1 def find return(ticker):
    data = yf.Ticker(ticker).history(period='10y', interval='3mo', actions=False)
    ret = ((data['Close'].iloc[-1] - data['Close'].iloc[0])/data['Close'].iloc[0])
    return ret
1 def find_annualized_return(ticker):
    data = yf.Ticker(ticker).history(period='10y', interval='3mo', actions=False)
    ret = ((data['Close'].iloc[-1] - data['Close'].iloc[0])/data['Close'].iloc[0])
    annualized_ret = ((1+ret)**(1/10)) - 1
    return annualized_ret
1 tickers = ['TSLA', 'AAPL', 'MSFT', 'AMZN', 'UNH', 'GOOGL', 'NVDA']
1 table1 = pd.DataFrame(columns=['Ticker', 'Portfolio Weight (equally weighted)', 'Annualized Volatility (using trailing 3-months)', 'Beta against SPY (using trailing 12-m
1 table1['Ticker'] = tickers
2 table1['Portfolio Weight (equally weighted)'] = [1.0 for i in range(len(tickers))]
3 table1['Annualized Volatility (using trailing 3-months)'] = list(map(find_volatility, tickers))
4 table1['Beta against SPY (using trailing 12-months)'] = list(map(find_beta_against_spy, tickers))
5 table1['Beta against IWM (using trailing 12-months)'] = list(map(find_beta_against_iwm, tickers))
 6 table1['Beta against DIA (using trailing 12-months)'] = list(map(find_beta_against_dia, tickers))
7 table1['Average Weekly Drawdown (52-week Low minus 52-week High) / 52-week High'] = list(map(find_average_weekly_drawdown, tickers)) 8 table1['Maximum Weekly Drawdown (52-week Low minus 52-week High) / 52-week High'] = list(map(find_maximum_weekly_drawdown, tickers))
 9 table1['Total Return (using trailing 10-years)'] = list(map(find_return, tickers))
10 table1['Annualized Total Return (using trailing 10-years)'] = list(map(find_annualized_return, tickers))
1 table1.head(len(tickers))
                                                                                      Maximum
                                                                           Average
                                                                           Weeklv
                                                                                      Weeklv
                                              Beta
                                                         Beta
                                                                   Beta
                                                                          Drawdown
                                                                                    Drawdown
                                                                                                   Total Annualized
                             Annualized
                                          against
                                                     against
                                                                against
                                                                                     (52-week
                                                                          (52-week
                 Portfolio
                                                                                                  Return
                                                                                                               Total
                             Volatility
                                               SPY
                                                         IWM
                                                                    DIA
                    Weight
                                                                                                               Return
                                            (using
                                                      (using
                                                                 (using
         Ticker
                                 (using
                  (equally
                                                                            minus
                                                                                       minus
                                                                                                trailing
                                                                                                              (using
                               trailing
                                          trailing
                                                    trailing
                                                               trailing
                                                                           52-week
                                                                                      52-week
                                                                                                             trailing
                  weighted)
                              3-months)
                                               12-
                                                         12-
                                                                    12-
                                                                           High) /
52-week
                                                                                     High) /
52-week
                                                                                                  vears)
                                                                                                           10-years)
                                           months)
                                                     months)
                                                                months)
                                                                              High
                                                                                        High
          TSLA
                        1.0
                               0.500054
                                         1.303787
                                                    1.037119
                                                              1.474275 -0.298248 -0.298248
                                                                                                9.893495
                                                                                                            0.269746
     1
          AAPL
                        1.0
                               0.294755
          MSFT
                        1.0
                               0.325976
                                         1.295466
                                                    0.998773
                                                               1.466260 -0.317698 -0.317698
                                                                                                8.188563
                                                                                                             0.248317
                                                                                                            0.221389
     3
          AMZN
                        1.0
                               0.417194 1.657138 1.361745 1.776833 -0.516759 -0.516759
                                                                                                6.388204
                                                                                                8.405080
                               0.251227
                        1.0
1 etfs = ['EWD', 'SLX', 'C000']
1 df = yf.Ticker(tickers[0]).history(period='10y', interval='1d', actions=False)['Close']
2 for t in tickers[1:]:
   df = df + yf.Ticker(t).history(period='10y', interval='1d', actions=False)['Close']
1 df.head()
     2013-02-28 00:00:00-05:00
    2013-03-01 00:00:00-05:00
                                   121.075366
     2013-03-04 00:00:00-05:00
     2013-03-05 00:00:00-05:00
                                   122.868421
     2013-03-06 00:00:00-05:00
    Name: Close, dtvpe: float64
1 df.tail()
    Date
2023-02-21 00:00:00-05:00
                                   1482.749992
     2023-02-22 00:00:00-05:00
                                   1485.150009
     2023-02-23 00:00:00-05:00
                                   1521.280006
    2023-02-24 00:00:00-05:00
                                   1492,629997
     2023-02-27 00:00:00-05:00
                                   1507.670013
    Name: Close, dtype: float64
1 df/=len(tickers)
1 df.tail()
     2023-02-21 00:00:00-05:00
                                   211.821427
     2023-02-22 00:00:00-05:00
                                   212.164287
     2023-02-23 00:00:00-05:00
                                   217.325715
     2023-02-27 00:00:00-05:00
                                   215.381430
     Name: Close, dtype: float64
```

```
1 def find_etf_correlation(ticker):
2 data = yf.Ticker(ticker).history(period='10y', interval='1d', actions=False)['Close']
3 corr = data.corr(df)
     return corr
1 def find_etf_covariance(ticker):
2 data = yf.Ticker(ticker).history(period='10y', interval='1d', actions=False)['Close']
3 cov = data.cov(df)
   return cov
1 def find_etf_tracking_errors(ticker):
2 data = yf.Ticker(ticker).history(period='10y', interval='1d', actions=False)['Close']
     tracking_error = np.std((df.values - data.values) / df.values)
4 return tracking_error
1 def find_etf_sharpe_ratio(ticker):
2 data = yf.Ticker(ticker).history(period='10y', interval='1d', actions=False)['Close'].pct_change().dropna()
     data = data.values
     risk_free_Rate = 0.0
     mean_daily_return = sum(data)/len(data)
6 std = np.std(data)
    daily_sharpe_ratio = (mean_daily_return - risk_free_Rate) / std
8 sharpe_ratio = 252**(1/2) * daily_sharpe_ratio
9 return sharpe_ratio
1 def find_etf_annualized_volatility_spread(ticker):
data = yf.Ticker(ticker).history(period='10y', interval='1d', actions=False)
     data['Log returns'] = np.log(data['Close']/data['Close'].shift())
     etf_volatility = data['Log returns'].std()*252**.5
5 portfolio_log_returns = np.log(df/df.shift())
6 portfolio_volatility = portfolio_log_returns.std()*252**.5
7 return portfolio_volatility - etf_volatility
1 table2 = pd.DataFrame(columns=['ETF Ticker', 'Correlation against ETF', 'Covariance of Portfolio against ETF', 'Tracking Errors (using trailing 10-years)', 'Sharpe Ratic
1 table2['ETF Ticker'] = etfs
2 table2['Correlation against ETF'] = list(map(find_etf_correlation, etfs))
3 table2['Covariance of Portfolio against ETF'] = list(map(find_etf_covariance, etfs))
4 table2['Tracking Errors (using trailing 10-years)'] = list(map(find_etf_tracking_errors, etfs))
5 table2['Sharpe Ratio (using current risk-free rate)'] = list(map(find_etf_sharpe_ratio, etfs))
6 table2['Annualized Volatility (252 days) Spread (Portfolio Volatility - ETF Volatility)'] = list(map(find_etf_annualized_volatility_spread, etfs))
1 table2.head()
                                                                                  Sharpe Ratio
Sharpe Ratio
Annualized Volatility (252
                                        Covariance of Tracking Errors
Portfolio (using trailing
                      Correlation
```

70819 0.30294	2 0.320300	0.003271
32297 0.38579	0.387549	-0.091469
33906 0.34095	3 0.375248	-0.073745
	33906 0.34095	33906 0.340953 0.375248

```
1 table3 = pd.DataFrame()
2 table3['EquallyWeightedPortfolio'] = df
3 correlation_matrix_tickers = etfs + tickers
4 for t in correlation_matrix_tickers:
5 table3[t] = yf.Ticker(t).history(period='10y', interval='1d', actions=False)['Close']
```

1 table3.corr()

₽		EquallyWeightedPortfolio	EWD	SLX	cąąą	TSLA	AAPL	MSFT	
	EquallyWeightedPortfolio	1.000000	0.874130	0.794261	0.667982	0.945736	0.991506	0.991801	C
	EWD	0.874130	1.000000	0.790014	0.787748	0.858171	0.859056	0.869914	С
	SLX	0.794261	0.790014	1.000000	0.458323	0.772197	0.785749	0.757554	C
	CQQQ	0.667982	0.787748	0.458323	1.000000	0.576914	0.641645	0.684314	C
	TSLA	0.945736	0.858171	0.772197	0.576914	1.000000	0.949035	0.917264	С
	AAPL	0.991506	0.859056	0.785749	0.641645	0.949035	1.000000	0.983074	С
	MSFT	0.991801	0.869914	0.757554	0.684314	0.917264	0.983074	1.000000	C
	AMZN	0.917271	0.845088	0.620575	0.830828	0.807319	0.892662	0.936343	1
	UNH	0.966159	0.773046	0.795778	0.590329	0.862377	0.956960	0.954339	С
	GOOGL	0.983111	0.903593	0.786078	0.693499	0.924280	0.961525	0.979088	С
	NVDA	0.975522	0.896157	0.803811	0.647499	0.938622	0.960729	0.962923	C
					+ Code	+ Te	ext		-
1 table3.head()									

	EquallyWeightedPortfolio	EWD	SLX	cąąą	TSLA	AAPL	MSFT	AMZN	
1									
2013-02- 28 00:00:00- 05:00	17.290014	22.332064	31.586666	21.968330	2.322000	13.615319	23.055151	13.2135	45
2013-03- 01 00:00:00-	17.296481	22.532320	31.019159	21.903473	2.310000	13.278174	23.179544	13.2870	45
4									>

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