import numpy as np import util as ut import matplotlib.pyplot as plt import datetime as dt import random from indicators import * from marketsimcode import * import ManualStrategy as ms import StrategyLearner as sl from marketsimcode import compute portvals def createBenchmark(symbol="JPM", sd=dt.datetime(2008, 1, 1), ed=dt.datetime(2009, 12, 31), sv=100000): trades = pd.DataFrame(columns=['Date', 'Symbol', 'Order', 'Shares']) syms = [symbol] dates = pd.date_range(sd, ed) prices all = ut.get data(syms, dates) prices = prices_all[syms] numRows = prices.shape[0]-1 trades.loc[0] = [prices.index[0].strftime('%Y-%m-%d'),symbol,'BUY', 1000] trades.loc[1] = [prices.index[numRows].strftime('%Y-%m-%d'),symbol,'SELL', -1000] trades.set index('Date', inplace=True) trades.index = pd.to_datetime(trades.index, format='%Y-%m-%d') return trades def computePerformance(port vals): # Summary of Performance of Manual Strategy dailyRets = (port_vals / port_vals.shift(1)) - 1 dailyRets = dailyRets[1:] dailyRiskFreeReturn = 0 sharpeRatio = (np.mean(dailyRets - dailyRiskFreeReturn)) / (np.std(dailyRets)) numberOfSamples = 252 # daily(252), weekly(52), monthly(12) k_factor = np.sqrt(numberOfSamples) sharpeRatioAnnualized = k_factor * sharpeRatio cr_in, adr_in, sddr_in, sr_in = [(port_vals[-1]/port_vals[0])-1, dailyRets.mean(), dailyRets.std(), sharpeRatioAnnualized, def toOrders(dailyTrades, symbol="JPM"): trades = pd.DataFrame(columns=['Date', 'Symbol', 'Order', 'Shares']) index = 0for i in range(0, dailyTrades.shape[0]): if dailyTrades[symbol].iloc[i] == 2000: trades.loc[index] = [dailyTrades.index[i].strftime('%Y-%m-%d'),symbol,'BUY',2000] index += 1elif dailyTrades[symbol].iloc[i] == 1000: trades.loc[index] = [dailyTrades.index[i].strftime('%Y-%m-%d'),symbol,'BUY',1000] index += 1elif dailyTrades[symbol].iloc[i] == -2000: trades.loc[index] = [dailyTrades.index[i].strftime('%Y-%m-%d'),symbol,'SELL',2000] elif dailyTrades[symbol].iloc[i] == -1000: trades.loc[index] = [dailyTrades.index[i].strftime('%Y-%m-%d'),symbol,'SELL',1000] index += 1trades.set_index('Date', inplace=True) trades.index = pd.to_datetime(trades.index, format='%Y-%m-%d') return trades def computePerformance(port vals): dailyRets = (port_vals / port_vals.shift(1)) - 1 dailyRets = dailyRets[1:] dailyRiskFreeReturn = 0 sharpeRatio = (np.mean(dailyRets - dailyRiskFreeReturn)) / (np.std(dailyRets)) numberOfSamples = 252 # daily(252), weekly(52), monthly(12) k factor = np.sqrt(numberOfSamples) sharpeRatioAnnualized = k_factor * sharpeRatio cr, adr, sddr, sr = [(port_vals[-1]/port_vals[0])-1, dailyRets.mean(), dailyRets.std(), sharpeRatioAnnualized, return (cr, adr, sddr, sr) In [58]: symbol="SPY" in_sd=dt.datetime(2018, 1, 1) in_ed=dt.datetime(2019, 12, 31) out_sd=dt.datetime(2020, 1, 1) out ed=dt.datetime(2020, 12, 31) sv=100000 commission=0.0 impact=0.0 In [59]: in benchmark = createBenchmark(symbol=symbol, sd=in sd, ed=in ed, sv=sv) in_ManualTrades, in_buys, in_sells, in_strongbuys, in_strongsells = ms.testPolicy(symbol=symbol, sd=in_sd, ed=in_ed, sv=sv) out benchmark = createBenchmark(symbol=symbol, sd=out sd, ed=out ed, sv=sv) out_ManualTrades, out_buys, out_sells, out_strongbuys, out_strongsells = ms.testPolicy(symbol=symbol, sd=out_sd, ed=out_ed, sv=sv) learner = sl.StrategyLearner(verbose = False, impact=impact) learner.add_evidence(symbol=symbol, sd=in_sd, ed=in_ed, sv=sv) in_MLTrades = learner.testPolicy(symbol=symbol, sd=in_sd, ed=in_ed, sv=sv) out_MLTrades = learner.testPolicy(symbol=symbol, sd=out_sd, ed=out_ed, sv=sv) in_MLTrades = toOrders(symbol=symbol, dailyTrades = in_MLTrades) out_MLTrades = toOrders(symbol=symbol, dailyTrades = out_MLTrades) in_df1 = compute_portvals(in_benchmark, start_date = in_sd, end_date = in_ed , start_val = sv, commission=commission, impact=impact) in df2 = compute portvals(in ManualTrades, start date = in sd, end date = in ed , start val = sv, commission=commission, impact=impact) in ml1 = compute portvals(in MLTrades, start date = in sd, end date = in ed , start val = sv, commission=commission, impact=impact) out df1 = compute portvals(out benchmark, start date = out sd, end date = out ed , start val = sv, commission=commission, impact=impact) out_df2 = compute_portvals(out_ManualTrades, start_date = out_sd, end_date = out_ed , start_val = sv, commission=commission, impact=impact) out_ml1 = compute_portvals(out_MLTrades, start_date = out_sd, end_date = out_ed , start_val = sv, commission=commission, impact=impact) <Figure size 1440x1008 with 0 Axes> <Figure size 1440x1008 with 0 Axes> In [60]: # Summary of Performance of Manual Strategy cr_in, adr_in, sddr_in, sr_in = computePerformance(in_df2) print('Performance stats of Manual Strategy') print(' In-Sample Period ') Cumulative Return : ', cr in) print(' Mean of Daily Return: ', adr in) print(' Standard Deviation : ', sddr in) print(' Sharpe Ratio : ', sr_in) print(' cr_out, adr_out, sddr_out, sr_out = computePerformance(out_df2) print(' Out-Sample Period ') Cumulative Return : ', cr_out) print(' Mean of Daily Return: ', adr_out) print(' Standard Deviation : ', sddr out) print(' print(' Sharpe Ratio : ', sr out) Performance stats of Manual Strategy In-Sample Period Cumulative Return : 0.43581418000000305 Mean of Daily Return: 0.0009304201710181016 Standard Deviation : 0.020609034820742913 Sharpe Ratio : 0.7173890279497704 Out-Sample Period Cumulative Return : -0.37537199000000254 Mean of Daily Return: 0.0021410804291693356 Standard Deviation : 0.09225675753041711 Sharpe Ratio : 0.36914626196693073 In [61]: in df1 = pd.DataFrame(in df1) in df2 = pd.DataFrame(in_df2) in_ml1 = pd.DataFrame(in_ml1) out_df1 = pd.DataFrame(out_df1) out df2 = pd.DataFrame(out_df2) out_ml1 = pd.DataFrame(out_ml1) in_df1_norm = in_df1 / in_df1.iloc[0] in_df2_norm = in_df2 / in_df2.iloc[0] in_ml1_norm = in_ml1 / in_ml1.iloc[0] out_df1_norm = out_df1 / out_df1.iloc[0] out_df2_norm = out_df2 / out_df2.iloc[0] out_ml1_norm = out_ml1 / out_ml1.iloc[0] perf_in_df1 = in_df1_norm.iloc[-1].values perf_in_df2 = in_df2_norm.iloc[-1].values perf_in_ml1 = in_ml1_norm.iloc[-1].values perf_out_df1 = out_df1_norm.iloc[-1].values perf_out_df2 = out_df2_norm.iloc[-1].values perf_out_ml1 = out_ml1_norm.iloc[-1].values print('Performance of Manual Strategy') print(' In-Sample Period ') print(' BenchMark : ', perf_in_df1) Manual Strategy : ', perf_in_df2) print(' Strategy Leraner: ', perf_in_ml1) print(' print(' Out-Sample Period ') BenchMark : ', perf_out_df1) print(' Manual Strategy : ', perf_out_df2) print(' Strategy Learner: ', perf_out_ml1) print(' plt.figure(figsize=(20,10)) plt.subplot(211) plt.plot(in_df1_norm.index, in_df1_norm['Total'], label='In_BenchMark', linewidth=0.5, color='green') plt.plot(in df2 norm.index, in df2 norm['Total'], label='In Indicators', color='red') for i in range(len(in buys)): plt.axvline(x=in_buys[i], color='blue') for i in range(len(in_strongbuys)): plt.axvline(x=in_strongbuys[i], color='blue') for i in range(len(in_sells)): plt.axvline(x=in_sells[i], color='black') for i in range(len(in_strongsells)): plt.axvline(x=in strongsells[i], color='black') plt.title('Manual Strategy vs. Benchmark') plt.ylabel('Normalized', fontsize=8) plt.legend(loc='best') plt.subplot(212) plt.plot(out_df1_norm.index, out_df1_norm['Total'], label='Out_BenchMark', linewidth=0.5, color='green') plt.plot(out_df2_norm.index, out_df2_norm['Total'], label='Out_Indicators', color='red') for i in range(len(out_buys)): plt.axvline(x=out_buys[i], color='blue') for i in range(len(out strongbuys)): plt.axvline(x=out_strongbuys[i], color='blue') for i in range(len(out_sells)): plt.axvline(x=out sells[i], color='black') for i in range(len(out_strongsells)): plt.axvline(x=out_strongsells[i], color='black') plt.xlabel('Date', fontsize=8) plt.ylabel('Normalized', fontsize=8) plt.legend(loc='best') plt.show() plt.clf plt.figure(figsize=(20,10)) plt.subplot(211) plt.plot(in df1 norm.index, in df1 norm['Total'], label='In BenchMark', linewidth=0.5, color='green') plt.plot(in_df2_norm.index, in_df2_norm['Total'], label='In_Indicators', color='red') plt.plot(in ml1 norm.index, in ml1 norm['Total'], label='In ML', color = 'black') plt.title('Benchmark vs. Manual Strategy vs. Strategy Learner') plt.ylabel('Normalized', fontsize=8) plt.legend(loc='best') plt.subplot(212) plt.plot(out_df1_norm.index, out_df1_norm['Total'], label='Out_BenchMark', linewidth=0.5, color='green') plt.plot(out_df2_norm.index, out_df2_norm['Total'], label='Out_Indicators', color='red') plt.plot(out_ml1_norm.index, out_ml1_norm['Total'], label='Out_ML', color = 'black') plt.xlabel('Date', fontsize=8) plt.ylabel('Normalized', fontsize=8) plt.legend(loc='best') plt.show() Performance of Manual Strategy In-Sample Period BenchMark : [1.61846786] Manual Strategy : [1.43581418] Strategy Leraner: [4.13950638] Out-Sample Period BenchMark : [1.54965698] Manual Strategy : [0.62462801] Strategy Learner: [1.85319826] Manual Strategy vs. Benchmark In BenchMark In Indicators 1.4 § 1.2 1.0 0.8 2018-01 2018-04 2018-07 2018-10 2019-01 2019-04 2019-07 2019-10 2020-01 Out BenchMark Out Indicators 1.2 1.0 0.8 <u>آ</u> 0.6 -0.4 0.2 0.0 2020-05 2020-07 2020-09 2020-11 2021-01 2020-01 2020-03 Benchmark vs. Manual Strategy vs. Strategy Learner In BenchMark 4.0 In Indicators 3.5 3.0 E 2.5 2.0 1.5 1.0 2018-01 2018-07 2019-01 2020-01 2018-04 2018-10 2019-04 2019-07 2019-10 Out BenchMark 1.75 Out Indicators 1.50 1.25 1.00 ≥ 0.75 0.50 0.25 0.00 2020-03 2020-07 2021-01 2020-01 2020-05 2020-09 2020-11 In []: In []: In []:

import pandas as pd