

## STRATEGY & ECONOMIC RATIONALE

The investment universe consists of stocks from NYSE, AMEX, and NASDAQ, while illiquid stocks a re removed from the investment universe. Cumulative total return index is then created for each stock (dividends included), and the starting price during the formation period is set to \$1 (price normalization).

Pairs are formed over twelve months (formation period) and are then traded in the next six-mont h period (trading period). The matching partner for each stock is found by looking for the security that minimizes the sum of squared deviations between two normalized price series.

Top 20 pairs with the smallest historical distance measure are then traded.

BUY	SELL
long-short position is opene d when pair prices have dive rged by two standard deviati ons	position is closed when prices revert.

## PARAMETER & VARIABLES

PARAMETER	VALUE		
MARKETS TRADED	Equity		
FINANCIAL INSTRUMENTS	Stock		
REGION	United States		
PERIOD OF REBALANCING	Daily		
NO. OF TRADED INSTRUMENTS	40		
WEIGHTING	Equal weighting		
LOOKBACK PERIODS	12 months		
LONG/SHORT	Long & short		

## **ALGORITHM**

```
import numpy as np
import itertools as it

class PairsTradingwithStocks(QCAlgorithm):

    def Initialize(self):
        self.SetStartDate(2005, 1, 1)
        self.SetCash(100000)

        self.symbol = self.AddEquity('SPY', Resolution.Daily).Symbol

        # Daily price data.
        self.history_price = {}
        self.period = 12 * 21

        # Equally weighted brackets.
        self.max traded pairs = 5
```

```
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        self.traded pairs = []
        self.traded_quantity = {}
        self.sorted_pairs = []
        self.coarse count = 500
        self.month = 6
        self.selection_flag = True
        self.UniverseSettings.Resolution = Resolution.Daily
        self.AddUniverse(self.CoarseSelectionFunction)
        self.Schedule.On(self.DateRules.MonthStart(self.symbol),
self.TimeRules.AfterMarketOpen(self.symbol), self.Selection)
    def OnSecuritiesChanged(self, changes):
        for security in changes.AddedSecurities:
            security.SetFeeModel(CustomFeeModel())
            security.SetLeverage(5)
        for security in changes.RemovedSecurities:
            symbol = security.Symbol
            if symbol in self.history_price:
                del self.history_price[symbol]
        symbols = [x for x in self.history_price.keys() if x != self.symbol]
        self.symbol_pairs = list(it.combinations(symbols, 2))
        # minimize the sum of squared deviations
        distances = {}
        for pair in self.symbol_pairs:
            if self.history price[pair[0]].IsReady and self.history price[pair[1]].IsReady:
                distances[pair] = self.Distance(self.history_price[pair[0]],
self.history_price[pair[1]])
        if len(distances) != 0:
            self.sorted_pairs = [x[0]] for x in sorted(distances.items(), key = lambda x:
x[1])[:20]]
        self.Liquidate()
        self.traded_pairs.clear()
        self.traded_quantity.clear()
    def CoarseSelectionFunction(self, coarse):
        # Update the rolling window every day.
        for stock in coarse:
            symbol = stock.Symbol
            if symbol in self.history price:
                self.history_price[symbol].Add(stock.AdjustedPrice)
        if not self.selection_flag:
```

return Universe. Unchanged

self.selection\_flag = False

```
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        selected = sorted([x for x in coarse if x.HasFundamentalData and x.Price > 5 and
x.Market == 'usa'],
            key=lambda x: x.DollarVolume, reverse=True)[:self.coarse count]
        # Warmup price rolling windows.
        for stock in selected:
            symbol = stock.Symbol
            if symbol in self.history_price:
                continue
            self.history_price[symbol] = RollingWindow[float](self.period)
            history = self.History(symbol, self.period, Resolution.Daily)
            if history.empty:
                self.Log(f"Not enough data for {symbol} yet")
                continue
            closes = history.loc[symbol].close
            for time, close in closes.iteritems():
                self.history_price[symbol].Add(close)
        return [x.Symbol for x in selected if self.history_price[x.Symbol].IsReady]
    def OnData(self, data):
        if self.sorted_pairs is None: return
        pairs to remove = []
        for pair in self.sorted_pairs:
            # Calculate the spread of two price series.
            price_a = [x for x in self.history_price[pair[0]]]
            price_b = [x for x in self.history_price[pair[1]]]
            norm_a = np.array(price_a) / price_a[-1]
            norm_b = np.array(price_b) / price_b[-1]
            spread = norm_a - norm_b
            mean = np.mean(spread)
            std = np.std(spread)
            actual_spread = spread[0]
            # Long-short position is opened when pair prices have diverged by two standard
deviations.
            traded_portfolio_value = self.Portfolio.TotalPortfolioValue /
self.max traded pairs
            if actual_spread > mean + 2*std or actual_spread < mean - 2*std:</pre>
                if pair not in self.traded_pairs:
                    # open new position for pair, if there's place for it.
                    if len(self.traded_pairs) < self.max_traded_pairs:</pre>
                         symbol_a = pair[0]
                         symbol_b = pair[1]
                         a_price_norm = norm_a[0]
                         b_price_norm = norm_b[0]
                         a_price = price_a[0]
```

b\_price = price\_b[0]

```
# a stock's price > b stock's price
                        if a price norm > b price norm:
                            long q = traded portfolio value / b price # long b stock
                            short_q = -traded_portfolio_value / a_price # short a stock
                            if self.Securities.ContainsKey(symbol_a) and
self.Securities.ContainsKey(symbol b) and \
                                self.Securities[symbol_a].Price != 0 and
self.Securities[symbol_a].IsTradable and \
                                self.Securities[symbol b].Price != 0 and
self.Securities[symbol_b].IsTradable:
                                self.MarketOrder(symbol_a, short_q)
                                self.MarketOrder(symbol b, long q)
                                self.traded_quantity[pair] = (short_q, long_q)
                                self.traded pairs.append(pair)
                        # b stock's price > a stock's price
                        else:
                            long_q = traded_portfolio_value / a_price
                            short q = -traded portfolio value / b price
                            if self.Securities.ContainsKey(symbol_a) and
self.Securities.ContainsKey(symbol_b) and \
                                self.Securities[symbol a].Price != 0 and
self.Securities[symbol_a].IsTradable and \
                                self.Securities[symbol_b].Price != 0 and
self.Securities[symbol b].IsTradable:
                                self.MarketOrder(symbol a, long q)
                                self.MarketOrder(symbol_b, short_q)
                                self.traded quantity[pair] = (long q, short q)
                                self.traded pairs.append(pair)
            # The position is closed when prices revert back.
            else:
                if pair in self.traded pairs and pair in self.traded quantity:
                    # make opposite order to opened position
                    self.MarketOrder(pair[0], -self.traded_quantity[pair][0])
                    self.MarketOrder(pair[1], -self.traded_quantity[pair][1])
                    pairs_to_remove.append(pair)
        for pair in pairs to remove:
            self.traded_pairs.remove(pair)
            del self.traded_quantity[pair]
    def Distance(self, price_a, price_b):
        # Calculate the sum of squared deviations between two normalized price series.
        price a = [x \text{ for } x \text{ in price } a]
        price b = [x for x in price b]
        norm_a = np.array(price_a) / price_a[-1]
        norm_b = np.array(price_b) / price_b[-1]
        return sum((norm_a - norm_b)**2)
   def Selection(self):
```

```
if self.month == 6:
        self.selection_flag = True

self.month += 1
    if self.month > 12:
        self.month = 1

# Custom fee model.

class CustomFeeModel(FeeModel):
    def GetOrderFee(self, parameters):
        fee = parameters.Security.Price * parameters.Order.AbsoluteQuantity * 0.00005
        return OrderFee(CashAmount(fee, "USD"))
```

## **BACKTESTING PERFORMANCE**



Fig 1. Overall Performance

PSR	1.307%	Sharpe Ratio	0.59	
Total Trades	8048	Average Win	0.45%	
Average Loss	-0.47%	Compounding Annual Return	6.732%	
Drawdown	27.300%	Expectancy	0.070	
Net Profit	227.058%	Loss Rate	45%	
Win Rate	55%	Profit-Loss Ratio	0.95	
Alpha	0.045	Beta	0.069	
Annual Standard Deviation	0.084	Annual Variance	0.007	
Information Ratio	-0.145	Tracking Error	0.171	
Treynor Ratio	0.723	Total Fees	\$18212.08	
Estimated Strategy Capacity	\$7900000.00	Lowest Capacity Asset	NVR R735QTJ8XC9X	

Fig 2. Performance Metrics

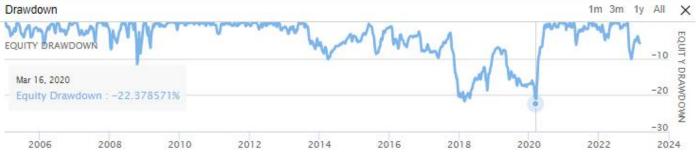


Fig 3. Drawdown

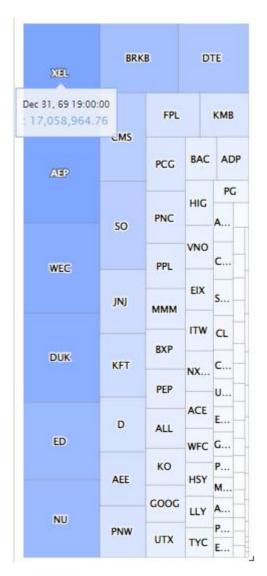


Fig 4. Assets Sales Volume