

# Not Over Thinking

## Return Asymmetry Effect in Commodity Futures

Algorithmic Trading Strategy with Full Code

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2024.01 | Vol 67.

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## STRATEGY & ECONOMIC RATIONALE

The investment universe consists of 22 commodity futures, namely: soybean oil, corn, cocoa, cotton, feeder cattle, gold, copper, heating oil, coffee, live cattle, lean hogs, natural gas, oats, orange juice, palladium, platinum, soybean, sugar, silver, soybean meal, wheat, and crude oil. Firstly, at the beginning of each month, construct the asymmetry measure (IE) for each commodity based on the latest 260 daily returns using the following formula (the formula originally consists of theoretical density and integrals, however the solution is simple when empirical distribution is utilized):  $IE = (\text{number of trading days when the daily return is greater than the average plus two standard deviations}) - (\text{number of trading days when the daily return is smaller than the average minus two standard deviations})$ .

Then rank the commodities according to their IE. Buy the bottom seven commodities with the lowest IE in the previous month and sell the top seven commodities with the highest IE in the previous month. Weigh the portfolio equally and rebalance monthly.

BUY	SELL
Buy the bottom seven commodities with the lowest IE in the previous month	sell the top seven commodities with the highest IE in the previous month.

## PARAMETER & VARIABLES

PARAMETER	VALUE
MARKETS TRADED	Commodity
FINANCIAL INSTRUMENTS	Futures
REGION	Global
PERIOD OF REBALANCING	Monthly
NO. OF TRADED INSTRUMENTS	22
WEIGHTING	Equal weighting
LOOKBACK PERIODS	Month
LONG/SHORT	Long & short

## ALGORITHM

```
from AlgorithmImports import *
class ReturnAsymmetryEffectInCommodityFutures(QCAlgorithm):

    def Initialize(self):
        self.SetStartDate(2000, 1, 1)
        self.SetCash(100000)

        self.tickers = [
            "CME_S1", # Soybean Futures, Continuous Contract
            "CME_W1", # Wheat Futures, Continuous Contract
            "CME_SM1", # Soybean Meal Futures, Continuous Contract
            "CME_BO1", # Soybean Oil Futures, Continuous Contract
            "CME_C1", # Corn Futures, Continuous Contract
            "CME_O1", # Oats Futures, Continuous Contract
            "CME_LC1", # Live Cattle Futures, Continuous Contract
            "CME_FC1", # Feeder Cattle Futures, Continuous Contract
            "CME_LN1", # Lean Hog Futures, Continuous Contract
            "CME_GC1", # Gold Futures, Continuous Contract
            "CME_SI1", # Silver Futures, Continuous Contract
```

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```
"CME_PL1", # Platinum Futures, Continuous Contract
"CME_CL1", # Crude Oil Futures, Continuous Contract
"CME_HG1", # Copper Futures, Continuous Contract
"CME_LB1", # Random Length Lumber Futures, Continuous Contract
# "CME_NG1", # Natural Gas (Henry Hub) Physical Futures, Continuous Contract
"CME_PA1", # Palladium Futures, Continuous Contract
"CME_RR1", # Rough Rice Futures, Continuous Contract
"CME_RB2", # Gasoline Futures, Continuous Contract
"CME_KW2", # Wheat Kansas, Continuous Contract

"ICE_CC1", # Cocoa Futures, Continuous Contract
"ICE_CT1", # Cotton No. 2 Futures, Continuous Contract
"ICE_KC1", # Coffee C Futures, Continuous Contract
"ICE_O1", # Heating Oil Futures, Continuous Contract
"ICE_OJ1", # Orange Juice Futures, Continuous Contract
"ICE_SB1", # Sugar No. 11 Futures, Continuous Contract
"ICE_RS1", # Canola Futures, Continuous Contract
"ICE_GO1", # Gas Oil Futures, Continuous Contract
"ICE_WT1", # WTI Crude Futures, Continuous Contract
]

self.data = {} # storing objects of SymbolData class keyed by comodity symbols

self.period = 261 # need 261 daily prices, to calculate 260 daily returns
self.buy_count = 7 # buy n comodities on each rebalance
self.sell_count = 7 # sell n comodities on each rebalance

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

# subscribe to futures contracts
for ticker in self.tickers:
    security = self.AddData(QuantpediaFutures, ticker, Resolution.Daily)
    security.SetFeeModel(CustomFeeModel())
    security.SetLeverage(5)

    self.data[security.Symbol] = SymbolData(self.period)

self.rebalance_flag = False
self.Schedule.On(self.DateRules.MonthStart(self.symbol), self.TimeRules.BeforeMarketClose(self.symbol, 0), self.Rebalance)

def OnData(self, data):
    # update daily closes
    for symbol in self.data:
        if symbol in data and data[symbol]:
            close = data[symbol].Value
            self.data[symbol].update_closes(close)

    # rebalance monthly
    if not self.rebalance_flag:
        return
    self.rebalance_flag = False

IE = {}

for symbol, symbol_obj in self.data.items():
    # check if comodity has ready prices
    if not symbol_obj.is_ready():
        continue

    # calculate IE
    IE_value = symbol_obj.calculate_IE()

    # store IE value under comodity symbol
```

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```
IE[symbol] = IE_value
```

```
# make sure, there are enough commodities for rebalance
```

```
if len(IE) < (self.buy_count + self.sell_count):  
    return
```

```
# sort commodities based on IE values
```

```
sorted_by_IE = [x[0] for x in sorted(IE.items(), key=lambda item: item[1])]
```

```
# select long and short parts
```

```
long = sorted_by_IE[:self.buy_count]
```

```
short = sorted_by_IE[-self.sell_count:]
```

```
# trade execution
```

```
invested = [x.Key for x in self.Portfolio if x.Value.Invested]
```

```
for symbol in invested:
```

```
    if symbol not in long + short:  
        self.Liquidate(symbol)
```

```
for symbol in long:
```

```
    self.SetHoldings(symbol, 1 / self.buy_count)
```

```
for symbol in short:
```

```
    self.SetHoldings(symbol, -1 / self.sell_count)
```

```
def Rebalance(self):
```

```
    self.rebalance_flag = True
```

```
        class SymbolData():
```

```
def __init__(self, period):
```

```
    self.closes = RollingWindow[float](period)
```

```
def update_closes(self, close):
```

```
    self.closes.Add(close)
```

```
def is_ready(self):
```

```
    return self.closes.IsReady
```

```
def calculate_IE(self):
```

```
    closes = np.array([x for x in self.closes])
```

```
    daily_returns = (closes[:-1] - closes[1:]) / closes[1:]
```

```
    average_daily_returns = np.average(daily_returns)
```

```
    two_daily_returns_std = 2 * np.std(daily_returns)
```

```
    avg_plus_two_std = average_daily_returns + two_daily_returns_std
```

```
    avg_minus_two_std = average_daily_returns - two_daily_returns_std
```

```
    over_avg_plus_two_std = 0 # counting number of daily returns, which were over avg_plus_  
two_std
```

```
    under_avg_minus_two_std = 0 # counting number of daily returns, which were under avg_mi  
nus_two_std
```

```
for daily_return in daily_returns:
```

```
    if daily_return > avg_plus_two_std:
```

```
        over_avg_plus_two_std += 1
```

```
    elif daily_return < avg_minus_two_std:
```

```
        under_avg_minus_two_std += 1
```

```
IE_value = over_avg_plus_two_std - under_avg_minus_two_std
```

```
return IE_value
```

```
# Quantpedia data.# NOTE: IMPORTANT: Data order must be ascending (datewise)class Quant
```

```
pediaFutures(PythonData):
```

```
def GetSource(self, config, date, isLiveMode):
```

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```
return SubscriptionDataSource("data.quantpedia.com/backtesting_data/futures/{0}.csv".format(config.Symbol.Value), SubscriptionTransportMedium.RemoteFile, FileFormat.Csv)
```

```
def Reader(self, config, line, date, isLiveMode):
    data = QuantpediaFutures()
    data.Symbol = config.Symbol

    if not line[0].isdigit(): return None
    split = line.split(';')

    data.Time = datetime.strptime(split[0], "%d.%m.%Y") + timedelta(days=1)
    data['back_adjusted'] = float(split[1])
    data['spliced'] = float(split[2])
    data.Value = float(split[1])

    return data
# 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

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Total Trades	3998	Average Win	0.80%
Average Loss	-0.78%	Compounding Annual Return	2.631%
Drawdown	48.200%	Expectancy	0.057
Net Profit	83.005%	Sharpe Ratio	0.204
Probabilistic Sharpe Ratio	0.000%	Loss Rate	48%
Win Rate	52%	Profit-Loss Ratio	1.03
Alpha	0.028	Beta	-0.014
Annual Standard Deviation	0.134	Annual Variance	0.018
Information Ratio	-0.145	Tracking Error	0.212
Treynor Ratio	-1.947	Total Fees	\$1986.76
Estimated Strategy Capacity	\$0	Lowest Capacity Asset	CME_LB1.QuantpediaFutures 2S
Portfolio Turnover	3.23%		

Fig 2. Performance Metrics