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STRATEGY & ECONOMIC RATIONALE

The investment universe consists of all stocks on NYSE, AMEX, and NASDAQ. At the end of June of each year t, ten portfolios are formed based on ranked values net pay-out yield.

| LONG | SHORT | | | | |
|------------------------------|----------------------------|--|--|--|--|
| Portfolio with the highest n | Portfolio with the lowest | | | | |
| et pay-out ratio and held fo | net pay-out ratio and held | | | | |
| r 1 year | for 1 year | | | | |

PARAMETER & VARIABLES

• The net pay-out yield is the ratio of dividends plus repurchases minus common share issu ances in year t to year-end market capitalization. There are two measures of pay-out yield, one based on the statement of cash flows, the other based on the change in Treasury stocks. For the net pay-out yield, we use the cash flow-based measure of repurchases.

| PARAMETER | VALUE |
|---------------------------|--------------------|
| MARKETS | NYSE, AMEX, NASDAQ |
| TRADED | |
| FINANCIAL INSTRUMENTS | Stocks |
| PERIOD OF REBALANCING | 1 year |
| NO. OF TRADED INSTRUMENTS | 500 |
| WEIGHTING | Equal weighting |
| HOLDING PERIODS | 1 year |
| LONG/SHORT | Long Only |

ALGORITHM

```
from AlgorithmImports import *

class NetPayoutYieldEffect(QCAlgorithm):

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

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

    self.coarse_count = 500 ## coarse universe will consist 500 stocks
    self.quantile = 10
    self.leverage = 5

    self.long = []

    self.selection_flag = False
    self.UniverseSettings.Resolution = Resolution.Daily
    self.AddUniverse(self.CoarseSelectionFunction, self.FineSelectionFunction)
    self.Schedule.On(self.DateRules.MonthEnd(self.symbol),
self.TimeRules.AfterMarketOpen(self.symbol), self.Selection)
```

```
def OnSecuritiesChanged(self, changes):
        for security in changes.AddedSecurities:
            security.SetFeeModel(CustomFeeModel())
            security.SetLeverage(self.leverage)
    def CoarseSelectionFunction(self, coarse):
        if not self.selection flag:
            return Universe.Unchanged ## make sure up till now there is no symbols selected
        selected = sorted([x for x in coarse if x.HasFundamentalData and x.Market == 'usa'],
            key=lambda x: x.DollarVolume, reverse=True)
        ## only symbols that (i) have fundamental data; (ii) traded in the US market are
selected; then sorted by trading volume
        return [x.Symbol for x in selected[:self.coarse count]]
        ## [:self.coarse count] means select from the first to the last item in the carse
universe
    def FineSelectionFunction(self, fine):
        fine = [x for x in fine if x.MarketCap !=0 and x.ValuationRatios.TotalYield != 0 and
x.FinancialStatements.CashFlowStatement.CommonStockIssuance.TwelveMonths != 0 and
            ((x.SecurityReference.ExchangeId == "NYS") or (x.SecurityReference.ExchangeId ==
"NAS") or (x.SecurityReference.ExchangeId == "ASE"))]
    ## only select the stocks that (i) actively traded (aka market cap is not 0); (ii) pay
dividend or generate cashflows (aka total yield is not 0); (iii) have stocks issued in the past
12 months; (iv) traded in NYS, NAS, or ASE stock exchange.
        # Sorting by net payout.
        sorted_by_payout = sorted(fine, key = lambda x: ( (x.ValuationRatios.TotalYield *
(x.MarketCap)) - \
                                                            (x.FinancialStatements.CashFlowStat
ement.CommonStockIssuance.TwelveMonths / (x.MarketCap))), reverse=True)
        if len(sorted_by_payout) >= self.quantile:
        ## check if there if more than 10 quantiles generated
            decile = int(len(sorted_by_payout) / self.quantile)
        ## if there is 32 items in "sorted_by_payout", divided by 10 quantiles, equals 3.2 >
then int( ) > equal to 3 > this is the number of items in the 1st quantile, that we will buy
            self.long = [x.Symbol for x in sorted_by_payout[:decile]]
        return self.long
        ## Fine universe will output with a list of securities that we will long
    def OnData(self, data):
        if not self.selection_flag:
            return
        self.selection_flag = False
        stocks_invested = [x.Key for x in self.Portfolio if x.Value.Invested]
        for symbol in stocks_invested:
            if symbol not in self.long:
                self.Liquidate(symbol)
```

```
for symbol in self.long:
    if symbol in data and data[symbol]:
        self.SetHoldings(symbol, 1 / len(self.long)) ## equal weighting

self.long.clear()

def Selection(self):
    if self.Time.month == 6:
        self.selection_flag = True ## only start to select stocks to long in June

# 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

Not Over Thinking – where I share my journey to algorithmic trading and investments in shortest words possible

| PSR | 0.020% | Sharpe Ratio | 0.399 | |
|-----------------------------|---------------|---------------------------|------------------|--|
| Total Trades | 1038 | Average Win | 0.48% | |
| Average Loss | -0.54% | Compounding Annual Return | 7.123% | |
| Drawdown | 54.700% | Expectancy | 0.437 | |
| Net Profit | 390.227% | Loss Rate | 24% | |
| Win Rate | 76% | Profit-Loss Ratio | 0.89 | |
| Alpha | 0.009 | Beta | 0.891 | |
| Annual Standard Deviation | 0.153 | Annual Variance | 0.023 | |
| Information Ratio | 0.059 | Tracking Error | 0.053 | |
| Treynor Ratio | 0.068 | Total Fees | \$136.30 | |
| Estimated Strategy Capacity | \$92000000.00 | Lowest Capacity Asset | NVS RULY784EQ6AT | |
| | | | | |

Fig 2. Performance Metrics

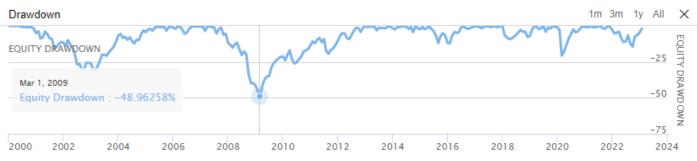


Fig 3. Drawdown

| NVS | ммм | A | | (P | SNY | | |
|-------|-----------|------|-----|------|-----|-----|---------|
| МО | TSM | TXN | | NFLX | | UPS | |
| MO | | ТМ | | DOW | ' | HD | |
| RIO | CHTR | | | GE | КМВ | | VOD |
| | GM | ι | JSB | RY A | | LL | CL |
| ко | | JCI | | MDT | T C | | CS |
| | INTC | BMY | UTX | FLS | | | |
| PEP | AMGN | 5, | | TRV | | Α | |
| | | IBM | | хом | BEL | | С |
| WFC | ORCL | MCD | | CMC | D | | MS |
| BPA | BA | н | HBC | SLB | L | 0 | |
| br.A. | | 1150 | | ABT | H | | C PG |
| QCOM | MRK GS | | DIS | STD | UNH | | L |
| | | A | | ВНР | H | | P |
| | | A | ZN | CAT | | | P |

Fig 4. Assets Sales Volume