

STRATEGY & ECONOMIC RATIONALE

The investment universe consists of stocks in the MSCI World Index. Paper uses MSCI ESG Ratings as the ESG database. The ESG Momentum strategy is built by overweighting, relative to the MSCI World Index, companies that increased their ESG ratings most during the recent past and underweight those with decreased ESG ratings, where the increases and decreases are based on a 12-mon th ESG momentum. The paper uses the Barra Global Equity Model (GEM3) for portfolio construction with constraints that can be found in Appendix 2. Therefore, this strategy is very specific, but we aim to present the idea, not the portfolio construction. The strategy is rebalanced month ly.

BUY	SELL
(see above)	(see above)

PARAMETER & VARIABLES

PARAMETER	VALUE
MARKETS	Equity
TRADED	
FINANCIAL INSTRUMENTS	Stocks
REGION	Global
PERIOD OF REBALANCING	Monthly
NO. OF TRADED INSTRUMENTS	1000
WEIGHTING	Equal weighting
LOOKBACK PERIODS	N/A
LONG/SHORT	Long only

ALGORITHM

```
from AlgorithmImports import *
from numpy import floor
#endregion
class ESGFactorMomentumStrategy(QCAlgorithm):
    def Initialize(self):
        self.SetStartDate(2009, 6, 1)
        self.SetCash(100000)
        # Decile weighting.
        # True - Value weighted
        # False - Equally weighted
        self.value_weighting = True
        self.symbol = self.AddEquity('SPY', Resolution.Daily).Symbol
        self.esg_data = self.AddData(ESGData, 'ESG', Resolution.Daily)
        self.tickers = []
        self.holding_period = 3
        self.managed_queue = []
        self.quantile = 10
        # Monthly ESG decile data.
        self.esg = {}
```

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Not Over Thinking – where I share my journey to algorithmic trading and investments in shortest words possible
    self.period = 14
    self.latest price = {}
    self.selection_flag = False
    self.UniverseSettings.Resolution = Resolution.Daily
    self.AddUniverse(self.CoarseSelectionFunction, self.FineSelectionFunction)
def OnSecuritiesChanged(self, changes):
    for security in changes.AddedSecurities:
        security.SetFeeModel(CustomFeeModel())
        security.SetLeverage(10)
def CoarseSelectionFunction(self, coarse):
    if not self.selection flag:
        return Universe. Unchanged
    self.latest price.clear()
    selected = [x for x in coarse if (x.Symbol.Value).lower() in self.tickers]
    for stock in selected:
        symbol = stock.Symbol
        self.latest_price[symbol] = stock.AdjustedPrice
    return [x.Symbol for x in selected]
def FineSelectionFunction(self, fine):
    fine = [x for x in fine if x.MarketCap != 0]
    momentum = \{\}
    # Momentum calc.
    for stock in fine:
        symbol = stock.Symbol
        ticker = symbol.Value
        # ESG data for 14 months is ready.
        if ticker in self.esg and self.esg[ticker].IsReady:
            esg_data = [x for x in self.esg[ticker]]
            esg decile 2 months ago = esg data[1]
            esg decile 14 months ago = esg data[13]
            if esg decile 14 months ago != 0 and esg decile 2 months ago != 0:
                # Momentum as difference.
                # momentum_ = esg_decile_2_months_ago - esg_decile_14_months_ago
                # Momentum as ratio.
                momentum_ = (esg_decile_2_months_ago / esg_decile_14_months_ago) - 1
                # Store momentum/market cap pair.
                momentum[stock] = momentum_
    if len(momentum) <= self.quantile:</pre>
        return Universe. Unchanged
    # Momentum sorting.
    sorted by momentum = sorted(momentum.items(), key = lambda x: x[1], reverse = True)
    quantile = int(len(sorted_by_momentum) / self.quantile)
    long = [x[0]] for x in sorted by momentum[:quantile]]
    short = [x[0] for x in sorted_by_momentum[-quantile:]]
    long_symbol_q = []
```

short_symbol_q = []

```
# ew
        if not self.value weighting:
            if len(long) != 0:
                long w = self.Portfolio.TotalPortfolioValue / self.holding period / len(long)
                long_symbol_q = [(x.Symbol, floor(long_w / self.latest_price[x.Symbol])) for x
in long]
            if len(short) != 0:
                short_w = self.Portfolio.TotalPortfolioValue / self.holding_period / len(short)
                short_symbol_q = [(x.Symbol, -floor(short_w / self.latest_price[x.Symbol])) for
x in short]
        # vw
        else:
            if len(long) != 0:
                total market cap long = sum([x.MarketCap for x in long])
                long_w = self.Portfolio.TotalPortfolioValue / self.holding_period
                long_symbol_q = [(x.Symbol, floor((long_w * (x.MarketCap / total_market_cap_lon
g))) / self.latest price[x.Symbol]) for x in long]
            short_symbol_q = []
            if len(short) != 0:
                total_market_cap_short = sum([x.MarketCap for x in short])
                short_w = self.Portfolio.TotalPortfolioValue / self.holding_period
                short_symbol_q = [(x.Symbol, -floor((short_w * (x.MarketCap / total_market_cap_
short))) / self.latest price[x.Symbol]) for x in short]
        self.managed_queue.append(RebalanceQueueItem(long_symbol_q + short_symbol_q))
        return [x.Symbol for x in long + short]
    def OnData(self, data):
        new_data_arrived = False
        if 'ESG' in data and data['ESG']:
            # Store universe tickers.
            if len(self.tickers) == 0:
                # TODO '_typename' in storage dictionary?
                self.tickers = [x.Key for x in self.esg_data.GetLastData().GetStorageDictionary
()][:-1]
            # Store history for every ticker.
            for ticker in self.tickers:
                ticker_u = ticker.upper()
                if ticker u not in self.esg:
                    self.esg[ticker_u] = RollingWindow[float](self.period)
                decile = self.esg_data.GetLastData()[ticker]
                self.esg[ticker_u].Add(decile)
                # trigger selection after new esg data arrived.
                if not self.selection_flag:
                    new_data_arrived = True
        if new data arrived:
            self.selection_flag = True
            return
        if not self.selection_flag:
        self.selection_flag = False
        # Trade execution
        remove_item = None
```

```
# Rebalance portfolio
        for item in self.managed queue:
            if item.holding_period == self.holding_period:
                for symbol, quantity in item.symbol_q:
                    self.MarketOrder(symbol, -quantity)
                remove item = item
            elif item.holding_period == 0:
                open_symbol_q = []
                for symbol, quantity in item.symbol q:
                    if quantity >= 1:
                        if symbol in data and data[symbol]:
                            self.MarketOrder(symbol, quantity)
                            open_symbol_q.append((symbol, quantity))
                # Only opened orders will be closed
                item.symbol_q = open_symbol_q
            item.holding_period += 1
        if remove_item:
            self.managed_queue.remove(remove_item)
class RebalanceOueueItem():
    def __init__(self, symbol_q):
        # symbol/quantity collections
        self.symbol_q = symbol_q
        self.holding period = 0
# ESG data.
class ESGData(PythonData):
    def __init__(self):
        self.tickers = []
    def GetSource(self, config, date, isLiveMode):
        return SubscriptionDataSource("data.quantpedia.com/backtesting data/economic/esg decile
s_data.csv", SubscriptionTransportMedium.RemoteFile, FileFormat.Csv)
    def Reader(self, config, line, date, isLiveMode):
        data = ESGData()
        data.Symbol = config.Symbol
        if not line[0].isdigit():
            self.tickers = [x for x in line.split(';')][1:]
            return None
        split = line.split(';')
        data.Time = datetime.strptime(split[0], "%Y-%m-%d") + timedelta(days=1)
        index = 1
        for ticker in self.tickers:
            data[ticker] = float(split[index])
            index += 1
        data.Value = float(split[1])
        return data
# Custom fee model.
class CustomFeeModel(FeeModel):
    def GetOrderFee(self, parameters):
```

BACKTESTING PERFORMANCE

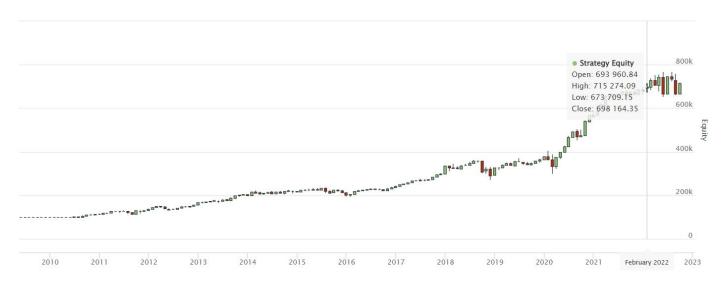


Fig 1. Overall Performance

Total Trades	14167	Average Win	0.06%
Average Loss	-0.04%	Compounding Annual Return	15.863%
Drawdown	26.100%	Expectancy	0.718
Net Profit	620.347%	Sharpe Ratio	0.843
Probabilistic Sharpe Ratio	18.964%	Loss Rate	36%
Win Rate	64%	Profit-Loss Ratio	1.67
Alpha	0.036	Beta	0.822
Annual Standard Deviation	0.14	Annual Variance	0.02
Information Ratio	0.233	Tracking Error	0.077
Treynor Ratio	0.143	Total Fees	\$1160.83
Estimated Strategy Capacity	\$150000000.00	Lowest Capacity Asset	UDL S3QTIGASA5WL

Fig 2. Performance Metrics