

# Work Sample

Screen the value stocks using Benjamin Graham value investment method

1. Back test the value stocks in China stock market for last 10 years
2. Adjust positions 4 times per year to satisfy Benjamin Graham value investment method
3. Compare the strategy return with the benchmark (SSE Composite)

Code:

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value_investing.py - /Users/tylerwang/Desktop/StudyQuant/量化编程 12_27王晨浩/value_investing.py (3.8.3)

from jddata import finance
from jddata import *
import pandas as pd
from dateutil.relativedelta import relativedelta
from datetime import datetime, timedelta
from dateutil.parser import parse
from jqfactor import winsorize_med, standardize, neutralize

def initialize(context):
    # set trading commission
    set_order_cost(OrderCost(open_commission=0.0013, close_commission=0.0013), type='stock')
    # set SSE composite as benchmark
    set_benchmark('000001.XSHG')
    # run the script once a month, adjust positions before the market open
    run_monthly(monthly, 1, time='9:00')
    set_option('use_real_price', True)
    # 3-year interest rate
    s = ['2002-02-21': 2.52, '2004-10-29': 3.24, '2006-08-19': 3.69, '2007-03-18': 3.96, '2007-05-19': 4.41,
        '2007-07-21': 4.68, '2007-08-22': 4.95, '2007-09-10': 5.22, '2007-12-21': 5.4, '2008-10-09': 5.13,
        '2008-10-30': 4.77, '2008-11-27': 3.6, '2008-12-23': 3.33, '2010-10-20': 3.85, '2010-12-26': 4.15,
        '2011-02-09': 4.5, '2011-06-06': 4.75, '2011-07-07': 5.08, '2012-06-08': 4.65, '2012-07-06': 4.25,
        '2014-11-22': 4.00, '2015-03-01': 3.75, '2015-05-11': 3.50, '2015-06-28': 3.25, '2015-08-26': 3.00,
        '2015-10-24': 2.75]
    s = pd.Series(s)
    g.bond_yield = s[s.index<context.current_dt.strftime('%Y-%m-%d')][-1]
    g.today = context.current_dt.strftime('%Y-%m-%d')
    # adjust positions 4 times per year
    g.month = 1
    g.period = 3
    # set the stock pool
    g.scu = get_index_stocks('000002.XSHG')

# Change the non-trading day data to previous trading day data
def shift_trading_day(date):
    tradingday = get_all_trade_days()
    date1 = datetime.date(date)
    for i in tradingday[::-1]:
        if i<date1:
            return i

# Current reciprocal of PE ratio is greater than twice of current 3-year interest rate
def condition_a(context):
    df = get_fundamentals(query(valuation.code, valuation.pe_ratio).
        filter(valuation.code.in(g.scu), 1/valuation.pe_ratio*100 > g.bond_yield*2), date=g.today)
    buylist = list(df['code'])
    return buylist

# Current PE ratio is lower than 40 percent of highest PE ratio in last 5 years
def condition_b(context):
    df = get_fundamentals(continuously(query(valuation.code, valuation.pe_ratio_lyr).
        filter(valuation.code.in(g.scu), end_date=g.today, count=250*5, panel=False)
        five_year_max = np.max(df['pe_ratio_lyr'])
    df = get_fundamentals(query(valuation.code, valuation.pe_ratio_lyr).
        filter(valuation.code.in(g.scu), valuation.pe_ratio_lyr<five_year_max*0.4), date=g.today)
    buylist = list(df['code'])
    return buylist

# Current dividend yield is greater than 2/3 of 3-year interest rate
def condition_c(context):
    frq = finance.run_query(query(finance.STK_XR_XD.code, finance.STK_XR_XD.bonus_ratio_rmb, finance.STK_XR_XD.report_date).
        filter(finance.STK_XR_XD.code.in(g.scu), finance.STK_XR_XD.bonus_ratio_rmb>g.bond_yield*(2/3)))
    pre_list = list(frq['code'])
    df = get_fundamentals(query(valuation.code).
        filter(valuation.code.in(pre_list), date=g.today)
    buylist = list(df['code'])
    return buylist

# Current market cap is lower than 2/3 of total tangible assets
def condition_d(context):
    df = get_fundamentals(query(valuation.code, valuation.market_cap, balance.total_current_assets, balance.fixed_assets, balance.total_liability).
        filter(valuation.code.in(g.scu), valuation.market_cap < (2/3)*(balance.total_current_assets+balance.fixed_assets+balance.total_liability)), date=g.today)
    buylist = list(df['code'])
    return buylist

# Current market cap is lower than 2/3 of (current assets - total liability)
def condition_e(context):
    df = get_fundamentals(query(valuation.code, valuation.market_cap, balance.total_current_assets, balance.total_liability).
        filter(valuation.code.in(g.scu), valuation.market_cap < (2/3)*(balance.total_current_assets- balance.total_liability)), date=g.today)
    buylist = list(df['code'])
    return buylist

# Total liability is lower than tangible assets
def condition_f(context):
    df = get_fundamentals(query(valuation.code, balance.total_current_assets, balance.total_liability, balance.fixed_assets).
        filter(valuation.code.in(g.scu), balance.total_liability < balance.fixed_assets+balance.total_current_assets+balance.total_liability), date=g.today)
    buylist = list(df['code'])
    return buylist

# (current assets / current liability) > 2
def condition_g(context):
    df = get_fundamentals(query(valuation.code, balance.total_current_assets, balance.total_current_liability).
        filter(valuation.code.in(g.scu), balance.total_current_assets/balance.total_current_liability>2), date=g.today)
    buylist = list(df['code'])
    return buylist

# Total liability < current liability * 2
def condition_h(context):
    df = get_fundamentals(query(valuation.code, balance.total_current_assets, balance.total_liability).
        filter(valuation.code.in(g.scu), balance.total_liability<(balance.total_current_assets - balance.total_liability)*2), date=g.today)
    buylist = list(df['code'])
    return buylist

# average yearly return in the past 10 years > 7%
def condition_i(context):
    stocks = get_index_stocks('000002.XSHG', date=None)
    now = get_price(stocks, start_date = shift_trading_day(context.current_dt-timedelta(days=1)), end_date = shift_trading_day(context.current_dt - timedelta(days=1)), frequency = 'daily', fields = ['close'], panel=False)
    ago = get_price(stocks, start_date = shift_trading_day(context.current_dt-timedelta(days=1)-relativedelta(years=10)), end_date = shift_trading_day(context.current_dt - timedelta(days=1)-relativedelta(years=10)), frequency = 'daily', fields = ['close'], panel=False)
    now['bool'] = now['close']/ago['close']*(1/10)>0.07
    buylist = list(now.loc[now['bool']] == True, 'code')
    return buylist

# Yearly return cannot below -5% twice in the last 10 years
def condition_j(context):
    stocks = get_index_stocks('000002.XSHG', date=None)
    pre_list=[]
    count = 0
    for j in range(10):
        now = get_price(stocks, start_date = shift_trading_day(context.current_dt-timedelta(days=1)-relativedelta(years=j)), end_date = shift_trading_day(context.current_dt- timedelta(days=1)-relativedelta(years=j)), frequency = 'daily', fields = ['close'], panel=False)
        ago = get_price(stocks, start_date = shift_trading_day(context.current_dt-timedelta(days=1)-relativedelta(years=j+1)), end_date = shift_trading_day(context.current_dt - timedelta(days=1)-relativedelta(years=j+1)), frequency = 'daily', fields = ['close'], panel=False)
        now['bool'] = (now['close']-ago['close'])/ago['close']< -0.05
        pre_list.append(now)
    new_df = pd.concat(pre_list)
    fil_df = new_df.groupby('code')['bool'].sum().to_frame()
    index_label = fil_df[fil_df['bool']==2].index.tolist()
    return index_label

# Winsorize, standardize and neutralize the factors
def clean_factor(factors, date):
    factors = factors.fillna(factors.mean())
    factors = winsorize_med(factors, scale=3, inclusive=True, inf2nan=True, axis=0)
    factors = standardize(factors, inf2nan=True, axis=0)
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factors = neutralize(factors,['sw_11', 'pe_ratio'],date=str(date),axis=0)
return factors

# Screen stocks and sort by PE ratio before Winsorize, standardize and neutralize the factors
def preprocess_data(buylist):
    df = get_fundamentals(query(valuation.code,valuation.pe_ratio),
        filter(valuation.code.in_(buylist)).dateeq(today).set_index('code'))
    new_df = clean_factor(df,g.today)
    final_df = new_df.sort_values('pe_ratio',ascending = False)
    final_list = list(final_df.index.values)
    return final_list

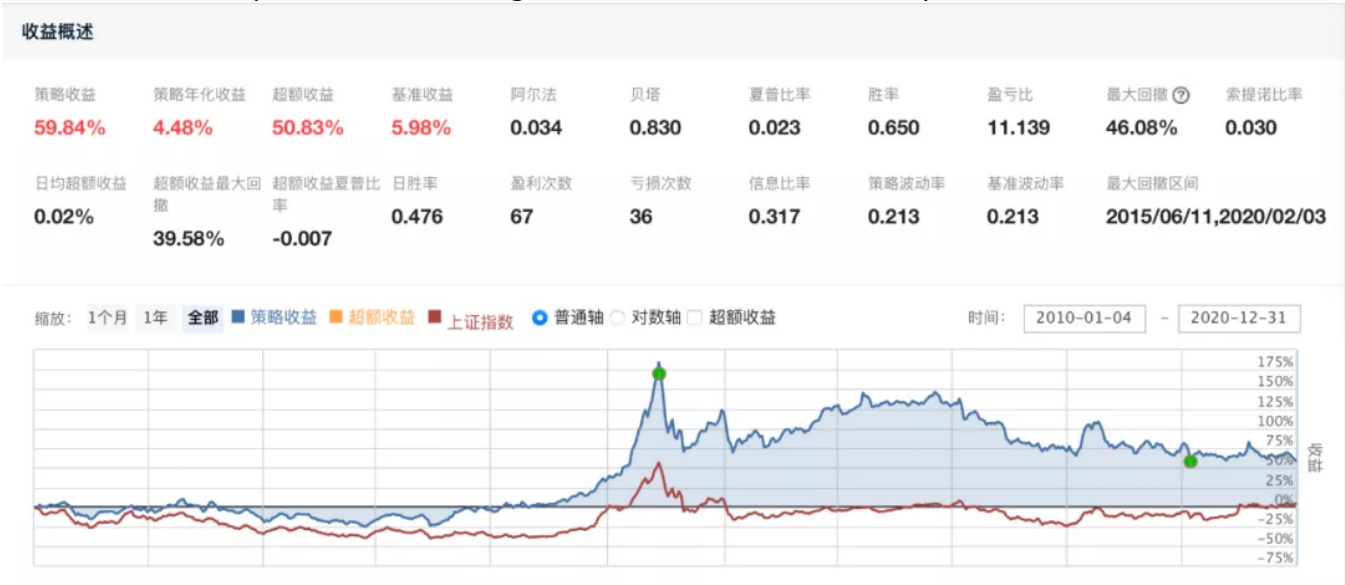
# Trade a list of stocks
def trade(context, buylist):
    for stock in context.portfolio.positions:
        if stock not in buylist:
            order_target(stock,0)
    # Sell all positions which are not in the list and buy all positions which are in the list (split funds evenly in every chosen stock)
    position_per_stk = context.portfolio.total_value/len(buylist)
    for stock in buylist:
        order_target_value(stock,position_per_stk)
    return

def monthly(context):
    # Adjust the positions in Jan, April, July, October
    if g.month % g.period == 1:
        # buylistA = condition_a(context)
        # buylistB = condition_b(context)
        # buylistC = condition_c(context)
        # buylistD = condition_d(context)
        # buylistE = condition_e(context)
        # buylistF = condition_f(context)
        # buylistG = condition_g(context)
        # buylistH = condition_h(context)
        # buylistI = condition_i(context)
        buylist = condition_j(context)
        final_list = preprocess_data(buylist)
        trade(context,final_list)
    else:
        pass
    g.month = g.month + 1

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Hardly a stock can satisfy all rules, so we back test each rule  
Output:

A. Current reciprocal of PE ratio is greater than twice of current 3-year interest rate



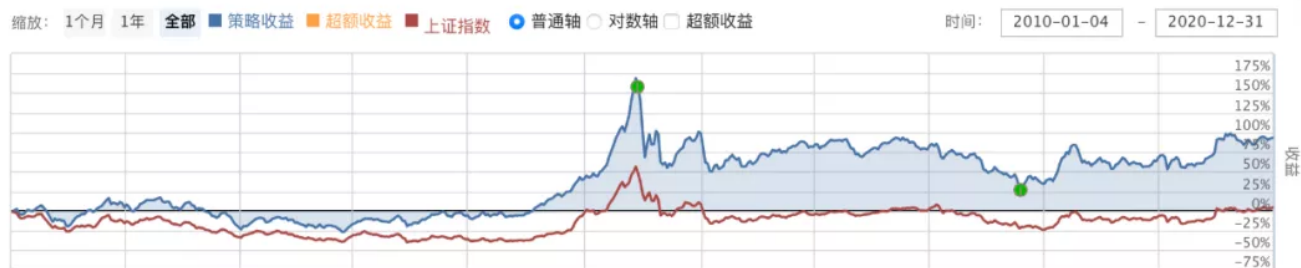
B. Current PE ratio is lower than 40 percent of highest PE ratio in last 5 years



### C. Current dividend yield is greater than 2/3 of 3-year interest rate

#### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤 ?	索提诺比率
94.84%	6.43%	83.84%	5.98%	0.060	1.031	0.102	0.683	9.961	54.53%	0.123
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	17.63%	0.204	0.567	3270	1518	0.648	0.238	0.213	2015/06/12,2018/10/18	



### D. Current market cap is lower than 2/3 of total tangible assets

#### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤 ?	索提诺比率
68.84%	5.02%	59.31%	5.98%	0.046	1.026	0.042	0.668	8.365	61.09%	0.049
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	26.96%	0.041	0.566	4365	2172	0.410	0.245	0.213	2015/06/12,2018/10/18	



E. Current market cap is lower than 2/3 of (current assets - total liability)

收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
79.44%	5.62%	69.32%	5.98%	0.052	1.031	0.065	--	--	62.00%	0.078
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	28.33%	0.091	--	--	--	0.440	0.248	0.213	2015/06/12,2018/10/18	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31



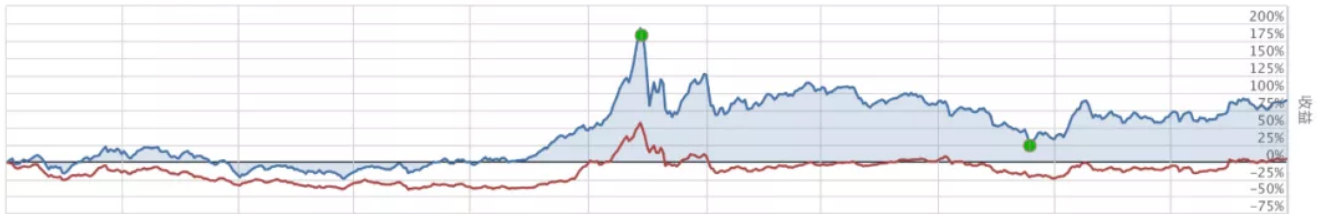
F. Total liability is lower than tangible assets

收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
90.17%	6.19%	79.44%	5.98%	0.057	1.007	0.090	0.667	9.135	59.29%	0.108
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	26.36%	0.139	0.562	1632	815	0.487	0.244	0.213	2015/06/12,2018/10/18	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31



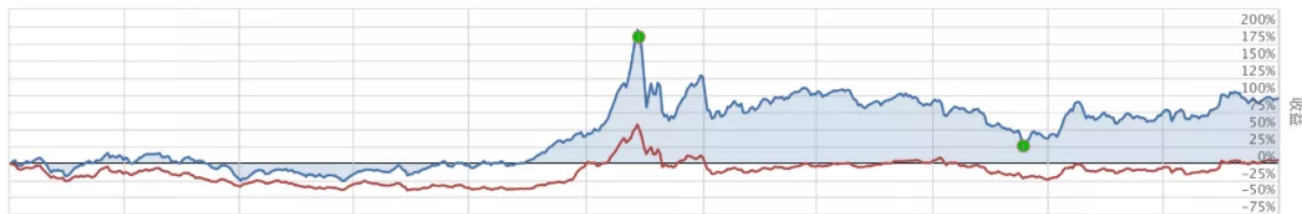
G.  $(\text{current assets} / \text{current liability}) > 2$

#### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
95.93%	6.49%	84.88%	5.98%	0.060	1.020	0.101	0.665	9.908	59.46%	0.121
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.03%	24.90%	0.166	0.565	1253	631	0.517	0.246	0.213	2015/06/12,2018/10/18	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31



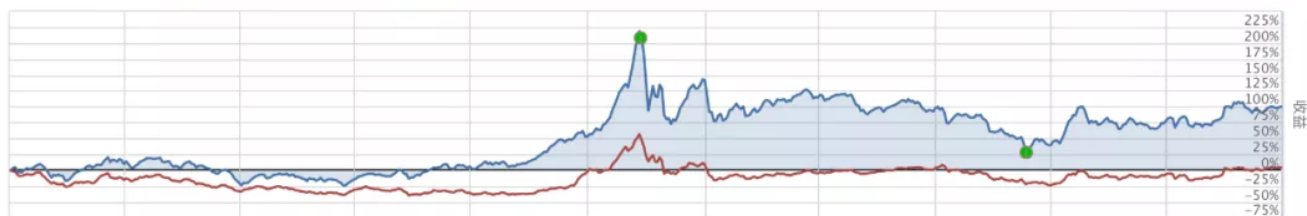
H.  $\text{Total liability} < \text{current liability} * 2$

#### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
101.77%	6.78%	90.39%	5.98%	0.063	1.024	0.111	0.679	9.820	61.38%	0.133
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.03%	27.95%	0.178	0.557	1480	701	0.505	0.251	0.213	2015/06/12,2018/10/18	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31



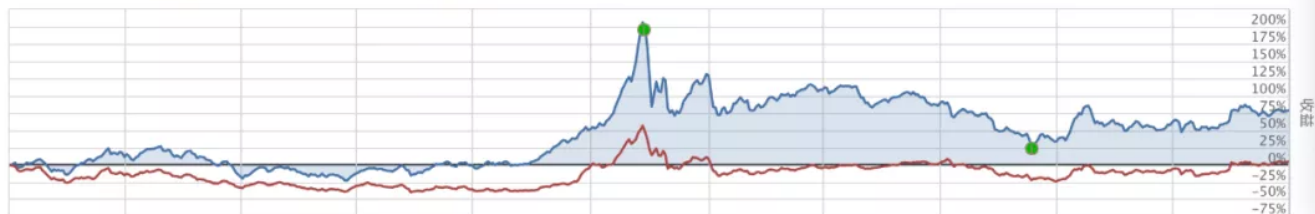
## I. average yearly return in the past 10 years > 7%

### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
80.39%	5.67%	70.21%	5.98%	0.052	1.013	0.069	--	--	61.14%	0.082
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	27.27%	0.104	--	--	--	0.485	0.241	0.213	2015/06/12,2018/10/18	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31



## J. Yearly return cannot below -5% twice in the last 10 years

### 收益概述

策略收益	策略年化收益	超额收益	基准收益	阿尔法	贝塔	夏普比率	胜率	盈亏比	最大回撤	索提诺比率
74.58%	5.35%	64.73%	5.98%	0.037	0.666	0.084	0.503	1.390	39.19%	0.102
日均超额收益	超额收益最大回撤	超额收益夏普比率	日胜率	盈利次数	亏损次数	信息比率	策略波动率	基准波动率	最大回撤区间	
0.02%	24.46%	0.075	0.525	1584	1567	0.466	0.161	0.213	2015/06/12,2019/01/03	

缩放: 1个月 1年 全部 策略收益 超额收益 上证指数 普通轴 对数轴 超额收益

时间: 2010-01-04 - 2020-12-31

