ABU 量化系统 简介(版本 0.1)

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第十部分 解决方案D

'非均衡胜负收益'带来的必然'非均衡胜负比例',目标由'因子'的能力解决一部分,'模式识别'提升关键的一部分

python import ZEnv import ZLog import ZCommonUtil import pandas as pd pd.options.display.max_columns = 100 %matplotlib
inline

使用11年后止今学习的数据作用在10年的回测,由于担心回测数据集中在周期内的学习虽然 对周期的训练集测试集生效,但是完全在周期之外的效果验证

```
run_factor_by_year生成10年的回测结果集
run_func: 回测不开启识别优化
run_func_with_filter: 回测开启识别优化
run_func_with_ml: 回测开启识别但止做数据记录不做优化
```

"python import FactorUnitTest import BuyGoldenFactor import SymbolPd from BuyGoldenFactor import BuyGoldenFactorClass def run_factor_by_year(enable_fiter, enable_filter_pipe_ml, symbols=None): SymbolPd.g_force_folds = 7 SymbolPd.g_force_n_year = 1

```
BuyGoldenFactor.g_enable_filter_ml = False
BuyGoldenFactor.g_enable_fiter = enable_fiter
BuyGoldenFactor.g_enable_filter_pipe_ml = enable_filter_pipe_ml
buy_factors = [{'XD': 42, 'class': BuyGoldenFactorClass, 'draw': True}]
sell_factors = □
parameters = {
                'stop_win_base_n': 4.5,
                'stop_loss_base_n': 2.0,
                'mv_close_atr': 3.5,
                'mv_pre_atr': 2.0,
if symbols is None:
              cap, results, orders_pd, action_pd, all_fit_symbols = FactorUnitTest.random_unit_test(ret_cnt_need=0,
                                          buy_factors=buy_factors, sell_factors=sell_factors, parameters=parameters, show=False)
               cap, \ results, \ orders\_pd, \ action\_pd, \ all\_fit\_symbols = FactorUnitTest.random\_unit\_test(ret\_cnt\_need=0, results) = FactorUnitTest.random\_unitTest(ret\_cnt\_need=0, results) = FactorUnitTest.random\_unitTest(ret\_cnt\_need=0, results) = FactorUnitTest.random\_unitTest(ret\_cnt\_need=0, results) = FactorUnitTest.random\_unitTest(ret\_cnt\_need=0, results) = FactorUnitTest(ret\_cnt\_need=0, results) = FactorUnitTest(
                                           symbols=symbols
                                           buy_factors=buy_factors, sell_factors=sell_factors, parameters=parameters, show=False)
return cap, results, orders_pd, action_pd, all_fit_symbols
```

from functools import partial

run_func = partial(run_factor_by_year, False, False) run_func_with_filter = partial(run_factor_by_year, True, False) run_func_with_ml = partial(run_factor_by_year, True, True)

ipcluster nbextension enable

ipcluster start

import ipyparallel as ipp

rc = ipp.Client()

rsc_filter = rc[0].apply(run_factor_by_year, True)

rsc = rc[1].apply(run_factor_by_year, False) rsc_ret = rsc.get() rsc filter ret = rsc filter.get()

```
makedirs data/DayKLine/2016-09-18

'``python
rsc_ret = run_func()
```

```
backSymbols = None
BuyGoldenFactorClass:42open down 0.07
```

python import MetricsManger from MetricsManger import metrics_rsc from FactorMetrics import METRICSTYPE from UmpMain import UmpMainClass from MlFiterGoldenPd import MlFiterGoldenPdClass

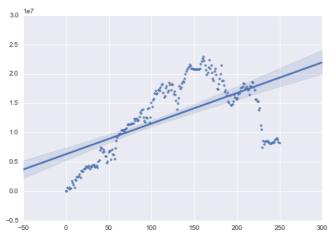
ZCommonUtil.dump_pickle(rsc, fn)

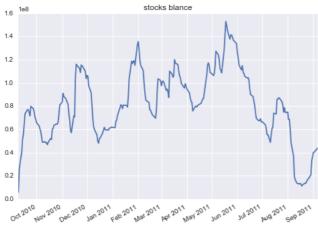
rsc = ZCommonUtil.load_pickle(fn) "

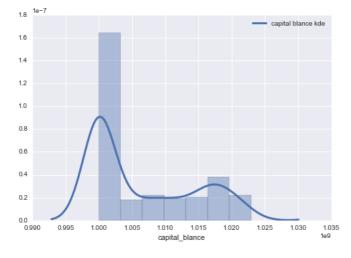
python rsc = metrics_rsc(*rsc_ret) MetricsManger.make_metrics_from_rsc(rsc, METRICSTYPE.SYSMBOL_R_SCORES_GOLDEN.value)
UmpMainClass(rsc.ordersPd, MlFiterGoldenPdClass).show_general()

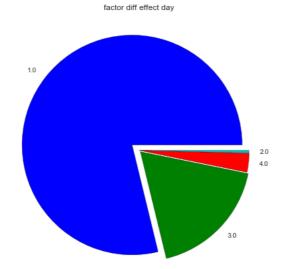


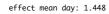
[&]quot;python fn = './data/cache/rsc_abu'

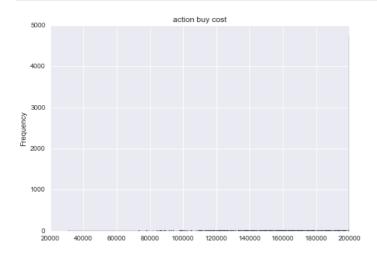




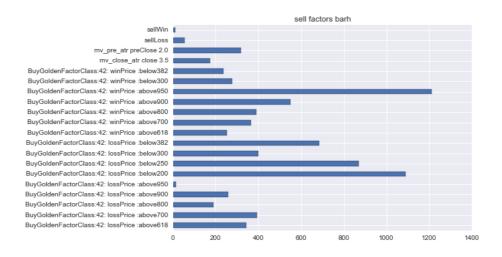




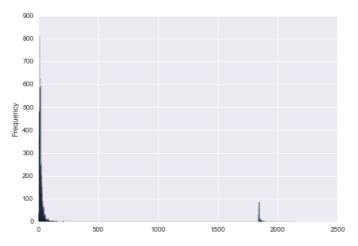




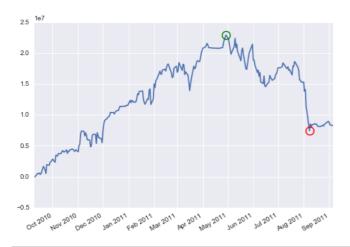
cost info: moments_tuple(mean=174773.97699474267, std=37823.181170398115, skewness=-1.4195782390197598, kurtosis=4.004374570313741)



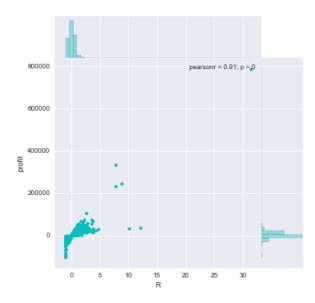
```
BuyGoldenFactorClass:42: lossPrice :above618
BuyGoldenFactorClass:42: lossPrice :above700
                                                       394.0
BuyGoldenFactorClass:42: lossPrice :above800
                                                       189.0
BuyGoldenFactorClass:42: lossPrice :above900
                                                       259.0
BuyGoldenFactorClass:42: lossPrice :above950
                                                       12.0
BuyGoldenFactorClass:42: lossPrice :below200
BuyGoldenFactorClass:42: lossPrice :below250
                                                      1090.0
                                                       869.0
BuyGoldenFactorClass:42: lossPrice :below300
                                                       399.0
BuyGoldenFactorClass:42: lossPrice :below382
                                                       684.0
BuyGoldenFactorClass:42: winPrice :above618
                                                       252.0
BuyGoldenFactorClass:42: winPrice :above700
                                                       365.0
                                                       390.0
Buy Golden Factor Class: 42: \ win Price: above 800
BuyGoldenFactorClass:42: winPrice :above900
BuyGoldenFactorClass:42: winPrice :above950
                                                       551.0
                                                      1211.0
BuyGoldenFactorClass:42: winPrice :below300
                                                       276.0
BuyGoldenFactorClass:42: winPrice :below382
                                                       237.0
mv_close_atr close 3.5
                                                       172.0
mv_pre_atr preClose 2.0
                                                       319.0
sellLoss
                                                        53.0
sellWin
                                                        11.0
dtype: float64
```



```
keep days mean: 20.6230035904
keep days median: buy Date
                                       2.011030e+07
buy Price
                    2.102000e+01
buy Cnt
                    8.458000e+03
Sell Price
                    2.115500e+01
                    1.812000e+01
MaxLoss
                    3.730000e+02
key
profit
                    2.394700e+02
result
                    1.000000e+00
R
                    1.458333e-02
profit_cg
                    1.205083e-03
{\tt profit\_cg\_hunder}
                    1.205083e-01
                    1.600000e+01
keep_days
dtype: float64
factor win effect = 0.0293425776897%
factor loss effect = 0.0846849077628%
```



max down rate: 0.0151533454858 {(Timestamp('2011-04-29 00:00:00'), Timestamp('2011-08-08 00:00:00')): 15500536.470498919}

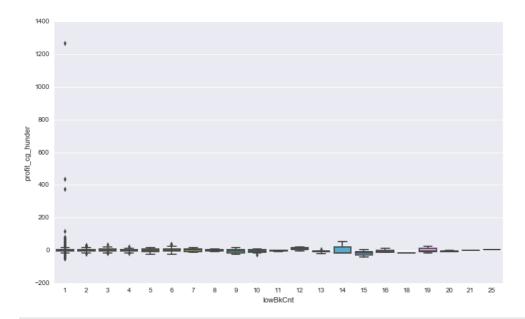


```
factor effect symbol rate: 1.0
factor gen order rate: 2.23976451699
R win rate: 0.512443136206
result win rate: 0.489605734767
R return: 0.125768478197
P return: 0.00781818323751
C return: 0.0082517410685
             52191.1770793
order win mean: 10800.2696193 cg: 7.10351737508
order win max: 788814.72 cg: 1266.29834254
Win Top 5
2011-01-20
                  1266.298343
2011-04-25
                  435.074627
2011-05-13
                   373.801558
                  116.088634
2011-05-16
2011-03-28
                   79.166667
Name: profit_cg_hunder, dtype: float64
order loss mean: -8946.80854235 cg: -5.72894385578
order loss max: -101690.05 cg: -53.8799757276
2011-07-25 -53.879976
2011-05-18 -50.856450
2011-01-21 -48.958950
2011-06-13 -48.476218
2011-01-21 -47.499217
```

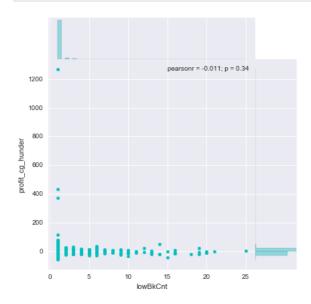
Name: profit_cg_hunder, dtype: float64

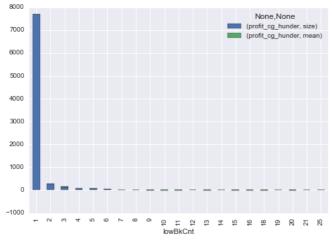


top 10 win profit_cg mean: 262.074517544 top 10 loss profit_cg mean: -44.8430390679

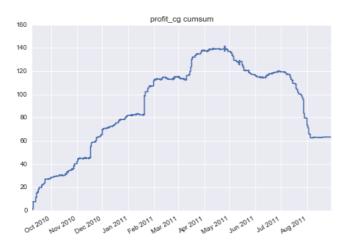


TOP 3 = [(1, 7694), (2, 289), (3, 145)]





```
all fit order = (8077, 30)
win rate = 0.507366596509
profit_cg.sum() = 63.1474660094
win mean = 0.0710351737508 loss_mean = -0.0575207358116
```



python rsc_filter_ret = run_func_with_filter()

```
backSymbols = None
BuyGoldenFactorClass:42open down 0.07
```

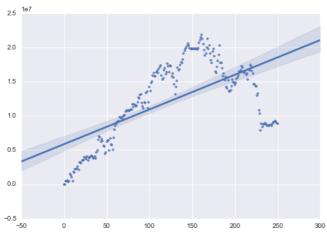
ZCommonUtil.dump_pickle(rsc_filter, fn)

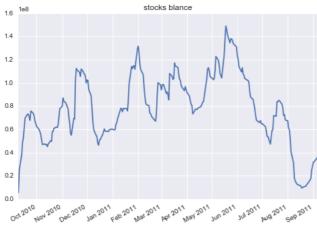
rsc_filter = ZCommonUtil.load_pickle(fn) ```

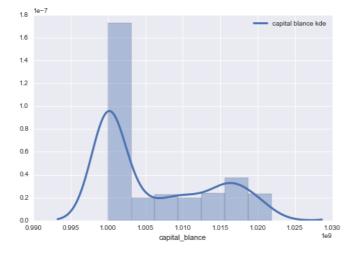
python rsc_filter = metrics_rsc(*rsc_filter_ret) MetricsManger.make_metrics_from_rsc(rsc_filter, METRICSTYPE.SYSMBOL_R_SCORES_GOLDEN.value) UmpMainClass(rsc_filter.ordersPd, MlFiterGoldenPdClass).show_general()

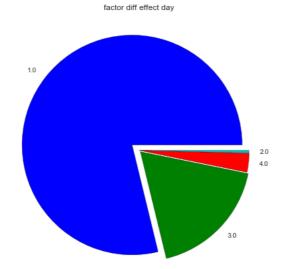


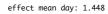
[&]quot;python fn = './data/cache/rsc_filter_abu'

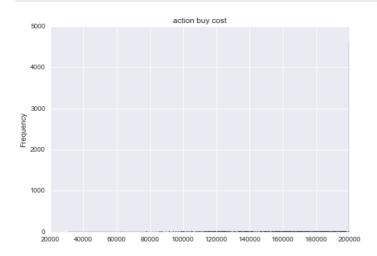




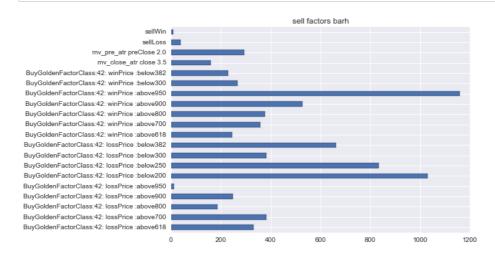




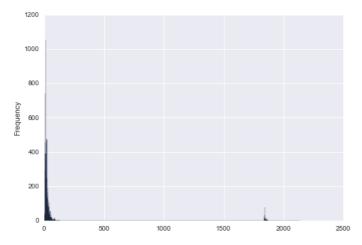




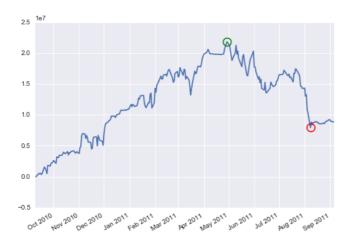
 $cost \ info: \\ moments_tuple(mean=176283.51773019947, \ std=36734.398287652264, \ skewness=-1.495698519948341, \ kurtosis=4.257294980994378)$



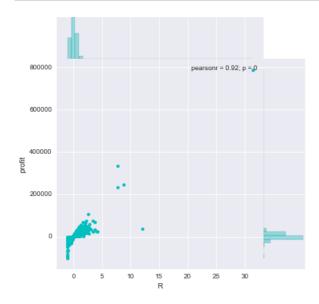
```
BuyGoldenFactorClass:42: lossPrice :above618
BuyGoldenFactorClass:42: lossPrice :above700
                                                       382.0
BuyGoldenFactorClass:42: lossPrice :above800
                                                       186.0
BuyGoldenFactorClass:42: lossPrice :above900
                                                       249.0
BuyGoldenFactorClass:42: lossPrice :above950
                                                       11.0
BuyGoldenFactorClass:42: lossPrice :below200
BuyGoldenFactorClass:42: lossPrice :below250
                                                      1030.0
                                                       835.0
BuyGoldenFactorClass:42: lossPrice :below300
                                                       383.0
BuyGoldenFactorClass:42: lossPrice :below382
                                                       662.0
BuyGoldenFactorClass:42: winPrice :above618
                                                       245.0
BuyGoldenFactorClass:42: winPrice :above700
                                                       357.0
                                                       376.0
Buy Golden Factor Class: 42: \ win Price: above 800
BuyGoldenFactorClass:42: winPrice :above900
BuyGoldenFactorClass:42: winPrice :above950
                                                       527.0
                                                      1159.0
BuyGoldenFactorClass:42: winPrice :below300
                                                       268.0
BuyGoldenFactorClass:42: winPrice :below382
                                                       230.0
mv_close_atr close 3.5
                                                       159.0
mv_pre_atr preClose 2.0
                                                       294.0
sellLoss
                                                        38.0
sellWin
                                                         9.0
dtype: float64
```



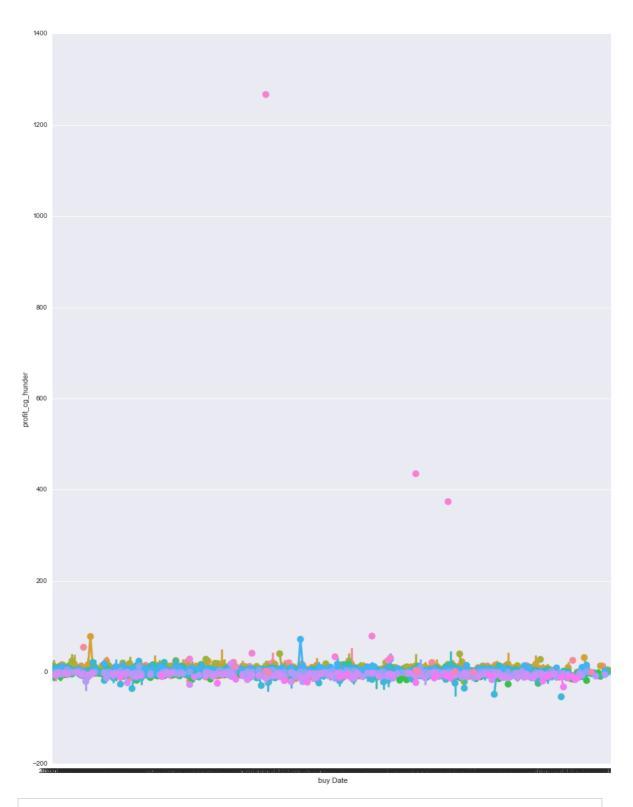
```
keep days mean: 20.6162700466
keep days median: buy Date
                                       2.011030e+07
buy Price
                    2.138375e+01
buy Cnt
                    8.401500e+03
Sell Price
                    2.147475e+01
                    1.844000e+01
MaxLoss
                    3.730000e+02
key
profit
                    3.470000e+02
result
                    1.000000e+00
R
                    1.904935e-02
profit_cg
                    1.824889e-03
{\tt profit\_cg\_hunder}
                    1.824889e-01
                    1.600000e+01
keep_days
dtype: float64
factor win effect = 0.0297465080186%
factor loss effect = 0.0856182100362%
```



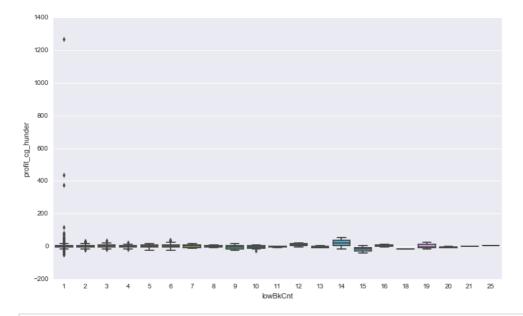
max down rate: 0.013615453933 {(Timestamp('2011-04-29 00:00:00'), Timestamp('2011-08-08 00:00:00')): 13912832.875998974}



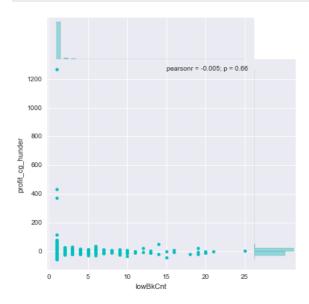
```
factor effect symbol rate: 1.0
factor gen order rate: 2.16259500543
R win rate: 0.518458197611
result win rate: 0.495669637254
R return: 0.124761621145
P return: 0.00851828820408
C return: 0.0088576081095
             50743.7860091
order win mean: 10671.6661611 cg: 6.95253920443
order win max: 788814.72 cg: 1266.29834254
Win Top 5
2011-01-20
                 1266.298343
2011-04-25
                  435.074627
2011-05-13
                   373.801558
                 116.088634
2011-05-16
2011-03-28
                   79.166667
Name: profit_cg_hunder, dtype: float64
order loss mean: -8761.14996828 cg: -5.51658389609
order loss max: -101690.05 cg: -53.8799757276
2011-07-25 -53.879976
2011-05-18 -50.856450
2011-01-21 -48.958950
2011-06-13 -48.476218
2011-01-21 -47.499217
Name: profit_cg_hunder, dtype: float64
```

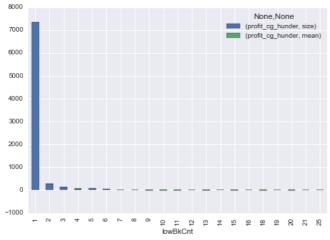


top 10 win profit_cg mean: 259.499429754 top 10 loss profit_cg mean: -44.8430390679



TOP 3 = [(1, 7337), (2, 270), (3, 135)]





```
all fit order = (7732, 30)
win rate = 0.510734609415
profit_cg.sum() = 65.8634043939
win mean = 0.0695253920443 loss_mean = -0.0553854481924
```



python rsc.ordersPd.shape[0] - rsc_filter.ordersPd.shape[0]

403

结果总共提高了1%的胜率,block了403个单子,看看orders diff

"python import TradeProxy reload(TradeProxy) diff_order = TradeProxy.find_unsame_in_2orders(rsc.ordersPd, rsc_filter.ordersPd) ZLog.info(diff_order.shape) diff_order.result.value_counts()

 $\label{limited_control} \mbox{diff_order_has_ret = diff_order[diff_order.result <> 0] xt = \mbox{diff_order_has_ret.result.value_counts() float(xt[-1])/xt.sum() $^{""}$ and $"" = 1$ in (-1) in$

```
(403, 31)
0.5681159420289855
```

只有0.56是生效的, ump失效了?看看pipe line的学习数据分类结果run_func_with_ml

python rsc_ml_ret = run_func_with_ml()

```
backSymbols = None
BuyGoldenFactorClass:42open down 0.07
```

""python rsc_ml = metrics_rsc(*rsc_ml_ret) MetricsManger.make_metrics_from_rsc(rsc_ml, METRICSTYPE.SYSMBOL_R_SCORES_GOLDEN.value, show=False)

import ast def map_str_dict(extra_info, key): try: map_ast = ast.literal_eval(extra_info)[key] except Exception, e: import pdb pdb.set_trace() raise e return map_ast def extra_info_to_pd(orders_pd): orders_pd['can_win'] = orders_pd['Extralnfo'].apply(map_str_dict, args=('can_win',)) orders_pd['edge'] = orders_pd['Extralnfo'].apply(map_str_dict, args=('edge_type',)) orders_pd['edge_type'] = orders_pd['edge_type']

```python

# fn = './data/cache/order\_pd\_ml\_abu' key = 'order\_pd\_ml\_abu' ZCommonUtil.dump\_hdf5(order\_pd\_ml, key, fn) order\_pd\_ml = ZCommonUtil.load\_hdf5(fn, key)

```python
can_win_pd = order_pd_ml.filter(['result', 'can_win', 'edge', 'edge_type'])
can_win_pd.head()

| | result | can_win | edge | edge_type |
|------------|--------|---------|------|-----------|
| 2010-11-09 | 0 | 1 | 1 | 8 |
| 2011-03-30 | 0 | 1 | 0 | 8 |
| 2011-07-12 | 0 | 0 | -1 | 0 |
| 2011-07-15 | 0 | 1 | 0 | 8 |
| 2011-07-28 | 0 | 1 | 0 | 8 |

对照UmpPipeLineClass中learn_pipe_line_predict的返回结果,从crosstab上看整个结果是 理想的,**问题应该出在edge_type = 0的有35个还持有的单子** python pd.crosstab([can_win_pd['result'], can_win_pd['can_win']], can_win_pd['edge_type'])

| | edge_type | 0 | 1 | 4 | 6 | 7 | 8 |
|--------|-----------|----|---|---|---|-----|------|
| result | can_win | | | | | | |
| -1 | 0 | 21 | 6 | 1 | 2 | 166 | 0 |
| | 1 | 0 | 0 | 0 | 0 | 0 | 3783 |
| 0 | 0 | 35 | 1 | 0 | 3 | 19 | 0 |
| | 1 | 0 | 0 | 0 | 0 | 0 | 235 |
| 1 | 0 | 11 | 3 | 0 | 0 | 135 | 0 |
| | 1 | 0 | 0 | 0 | 0 | 0 | 3949 |

以周期内最后一天为限制日,查看这35个单子的盈亏情况,结果喜人阿!

"python from Capital import CapitalClass orders_pd_n_ret = order_pd_ml[(order_pd_ml.result == 0) & (order_pd_ml.edge_type == 0)] cap = CapitalClass(1000000000) def calc_last_loss(order): kl_pd = SymbolPd.make_kfold_pd(order.Symbol, cap=cap) return kl_pd.iloc[-1].close - order['buy Price']

last_loss_result = orders_pd_n_ret.apply(calc_last_loss, axis=1) last_loss_result.sum() import numpy as np dummies_result = pd.Series(np.where(last_loss_result > 0, 1, 0)) float(dummies_result.value_counts()[0])/dummies_result.value_counts().sum() ```

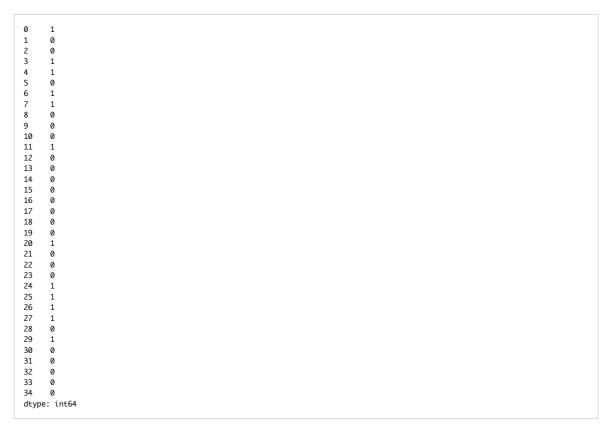
0.6571428571428571

 $python \ dummies_result.value_counts() [\emptyset], \ dummies_result.value_counts() [1]$

(23, 12)

0,0对应的单子最多 从趋势图上看最后一段的下跌趋势中裁判优化祈祷了作用,且判断的正确,综合结果比较接近目标

python dummies_result



'非均衡胜负收益'带来的必然'非均衡胜负比例',目标由'因子'的能力解决一部分,'模式识别'提升关键的一部分

"`python