# 《机器学习驱动的基本面量化投资研究》

# 程序列表

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### 1. 数据及代码说明

#### 1) 数据说明:

- a) factor 文件夹中包含初始 96 项因子数据
- b) factorselect 文件夹中包含筛选完成的 16 项因子数据
- c) returnseries 文件夹中包含 3/12/24/36 个月滑动窗口下各算法构建投资组合月度收益序列
- d) ff3/ff5 分别为 Fama-French3 因子数据, RF 为月度无风险利率数据, final\_return 为股票月度收益数据
- e) ff30 为去掉市值最小的 30%股票后分别根据市值 (size)和收益价格比 (EP) 分组后构 建 MKT/SMB/VMG 3 因子数据
- f) factorEW/factorVW 分别为单因子检验 10-1 多空组合(等权重/市值加权)月度收益 序列

#### 2) 代码说明:

#### a) MainFile.py

主程序, 运行该函数即可得到各主要结果

#### b) DataTransfrom.py

导入基础数据并进行预处理,最后将原始数据转换成每个截面一个  $\frac{1}{2}$   $\frac{1$ 

#### c) NWttest.py

定义对某一序列是否异于 0 进行 Newey and West (1987) t 检验函数

#### d) ReturnSeriesTest.py

在获取各个算法构建多空组合月度收益序列后,对各个收益序列与 OLS 回归 (benchmark) 和 DFN (表现最好的深度算法) 序列是否存在显著差异进行 NW-T 检验

#### e) StrategyConstruct.py

- i. 投资组合构建通用函数 (output),
- ii. FC 和 ensemble 无内置算法包,故单独构建 FC 和 ensemblenn

#### f) FactorTest.py

因子检验补充结果,内容包含:

- i. 在去掉市值最小的 30%股票后分别根据市值(size)和收益价格比(EP)分组后构建 MKT/SMB/VMG 3 因子
- ii. 单因子 10 分组 10-1/1-10 多空组合因子调整收益
- iii. 单因子 10 分组检验各组因子调整收益
- iv. 各因子与 size 因子独立双变量分组检验结果
- v. 各因子与 BM 因子独立双变量分组检验结果
- vi. 6.96 项因子 fama macbeth 回归检验结果

#### g) DFN.py

深度前馈网络核心库文件,包含深度前馈网络核心函数 DFN(),需在 GPU 环境下运行

#### h) RNNMODEL.py

循环神经网络库文件, 包含如下内容:

- i. 基础循环神经网络单元:BaseRnn()
- ii. 可用于训练的循环神经网络整体架构:Istmmodule(),该架构基于 BaseRnn()构建循环神经网络。

该文件需在 GPU 环境下运行

#### i) Ensembleall.py

集成深度学习模型库文件,包含集成模型:EnsembleIr()。

因其中集成深度学习模型,需在 GPU 环境下使用,如不集成深度学习模型,则无 GPU 限制。

#### j) transecfee.py

计算不同交易成本下的投资组合绩效变化,包含核心函数 transecfee()以及 showtransecfee()

直接调用 showtransecfee()即可

#### k) selectFactor.py

## 用于筛选重要因子,包含:

- i. 非循环神经网络模型所用的 dropimportant()函数
- ii. 循环神经网络模型所用的 dropimportant2()函数
- iii. FC 方法的筛选函数 FCselect()函数

注:获取结果需要运行的 python 文件仅为:MainFile.py 和 FactorTest.py

### 2. 代码整理

#### a) MainFile.py

```
1
     #!/usr/bin/env python
 2
     # -*- coding: utf-8 -*-
 3
 4
     @description:
 5
         构建机器学习驱动多因子投资策略主函数
         1. 首先输入各算法参数(参数根据第一个滑动窗口网格调参确定,此处直接输入)
 6
 7
        2. 全样本 3/12/24/36 个月滑动窗口函数运行, 最终直接输出多空组合月度收益,FF3/5-alpha, sharpe
     ratio, 并将收益序列保存在' output'文件夹内
 8
 9
        3. 全样本 3/12/24/36 个月滑动窗口各个算法多空组合月度收益序列是否存在显著差异 NW-T 检验
        4. 不同交易费率下的多空组合绩效结果
10
11
        5. 全样本 12 个月华东窗口下各个算法的特征筛选
         6. 特征筛选后 16 项因子 12 个月滑动窗口函数运行, 最终输出多空组合月度收益, FF3/5-
12
13
     alpha, sharpe ratio, 并将收益序列保存在' output'文件夹内
14
        注:深度学习算法要在使有 GPU 的环境下进行训练
15
16
     from StrategyConstruct import FC, output, output2, comboutput, ensemblenn
17
     from selectFactor import dropimportant, dropimportant2, FCselect
18
     from DataTransfrom import datatransfrom, datatransfrom2
19
     from xgboost.sklearn import XGBRegressor
20
     from sklearn.ensemble import GradientBoostingRegressor
21
     from sklearn.linear_model import LinearRegression, Lasso, ElasticNet, Ridge
22
     from sklearn.cross_decomposition import PLSRegression
23
     from sklearn.neural_network import MLPRegressor
24
     from sklearn.svm import SVR
25
     import DFN
26
     import RNNMODEL as rm
27
     import Ensembleall as ea
28
     import warnings
29
     from mxnet import gpu
30
     import os
31
     from transecfee import showtrasecfee
     from ReturnSeriesTest import returnseriestest
32
     warnings. filterwarnings ('ignore')
33
34
35
36
     #各个算法参数(根据第一个窗口网格调参确定)
     window=[3, 12, 24, 36]
37
     PLS_params=[2, 2, 1, 1]
38
```

```
39
           lasso_params=[1e-3, 5e-4, 0.01, 0.01]
40
           ridge_params=[0.1, 0.005, 0.01, 0.005]
           elasticnet_params={'alpha': [0.01, 1e-3, 0.01, 0.1], '11_ratio': [0.3, 0.3, 0.7, 0.3]}
41
           SVR_params={'kernel':['linear', 'linear', 'rbf', 'rbf'], 'gamma':[1e-3, 1e-3, 
42
43
           4], 'C': [0.01, 0.001, 0.01, 1e-4]}
44
           GBDT_params={'learning_rate': [0.1, 0.1, 0.1, 0.1], 'maxdepth': [2, 3, 2, 2], 'n_estimators': [100, 100,
45
           100, 100]} #XGBOOST 与 GBDT 相同 此处共用
           ENANN_params = {'max_iter': [100, 100, 200, 300], 'p': [0.3, 0.5, 0.7, 0.5]}
46
           DFN_params = {'learning_rate':[0.1, 0.1, 0.01], 'batch': [300, 400, 300, 400]}
47
48
          LSTM_params = {'learning_rate': [1e-4, 1e-5, 1e-4, 1e-6], 'depth': [2, 2, 1, 2],
49
           'hidden number': [256]*4}
50
           RNN_params = {'learning_rate': [0.1, 0.1, 0.001], 'depth': [1, 1, 2, 1],
51
           'hidden_number': [256]*4}
52
53
54
           55
           *********************************#
           path = r'...\DataBase\factor'#96 项因子所在路径
56
           factorname = [x[1:-4] for x in os. listdir(path)]
57
58
           riskfree, timeseries, factor, timeseries2, index = datatransfrom(path)[0],
59
           datatransfrom(path)[1], datatransfrom(path)[2], datatransfrom2(path)[0],
60
           datatransfrom2(path)[1]
           for i in range (4):
61
                  i = 0
62
                  output (window[i], LinearRegression(), 'OLS' +str(window[i]), riskfree[i], timeseries)
63
64
                  FC(window[i], riskfree[i], timeseries, 96, 'FC')
65
                  output(window[i], PLSRegression(PLS_params[i]), 'PLS' + str(window[i]), riskfree[i],
           timeseries)
66
                  output(window[i], Lasso(alpha=lasso_params[i]), 'Lasso' + str(window[i]), riskfree[i],
67
68
69
                  output(window[i], Ridge(alpha=ridge_params[i]), 'Ridge'+str(window[i]), riskfree[i],
70
           timeseries)
71
                  output (window[i], ElasticNet (alpha= elasticnet_params['alpha'] [i], l1_ratio=
           elasticnet params['11 ratio'][i]), 'ElasticNet'+str(window[i]), riskfree[i], timeseries)
72
73
                  output (window[i], SVR (kernel=SVR_params['kernel'][i], gamma= SVR_params ['gamma'][i], C=
           SVR_params ['C'][i] ), 'SVR'+str(window[i]), riskfree[i], timeseries)
74
75
                  output (window[i],
           GradientBoostingRegressor(n estimators=GBDT params['n estimators'][i], max depth=GBDT params[
76
77
           'maxdepth'][i], learning_rate=GBDT_params['learning_rate'][i]), 'GBDT' +
78
           str(window[i]), riskfree[i], timeseries)
79
                  output (window[i],
80
           XGBRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=GBDT_params['maxdepth'][i
```

```
81
      ], learning_rate=GBDT_params['learning_rate'][i]), 'XGBOOST' + str(window[i]), riskfree[i],
82
      timeseries)
          output(window[i], ensemblenn(5, modeluse = MLPRegressor(solver = 'lbfgs',
83
      max_iter=ENANN_params['max_iter'][i]), pickpercent=ENANN_params['p'][i]), 'ENANN' +
84
      str(window[i]), riskfree[i], timeseries)
85
          output (window[i], DFN. DFN (outputdim=1, neural set=[96, 50, 25, 10, 5, 2], ctx=gpu(0),
86
87
      epoch=10, batch_size=DFN_params['batch'][i], lr=DFN_params['learning_rate'][i]), 'DFN' +
      str(window[i]), riskfree[i], timeseries)
88
          output2(window[i], rm.lstmmodule(96, LSTM_params['hidden_number'][i],
89
90
      LSTM_params['depth'][i], 100, 3571, lr=LSTM_params['learning_rate'][i]), 'LSTM'+
91
      str(window[i]) ,riskfree[i], timeseries2)
92
          output2(window[i], rm.lstmmodule(96, RNN params['hidden_number'][i],
      RNN_params['depth'][i], 100, 3571, lr=RNN_params['learning_rate'][i], ntype='RNN'), 'RNN'+
93
      str(window[i]), riskfree[i], timeseries2)
94
95
          modellist = [DFN.DFN(outputdim=1, neuralset=[96, 50, 25, 10, 5, 2], ctx=gpu(0),
96
      epoch=10, batch_size=DFN_params['batch'][i], lr=DFN_params['learning_rate'][i]),
97
                       ensemblenn (5, modeluse = MLPRegressor (solver = 'lbfgs',
      max_iter=ENANN_params['max_iter'][i]), pickpercent=ENANN_params['p'][i]),
98
99
100
      XGBRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=GBDT_params['maxdepth'][i
      ], learning_rate=GBDT_params['learning_rate'][i]),
101
102
103
      GradientBoostingRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=GBDT_params[
104
      'maxdepth'][i], learning_rate=GBDT_params['learning_rate'][i]),
105
                       PLSRegression(PLS_params[i]),
106
                       Ridge(alpha=ridge_params[i]),
107
                       SVR(kernel=SVR_params['kernel'][i], gamma= SVR_params ['gamma'][i], C=
108
      SVR_params ['C'][i])]# PLS 一定要放在倒数第三个 (PLS 输出形式为 list,故进行了进一步处理)
109
          nmolist = [rm. lstmmodule (96, LSTM params ['hidden number'] [i], LSTM params ['depth'] [i],
110
      100, 3571, lr=LSTM_params['learning_rate'][i]),
                     rm. lstmmodule (96, RNN params ['hidden_number'][i], RNN params ['depth'][i],
111
112
      100, 3571, lr=RNN params['learning rate'][i], ntype='RNN')]# 循环神经网络模型
          modelname = ['DFN', 'En-ann', 'xgboost', 'GBDT', 'lasso', 'Elasticnet', 'pls', 'Ridge',
113
      'svm', 'LSTM', 'RNN']
114
          ensemblemodel = ea.Ensemblelr(modellist, nmolist, modelname)
115
116
          comboutput(window[i], ensemblemodel, 'Ensemble' +str(window[i]), riskfree[i], timeseries2,
117
118
       119
       ***********
120
121
      for i in window:
122
          returnseriestest(i)
```

```
123
124
      #******* 不同交易费率情形
125
      ************
126
127
      showtrasecfee (0.005)
128
      showtrasecfee (0.0075)
129
      showtrasecfee (0.01)
130
131
      132
      133
134
      i = 1#选取 12 个月滑动窗口筛选因子
135
      dropimportant(window[i] ,LinearRegression(), 'OLS'+str(window[i]), factorname,
      timeseries, 0.0201)
136
137
      FCselect(factorname, timeseries)
138
      dropimportant (window[i], PLSRegression (PLS_params[i]), 'PLS', factorname, timeseries,
139
      0.0230)
      dropimportant(window[i], Lasso(alpha=lasso_params[i]), 'Lasso', factorname, timeseries,
140
141
      0.0208)
142
      dropimportant(window[i], Ridge(alpha=ridge_params[i]), 'Ridge', factorname, timeseries,
143
      0.0208)
144
      dropimportant(window[i], ElasticNet(alpha= elasticnet params['alpha'] [i], 11 ratio=
145
      elasticnet_params['11_ratio'][i]), 'ElasticNet', factorname, timeseries, 0.0212)
146
      dropimportant(window[i], SVR(kernel=SVR_params['kernel'][i], gamma= SVR_params
147
      ['gamma'][i], C= SVR_params ['C'][i] ), 'SVR', factorname, timeseries, 0.0225)
148
      dropimportant(window[i],
149
      GradientBoostingRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=GBDT_params[
      'maxdepth'][i], learning rate=GBDT params['learning_rate'][i]), 'GBDT', factorname,
150
151
      timeseries, 0.0268)
152
      dropimportant(window[i], XGBRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=G
      BDT_params['maxdepth'][i], learning_rate=GBDT_params['learning_rate'][i]), 'XGBOOST',
153
154
      factorname, timeseries, 0.0273)
155
      dropimportant (window[i], ensemblenn (5, modeluse = MLPRegressor (solver = 'lbfgs',
      max iter=ENANN params['max iter'][i]), pickpercent=ENANN params['p'][i]), 'ENANN',
156
157
      factorname, timeseries, 0.0234)
      dropimportant(window[i], DFN.DFN(outputdim=1, neuralset=[96, 50, 25, 10, 5, 2], ctx=gpu(0),
158
159
      epoch=10, batch_size=DFN_params['batch'][i], lr=DFN_params['learning_rate'][i]), 'DFN',
160
      factorname, timeseries, 0.0278)
161
      dropimportant2(window[i], rm.lstmmodule(95, LSTM_params['hidden_number'][i],
162
      LSTM params ['depth'][i], 100, 3571, lr=LSTM params ['learning rate'][i]), 'LSTM', factorname,
163
      timeseries2, 0.0257)
164
      dropimportant2(window[i], rm.lstmmodule(95, RNN_params['hidden_number'][i],
```

```
165
       RNN params['depth'][i], 100, 3571, lr=RNN params['learning_rate'][i], ntype='RNN'), 'RNN',
166
       factorname, timeseries2, 0.0210)
167
168
169
       170
       ***********
171
       path = r'...\DataBase\factorselect'#经过筛选后因子集合所在路径
172
       riskfree, timeseries, factor, timeseries2-datatransfrom(path)[0], datatransfrom(path)[1], datatra
173
       nsfrom(path)[2], datatransfrom2(path, after=True)[0]
174
       i=1 #选取 12 个月滑动窗口测试筛选后因子集合绩效表现
175
       output (window[i], LinearRegression(), 'OLS' +str(window[i]), riskfree[i], timeseries)
       FC(window[i], riskfree[i], timeseries, 11, 'FC')
176
       output(window[i], PLSRegression(PLS_params[i]), 'PLS' + str(window[i]), riskfree[i],
177
178
       timeseries)
179
       output (window[i], Lasso (alpha=lasso_params[i]), 'Lasso' + str(window[i]), riskfree[i],
180
       timeseries)
181
       output (window[i], Ridge (alpha=ridge_params[i]), 'Ridge' +str(window[i]), riskfree[i],
182
       timeseries)
183
       output(window[i], ElasticNet(alpha= elasticnet_params['alpha'] [i], l1_ratio=
184
       elasticnet_params['11_ratio'][i]), 'ElasticNet'+str(window[i]), riskfree[i], timeseries)
       output(window[i], SVR(kernel=SVR params['kernel'][i], gamma= SVR_params ['gamma'][i], C=
185
186
       SVR params ['C'][i]), 'SVR'+str(window[i]), riskfree[i], timeseries)
187
       output (window[i],
188
      GradientBoostingRegressor(n_estimators=GBDT_params['n_estimators'][i], max_depth=GBDT_params[
       'maxdepth'][i], learning_rate=GBDT_params['learning_rate'][i]), 'GBDT' +
189
190
       str(window[i]), riskfree[i], timeseries)
191
       output (window[i],
       XGBRegressor(n estimators=GBDT params['n_estimators'][i], max depth=GBDT params['maxdepth'][i
192
       ], learning_rate=GBDT_params['learning_rate'][i]), 'XGBOOST' + str(window[i]), riskfree[i],
193
194
       timeseries)
       output(window[i], ensemblenn(5, modeluse = MLPRegressor(solver = 'lbfgs',
195
196
       max iter=ENANN params['max iter'][i]), pickpercent=ENANN params['p'][i]), 'ENANN' +
197
       str(window[i]), riskfree[i], timeseries)
198
       output (window[i], DFN. DFN (outputdim=1, neural set=[16, 50, 25, 10, 5, 2], ctx=gpu(0),
       epoch=10, batch_size=DFN_params['batch'][i], lr=DFN_params['learning_rate'][i]), 'DFN' +
199
200
       str(window[i]), riskfree[i], timeseries)
201
       output2(window[i], rm.lstmmodule(11, LSTM_params['hidden_number'][i],
       LSTM params ['depth'][i], 100, 3571, 1r=LSTM params ['learning rate'][i]), 'LSTM'+
202
203
       str(window[i]) ,riskfree[i], timeseries2)
       output2(window[i], rm.lstmmodule(11, RNN_params['hidden_number'][i],
204
205
       RNN_params['depth'][i], 100, 3571, lr=RNN_params['learning_rate'][i], ntype='RNN'), 'RNN'+
206
       str(window[i]), riskfree[i], timeseries2)
```

#### b) DataTransfrom.py

```
#!/usr/bin/env python
 1
2
     # -*- coding: utf-8 -*-
 3
 4
     @description:
 5
         导入基础数据并进行一些预处理,最后变成每个截面一个 Dataframe, 列名为[各因子名称+
     stock'+'ret'], index 代表单只股票
6
7
     import glob, os
8
9
     import pandas as pd
10
     import warnings
11
12
     #***************************1. 导入因子数据,无风险利率,股票月度收益数据
13
     14
     warnings. filterwarnings ('ignore')
15
     def datatransfrom(datapath):
16
         path=datapath
17
         file = glob. glob (os. path. join (path, "*. csv"))
18
        k=[]
         for i in range(len(file)):
19
            k.append(pd.read_csv(file[i]))
20
21
         #股票月度收益
         ret=pd. read_csv('...\DataBase\\final_return. csv')
22
23
         #无风险利率
         rf=pd.read_csv('..\DataBase\\RF.csv')
24
         rf3=rf.iloc[3:-1,:]
25
         rf12=rf.iloc[12:-1,:]
26
27
         rf24=rf.iloc[24:-1,:]
         rf36=rf.iloc[36:-1,:]
28
         riskfree = [rf3, rf12, rf24, rf36]
29
30
         #因子名称
31
         factor=[]
32
         for i in range(len(file)):
             factor.append(file[i][20:-4])
33
         factor.append('stock')
34
35
36
37
         #*****对原始数据进行预处理,每个截面一个 Dataframe, 列名为 {96 因子名称+' stock'+'ret'},
     index 代表单只股票*****#
38
         timeseries=[]
39
         for i in range(len(ret.columns)-1):
40
```

```
41
             kl=pd.concat([k[j].iloc[:,i+1] for j in range(len(file))], axis=1)
             kl['stock'] = ret.iloc[:, 0]
42
             kl.columns = factor
43
             kl=kl.iloc[:-2,:]
44
             timeseries.append(kl)
45
46
         #删除月度收益不存在的数据条
47
         for i in range(len(timeseries)):
             timeseries[i]['ret']=ret.iloc[:,i+1]
48
             timeseries[i]['ret']=timeseries[i]['ret'].fillna('null')
49
             timeseries[i]=timeseries[i][~timeseries[i]['ret'].isin(['null'])]
50
51
         return riskfree, timeseries, factor
52
53
     ## 为 LSTM\RNN 设计的数据读取函数
54
     def datatransfrom2(datapath, after=False):
55
         path=datapath
56
         file = glob. glob (os. path. join (path, "*. csv"))
57
         k=[]
         for i in range(len(file)):
58
             k.append(pd.read_csv(file[i]))
59
60
         #股票月度收益
         ret=pd. read_csv('..\DataBase\\final_return. csv')
61
62
         #因子名称
         factor=[]
63
         for i in range(len(file)):
64
             factor.append(file[i][20:-4])
65
         factor.append('stock')
66
67
68
         #*****对原始数据进行预处理,每个截面一个 Dataframe, 列名为 {96 因子名称+' stock'+'ret'},
      index 代表单只股票*****#
69
         timeseries2=[]
70
         index = []
71
         for i in range(len(ret.columns)-1):
72
73
             kl=pd.concat([k[j].iloc[:,i+1] for j in range(len(file))], axis=1)
             kl['stock'] = ret.iloc[:,0]
74
             kl.columns = factor
75
             if after:# 保证筛选后因子个数为 3571 个
76
                 kl = kl.iloc[:, :]
77
             else:
78
                 kl = kl. iloc[:-2,:]
79
             timeseries2.append(kl)
80
81
         # 加入月度收益, 令月度收益不存在为 null, 方便下一步函数处理
82
         for i in range(len(timeseries2)):
```

```
timeseries2[i]['ret'] = ret.iloc[:, i + 1]

timeseries2[i]['ret'] = timeseries2[i]['ret'].fillna('null')

index.append(timeseries2[i]['ret'].isin(['null']))

return timeseries2, index
```

#### c) NWttest.py

```
1
      # -*- coding:utf-8 -*-
      ,,,
2
3
      @description:
4
     NW-t 检验所用包
5
6
7
      import numpy as np
8
     from collections import namedtuple
9
     from scipy.stats import distributions
10
11
     def _ttest_finish(df, t):
12
13
      :param df:自由度
14
         :param t: t值
15
         :return: 输出 t 和对应 p 值
16
         prob = distributions. t. sf(np. abs(t), df) * 2 # use np. abs to get upper tail
17
         if t. ndim == 0:
18
             t = t[()]
19
20
         return t, prob
21
     NWt_1sampleResult = namedtuple('NWT_1sampResult', ('statistic', 'pvalue'))
22
23
      def nwttest_1samp(a, popmean, axis=0, L=1):
         111
24
25
         主函数
26
         :param a: 数据列表
27
         :param popmean: 原假设值 u0
         :param axis: 行还是列, 默认行
28
29
         :param L: lag, 滞后多少, 默认 l
30
          :return: 输出 nw-t 和对应 p 值
31
32
         a = np. array(a)
33
         N = len(a)
34
         df = N-1
35
         e = a - np. mean(a)
         residuals = np. sum(e**2)
36
37
         Q = 0
38
         for i in range(L):
39
             w_1 = 1 - (i+1)/(1+L)
40
             for j in range (1, N):
                 Q += w_1*e[j]*e[j-(i+1)]
41
         S = residuals + 2*Q
42
```

```
nw_var = S/N

d = np. mean(a, axis) - popmean

nw_sd = np. sqrt(nw_var / float(df))

with np. errstate(divide='ignore', invalid='ignore'):

t = np. divide(d, nw_sd)

t, prob = _ttest_finish(df, t)

return NWt_1sampleResult(t, prob)
```

#### d) ReturnSeriesTest.py

```
1
      #!/usr/bin/env python
2
      # -*- coding: utf-8 -*-
 3
 4
      @description:
 5
          在获取各个算法构建多空组合月度收益序列后,对各个收益序列与 OLS 回归 (benchmark)
6
          和 DFN (表现最好的深度算法) 序列是否存在显著差异进行 NW-T 检验
7
8
9
      import glob, os
10
      import pandas as pd
11
      from NWttest import nwttest_1samp
12
      import warnings
13
      warnings.filterwarnings('ignore')
14
15
     def returnseriestest(length):
16
         path = r'..\DataBase\returnseries'+'\\'+str(length)
17
         file = glob.glob(os.path.join(path, "*.csv"))
18
         ols = pd. read_csv(path+'\\OLS'+str(length)+'.csv')
19
20
         dfn=pd. read_csv(path+' \\DFN' +str(length) +' . csv')
21
         k = [] # 每个算法一个 df
         for i in range(len(file)):
22
              k. append (pd. read csv(file[i]))
23
          #OLS 与其他算法区别
24
         for i in range(len(k)):
25
26
              t = []
              t1 = nwttest_1samp(k[i].iloc[:, 1] - ols['long-short'], 0)
27
              t. append(t1. statistic)
28
              t2 = nwttest_1samp(k[i].iloc[:, 2] - ols['long'], 0)
29
              t. append(t2. statistic)
30
              t3 = nwttest_1samp(k[i].iloc[:, 3] - ols['short'], 0)
31
32
              t. append(t3. statistic)
              print('ols-'+file[i][27:-4], t)
33
          #DFN 与其他算法区别
34
          for i in range(len(k)):
35
              t = []
36
              t1 = nwttest_1samp(-k[i].iloc[:, 1] + dfn['long-short'], 0)
37
              t. append(t1. statistic)
38
39
              t2 = nwttest_1samp(-k[i].iloc[:, 2] + dfn['long'], 0)
              t.append(t2.statistic)
40
              t3 = nwttest_1samp(-k[i].iloc[:, 3] + dfn['short'], 0)
41
              t. append(t3. statistic)
42
```

print('dfn-'+file[i][27:-4], t)

44 return

#### e) StrategyConstruct.py

```
1
     #!/usr/bin/env python
 2
     # -*- coding: utf-8 -*-
 3
 4
     @description:
 5
         1. 投资组合构建通用函数 (output),函数最终输出多空组合月度收益, FF3/5-alpha, sharpe ratio
6
         2. FC 和 ensemble 无内置算法包,此处单独构建 FC 和 ensemblenn
 7
 8
9
     import glob, os
10
     import pandas as pd
11
     import numpy as np
     from scipy import stats
12
13
     import statsmodels.api as sm
14
     from xgboost.sklearn import XGBRegressor
15
     from sklearn.ensemble import GradientBoostingRegressor
16
     from sklearn.preprocessing import scale
     from sklearn.linear_model import LinearRegression
17
     from sklearn.cross_decomposition import PLSRegression
18
19
     from sklearn. linear model import Lasso, ElasticNet, Ridge
20
     from NWttest import nwttest_lsamp
21
     from sklearn.svm import SVR
22
     import warnings
23
     from sklearn.neural_network import MLPRegressor
24
     import gc
25
26
27
      28
     #python 里没有 EN-ANN 和 FC 的对应算法包 此处先定义算法计算方式
     #EN-ANN
29
     class ensemblenn(object):
30
31
         def __init__(self, ensemblenumbers, modeluse = MLPRegressor(solver = 'lbfgs'), pickpercent
32
     = 0.5):
33
             self.ensemblenumbers = ensemblenumbers
             self.modellist = []
34
             self.score = [0]*ensemblenumbers
35
             for i in range(self.ensemblenumbers):
36
                 self.modellist.append(modeluse)
37
38
             self.pickpercent = pickpercent
39
         def fit(self, X, Y):
             for i in range(self.ensemblenumbers):
40
41
                 self.modellist[i].fit(X, Y)
                 self.score[i] = self.modellist[i].loss
42
```

```
43
          def predict(self, xtest):
44
              usemodel = np. array(self. modellist)[np. argsort(np. array(self. score))]
45
              usemodel = usemodel[0:self.ensemblenumbers//2]
              predict = []
46
              for i in range(self.ensemblenumbers//2):
47
                  predict.append(usemodel[i].predict(xtest))
48
              return list(np.mean(predict, axis=0))
49
      #FC
50
51
      def FC(length, rf, timeseries, lenn=96, na='FC'):
52
          #length 为滑动窗口长度: 取值{3, 12, 24, 36}
53
          #na 为输出文件名称
54
          #rf 为无风险利率,取值与 length 对应 {rf3, rf12, rf24, rf36}
55
         Long_Short = []
          Long = []
56
57
          Short = []
58
          for i in range(len(timeseries) - length):
              print(i)
59
              FINALm = pd. concat(timeseries[i:i + (length +1)], axis=0)
60
              FINALm = FINALm. fillna(0)
61
              FINAL_X = FINALm.iloc[:, :-2]
62
63
              FINAL x = scale(FINAL X)
64
              final = pd. concat(timeseries[i:i + length], axis=0)
              x_train = FINAL_x[:len(final)]
65
              x_{test} = FINAL_x[len(final):]
66
67
              y_{train} = final.iloc[:, -1]
              test = timeseries[i + length]
68
              clf = LinearRegression()
69
70
              k = []
              for i in range(lenn):
71
72
                  x = x_{train}[:, i]. reshape(-1, 1)
                  clf.fit(x, y_train)
73
                  k.append(clf.coef [0])
74
              PREDICTION = []
75
              for i in range(len(x_test)):
76
77
                  test0 = np. array(x_test[i])
                  y = 0
78
79
                  for j in range(lenn):
                      y = y + test0[j] * k[j]
80
                  PREDICTION. append (y)
81
82
              y_{test} = test.iloc[:, -1]
83
              # 构建投资组合
              r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
84
85
              r_ture = pd. DataFrame(y_test)
              r ture.columns = ['ture']
86
```

```
87
               r_ture.index = r_predict.index
               FINAL = pd. concat([r_predict, r_ture], axis=1)
88
89
               FINAL_sort = FINAL. sort_values(by='predict', axis=0)
90
               r_final = np. array(FINAL_sort['ture'])
91
               m = int(len(r_final) * 0.1) + 1
               r_final = r_final.tolist()
92
               long = r final[-m:]
93
94
               short = r_final[:m]
95
               r_{end} = (np. sum(long) - np. sum(short)) / m
               Long_Short.append(r_end)
96
97
               Long. append (np. average (long))
98
               Short. append (np. average (short))
           T_value = []
99
           Mean = []
100
101
           p_value = []
102
           sharpratio = []
           Std = []
103
104
           TO = [Long Short, Long, Short]
105
           for 1 in TO:
106
               t_{tst} = nwttest_{1samp}(1, 0)
107
               mean = np. average (1) * 12
108
               STD = np. std(1) * np. sqrt(12)
109
               sharp = (mean - rf.mean().tolist()[0] * 12 / 100) / STD
110
               T_value.append(t_test.statistic)
111
               p_value.append(t_test.pvalue)
               Mean. append (mean)
112
113
               Std. append (STD)
114
               sharpratio.append(sharp)
115
           name = na
116
           length = length
           print(name, 'long-short', 'long', 'short')
117
           print('mean', Mean[0] / 12, Mean[1] / 12, Mean[2] / 12)
118
           print('t-statistic', '(' + str(round(T_value[0], 4)) + ')', '(' + str(round(T_value[1],
119
       4)) + ')',
120
121
                 '(' + str(round(T_value[2], 4)) + ')')
           A = pd. DataFrame (Long_Short, columns=['long-short'])
122
           B = pd. DataFrame (Long, columns=['long'])
123
           C = pd. DataFrame(Short, columns=['short'])
124
           M = pd. concat([A, B, C], axis=1)
125
           M. to_csv('...\output\\'+name + '.csv')
126
127
           ff3 = pd. read_csv('..\DataBase\\ff3.csv')
           ff5 = pd. read_csv('...\DataBase\\ff5.csv')
128
129
           alpha3 = []
           t3 = []
130
```

```
131
          t5 = []
          alpha5 = []
132
133
          for i in range(3):
              X1 = ff3.iloc[length:, 1:]
134
              X2 = ff5.iloc[length:, 1:]
135
              Y = M. iloc[:-2, i]
136
              Y. index = X1. index
137
              Y = Y - rf.RF[:-1] / 100
138
139
              x1 = sm. add_constant(X1)
              reg = sm. OLS(Y, x1). fit()
140
141
              t3. append (reg. tvalues [0])
142
              alpha3. append (reg. params [0] * 12)
143
              x2 = sm. add_constant(X2)
144
              reg = sm. OLS(Y, x2). fit()
145
              t5. append (reg. tvalues [0])
146
              alpha5. append (reg. params [0] * 12)
          print('alpha-FF3', alpha3[0] / 12, alpha3[1] / 12, alpha3[2] / 12)
147
          print('t-statistic', '('+str(round(t3[0], 4)) + ')', '('+str(round(t3[1], 4)) + ')',
148
                '(' + str(round(t3[2], 4)) + ')')
149
          print('alpha-FF5', alpha5[0] / 12, alpha5[1] / 12, alpha5[2] / 12)
150
          print('t-statistic', '(' + str(round(t5[0], 4)) + ')', '(' + str(round(t5[1], 4)) + ')',
151
152
                '(' + str(round(t5[2], 4)) + ')')
          print('sharpe', sharpratio[0], sharpratio[1], sharpratio[2])
153
154
155
      156
157
      def output(length, CLF, name, rf, timeseries):
158
          #length 为滑动窗口长度: 取值{3,12,24,36}
159
          #CLF 为预测选取的机器学习模型
160
          #name 为输出文件名称 (type:string)
          #rf 为无风险利率,取值与 length 对应 {rf3, rf12, rf24, rf36}
161
162
          Long Short = []
163
          Long = []
          Short = []
164
          for i in range(len(timeseries) - (length)):
165
166
              print(i)
167
              FINALm = pd. concat(timeseries[i:i + (length+1)], axis=0)
              FINALm = FINALm. fillna(0) #因子缺失值以 0 填充
168
              FINAL X = FINALm.iloc[:, :-2]
169
170
              FINAL_x = scale(FINAL_X)
171
              final = pd.concat(timeseries[i:i + length], axis=0)
172
              x_{train} = FINAL_x[:len(final)]
              x_{test} = FINAL_x[len(final):]
173
              y train = final.iloc[:, -1]
174
```

```
175
                test = timeseries[i + length]
               y_{test} = test.iloc[:, -1]
176
177
                # 基准-linear
178
               clf = CLF
179
               clf.fit(x_train, y_train)
               PREDICTION = clf.predict(x_test)
180
181
                # 构建投资组合
                prediction = pd. DataFrame (PREDICTION)
182
                r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
183
184
                r_ture = pd. DataFrame(y_test)
185
                r ture.columns = ['ture']
186
                r_ture.index = r_predict.index
               FINAL = pd.concat([r_predict, r_ture], axis=1)
187
188
               FINAL_sort = FINAL. sort_values(by='predict', axis=0)
189
                r_final = np. array(FINAL_sort['ture'])
190
               m = int(len(r final) * 0.1) + 1
               r_final = r_final.tolist()
191
192
                long = r final[-m:]
                short = r_final[:m]
193
194
                r_{end} = (np. sum(long) - np. sum(short)) / m
195
               Long Short. append (r end)
196
               Long. append (np. average (long))
197
               Short. append (np. average (short))
           T_value = []
198
199
           Mean = []
           p_value = []
200
201
           sharpratio = []
           Std = []
202
203
           TO = [Long_Short, Long, Short]
204
           for 1 in TO:
205
                t_{test} = nwttest_{1samp}(1, 0)
206
               mean = np. average(1) * 12
207
               STD = np. std(1) * np. sqrt(12)
                sharp = (mean - rf.mean().tolist()[0]*12/100) / STD
208
209
               T_value.append(t_test.statistic)
210
                p_value.append(t_test.pvalue)
211
               Mean. append (mean)
212
               Std. append (STD)
213
                sharpratio.append(sharp)
           print(name, 'long-short', 'long', 'short')
214
215
           print('mean', Mean[0]/12, Mean[1]/12, Mean[2]/12)
           print('t-statistic', '('+str(round(T_value[0], 4))+')', '('+str(round(T_value[1], 4))+')',
216
       '('+str(round(T_value[2], 4))+')')
217
           A = pd. DataFrame (Long Short, columns=['long-short'])
218
```

```
219
           B = pd. DataFrame (Long, columns=['long'])
220
           C = pd. DataFrame (Short, columns=['short'])
221
           M = pd. concat([A, B, C], axis=1)
222
           M. to_csv('...\output\\'+name+'.csv')
           ff3 = pd. read_csv('..\DataBase\\ff3.csv')
223
           ff5 = pd. read_csv('...\DataBase\\ff5.csv')
224
           alpha3 = []
225
           t3 = []
226
227
           t5 = []
           alpha5 = []
228
229
           for i in range (3):
230
               X1 = ff3.iloc[length:, 1:]
231
               X2 = ff5.iloc[length:, 1:]
232
               Y = M. iloc[:-2, i]
233
               Y. index = X1. index
234
               Y = Y - rf.RF[:-1] / 100
               x1 = sm. add_constant(X1)
235
               reg = sm. OLS(Y, x1). fit()
236
               t3. append (reg. tvalues [0])
237
238
               alpha3.append(reg.params[0] * 12)
239
               x2 = sm. add constant(X2)
240
               reg = sm. OLS(Y, x2). fit()
               t5. append (reg. tvalues [0])
241
               alpha5. append (reg. params [0] * 12)
242
           print('alpha-FF3', alpha3[0]/12, alpha3[1]/12, alpha3[2]/12)
243
           print('t-statistic', '('+str(round(t3[0],4))+')', '('+str(round(t3[1],4))+')',
244
245
       '('+str(round(t3[2],4))+')')
246
           print ('alpha-FF5', alpha5[0]/12, alpha5[1]/12, alpha5[2]/12)
           print('t-statistic', '('+str(round(t5[0], 4))+')', '('+str(round(t5[1], 4))+')',
247
248
       '('+str(round(t5[2], 4))+')')
           print('sharpe', sharpratio[0], sharpratio[1], sharpratio[2])
249
250
251
       # 因为 LSTM 与 RNN 一个步长内所用数据形状必须一致,设置专用的主函数供使用
252
       def output2(length, CLF, name, rf, timeseries2):
253
           # length 为滑动窗口长度: 取值{3,12,24,36}
254
           # CLF 为预测选取的机器学习模型
255
           # name 为输出文件名称 (type:string)
           # rf 为无风险利率,取值与 length 对应 {rf3, rf12, rf24, rf36}
256
           Long Short = []
257
258
           Long = []
259
           Short = []
           for i in range(len(timeseries2) - length):
260
261
               FINALm = pd. concat(timeseries2[i:(i + length + 1)], axis=0)
               FINALm[~FINALm['ret'].isin(['null'])] =
262
```

```
FINALm[~FINALm['ret']. isin(['null'])]. fillna(0)
263
264
                                FINAL X = FINALm.iloc[:, :-2]
265
                                FINAL_X = FINAL_X
266
                                FINAL_x[~FINALm['ret'].isin(['null'])] =
               scale(FINAL_X[~FINALm['ret'].isin(['null'])])
267
                                FINAL_x[FINALm['ret'].isin(['null'])] = 0
268
269
                                FINALm[FINALm['ret'].isin(['null'])] = 0
270
                                 x_{train} = [FINAL_x.iloc[j * 3571:(j + 1) * 3571, :].values for j in range(length)]
271
                                 y_{train} = [FINALm.iloc[j * 3571:(j + 1) * 3571, -1].values for j in range(length)]
272
                                 x_{test} = np. array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1, state test) array([FINAL_x. iloc[j * 357
273
               length + 1)
274
                                 y_{test} = list(FINALm.iloc[(length) * 3571:(length + 1) * 3571, -1].values)
275
                                 # 基准-linear
276
                                 clf = CLF
277
                                 clf.fit(x_train, y_train)
278
                                 PREDICTION = clf.predict(x test)
279
                                 PREDICTION = [PREDICTION[m][0] for m in range (3571 * (length - 1), 3571 * length)]
                                 # 构建投资组合
280
                                 r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
281
282
                                 r_ture = pd. DataFrame(y_test)
283
                                 r ture.columns = ['ture']
284
                                 r_ture.index = r_predict.index
                                FINAL = pd. concat([r predict, r ture], axis=1)
285
                                FINAL = FINAL["timeseries2[i + length]['ret'].isin(['null'])]
286
                                 FINAL_sort = FINAL. sort_values(by='predict', axis=0)
287
                                 r_final = np. array(FINAL_sort['ture'])
288
289
                                 m = int(len(r_final) * 0.1) + 1
290
                                 r final = r final.tolist()
291
                                 long = r_final[-m:]
292
                                 short = r_final[:m]
293
                                 r_{end} = (np. sum(long) - np. sum(short)) / m
294
                                 Long Short. append (r end)
295
                                 Long. append (np. average (long))
296
                                Short. append (np. average (short))
297
                                 gc. collect()
                       T_value = []
298
                       Mean = []
299
                       p_value = []
300
301
                        sharpratio = []
302
                       Std = []
303
                       TO = [Long_Short, Long, Short]
304
                        for 1 in TO:
305
                                 t_{tst} = nwttest_{1samp}(1, 0, L=1)
                                 mean = np. average(1) * 12- rf. mean(). tolist()[0] * 12 / 100
306
```

```
307
               STD = np. std(1) * np. sqrt(12)
308
                sharp = (mean ) / STD
309
                T_value.append(t_test.statistic)
                p_value.append(t_test.pvalue)
310
               Mean. append (mean)
311
               Std. append (STD)
312
313
                sharpratio.append(sharp)
           print(name, 'long-short', 'long', 'short')
314
315
           print('mean', Mean[0] / 12 + rf.mean().tolist()[0] / 100, Mean[1] / 12 +
       rf.mean().tolist()[0] / 100, Mean[2] / 12 + rf.mean().tolist()[0] / 100)
316
           print('t-statistic', '('+str(round(T_value[0], 4))+')', '('+str(round(T_value[1], 4))+')',
317
       '('+str(round(T value[2],4))+')')
318
319
           A = pd. DataFrame (Long_Short, columns=['long-short'])
320
           B = pd. DataFrame (Long, columns=['long'])
321
           C = pd. DataFrame(Short, columns=['short'])
322
           M = pd.concat([A, B, C], axis=1)
           M. to_csv('...\output\\' + name + '.csv')
323
           ff3 = pd. read csv('..\DataBase\\ff3.csv')
324
           ff5 = pd. read_csv('..\DataBase\\ff5.csv')
325
           alpha3 = []
326
           t3 = []
327
328
           t5 = []
           alpha5 = []
329
           for i in range (3):
330
331
               X1 = ff3.iloc[length:, 1:]
332
               X2 = ff5.iloc[length:, 1:]
               Y = M. iloc[:-2, i]
333
334
               Y. index = X1. index
               Y = Y - rf.RF[:-1] / 100
335
                x1 = sm. add constant(X1)
336
                reg = sm. OLS(Y, x1). fit()
337
                t3. append (reg. tvalues [0])
338
339
               alpha3. append (reg. params [0] * 12)
340
                x2 = sm. add_constant(X2)
                reg = sm. OLS(Y, x2). fit()
341
                t5. append (reg. tvalues [0])
342
343
               alpha5. append (reg. params [0] * 12)
           print('alpha-FF3', alpha3[0] / 12, alpha3[1] / 12, alpha3[2] / 12)
344
           print('t-statistic', '(' + str(round(t3[0], 4)) + ')', '(' + str(round(t3[1], 4)) + ')',
345
                  '(' + str(round(t3[2], 4)) + ')')
346
347
           print('alpha-FF5', alpha5[0] / 12, alpha5[1] / 12, alpha5[2] / 12)
           print('t-statistic', '(' + str(round(t5[0], 4)) + ')', '(' + str(round(t5[1], 4)) + ')',
348
                  '(' + str(round(t5[2], 4)) + ')')
349
           print('sharpe', sharpratio[0], sharpratio[1], sharpratio[2])
350
```

```
351
352
       # 针对所有机器学习模型集成所设置的主函数
353
       def comboutput (length, clf, name, rf, timeseries2, index):
354
           Long_Short = []
           Long = []
355
           Short = []
356
357
           for i in range(len(timeseries2) - length):
               print(i)
358
               # LSTM 数据
359
360
               FINALm = pd. concat(timeseries2[i:(i + length + 1)], axis=0)
361
               FINALm[~FINALm['ret']. isin(['null'])] =
       FINALm[~FINALm['ret'].isin(['null'])].fillna(0)
362
               FINAL_X = FINALm.iloc[:, :-2]
363
364
               FINAL x = FINAL X
365
               FINAL_x[~FINALm['ret'].isin(['null'])] =
366
       scale(FINAL X[~FINALm['ret'].isin(['null'])])
               FINAL_x[FINALm['ret'].isin(['null'])] = 0
367
               FINALm[FINALm['ret'].isin(['null'])] = 0
368
               Nx_{train} = [FINAL_x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(length)]
369
               Ny_{train} = [FINALm.iloc[j * 3571:(j + 1) * 3571, -1].values for j in range(length)]
370
               Nx_{test} = [FINAL_x.iloc[j * 3571:(j + 1) * 3571, :].values for j in range(1, length)
371
372
       + 1)]
373
               ## 传统数据
374
               dl = timeseries2[i:i + (length + 1)]
375
               p = [dl[j][^{\sim}index[i+j]] for j in range(len(dl))]
376
               TTX = pd. concat(p, axis=0)
377
               TTX = TTX. fillna(0)
378
               TXX = TTX. iloc[:, :-2]
379
               TXx = scale(TXX)
380
               final = pd. concat(p[:length], axis=0)
               Tx_{train} = TXx[:len(final)]
381
               Tx test = TXx[len(final):]
382
               Ty_train = final.iloc[:, -1]
383
384
               test = p[-1]
               Ty_test = test.iloc[:, -1]
385
386
               # 基准-linear
387
               clf = clf
               clf.fit(Tx_train, Ty_train, Nx_train, Ny_train)
388
389
               PREDICTION = clf.predict(Tx_test, Nx_test, index[i+length], length)
390
               r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
391
               r_ture = pd. DataFrame(Ty_test)
               r ture.columns = ['ture']
392
393
               r_ture.index = r_predict.index
               FINAL = pd. concat([r predict, r ture], axis=1)
394
```

```
395
               FINAL_sort = FINAL.sort_values(by='predict', axis=0)
396
               r_final = np. array(FINAL_sort['ture'])
397
                m = int(len(r_final) * 0.1) + 1
398
               r_final = r_final.tolist()
399
                long = r_final[-m:]
                short = r_final[:m]
400
401
               r_{end} = (np. sum(long) - np. sum(short)) / m
402
               Long_Short. append (r_end)
403
               Long. append (np. average (long))
               Short. append (np. average (short))
404
405
                gc. collect()
406
           T value = []
           Mean = []
407
408
           p value = []
409
           sharpratio = []
410
411
           TO = [Long_Short, Long, Short]
           for 1 in TO:
412
413
                t_{t} = nwttest_{samp}(1, 0, L=1)
414
               mean = np. average(1) * 12 - rf.mean().tolist()[0] * <math>12 / 100
415
               STD = np. std(1) * np. sqrt(12)
416
                sharp = (mean) / STD
               T_value.append(t_test.statistic)
417
               p_value.append(t_test.pvalue)
418
419
               Mean. append (mean)
               Std. append (STD)
420
421
                sharpratio.append(sharp)
422
           print(name, 'long-short', 'long', 'short')
           print('mean', Mean[0] / 12, Mean[1] / 12, Mean[2] / 12)
423
           print('t-statistic', '(' + str(round(T_value[0], 4)) + ')', '(' + str(round(T_value[1],
424
       4)) + ')',
425
426
                 '(' + str(round(T value[2], 4)) + ')')
           A = pd. DataFrame (Long_Short, columns=['long-short'])
427
           B = pd. DataFrame(Long, columns=['long'])
428
429
           C = pd. DataFrame(Short, columns=['short'])
           M = pd. concat([A, B, C], axis=1)
430
           M. to csv('..\output\\' + name + '.csv')
431
           ff3 = pd. read_csv('..\DataBase\\ff3.csv')
432
           ff5 = pd. read csv('..\DataBase\\ff5.csv')
433
434
           alpha3 = []
435
           t3 = []
436
           t5 = []
437
           alpha5 = []
           for i in range (3):
438
```

```
439
               X1 = ff3.iloc[length:, 1:]
440
               X2 = ff5.iloc[length:, 1:]
441
               Y = M. iloc[:-2, i]
442
               Y.index = X1.index
               Y = Y - rf.RF[:-1] / 100
443
               x1 = sm.add_constant(X1)
444
445
               reg = sm. OLS(Y, x1). fit()
446
               t3. append (reg. tvalues [0])
447
               alpha3.append(reg.params[0] * 12)
               x2 = sm. add_constant(X2)
448
449
               reg = sm. OLS(Y, x2). fit()
               t5. append (reg. tvalues [0])
450
451
               alpha5.append(reg.params[0] * 12)
           print('alpha-FF3', alpha3[0] / 12, alpha3[1] / 12, alpha3[2] / 12)
452
           print('t-statistic', '(' + str(round(t3[0], 4)) + ')', '(' + str(round(t3[1], 4)) + ')',
453
                 '(' + str(round(t3[2], 4)) + ')')
454
           print('alpha-FF5', alpha5[0] / 12, alpha5[1] / 12, alpha5[2] / 12)
455
           print('t-statistic', '(' + str(round(t5[0], 4)) + ')', '(' + str(round(t5[1], 4)) + ')',
456
                 '(' + str(round(t5[2], 4)) + ')')
457
           print('sharpe', sharpratio[0], sharpratio[1], sharpratio[2])
458
```

#### f) FactorTest.py

```
1
     #!/usr/bin/env python
 2
     # -*- coding: utf-8 -*-
 3
 4
     @description:
 5
     1. 在去掉市值最小的 30%股票后分别根据市值 (size)和收益价格比 (EP) 分组后构建 MKT/SMB/VMG 3 因
 6
     子
 7
     2. 单因子 10 分组 10-1/1-10 多空组合因子调整收益
     3. 单因子 10 分组检验各组因子调整收益
8
9
     4. 各因子与 size 因子独立双变量分组检验结果
     5. 各因子与 BM 因子独立双变量分组检验结果
10
     6.96 项因子 fama macbeth 回归检验结果
11
12
13
14
      #*****1.在去掉市值最小的30%股票后分别根据市值(size)和收益价格比(EP)分组后构建MKT/SMB/VMG
15
     3 因子****#
16
17
     import pandas as pd
18
     import numpy as np
19
     ret=pd. read_csv('...\DataBase\\final_return.csv')
20
     size=pd. read_csv('...\DataBase\\factor\\01size.csv')
21
     EP=pd. read csv('..\DataBase\\factor\\32EP.csv')
22
     faceq=pd.DataFrame(columns=['MKT','SMB','VMG'],index=ret.columns[1:])
23
     facvw=pd.DataFrame(columns=['MKT','SMB','VMG'], index=ret.columns[1:])
24
25
     rf=pd. read_csv('RF. csv')
26
     for i in range (len (ret. columns) -1):
         total=pd.concat([ret.iloc[:, i+1], size.iloc[:, i+1], EP.iloc[:, i+1]], axis=1)
27
         total.columns=['ret', 'size', 'EP']
28
         total=total.dropna()
29
         total = total.sort values(by='size')
30
         final = total.iloc[int(len(total) * 0.3):, :]#去掉 30%/小市值
31
32
         faceq. iloc[i, 0] = final. ret. mean()-rf. RF[i]/100
33
         final['VW'] = final.apply(lambda x: x['size'] * x['ret'], axis=1)
34
         facvw. iloc[i, 0] = final. VW. sum()/final. <math>iloc[:, 1]. sum()-rf. RF[i]/100
35
         #SMB/VMG
36
         SIZE1=final.iloc[:int(len(final)/2),:]#//
37
38
         SIZE1=SIZE1. sort_values(by='EP')
39
         SV=SIZE1.iloc[int(2*len(SIZE1)/3):,:]
         sveq=SV['ret'].mean()
40
41
         svvw=SV. VW. sum()/SV. iloc[:, 1]. sum()
         SM=SIZE1. iloc[int(1*len(SIZE1)/3):int(2*len(SIZE1)/3),:]
42
```

```
43
          smeq=SM['ret'].mean()
44
          smvw=SM. VW. sum()/SM. iloc[:, 1]. sum()
45
          SG=SIZE1.iloc[:int(1*len(SIZE1)/3),:]
          sgeq=SG['ret'].mean()
46
          sgvw=SG. VW. sum()/SG.iloc[:,1].sum()
47
48
49
          SIZE2 = final.iloc[int(len(final) / 2):,:]#大
          SIZE2 = SIZE2. sort_values(by='EP')
50
          BV=SIZE2. iloc[int(2*len(SIZE2)/3):,:]
51
          bveq=BV['ret'].mean()
52
          bvvw=BV. VW. sum()/BV.iloc[:,1].sum()
53
54
          BM=SIZE2. iloc[int(1*len(SIZE2)/3):int(2*len(SIZE2)/3),:]
          bmeq=BM['ret'].mean()
55
          bmvw=BM. VW. sum()/BM. iloc[:, 1]. sum()
56
57
          BG=SIZE2.iloc[:int(1*len(SIZE2)/3),:]
58
          bgeq=BG['ret'].mean()
          bgvw=BG.VW.sum()/BG.iloc[:,1].sum()
59
          faceq.iloc[i, 1] = (sveq +smeq +sgeq )/3-(bveq +bmeq +bgeq )/3
60
          facvw.iloc[i, 1] = (svvw + smvw + sgvw)/3 - (bvvw + bmvw + bgvw)/3
61
          faceq.iloc[i, 2] = (sveq +bveq)/2 - (sgeq+bgeq)/2
62
          facvw.iloc[i, 2] = (svvw + bvvw)/2 - (sgvw + bgvw)/2
63
64
          print(i)
65
      #********************2. 单因子 10 分组 10-1 多空组合因子调整收益*************************
66
67
      import pandas as pd
68
      import numpy as np
69
      import statsmodels.api as sm
70
      from scipy import stats
71
      from NWttest import nwttest_1samp
      factoreq=pd. read_csv('...\DataBase\\factorEW.csv')
72
      factorvw=pd. read_csv('...\DataBase\\factorVW.csv')
73
74
      resultew=pd. DataFrame(columns=factoreq.columns[1:], index=['ret', 't', 'exret', 't', 'capmret', 't
      ','ff3ret','t','ff30ret','t'])
75
      resultvw=pd. DataFrame(columns=factoreq.columns[1:], index=['ret', 't', 'exret', 't', 'capmret', 't
76
77
      ','ff3ret','t','ff30ret','t'])
      rf=pd. read_csv('RF. csv')
78
79
      for i in range (len (factoreq. columns) -1):
80
          if factoreq. iloc[:, i+1]. mean()>0:
              faceq=factoreq.iloc[:,i+1]
81
82
          else:
83
              faceq = -factoreq.iloc[:, i + 1]
          if factorvw.iloc[:, i+1].mean()>0:
84
              facvw=factorvw.iloc[:, i+1]
85
86
          else:
```

```
87
               facvw = -factorvw.iloc[:, i + 1]
88
           ####return
89
           reteq0 = faceq.dropna()
90
           resultew.iloc[0, i] = reteq0.mean()
91
           ttest = nwttest_1samp(reteq0, 0)
           resultew.iloc[1, i] = ttest.statistic
92
           retvw0 = facvw.dropna()
93
94
           resultvw.iloc[0, i] = retvw0.mean()
95
           ttest = nwttest_1samp(retvw0, 0)
96
           resultvw.iloc[1, i] = ttest.statistic
97
98
           ##excess return
99
           exeq=faceq-rf.RF/\frac{100}{100}
100
           exvw=facvw-rf.RF/100
101
           exeq0=exeq. dropna()
102
           resultew.iloc[2, i] = exeq0.mean()
103
           #ttest=stats. ttest_1samp(exeq0, 0)
104
           ttest=nwttest 1samp(exeq0, 0)
105
           resultew.iloc[3, i]=ttest.statistic
           exvw0 = exvw.dropna()
106
           resultvw.iloc[2, i] = exvw0.mean()
107
108
           # ttest=stats. ttest_1samp(exvw0, 0)
109
           ttest = nwttest 1samp(exvw0, 0)
           resultvw.iloc[3, i] = ttest.statistic
110
           ##CAPM
111
           ff3 = pd. read_csv('ff3. csv')
112
113
           capm=ff3.iloc[:,1]
114
           x1 = sm. add constant(capm)
           X1=pd. concat([exeq[:-2], x1], axis=1)
115
116
           X1 = X1. dropna()
           X2 = pd. concat([exvw[:-2], x1], axis=1)
117
           X2 = X2. dropna()
118
           regeq = sm. OLS(X1.iloc[:, 0], X1.iloc[:, 1:]).fit()
119
           regvw = sm. OLS(X2.iloc[:, 0], X2.iloc[:, 1:]).fit()
120
121
           resultew.iloc[4, i]=regeq.params[0]
           resultew.iloc[5, i]=regeq.tvalues[0]
122
           resultvw.iloc[4, i] = regvw.params[0]
123
           resultvw.iloc[5, i] = regvw.tvalues[0]
124
125
126
           x1 = sm. add_constant(ff3.iloc[:, 1:])
127
           X1 = pd.concat([exeq[:-2], x1], axis=1)
128
           X1 = X1. dropna()
           X2 = pd.concat([exvw[:-2], x1], axis=1)
129
           X2 = X2. dropna()
130
```

```
131
          regeq = sm. OLS(X1.iloc[:, 0], X1.iloc[:, 1:]).fit()
          regvw = sm. OLS(X2.iloc[:, 0], X2.iloc[:, 1:]).fit()
132
133
          resultew. iloc[6, i] = regeq. params[0]
134
          resultew.iloc[7, i] = regeq.tvalues[0]
          resultvw.iloc[6, i] = regvw.params[0]
135
          resultvw.iloc[7, i] = regvw.tvalues[0]
136
137
          ff30=pd. read_csv('..\DataBase\\ff30.csv')
138
139
          x1 = sm. add constant(ff30. iloc[:, 1:])
          X1 = pd. concat([exeq, x1], axis=1)
140
141
          X1 = X1. dropna()
          X2 = pd. concat([exvw, x1], axis=1)
142
143
          X2 = X2. dropna()
144
          regeq = sm. OLS(np. asarray(X1. iloc[:, 0]), np. asarray(X1. iloc[:, 1:])). fit()
145
          regvw = sm.OLS(np.asarray(X2.iloc[:, 0]), np.asarray(X2.iloc[:, 1:])).fit()
146
          resultew.iloc[8, i] = regeq.params[0]
          resultew.iloc[9, i] = regeq.tvalues[0]
147
148
          resultvw.iloc[8, i] = regvw.params[0]
          resultvw.iloc[9, i] = regvw.tvalues[0]
149
150
          print(i)
151
      152
153
      import pandas as pd
154
      import numpy as np
155
      import glob, os
156
      import statsmodels.api as sm
157
      from scipy import stats
158
      from NWttest import nwttest 1samp
      ret=pd. read_csv('final_return.csv')
159
160
      size=pd. read_csv('...\DataBase\\factor\\01size.csv')
      path = r'..\DataBase\\factor'
161
162
      file = glob. glob (os. path. join (path, "*. csv"))
163
      rf=pd. read_csv('RF. csv')
164
      k = [] # 每个因子一个表
165
      for i in range (96):
166
          k.append(pd.read_csv(file[i]))
167
      factor = []#因子名称
168
      for i in range(len(file)):
169
          factor.append(file[i][29:-4])
170
      for i in range (10):
171
          resultew=pd.DataFrame
      (columns=factor, index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't', 'ff30ret', 't'])
172
          resultvw = pd. DataFrame(columns=factor, index=['ret', 't', 'exret', 't', 'capmret', 't',
173
      'ff3ret', 't', 'ff30ret', 't'])
174
```

```
175
           for j in range (96):
                totalew =[]
176
177
                totalvw =[]
178
                for m in range(len(ret.columns)-1):
179
                    final=pd.concat([size.iloc[:, m+1], k[j].iloc[:, m+1], ret.iloc[:, m+1]], axis=1)
                    final.columns=['size', 'factor', 'ret']
180
                    final=final.dropna()
181
                    if len(final) == 0:
182
183
                        totalew. append (np. nan)
184
                        total vw. append (np. nan)
185
                    else:
186
                        final = final. sort values (by='factor')
                        final['VW'] = final.apply(lambda x: x['size'] * x['ret'], axis=1)
187
188
                        total = final.iloc[int(len(final) / 10) * i:int(len(final) / 10) * (i +
189
       1), :]
190
                        totalew.append(total['ret'].mean())
                        totalvw.append(total['VW'].sum() / total['size'].sum())
191
                faceq = pd. Series(totalew)
192
                facvw = pd. Series (totalvw)
193
194
195
                ####return
196
                reteq0 = faceq.dropna()
                resultew.iloc[0, j] = reteq0.mean()
197
                ttest = nwttest_1samp(reteq0, 0)
198
                resultew.iloc[1, j] = ttest.statistic
199
                retvw0 = facvw.dropna()
200
201
                resultvw.iloc[0, j] = retvw0.mean()
202
                ttest = nwttest 1samp(retvw0, 0)
203
                resultvw.iloc[1, j] = ttest.statistic
204
205
                ##excess return
206
                exeq = faceq - rf.RF / 100
207
                exvw = facvw - rf.RF / 100
                exeq0 = exeq. dropna()
208
209
                resultew. iloc[2, j] = exeq0. mean()
                # ttest=stats.ttest_1samp(exeq0,0)
210
211
                ttest = nwttest 1samp(exeq0, 0)
                resultew.iloc[3, j] = ttest.statistic
212
                exvw0 = exvw.dropna()
213
214
                resultvw.iloc[2, j] = exvw0.mean()
215
                # ttest=stats.ttest_1samp(exvw0,0)
216
                ttest = nwttest_1samp(exvw0, 0)
                resultvw.iloc[3, j] = ttest.statistic
217
                ##CAPM-alpha/FF3-alpha/去掉市值最小 30%股票后 adj-FF3-alpha
218
```

```
219
               ff3 = pd. read_csv('ff3. csv')
               capm = ff3.iloc[:, 1]
220
221
               x1 = sm. add_constant(capm)
222
               X1 = pd. concat([exeq[:-2], x1], axis=1)
               X1 = X1. dropna()
223
               X2 = pd. concat([exvw[:-2], x1], axis=1)
224
225
               X2 = X2. dropna()
               regeq = sm.OLS(X1.iloc[:, 0], X1.iloc[:, 1:]).fit()
226
               regvw = sm. OLS(X2.iloc[:, 0], X2.iloc[:, 1:]).fit()
227
               resultew. iloc[4, j] = regeq. params[0]
228
229
               resultew.iloc[5, j] = regeq.tvalues[0]
               resultvw. iloc[4, j] = regvw. params[0]
230
               resultvw.iloc[5, j] = regvw.tvalues[0]
231
232
233
               x1 = sm. add_constant(ff3. iloc[:, 1:])
234
               X1 = pd. concat([exeq[:-2], x1], axis=1)
               X1 = X1. dropna()
235
               X2 = pd. concat([exvw[:-2], x1], axis=1)
236
               X2 = X2. dropna()
237
               regeq = sm.OLS(X1.iloc[:, 0], X1.iloc[:, 1:]).fit()
238
239
               regvw = sm. OLS(X2.iloc[:, 0], X2.iloc[:, 1:]).fit()
240
               resultew.iloc[6, j] = regeq.params[0]
               resultew.iloc[7, j] = regeq.tvalues[0]
241
               resultvw.iloc[6, j] = regvw.params[0]
242
               resultvw.iloc[7, j] = regvw.tvalues[0]
243
244
245
               ff30 = pd. read_csv('...\DataBase\\ff30.csv')
246
               x1 = sm. add constant(ff30. iloc[:, 1:])
               X1 = pd. concat([exeq, x1], axis=1)
247
248
               X1 = X1. dropna()
               X2 = pd. concat([exvw, x1], axis=1)
249
               X2 = X2. dropna()
250
251
               regeq = sm. OLS(X1.iloc[:, 0], X1.iloc[:, 1:]).fit()
               regvw = sm. OLS(X2.iloc[:, 0], X2.iloc[:, 1:]).fit()
252
               resultew.iloc[8, j] = regeq.params[0]
253
               resultew.iloc[9, j] = regeq.tvalues[0]
254
               resultvw.iloc[8, j] = regvw.params[0]
255
               resultvw.iloc[9, j] = regvw.tvalues[0]
256
           resultew. to csv('RESLew'+str(i)+'.csv')
257
258
           resultvw.to_csv('RESLvw' + str(i) + '.csv')
259
       260
261
       ***********************************#
262
       import pandas as pd
```

```
263
       import numpy as np
264
       import glob, os
265
       import statsmodels.api as sm
266
       from NWttest import nwttest_lsamp
267
       ret=pd. read_csv('...\DataBase\\final_return. csv')
       size=pd.read_csv('..\DataBase\\factor\\01size.csv')
268
269
       size=size.iloc[:,:-1]
270
       size.columns=ret.columns
       path = r'..\DataBase\\factor'
271
       file = glob. glob (os. path. join (path, "*. csv"))
272
273
       rf=pd. read csv('..\DataBase\\RF.csv')
274
       k = [] # 每个因子一个表
275
       for i in range (96):
276
           k.append(pd.read_csv(file[i]))
277
       factor = []#因子名称
278
       for i in range(len(file)):
279
           factor.append(file[i][29:-4])
280
281
       def gendata (x1, x2, ret, size, m=5, n=5):
           x1 = x1[^np. isnan(ret)]
282
           x1=x1[^np.isnan(size)]
283
284
           x2 = x2[^np. isnan(ret)]
           x2 = x2[^np. isnan(size)]
285
286
           x1sort = x1.apply(lambda x: np.argsort(x), axis=0)
287
           x2sort = x2.apply(lambda x: np.argsort(x), axis=0)
           datacolumn = []
288
289
           for a in range(m):
290
               for b in range(n):
291
                   datacolumn.append(str(a + 1) + X' + str(b + 1))
292
           dfeq=pd. DataFrame (columns=datacolumn, index=[i for i in range(len(ret.columns)-1)])
           dfvw = pd. DataFrame(columns=datacolumn, index=[i for i in range(len(ret.columns) - 1)])
293
294
           for i in range(len(ret.columns)-1):
295
               truelistx1 = x1sort.iloc[:, i+1 ]
               truelistx2 = x2sort.iloc[:, i+1]
296
297
               truelistx1 = truelistx1[~(truelistx1 == -1)].sort_values()
               truelistx1=truelistx1.index
298
               truelistx2 = truelistx2[^(truelistx2 == -1)]. sort values()
299
               truelistx2 = truelistx2.index
300
301
               if len(truelistx1) <5 or len(truelistx2) <5:</pre>
302
                   dfeq.iloc[i,:] =np.nan
303
                   dfvw.iloc[i,:] =np.nan
304
               else:
305
                   x1lines = np.array_split(np.array(truelistx1), m)
                    x2lines = np. array split(np. array(truelistx2), n)
306
```

```
307
                   for a in range(m):
308
                       xluse = xllines[a]
309
                       for b in range(n):
310
                            x2use = x21ines[b]
                            tempindex = np.intersect1d(x1use, x2use)
311
                            if len(tempindex)==0:
312
                                dfeq.iloc[i, a * m + b]=np.nan
313
                                dfvw.iloc[i, a * m + b] = np.nan
314
315
                            else:
                                dfeq.iloc[i, a * m + b] = ret.iloc[tempindex, i + 1].mean()
316
317
                                total = pd.concat([ret.iloc[tempindex, i + 1], size.iloc[tempindex,
318
       i + 1]], axis=1)
                                total.columns = ['ret', 'size']
319
320
                                total['vw'] = total.apply(lambda x: x['size'] * x['ret'], axis=1)
                                dfvw.iloc[i, a * m + b] = total['vw'].sum() / total['size'].sum()
321
322
           return dfeg, dfvw
323
324
325
       RETEQ, RETVQ=[], []
       for i in range(len(k)):
326
           k[i] = k[i].iloc[:, :264]
327
328
           k[i].columns = ret.columns
329
           reteq, retvw = gendata(size, k[i], ret, size)
330
           RETEQ. append (reteq)
           RETVQ. append (retvw)
331
332
333
       #删除 SIZE 因子本身(不与自身分组)
334
       del RETEQ[0]
335
       del RETVQ[0]
336
       del factor[0]
337
338
       for 1 in range (5):
           longew=pd.concat([i.iloc[:, 1*5+4] for i in RETEQ], axis=1)
339
           shortew = pd.concat([i.iloc[:, 1 * 5 ] for i in RETEQ], axis=1)
340
341
           longvw = pd. concat([i.iloc[:, 1 * 5 + 4] for i in RETVQ], axis=1)
           shortvw = pd.concat([i.iloc[:, 1 * 5] for i in RETVQ], axis=1)
342
343
           longew.columns = factor
344
           longvw.columns = factor
345
           shortew.columns = factor
346
           shortvw.columns = factor
347
           1sew=longew-shortew
348
           1svw=longvw-shortvw
349
           resultew = pd. DataFrame (columns=factor,
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
350
```

```
351
       'ff30ret', 't'])
           resultvw = pd. DataFrame (columns=factor,
352
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
353
354
       'ff30ret', 't'])
           for j in range(len(factor)):
355
               faceq, facvw = lsew.iloc[:, j], lsvw.iloc[:, j]
356
357
               reteq0 = faceq.dropna()
               resultew.iloc[0, j] = reteq0.mean()
358
359
               ttest = nwttest_1samp(reteq0, 0)
               resultew.iloc[1, j] = ttest.statistic
360
               retvw0 = facvw.dropna()
361
               resultvw.iloc[0, j] = retvw0.mean()
362
               ttest = nwttest_1samp(retvw0, 0)
363
               resultvw.iloc[1, j] = ttest.statistic
364
365
               ##excess return
366
               exeq = faceq - rf.RF / 100
               exvw = facvw - rf.RF / 100
367
               exeq0 = exeq. dropna()
368
               resultew. iloc[2, j] = exeq0. mean()
369
370
               # ttest=stats.ttest_1samp(exeq0,0)
371
               ttest = nwttest 1samp(exeq0, 0)
372
               resultew.iloc[3, j] = ttest.statistic
               exvw0 = exvw.dropna()
373
               resultvw.iloc[2, j] = exvw0.mean()
374
375
               # ttest=stats. ttest 1samp(exvw0, 0)
376
               ttest = nwttest_1samp(exvw0, 0)
377
               resultvw.iloc[3, j] = ttest.statistic
378
               ##CAPM
               ff3 = pd. read_csv('...\DataBase\\ff3.csv')
379
380
               capm = ff3.iloc[:, 1]
               x1 = sm. add_constant(capm)
381
               X1 = pd. concat([exeq[:-2], x1], axis=1)
382
383
               X1 = X1. dropna()
384
               X2 = pd. concat([exvw[:-2], x1], axis=1)
               X2 = X2. dropna()
385
               regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
386
               regvw = sm.OLS(np.asarray(X2.iloc[:, 0]), np.asarray(X2.iloc[:, 1:])).fit()
387
               resultew.iloc[4, j] = regeq.params[0]
388
               resultew. iloc[5, j] = regeq. tvalues[0]
389
               resultvw.iloc[4, j] = regvw.params[0]
390
391
               resultvw.iloc[5, j] = regvw.tvalues[0]
               x1 = sm. add_constant(ff3. iloc[:, 1:])
392
393
               X1 = pd. concat([exeq[:-2], x1], axis=1)
               X1 = X1. dropna()
394
```

```
395
                X2 = pd. concat([exvw[:-2], x1], axis=1)
               X2 = X2. dropna()
396
397
                regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])). fit()
                regvw = sm. OLS(np. asarray(X2. iloc[:, 0]), np. asarray(X2. iloc[:, 1:])). fit()
398
                resultew. iloc[6, j] = regeq. params[0]
399
                resultew.iloc[7, j] = regeq.tvalues[0]
400
                resultvw.iloc[6, j] = regvw.params[0]
401
402
                resultvw.iloc[7, j] = regvw.tvalues[0]
403
                ff30 = pd. read_csv('..\DataBase\\ff30.csv')
404
                x1 = sm. add_constant(ff30. iloc[:, 1:])
405
               X1 = pd. concat([exeq, x1], axis=1)
406
               X1 = X1. dropna()
407
               X2 = pd.concat([exvw, x1], axis=1)
408
               X2 = X2. dropna()
409
                regeq = sm.OLS(np.asarray(X1.iloc[:, 0]), np.asarray(X1.iloc[:, 1:])).fit()
410
                regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])).fit()
                resultew. iloc[8, j] = regeq. params[0]
411
               resultew.iloc[9, j] = regeq.tvalues[0]
412
                resultvw.iloc[8, j] = regvw.params[0]
413
                resultvw.iloc[9, j] = regvw.tvalues[0]
414
           resultew.to_csv('longshort55SIZEEW' + str(l) + '.csv')
415
416
           resultvw.to_csv('longshort55SIZEVW' + str(l) + '.csv')
417
418
       for u in range (25):
419
           DATAEQ=pd. concat([i.iloc[:,u] for i in RETEQ], axis=1)
           DATAVQ = pd. concat([i.iloc[:, u] for i in RETVQ], axis=1)
420
421
           DATAEQ. columns=factor
422
           DATAVQ. columns = factor
           resultew = pd. DataFrame (columns=factor,
423
                                         index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret',
424
       't', 'ff30ret', 't'])
425
426
           resultvw = pd. DataFrame(columns=factor,
                                         index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret',
427
       't', 'ff30ret', 't'])
428
429
           for j in range(len(factor)):
                faceq, facvw=DATAEQ.iloc[:,j], DATAVQ.iloc[:,j]
430
431
                reteq0 = faceq.dropna()
                resultew.iloc[0, j] = reteq0.mean()
432
                ttest = nwttest 1samp(reteq0, 0)
433
434
                resultew.iloc[1, j] = ttest.statistic
435
                retvw0 = facvw.dropna()
                resultvw.iloc[0, j] = retvw0.mean()
436
437
                ttest = nwttest_1samp(retvw0, 0)
                resultvw.iloc[1, j] = ttest.statistic
438
```

```
439
440
                ##excess return
441
                exeq = faceq - rf.RF / 100
442
                exvw = facvw - rf.RF / 100
                exeq0 = exeq. dropna()
443
                resultew. iloc[2, j] = exeq0. mean()
444
445
                # ttest=stats. ttest 1samp(exeq0, 0)
                ttest = nwttest_1samp(exeq0, 0)
446
447
                resultew.iloc[3, j] = ttest.statistic
                exvw0 = exvw.dropna()
448
449
                resultvw.iloc[2, j] = exvw0.mean()
450
                # ttest=stats. ttest 1samp(exvw0, 0)
451
                ttest = nwttest_1samp(exvw0, 0)
452
                resultvw.iloc[3, j] = ttest.statistic
453
                ##CAPM
454
                ff3 = pd. read_csv('...\DataBase\\ff3.csv')
                capm = ff3.iloc[:, 1]
455
                x1 = sm. add constant(capm)
456
                X1 = pd. concat([exeq[:-2], x1], axis=1)
457
458
                X1 = X1. dropna()
459
                X2 = pd. concat([exvw[:-2], x1], axis=1)
460
                X2 = X2. dropna()
                regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
461
                regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])).fit()
462
                resultew.iloc[4, j] = regeq.params[0]
463
                resultew. iloc[5, j] = regeq. tvalues[0]
464
465
                resultvw.iloc[4, j] = regvw.params[0]
466
                resultvw.iloc[5, j] = regvw.tvalues[0]
467
468
                x1 = sm. add_constant(ff3. iloc[:, 1:])
                X1 = pd. concat([exeq[:-2], x1], axis=1)
469
470
                X1 = X1. dropna()
                X2 = pd. concat([exvw[:-2], x1], axis=1)
471
472
                X2 = X2. dropna()
                regeq = sm.OLS(np.asarray(X1.iloc[:, 0]), np.asarray(X1.iloc[:, 1:])).fit()
473
                regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])).fit()
474
                resultew. iloc[6, j] = regeq. params[0]
475
                resultew.iloc[7, j] = regeq.tvalues[0]
476
                resultvw. iloc[6, j] = regvw. params[0]
477
                resultvw.iloc[7, j] = regvw.tvalues[0]
478
479
                ff30 = pd. read_csv('C:\\Users\\15083\Desktop\\ff30.csv')
480
481
                x1 = sm. add_constant(ff30. iloc[:, 1:])
                X1 = pd. concat([exeq, x1], axis=1)
482
```

```
483
               X1 = X1. dropna()
484
               X2 = pd. concat([exvw, x1], axis=1)
485
               X2 = X2. dropna()
               regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
486
487
               regvw = sm. OLS(np. asarray(X2. iloc[:, 0]), np. asarray(X2. iloc[:, 1:])). fit()
               resultew.iloc[8, j] = regeq.params[0]
488
489
               resultew.iloc[9, j] = regeq.tvalues[0]
490
               resultvw.iloc[8, j] = regvw.params[0]
               resultvw.iloc[9, j] = regvw.tvalues[0]
491
           resultew. to_csv('EW55size'+reteq.columns[u]+'.csv')
492
493
           resultvw.to_csv('VW55size' + reteq.columns[u] + '.csv')
494
495
496
       #************************5. 各因子与 BM 因子独立双变量分组检验结果
497
       498
       import pandas as pd
499
       import numpy as np
500
       import glob, os
501
       import statsmodels.api as sm
502
       from NWttest import nwttest_1samp
503
504
       ret = pd. read_csv('...\DataBase\\final_return. csv')
505
       size = pd. read csv('..\DataBase\\factor\\01size.csv')
       BM = pd.read_csv('..\DataBase\\factor\\28BM.csv')
506
       size = size.iloc[:, :-1]
507
       size.columns = ret.columns
508
509
       path = r'..\DataBase\\factor'
510
       file = glob. glob (os. path. join (path, "*. csv"))
       rf = pd. read_csv('...\DataBase\\RF.csv')
511
       k = [] # 每个因子一个表
512
       for i in range (96):
513
514
           k.append(pd.read csv(file[i]))
515
       factor = [] # 因子名称
516
       for i in range(len(file)):
517
           factor.append(file[i][29:-4])
518
519
       BM = BM.iloc[:, :-1]
       BM. columns = ret. columns
520
521
522
523
       def gendata(x1, x2, ret, size, m=5, n=5):
           x1 = x1[^np. isnan(ret)]
524
       x1 = x1[^np.isnan(size)]
525
           x2 = x2[^np. isnan(ret)]
526
```

```
527
           x2 = x2[^{\sim}np. isnan(size)]
           x1sort = x1.apply(lambda x: np.argsort(x), axis=0)
528
529
           x2sort = x2.apply(lambda x: np.argsort(x), axis=0)
           datacolumn = []
530
531
532
           for a in range(m):
533
               for b in range(n):
                   datacolumn.append(str(a + 1) + X' + str(b + 1))
534
535
           dfeq = pd. DataFrame(columns=datacolumn, index=[i for i in range(len(ret.columns) - 1)])
536
           dfvw = pd. DataFrame(columns=datacolumn, index=[i for i in range(len(ret.columns) - 1)])
           for i in range(len(ret.columns) - 1):
537
               truelistx1 = x1sort.iloc[:, i + 1]
538
               truelistx2 = x2sort.iloc[:, i + 1]
539
               truelistx1 = truelistx1[~(truelistx1 == -1)].sort_values()
540
541
               truelistx1 = truelistx1.index
542
               truelistx2 = truelistx2[^(truelistx2 == -1)]. sort values()
               truelistx2 = truelistx2.index
543
               if len(truelistx1) < 5 or len(truelistx2) < 5:</pre>
544
                   dfeq.iloc[i, :] = np.nan
545
                   dfvw.iloc[i, :] = np.nan
546
547
               else:
548
                   xllines = np. array_split(np. array(truelistx1), m)
                   x2lines = np.array_split(np.array(truelistx2), n)
549
550
                    for a in range(m):
                        xluse = xllines[a]
551
552
                        lsew=[]
                        1svw = []
553
554
                        for b in range(n):
555
                            x2use = x2lines[b]
556
                            tempindex = np.intersect1d(x1use, x2use)
                            if len(tempindex) == 0:
557
                                dfeq.iloc[i, a * m + b] = np.nan
558
559
                                dfvw.iloc[i, a * m + b] = np.nan
560
                                1 sew. append (0)
                                1svw.append(0)
561
562
                            else:
                                dfeq.iloc[i, a * m + b] = ret.iloc[tempindex, i + 1].mean()
563
                                total = pd.concat([ret.iloc[tempindex, i + 1], size.iloc[tempindex,
564
       i + 1]], axis=1)
565
566
                                total.columns = ['ret', 'size']
567
                                total['vw'] = total.apply(lambda x: x['size'] * x['ret'], axis=1)
                                dfvw.iloc[i, a * m + b] = total['vw'].sum() / total['size'].sum()
568
                                lsew.append(ret.iloc[tempindex, i + 1].mean())
569
                                lsvw.append(total['vw'].sum() / total['size'].sum())
570
```

```
571
           return dfeq, dfvw
572
573
574
       RETEQ, RETVQ = [], []
575
       for i in range(len(k)):
           k[i] = k[i].iloc[:, :264]
576
           k[i].columns = ret.columns
577
578
           reteq, retvw = gendata(BM, k[i], ret, size)
579
           RETEQ. append (reteq)
           RETVQ. append (retvw)
580
581
582
       #删除 BM 因子本身 (不与自身分组)
583
       del RETEQ[27]
584
       del RETVQ[27]
585
       del factor[27]
586
       for 1 in range (5):
587
           longew=pd.\,concat([i.iloc[:,\ 1*5+4]\ \textbf{for}\ i\ \textbf{in}\ RETEQ],\ axis=1)
588
           shortew = pd. concat([i.iloc[:, 1 * 5] for i in RETEQ], axis=1)
589
           longvw = pd.concat([i.iloc[:, 1 * 5 + 4] for i in RETVQ], axis=1)
590
           shortvw = pd. concat([i.iloc[:, 1 * 5] for i in RETVQ], axis=1)
591
592
           longew.columns = factor
593
           longvw.columns = factor
594
           shortew.columns = factor
595
           shortvw.columns = factor
596
           1sew=longew-shortew
597
           1svw=longvw-shortvw
598
           resultew = pd. DataFrame (columns=factor,
599
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
       'ff30ret', 't'])
600
           resultvw = pd. DataFrame(columns=factor,
601
602
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
       'ff30ret', 't'])
603
604
           for j in range(len(factor)):
605
               faceq, facvw = lsew.iloc[:, j], lsvw.iloc[:, j]
               reteq0 = faceq.dropna()
606
               resultew.iloc[0, j] = reteq0.mean()
607
               ttest = nwttest_1samp(reteq0, 0)
608
609
               resultew.iloc[1, j] = ttest.statistic
610
               retvw0 = facvw.dropna()
611
               resultvw.iloc[0, j] = retvw0.mean()
               ttest = nwttest_1samp(retvw0, 0)
612
613
               resultvw.iloc[1, j] = ttest.statistic
614
               ##excess return
```

```
615
                exeq = faceq - rf.RF / 100
                exvw = facvw - rf.RF / 100
616
617
                exeq0 = exeq. dropna()
618
                resultew.iloc[2, j] = exeq0.mean()
                # ttest=stats. ttest_1samp(exeq0, 0)
619
                ttest = nwttest_1samp(exeq0, 0)
620
621
                resultew.iloc[3, j] = ttest.statistic
                exvw0 = exvw.dropna()
622
623
                resultvw.iloc[2, j] = exvw0.mean()
624
                # ttest=stats. ttest_1samp(exvw0, 0)
625
                ttest = nwttest 1samp(exvw0, 0)
626
                resultvw.iloc[3, j] = ttest.statistic
                ##CAPM
627
                ff3 = pd. read_csv('..\DataBase\\ff3.csv')
628
629
                capm = ff3.iloc[:, 1]
630
                x1 = sm. add constant (capm)
                X1 = pd. concat([exeq[:-2], x1], axis=1)
631
                X1 = X1. dropna()
632
                X2 = pd. concat([exvw[:-2], x1], axis=1)
633
634
                X2 = X2. dropna()
635
                regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])). fit()
636
                regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])). fit()
                resultew. iloc[4, j] = regeq. params[0]
637
                resultew. iloc[5, j] = regeq. tvalues[0]
638
                resultvw.iloc[4, j] = regvw.params[0]
639
                resultvw.iloc[5, j] = regvw.tvalues[0]
640
641
                x1 = sm. add_constant(ff3. iloc[:, 1:])
642
                X1 = pd. concat([exeq[:-2], x1], axis=1)
643
                X1 = X1. dropna()
644
                X2 = pd. concat([exvw[:-2], x1], axis=1)
645
                X2 = X2. dropna()
                regeq = sm. OLS(np. asarray(X1. iloc[:, 0]), np. asarray(X1. iloc[:, 1:])). fit()
646
647
                regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])). fit()
                resultew.iloc[6, j] = regeq.params[0]
648
                resultew.iloc[7, j] = regeq.tvalues[0]
649
                resultvw.iloc[6, j] = regvw.params[0]
650
                resultvw.iloc[7, j] = regvw.tvalues[0]
651
                ff30 = pd. read_csv('..\DataBase\\ff30.csv')
652
                x1 = sm. add constant(ff30. iloc[:, 1:])
653
654
                X1 = pd. concat([exeq, x1], axis=1)
655
                X1 = X1. dropna()
                X2 = pd. concat([exvw, x1], axis=1)
656
                X2 = X2. dropna()
657
                regeq = sm. OLS(np. asarray(X1. iloc[:, 0]), np. asarray(X1. iloc[:, 1:])). fit()
658
```

```
659
               regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])). fit()
               resultew.iloc[8, j] = regeq.params[0]
660
661
               resultew.iloc[9, j] = regeq.tvalues[0]
662
               resultvw.iloc[8, j] = regvw.params[0]
               resultvw.iloc[9, j] = regvw.tvalues[0]
663
           resultew.to_csv('longshort55BMEW' + str(l) + '.csv')
664
           resultvw.to_csv('longshort55BMVW' + str(l) + '.csv')
665
666
667
       for u in range (25):
           DATAEQ = pd. concat([i.iloc[:, u] for i in RETEQ], axis=1)
668
           DATAVQ = pd. concat([i.iloc[:, u] for i in RETVQ], axis=1)
669
670
           DATAEQ. columns = factor
671
           DATAVQ.columns = factor
672
           resultew = pd. DataFrame (columns=factor,
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
673
674
       'ff30ret', 't'])
           resultvw = pd. DataFrame(columns=factor,
675
                                    index=['ret', 't', 'exret', 't', 'capmret', 't', 'ff3ret', 't',
676
       'ff30ret', 't'])
677
678
           for j in range(len(factor)):
               faceq, facvw = DATAEQ.iloc[:, j], DATAVQ.iloc[:, j]
679
680
               reteq0 = faceq.dropna()
               resultew.iloc[0, j] = reteq0.mean()
681
               ttest = nwttest_1samp(reteq0, 0)
682
               resultew.iloc[1, j] = ttest.statistic
683
               retvw0 = facvw.dropna()
684
685
               resultvw.iloc[0, j] = retvw0.mean()
686
               ttest = nwttest 1samp(retvw0, 0)
               resultvw.iloc[1, j] = ttest.statistic
687
688
               ##excess return
               exeq = faceq - rf.RF / 100
689
690
               exvw = facvw - rf.RF / 100
691
               exeq0 = exeq. dropna()
692
               resultew.iloc[2, j] = exeq0.mean()
               # ttest=stats. ttest_1samp(exeq0, 0)
693
694
               ttest = nwttest_1samp(exeq0, 0)
695
               resultew.iloc[3, j] = ttest.statistic
               exvw0 = exvw.dropna()
696
               resultvw.iloc[2, j] = exvw0.mean()
697
698
               # ttest=stats. ttest_1samp(exvw0, 0)
699
               ttest = nwttest_1samp(exvw0, 0)
               resultvw.iloc[3, j] = ttest.statistic
700
701
               ##CAPM
               ff3 = pd. read csv('..\DataBase\\ff3.csv')
702
```

```
703
               capm = ff3.iloc[:, 1]
704
               x1 = sm. add_constant(capm)
705
               X1 = pd. concat([exeq[:-2], x1], axis=1)
706
               X1 = X1. dropna()
707
               X2 = pd. concat([exvw[:-2], x1], axis=1)
               X2 = X2. dropna()
708
               regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
709
               regvw = sm. OLS(np. asarray(X2.iloc[:, 0]), np. asarray(X2.iloc[:, 1:])).fit()
710
711
               resultew. iloc[4, j] = regeq. params[0]
               resultew. iloc[5, j] = regeq. tvalues[0]
712
               resultvw.iloc[4, j] = regvw.params[0]
713
               resultvw.iloc[5, j] = regvw.tvalues[0]
714
               x1 = sm. add_constant(ff3. iloc[:, 1:])
715
               X1 = pd. concat([exeq[:-2], x1], axis=1)
716
717
               X1 = X1. dropna()
718
               X2 = pd. concat([exvw[:-2], x1], axis=1)
               X2 = X2. dropna()
719
               regeq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
720
               regvw = sm. OLS(np. asarray(X2. iloc[:, 0]), np. asarray(X2. iloc[:, 1:])). fit()
721
               resultew.iloc[6, j] = regeq.params[0]
722
               resultew.iloc[7, j] = regeq.tvalues[0]
723
724
               resultvw.iloc[6, j] = regvw.params[0]
               resultvw.iloc[7, j] = regvw.tvalues[0]
725
               ff30 = pd. read_csv('...\DataBase\\ff30.csv')
726
727
               x1 = sm. add_constant(ff30. iloc[:, 1:])
               X1 = pd. concat([exeq, x1], axis=1)
728
               X1 = X1. dropna()
729
730
               X2 = pd. concat([exvw, x1], axis=1)
               X2 = X2. dropna()
731
732
               regeqq = sm. OLS(np. asarray(X1.iloc[:, 0]), np. asarray(X1.iloc[:, 1:])).fit()
               regvwq = sm.OLS(np.asarray(X2.iloc[:, 0]), np.asarray(X2.iloc[:, 1:])).fit()
733
               resultew.iloc[8, j] = regeqq.params[0]
734
               resultew.iloc[9, j] = regeqq.tvalues[0]
735
               resultvw.iloc[8, j] = regvwq.params[0]
736
               resultvw.iloc[9, j] = regvwq.tvalues[0]
737
           resultew.to_csv('EW55BM' + reteq.columns[u] + '.csv')
738
           resultvw. to csv('VW55BM' + reteq.columns[u] + '.csv')
739
740
741
       #***** fama macbeth 回归检验结果
742
743
       744
       import pandas as pd
745
       import numpy as np
746
       import glob, os
```

```
747
       import statsmodels.api as sm
748
       from sklearn.preprocessing import scale
749
       from factorNWttest import nwttest_1samp
750
       ret=pd. read_csv('final_return.csv')
751
       size=pd. read_csv('...\DataBase\\factor\\01size.csv')
       path = r'...\DataBase\\factor'
752
753
       file = glob. glob (os. path. join (path, "*. csv"))
754
       rf=pd.read_csv('..\DataBase\\RF.csv')
755
       k = [] # 每个因子一个表
       for i in range (96):
756
757
           k.append(pd.read csv(file[i]))
       factor = []#因子名称
758
       for i in range(len(file)):
759
760
           factor.append(file[i][29:-4])
761
       factor.insert(0, 'constant')
762
       xishu=pd. DataFrame(index=ret. columns[1:], columns=factor)
763
       for i in range(ret.shape[1]-1):
764
           final=pd.concat([j.iloc[:,i+1] for j in k], axis=1)
           final.columns=factor[1:]
765
           final = sm.add_constant(final)
766
           final['ret']=ret.iloc[:, i+1]
767
768
           X1=final.dropna()
769
           if 1en(X1) == 0:
               xishu.iloc[i,:]=np.nan
770
           else:
771
               X1.iloc[:, 1:-1] = scale(X1.iloc[:, 1:-1], axis=0)
772
               reg = sm.OLS(np.asarray(X1.iloc[:, 0]), np.asarray(X1.iloc[:, 1:])).fit()
773
774
               xishu.iloc[i,:]=reg.params
775
       FM=pd. DataFrame(columns=xishu.columns, index=['mean', 't-stats'])
776
       gen=xishu.dropna()
       for i in range(xishu.shape[1]):
777
778
           fa = gen.iloc[:, i]
779
           FM. iloc[0, i] = fa. mean() #截面系数时序均值
           ttest = nwttest_1samp(fa, 0)#t 检验-因子系数是否异于 0
780
           FM. iloc[1, i] = ttest. statistic
781
782
       FM. to csv ('FM. csv')
783
```

# g) DFN.py

```
1
     # -*- coding:utf-8 -*-
2
3
     @description:
 4
     深度前馈网络库函数,依赖 mxnet 和 gpu, 在有 GPU 的环境下运行。
 5
     from mxnet import gluon, nd, gpu, autograd
6
7
     from mxnet.gluon import loss as gloss, nn, data as gdata
8
     import mxnet as mx
9
     import numpy as np
10
11
     class DFN(nn. Block):
12
         def __init__(self, outputdim, batch_size=10, epoch=10, depth=1, neuralset=[],
13
                      activiatemethod='relu', ctx=gpu(), trainmethod='sgd', lr=0.5, **kwargs):
14
             主函数, 用于深度前馈网络训练
15
             :param outputdim: 輸出维度
16
             :param batch_size: 小批量大小
17
18
             :param epoch: 训练论述
19
             :param depth: 神经网络深度
20
             :param neuralset: 各层神经元数, list 形式 eg. [96, 50, 10]
21
             :param activiatemethod: 激活函数
             :param ctx: 训练用设备 默认 gpu
22
23
             :param trainmethod: 训练器
24
             :param lr: 学习率
25
             :param kwargs: 其他参数
26
             super(DFN, self). __init__(**kwargs)
27
28
             mx. random. seed (521, ctx=gpu())
             mx. random. seed (521)
29
             self.ctx = ctx
30
31
             self.outputdim = outputdim
32
             self.net = nn.Sequential()
33
             self.neuralset = neuralset
34
             self.activiatemethod = activiatemethod
35
             self.depth = depth
             self.trainmethod = trainmethod
36
             self.lr = lr
37
38
             if neuralset:
39
                 pass
40
             else:
                 neuralset = [256]*depth
41
             for i in range(depth):
42
```

```
43
                  self.net.add(nn.Dense(neuralset[i], activation=activiatemethod))
              self.net.add(nn.Dense(outputdim))
44
45
              self.net.initialize(ctx=self.ctx)
              self.Trainer = gluon.Trainer(self.net.collect_params(), trainmethod,
46
      {'learning rate': lr})
47
              self.loss = gloss.L2Loss()
48
49
              self.batch size = batch size
              self.epoch = epoch
50
51
              print('network initializing finished')
52
          def retrain(self, Xnd, Ynd, data iter, k):
53
54
              knew = 0
55
              self.net.initialize(ctx=self.ctx, force_reinit=True)
              self.Trainer = gluon.Trainer(self.net.collect_params(), self.trainmethod,
56
57
      {'learning_rate': self.lr/2})
58
              print('reinitialize finished and retrain begin')
              for i in range(k - 1):
59
                   # with mx. autograd. record():
60
                   # 1 = self. loss(self.net(Xnd), Ynd).mean()
61
                   # 1. backward()
62
63
                   # self. Trainer. step (4400)
64
                  for xtrain, ytrain in data_iter:
                       ytrain = nd. array(ytrain, ctx=self.ctx)
65
                       with mx. autograd. record():
66
                           1 = self.loss(self.net(xtrain), ytrain).mean()
67
68
                       1. backward()
69
                       self.Trainer.step(1)
70
                       print('#')
                  knew += 1
71
72
                   lossout = self.loss(self.net(Xnd), Ynd)
                  print('epoch' + str(i) + ': loss is %f' % lossout.mean().asnumpy())
73
74
                   if lossout.mean().asnumpy()>1:
75
                       self.retrain(Xnd, Ynd, data_iter, knew)
76
                       break
77
                   elif np.isnan(lossout.mean().asnumpy()):
78
                       self.retrain(Xnd, Ynd, data_iter, knew)
79
80
81
82
              # print (lossout)
83
84
85
          def fit(self, X, Y):
              Xnd = nd. array(X, ctx = self. ctx)
86
```

```
87
               Ynd = nd. array(Y, ctx = self. ctx)
               dataset = gdata.ArrayDataset(Xnd, Ynd)
88
89
               # 随机读取小批量
90
               data_iter = gdata.DataLoader(dataset, self.batch_size, shuffle=False)
91
92
               for i in range(self.epoch):
93
                   # with mx. autograd. record():
94
                   # 1 = self.loss(self.net(Xnd), Ynd).mean()
                   # 1. backward()
95
                   # self. Trainer. step (4400)
96
97
                   for xtrain, ytrain in data_iter:
98
                       ytrain = nd. array(ytrain, ctx=self. ctx)
99
                       with mx. autograd. record():
100
                            1 = self.loss(self.net(xtrain), ytrain).mean()
101
                       1. backward()
102
                       self. Trainer. step(1)
                       print('#', )
103
                   k += 1
104
105
                   lossout = self.loss(self.net(Xnd), Ynd)
106
                   print('epoch' + str(i) + ': loss is %f' % lossout.mean().asnumpy())
                   if lossout.mean().asnumpy()>1:
107
108
                       retrainNN = True
109
                   elif np.isnan(lossout.mean().asnumpy()):
110
                       retrainNN = True
111
                       break
112
113
                   else:
114
                       retrainNN = False
               if retrainNN:
115
116
                   self.retrain(Xnd, Ynd, data_iter, k)
               print('all finished')
117
118
           def predict(self, Xtest):
119
               Xtestnd = nd. array(Xtest, ctx = self.ctx)
120
121
               return list(map(lambda x:x[0], nd. array(self. net(Xtestnd), ctx=mx. cpu()). asnumpy()))
```

## h) RNNMODEL.py

```
1
      # -*- coding:utf-8 -*-
      111
2
 3
      @description:
 4
        用于构建循环神经网络
 5
 6
7
      import mxnet as mx
8
      from mxnet import nd, gluon, gpu
9
     from mxnet. gluon import rnn, nn, loss as gloss, data as gdata
10
      import numpy as np
11
12
      class BaseRnn(nn. Block):
          def __init__(self, num_feature, num_hidden, num_layers, ntype='RNN', bidirectional =
13
14
     False, dropout=0, **kwargs):
15
16
              基本循环神经网络单元
17
              :param num_feature: 特征个数
              :param num_hidden: 隐层神经元个数
18
              :param num_layers: 隐层数目
19
20
              :param ntype: 神经网络类型, 默认 RNN
21
              :param bidirectional: 是否双向, 默认否
              :param dropout: 丢弃率, 即神经元多大概率不激活
22
23
              :param kwargs: 其他参数
24
25
              super(BaseRnn, self). __init__(**kwargs)
26
              self.num hidden = num hidden
              if ntype == 'RNN':
27
28
                  with self.name scope():
29
                      self.rnn =
30
      rnn. RNN (num hidden, num layers, input size=num feature, bidirectional=bidirectional,
31
      layout='TNC', dropout=dropout)
                      self.decoder = nn.Dense(1, in_units=num_hidden)
32
33
              elif ntype == 'LSTM':
34
                  with self.name_scope():
35
                      self.rnn =
36
      rnn. LSTM (num_hidden, num_layers, input_size=num_feature, bidirectional=bidirectional,
37
      layout='TNC', dropout=dropout)
38
                      self.decoder = nn.Dense(1, in_units=num_hidden)
39
              elif ntype == 'GRU':
40
                  with self.name_scope():
41
                      self.rnn =
42
      rnn. GRU (num hidden, num layers, input size=num feature, bidirectional=bidirectional,
```

```
43
     layout='TNC', dropout=dropout)
                     self.decoder = nn.Dense(1, in_units=num_hidden)
44
45
         def forward(self, inputs, hidden):
46
             output, hidden = self.rnn(inputs, hidden)
47
             decoded = self.decoder(output.reshape((-1, self.num_hidden)))
48
49
             return decoded, hidden
50
51
         def begin_state(self, *args, **kwargs):
52
             return self.rnn.begin_state(*args, **kwargs)
53
54
55
     class lstmmodule(object):
56
57
     __init__(self, num_feature, num_hidden, num_layers, epoch, batch_size, ntype='LSTM', bidirectional
58
     = False, dropout=0, trainmethod = 'adam', \
                      1r=0.01, loss = gloss. L1Loss(), ctx =
59
     gpu(0), datashuffle=False, initialfunc=mx. init. Xavier(), prin=False, **kwargs):
60
61
62
             用于循环神经网络的训练和预测, 主函数
63
             :param num_feature: 特征个数
64
             :param num_hidden: 隐层神经元数目
65
             :param num layers: 隐层数
             :param epoch: 训练轮数
66
             :param batch_size: 小批量大小, 在这里是股票实体数目多少
67
             :param ntype: 神经网络类型,默认 LSTM
68
69
             :param bidirectional: 是否双向神经网络, 默认否
70
             :param dropout: 丢弃率, 默认不丢弃
71
             :param trainmethod: 训练器, 默认 adam
72
             :param 1r: 学习率
73
             :param loss: 损失函数
74
             :param ctx: 训练设备, 默认 gpu
75
             :param datashuffle: 是否随机打乱数据, 默认否
             :param initialfunc: 初始化方法, 默认 Xaiver
76
77
             :param prin: 是否输出训练指示, 默认否
             :param kwargs: 其他参数
78
79
             super(lstmmodule, self). __init__(**kwargs)
80
             mx. random. seed (521, ctx=gpu())
81
82
             mx. random. seed (521)
83
             self.ctx = ctx
             self.net = BaseRnn(num_feature, num_hidden, num_layers, ntype, bidirectional, dropout)
84
85
             self.trainmethod = trainmethod
             self.lr = lr
86
```

```
87
               self.intialfunc = initialfunc
               self.net.collect_params().initialize(self.intialfunc, ctx=self.ctx)
88
89
               self.Trainer = gluon.Trainer(self.net.collect_params(), trainmethod,
90
       {'learning_rate': lr})
91
               self.loss = loss
92
               self.datashuffle = datashuffle
93
               self.epoch = epoch
94
               self.batch_size = batch_size
95
               self.prin = prin
96
97
           def fit(self, X, Y):
98
               Xnd = nd. array(X, ctx = self. ctx)
               Ynd = nd. array(Y, ctx = self. ctx)
99
100
               for i in range(self.epoch):
101
                    state = self.net.begin_state(batch_size=self.batch_size, ctx=self.ctx)
102
                   for s in state:
103
                        s. detach()
104
                   with mx. autograd. record():
105
                        (output, state) = self.net(Xnd, state)
106
                        1 = self.loss(output.reshape(len(output)//self.batch_size, self.batch_size),
107
       Ynd).mean()
108
                   1. backward()
109
                   self. Trainer. step(1)
                   if self.prin:
110
                        print('#')
111
                        if i == self. epoch -1:
112
113
                            print('the last epoch loss is ', l.asnumpy())
114
               if self.prin:
                   print('all finished')
115
116
           def predict(self, Xtest):
117
118
               state = self.net.begin state(batch size=self.batch size, ctx=self.ctx)
119
               Xtestnd = nd. array(Xtest, ctx = self. ctx)
               (Y, statenew) = self.net(Xtestnd, state)
120
121
               return nd. array (Y, ctx=mx. cpu()). asnumpy()
```

## i) Ensembleall.py

```
1
      # -*- coding:utf-8 -*-
 2
 3
      @description:
 4
          集成所有函数模型模块
 5
 6
      import numpy as np
 7
      class Ensemblelr(object):
 8
          def __init__(self, classifymodellist, newmodellist, modelname):
 9
10
              :param classifymodellist: 非 RNN/LSTM 模型
11
12
              :param newmodellist: RNN/LSTM 模型
              :param modelname: 所有模型名称,
13
14
              注:PLS 放到传统模型倒数第三位
15
              self.cmodelist = classifymodellist
16
              self.nml = newmodellist
17
              self.modelname = modelname
18
19
              self.cmodelnum = len(classifymodellist)
20
              self.nmodelnum = len(newmodellist)
              print('Ensembleall model initializing')
21
22
23
          def fit(self, TX, TY, NX, NY):
              for i in range(self.cmodelnum):
24
                  self.cmodelist[i].fit(TX, TY)
25
26
                  print(self.modelname[i]+' finished')
              for i in range(self.nmodelnum):
27
                  self.nml[i].fit(NX, NY)
28
                  print(self.modelname[self.cmodelnum + i])
29
              print('fit finished')
30
31
          def predict(self, TXtest, NXtest, indlist, length):
32
              predictlist = []
33
              for i in range(self.cmodelnum):
34
                  predictlist.append(self.cmodelist[i].predict(TXtest))
35
              predictlist[-3] = list(map(lambda x: x[0], predictlist[-3])) # 为了PLS 的輸出结果变
36
37
      成数值
38
              for i in range(self.nmodelnum):
39
                  KP = self.nml[i].predict(NXtest)
                  predictlist.append(np.array([KP[m][0] for m in range(3571 * (length - 1), 3571 *
40
41
      length)])[~indlist])
              return list(np.mean(predictlist, axis=0))
42
```

## j) transecfee.py

```
1
      # -*- coding:utf-8 -*-
      ,,,
 2
      @description
 3
 4
        用于输出交易费用之后的结果
 5
 6
      import numpy as np
 7
      import pandas as pd
 8
      import os
 9
      import gc
10
      from NWttest import nwttest_1samp
11
      import statsmodels.formula.api as smf
12
      factorlist = os.listdir('../Database/returnseries/12')
13
      rf=pd. read_csv('.../Database/RF.csv')#无风险 rf
14
      rf12=rf.iloc[12:-1,:]
15
16
      def trasecfee(feerate):
17
18
          :param feerate: 交易费率
          :return:不輸出只打印结果
19
20
21
          for i in range(len(factorlist)):
              temp = pd.read_csv('.../Database/returnseries/12/' + factorlist[i])
22
              long_short = np. array(temp['long-short'])
23
              afterLS = list(long_short - feerate*2)
24
25
              length = 12
26
              name = factorlist[i]
              T_value = []
27
              Mean = []
28
              p_value = []
29
30
              sharpratio = []
              Std = []
31
              T0 = [afterLS]
32
33
              for 1 in TO:
                  t_test = nwttest_1samp(1, 0, L=1)
34
                  mean = np. average(1) * 12- rf12. mean(). tolist()[0] * 12 / 100
35
                  STD = np. std(1) * np. sqrt(12)
36
37
                  sharp = (mean) / STD
38
                  T_value.append(t_test.statistic)
39
                  p_value.append(t_test.pvalue)
40
                  Mean. append (mean)
41
                  Std. append (STD)
                  sharpratio.append(sharp)
42
```

```
43
              print(name, 'long-short')
44
              print('mean', Mean[0] / 12)
45
              print('t-statistic', '('+str(round(T_value[0], 4))+')')
              ff3 = pd. read_csv('../Database/ff3.csv')
46
              ff5 = pd. read_csv('../Database/ff5.csv')
47
              A = pd. DataFrame(afterLS, columns=['long-short'])
48
              M = pd. concat([A], axis=1)
49
              alpha3 = []
50
              t3 = []
51
              t5 = []
52
              alpha5 = []
53
54
              for i in range (1):
                  X1 = ff3.iloc[length:, 1:]
55
                  X2 = ff5.iloc[length:, 1:]
56
57
                  Y = M. iloc[:-2, i]
58
                  Y.index = X1.index
                  Y = Y - rf12.RF[:-1] / 100
59
                  used1 = {'X' : X1, 'Y' : Y}
60
                  reg = smf.ols(formula='Y~1+X', data=used1).fit(cov_type='HAC',
61
      cov_kwds={'maxlags': 1})
62
                  t3. append (reg. tvalues [0])
63
64
                  alpha3. append (reg. params [0] * 12)
                  used2 = {'X' : X2, 'Y' : Y}
65
                  reg = smf.ols(formula='Y~1+X', data=used2).fit(cov_type='HAC',
66
      cov kwds={'maxlags': 1})
67
                  t5. append (reg. tvalues [0])
68
69
                  alpha5.append(reg.params[0] * 12)
70
              print('alpha-FF3', alpha3[0]/12)
              print('t-statistic', '('+str(round(t3[0],4))+')')
71
              print('alpha-FF5', alpha5[0]/12,)
72
              print('t-statistic','('+str(round(t5[0],4))+')')
73
74
              print('sharpe', sharpratio[0])
              gc. collect()
75
              print ('*'*30) #分隔开不同收益序列
76
77
78
79
      def showtrasecfee(transectionfee):
          ,,,
80
81
          :param transectionfee: 交易费用比例
82
          :return: 无返回直接打印交易费用调整后的结果
83
          trasecfee(transectionfee)# 方便调用
84
```

## k) selectFactor.py

```
1
      # -*- coding:utf-8 -*-
2
3
      @description
 4
      用于筛选特征
 5
      import pandas as pd
6
7
      import numpy as np
8
     from sklearn. preprocessing import scale
9
     from sklearn. linear model import LinearRegression
10
      import gc
11
12
13
     def dropimportant (length, CLF, name, factor, timeseries, base = 0.028813732000000005, inpath =
14
      'output'):
15
16
         主函数 1, 用于非循环神经网络特征筛选
17
          :param length: 滑动窗口长度
18
         :param CLF: 模型
19
          :param name: 筛选用模型名称
20
          :param timeseries: 数据
21
          :param factor: 因子名称
22
          :param base: 多空对应月度收益
23
          :param inpath:输出路径
          :return:月度收益差
24
25
26
         meanlist = []
         for j in range (96):
27
28
             print(j)
             print(factor[j])
29
30
             Long Short = []
             Long = []
31
             Short = []
32
33
             for i in range(len(timeseries) - (length+1)):
34
                 FINALm = pd. concat(timeseries[i:i + (length+1)], axis=0)
35
                 FINALm = FINALm. fillna(0)
36
37
                 FINAL X= FINALm.iloc[:, :-2].copy()
38
                 FINAL_X.drop(columns=factor[j], inplace=True)
39
                 FINAL_x = scale(FINAL_X)
                 final = pd.concat(timeseries[i:i + length], axis=0)
40
41
                 x_train = FINAL_x[:len(final)]
                 x test = FINAL x[len(final):]
42
```

```
43
                  y_{train} = final.iloc[:, -1]
                  test = timeseries[i + length]
44
45
                  y_{test} = test.iloc[:, -1]
46
                  # 基准-linear
47
                  clf = CLF
                  clf.fit(x_train, y_train)
48
                  PREDICTION = clf.predict(x test)
49
50
                  # 构建投资组合
51
                  prediction = pd. DataFrame (PREDICTION)
                  r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
52
                  r_ture = pd. DataFrame(y_test)
53
54
                  r_ture.columns = ['ture']
55
                  r_ture.index = r_predict.index
                  FINAL = pd. concat([r_predict, r_ture], axis=1)
56
57
                  FINAL_sort = FINAL.sort_values(by='predict', axis=0)
58
                  r_final = np. array(FINAL_sort['ture'])
                  m = int(len(r_final) * 0.1) + 1
59
                  r_final = r_final.tolist()
60
61
                  long = r_final[-m:]
                  short = r_final[:m]
62
63
                  r_{end} = (np. sum(long) - np. sum(short)) / m
64
                  Long_Short.append(r_end)
                  Long. append (np. average (long))
65
                  Short. append (np. average (short))
66
67
                  gc. collect()
              A = pd. DataFrame (Long_Short, columns=['long-short'])
68
              B = pd. DataFrame(Long, columns=['long'])
69
70
              C = pd. DataFrame (Short, columns=['short'])
              M = pd. concat([A, B, C], axis=1)
71
72
              meanlist.append(np.average(Long_Short))
              if (i\%10) == 0:
73
74
                  print(i)
75
          gc. collect()
76
          u = base - np. array(meanlist)
77
          u = pd. DataFrame(u)
          u. to_csv('.../' + inpath + '/' +name+' returngap. csv')
78
79
          return u
80
81
      def dropimportant2(length, CLF, name, factor, timeseries, base = 0.028813732000000005, inpath =
82
      'output', a=0, b=96):
          ,,,
83
84
          主函数 2 用于循环神经网络特征筛选
          :param length: 滑动窗口大小
85
86
          :param CLF: 筛选用模型
```

```
87
           :param name: 模型名称
88
           :param factor: 因子名称
89
           :param timeseries: 数据
90
           :param base: 月度收益
91
           :param inpath: 輸出位置
92
           :param a: 起始因子
93
           :param b: 终止因子
94
           :return: 返回收益差
95
96
          meanlist = []
97
          for j in range(a, b):
98
               print(j)
99
               print(factor[j])
              Long_Short = []
100
101
              Long = []
102
              Short = []
103
               for i in range(len(timeseries) - length):
104
                   FINALm = pd. concat(timeseries[i:(i + length + 1)], axis=0)
                   FINALm[~FINALm['ret'].isin(['null'])] =
105
      FINALm[~FINALm['ret'].isin(['null'])].fillna(0)
106
                   FINAL X = FINALm.iloc[:, :-2]
107
108
                   FINAL_X. drop(columns=factor[j], inplace=True)
109
                   FINAL x = FINAL X
                   FINAL_x[~FINALm['ret'].isin(['null'])] =
110
       scale(FINAL_X[~FINALm['ret'].isin(['null'])])
111
                   FINAL_x[FINALm['ret'].isin(['null'])] = 0
112
                   FINALm[FINALm['ret'].isin(['null'])] = 0
113
                   x_{train} = [FINAL_x.iloc[j * 3571:(j + 1) * 3571, :].values for j in
114
115
      range(length)]
116
                   y_{train} = [FINALm. iloc[j * 3571:(j + 1) * 3571, -1]. values for j in
117
      range(length)]
                   x test = [FINAL \ x. iloc[j * 3571:(j + 1) * 3571, :]. values for j in range(1,
118
119
      length + 1)
                   y_{test} = list(FINALm.iloc[(length) * 3571:(length + 1) * 3571, -1].values)
120
121
                   # 基准-linear
                   clf = CLF
122
123
                   clf.fit(x train, y train)
                   PREDICTION = clf.predict(x_test)
124
                   PREDICTION = [PREDICTION[m][0] for m in range (3571 * (length - 1), 3571 *
125
126
      length)]
127
                   # 构建投资组合
                   r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
128
129
                   r_ture = pd. DataFrame(y_test)
                   r ture.columns = ['ture']
130
```

```
131
                   r_ture.index = r_predict.index
132
                   FINAL = pd. concat([r_predict, r_ture], axis=1)
133
                   FINAL = FINAL[~timeseries[i + length]['ret'].isin(['null'])]
134
                   FINAL_sort = FINAL.sort_values(by='predict', axis=0)
                   r_final = np. array(FINAL_sort['ture'])
135
                   m = int(len(r_final) * 0.1) + 1
136
                   r_final = r_final.tolist()
137
138
                   long = r_final[-m:]
                   short = r_final[:m]
139
                   r_{end} = (np. sum(long) - np. sum(short)) / m
140
141
                   Long_Short.append(r_end)
142
                   Long. append (np. average (long))
                   Short.append(np.average(short))
143
144
                   gc. collect()
145
               A = pd. DataFrame (Long_Short, columns=['long-short'])
146
               B = pd. DataFrame (Long, columns=['long'])
               C = pd. DataFrame(Short, columns=['short'])
147
148
               M = pd. concat([A, B, C], axis=1)
               meanlist.append(np.average(Long_Short))
149
               if (i\%10) == 0:
150
151
                   print(i)
152
           gc.collect()
           u = base - np. array(meanlist)
153
           u = pd. DataFrame(u)
154
           u. to_csv('.../' + inpath + '/' + name + 'returngap.csv')
155
156
           return u
157
158
       def FCselect(factor, timeseries):
159
160
           用于FC筛选因子
161
           :param factor:因子名称
162
           :param timeseries:数据
163
           :return:
164
165
           length = 12
           base = 0.0228
166
           name = 'FC'
167
           inpath = 'output'
168
169
           meanlist = []
170
           for j in range (96):
171
               print(j)
               print(factor[j])
172
               Long_Short = []
173
               Long = []
174
```

```
175
               Short = []
                for i in range(len(timeseries) - (length + 1)):
176
177
                    FINALm = pd. concat(timeseries[i:i + (length + 1)], axis=0)
178
                    FINALm = FINALm. fillna(0)
179
                    FINAL_X = FINALm.iloc[:, :-2].copy()
                    FINAL_X.drop(columns=factor[j], inplace=True)
180
                    FINAL x = scale(FINAL X)
181
                    final = pd.concat(timeseries[i:i + length], axis=0)
182
183
                    x_train = FINAL_x[:len(final)]
                    x_test = FINAL_x[len(final):]
184
185
                    y_{train} = final.iloc[:, -1]
186
                    test = timeseries[i + length]
187
                    y_{test} = test.iloc[:, -1]
188
                    clf = LinearRegression()
189
                    k = []
190
                    for ax in range (95):
191
                        x = x_{train}[:, ax]. reshape(-1, 1)
192
                        clf.fit(x, y_train)
                        k. append (clf. coef_[0])
193
194
                    PREDICTION = []
195
                    for ap in range(len(x test)):
196
                        test0 = np. array(x_test[ap])
197
                        v = 0
198
                        for ass in range (95):
199
                            y = y + test0[ass] * k[ass]
                        PREDICTION. append(y)
200
201
                    y_{test} = test.iloc[:, -1]
202
                    prediction = pd. DataFrame (PREDICTION)
203
                    r_predict = pd. DataFrame (PREDICTION, columns=['predict'])
204
                    r_ture = pd. DataFrame(y_test)
                    r_ture.columns = ['ture']
205
206
                    r ture.index = r predict.index
207
                    FINAL = pd. concat([r_predict, r_ture], axis=1)
                    FINAL_sort = FINAL.sort_values(by='predict', axis=0)
208
209
                    r_final = np. array(FINAL_sort['ture'])
                    m = int(len(r_final) * 0.1) + 1
210
                    r final = r final.tolist()
211
                    long = r_final[-m:]
212
                    short = r final[:m]
213
214
                    r_{end} = (np. sum(long) - np. sum(short)) / m
215
                    Long_Short.append(r_end)
                    Long. append (np. average (long))
216
217
                    Short. append (np. average (short))
                    gc. collect()
218
```