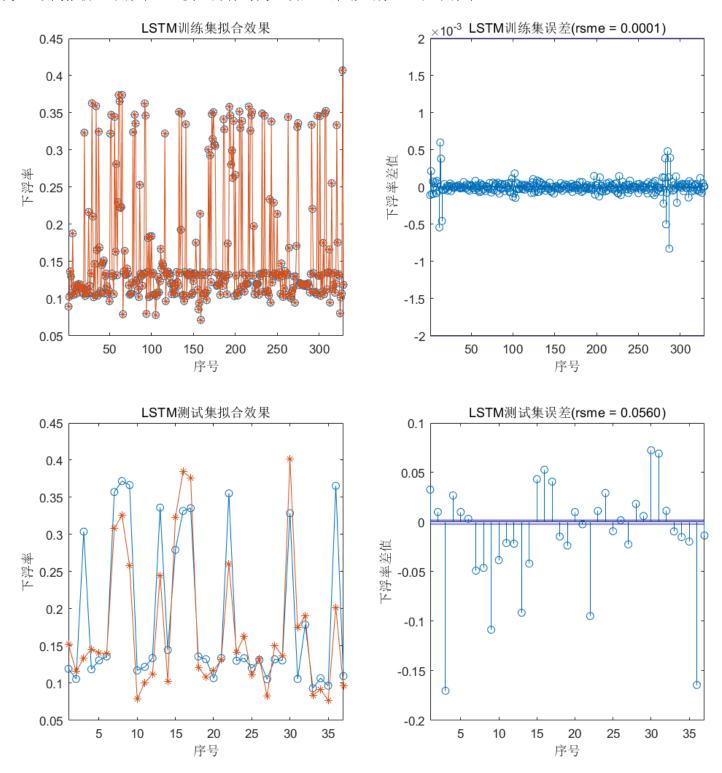
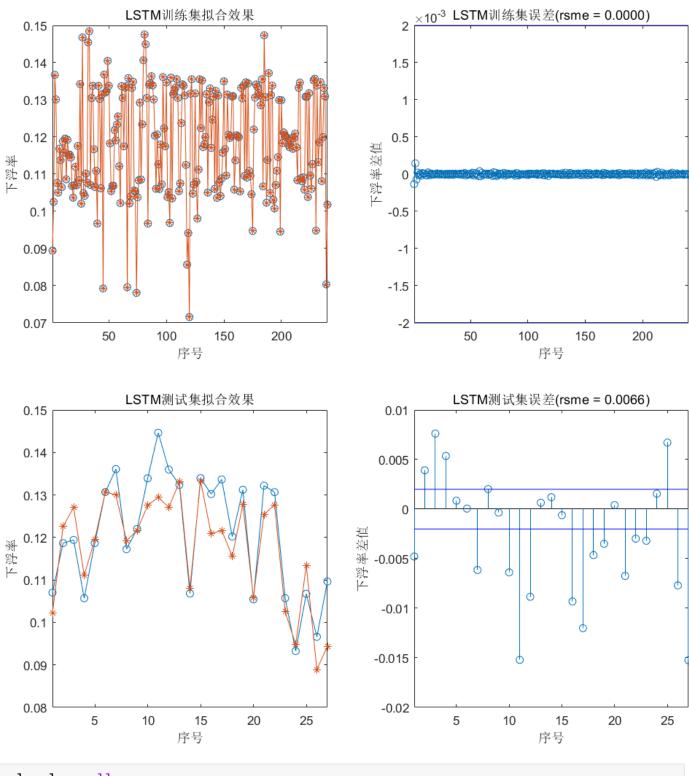
下浮率预测模型: LSTM

资质要求、最高报价、下浮率、工资、开标时间、镇区、计划工期 \rightarrow \mathbf{P} 值下浮率



选择下浮率小于 0.15 部分:资质要求、最高报价、下浮率、工资、开标时间、镇区、计划工期 \rightarrow P值下浮率



```
clear;clc;close all;
%%%%%%%%%%%%%%%%%%%%% 加载数据
filename = '\\longde\longde\008-投标中心\006-数据分析中心\分析结果汇总\zb_ML.xlsx';
file = readtable(filename, 'Sheet', 1, 'VariableNamingRule', 'preserve');
data = table2array(file(:, 2:end));
% 数据筛选
indx = find(data(:, end) <= 0.15);
data = data(indx, :);
% 输入、输出数据提取
```

```
% 方式一:已知信息 → 未知信息
output = data(:, end);
input = data(:, 1:end-2);
% 方式二:已知信息 + 之前的结果信息 → 未知信息
% output = data(2 : end, end);
% input = [data(2 : end, 1 : end-2), data(1 : end-1, end)];
% 数据分成训练集 (90%) 和测试集 (10%)
train_number = floor(0.9 * size(data, 1));
test_number = size(data, 1) - train_number;
xTrain = input(1:train number, :);
yTrain = output(1:train_number, :);
xTest = input(train_number+1 : end, :);
yTest = output(train number+1 : end, :);
[XTrain, PSx] = mapminmax(xTrain');
[YTrain, PSy] = mapminmax(yTrain');
XTest = mapminmax('apply', xTest', PSx);
%%%%%%%%%%%%%%%%% 构建 LSTM 网络
% 输入特征的维度
numFeatures = size(input, 2);
% LSTM 网路包含的隐藏单元数目,越大拟合越好,速度越慢
numHiddenUnits = 100;
% 输出响应维度
numResponses = 1;
layers = [sequenceInputLayer(numFeatures)
         lstmLayer(numHiddenUnits)
         fullyConnectedLayer(numResponses)
         regressionLayer];
miniBatchSize = 64;
options = trainingOptions('adam', ...
                         'ExecutionEnvironment', 'cpu', ...
                         'MaxEpochs', 1000, ...
                         'MiniBatchSize', miniBatchSize, ...
                         'GradientThreshold', 1, ...
                         'InitialLearnRate', 0.01, ...
                         'LearnRateSchedule', 'piecewise', ...
                         'LearnRateDropPeriod', 250, ...
                         'LearnRateDropFactor', 0.2, ...
                         'Verbose', false, ...
                         'Plots', 'training-progress');
% 训练
net = trainNetwork(XTrain, YTrain, layers, options);
% 训练效果
yTrain_pre = predict(net, XTrain, 'MiniBatchSize', miniBatchSize, 'SequenceLength', 'longest');
% 反归一化
yTrain pre = mapminmax('reverse', yTrain pre, PSy);
yTrain_pre = yTrain_pre';
% 差值
err_train = yTrain_pre - yTrain;
% 均方误差
rmse_train = rms(err_train);
line_up_train = 0.002 * ones(1, train_number);
line_down_train = -0.002 * ones(1, train_number);
% 画图
```

```
figure('Position', [10, 10, 900, 400]);
subplot(1, 2, 1);
plot(yTrain, '-o');
hold on
plot(yTrain_pre, '-*');
hold off
xlabel('序号');
ylabel('下浮率');
title('LSTM训练集拟合效果');
xlim([1, train_number]);
subplot(1, 2, 2);
stem(err_train);
hold on
plot(line_up_train, 'b');
plot(line_down_train, 'b');
hold off
xlim([1, train_number]);
xlabel('序号');
ylabel('下浮率差值');
title(sprintf('LSTM训练集误差(rsme = %.4f)', rmse_train));
yTest_pre = predict(net, XTest, 'MiniBatchSize', miniBatchSize, 'SequenceLength', 'longest');
% 反归一化
yTest_pre = mapminmax('reverse', yTest_pre, PSy);
yTest_pre = yTest_pre';
% 差值
err_test = yTest_pre - yTest;
% 均方误差
rmse test = rms(err test);
line_up_test = 0.002 * ones(1, test_number);
line_down_test = -0.002 * ones(1, test_number);
% 画图
figure('Position', [10, 10, 900, 400]);
subplot(1, 2, 1);
plot(yTest, '-o');
xlabel('序号');
ylabel('下浮率');
title('LSTM测试集拟合效果');
hold on
plot(yTest pre, '-*');
hold off
xlim([1, test_number]);
subplot(1, 2, 2);
stem(err_test);
xlabel('序号');
ylabel('下浮率差值');
title(sprintf('LSTM测试集误差(rsme = %.4f)', rmse_test));
hold on
plot(line_up_test, 'b');
plot(line_down_test, 'b');
hold off
xlim([1, test number]);
```