

机器学习

赶开东 前花 * 编新



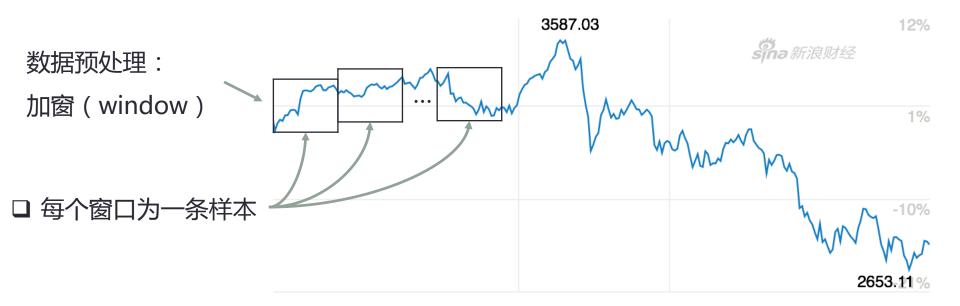
机器学习 卷积神经网络的股票预测

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基于卷积的股票预测-数据预处理



统计窗口内涨跌次数作为窗口的标记:

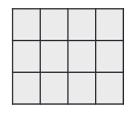
窗内涨多跌少,则标记为1;反之标记为-1

将股票走势预测转化为分类问题

基于卷积的股票预测

通常的做法:

按照图片的处理方法

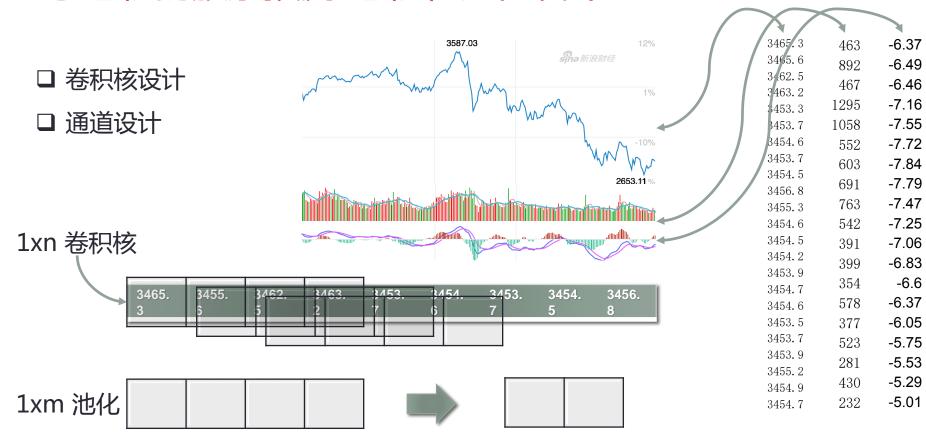




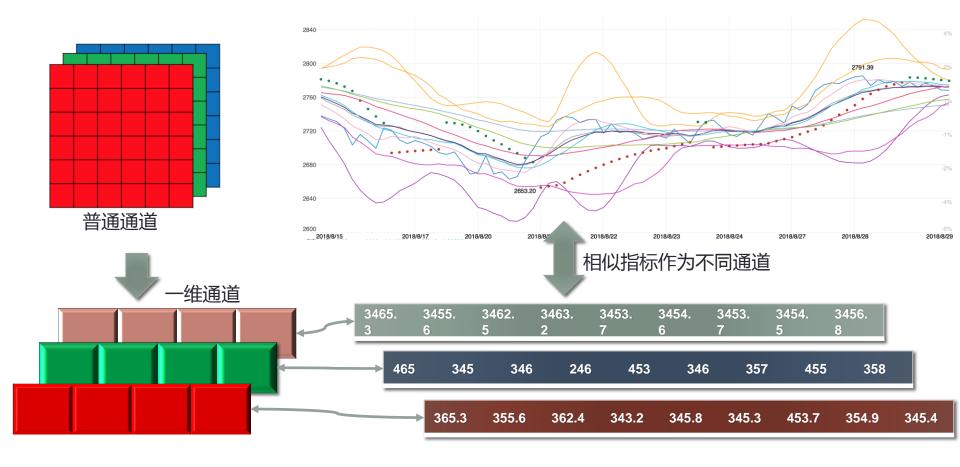
传统的卷积核



基于卷积的股票预测-卷积和池化设计



基于卷积的股票预测-通道设计



基于卷积的股票预测效果

□ 数据条数6000+

□ 训练Epoch:10次

□ 网络层数:5层

优化空间:

增加信息量(大数据方面)

。 调整网络结构

实际交易风险控制

```
Epoch:
           Training Loss:
                            6.705815553665161
                                                Training Accuracy:
                                                                     0.6666667
Epoch:
           Training Loss:
                            14.391897320747375
                                                 Training Accuracy:
                                                                      0.6666667
Epoch:
                                                 Training Accuracy:
                                                                      0.6666667
           Training Loss:
                            21.763847160339356
Epoch:
           Training Loss:
                            28.873583579063414
                                                 Training Accuracy:
                                                                      0.6666667
                                                Training Accuracy:
Epoch:
           Training Loss:
                            35.78049612045288
                                                                     0.6666667
           Training Loss:
                                                Training Accuracy:
Epoch:
                            42.51658825874328
                                                                     0.6666667
Epoch:
           Training Loss:
                            49.10774512290955
                                                Training Accuracy:
                                                                     0.6666667
Epoch:
                                                Training Accuracy:
           Training Loss:
                            55.57431836128235
                                                                     0.67676765
Epoch:
           Training Loss:
                                                Training Accuracy:
                            61.93427429199219
                                                                     0.67676765
           Training Loss:
                                                Training Accuracy:
Epoch:
                            68.19690041542053
                                                                     0.67676765
Testing Accuracy: 0.6888889
```

