实验一报告.md 2023-12-16

基于前馈神经网络的分类任务设计

介绍

本实验要求设计一个前馈神经网络,对一组二维高斯样本数据实现分类任务

我的电脑配置如下

- CPU R7 4800h 主频 2.9Ghz
- 内存 16G
- python版本 3.10
- 网络框架 pytorch

我的前馈神经网络使用了一个隐藏层,输入是2维的数据,输出是4维的数据,隐藏层我分别选择了8层和16层作为超参数进行测试。选择的激活函数为Relu和Sigmoid。

学习率选择为0.001, batch size设置为10, 训练了100轮。

因此做了四次对比实验。

结果

1. 使用8层隐藏层,激活函数选择Relu的结果如下

```
Epoch 10/100 - Train Loss: 0.0241 - Val Loss: 0.2112 - Val Accuracy: 0.9325
Epoch 20/100 - Train Loss: 0.0222 - Val Loss: 0.1956 - Val Accuracy: 0.9350
Epoch 30/100 - Train Loss: 0.0218 - Val Loss: 0.1931 - Val Accuracy: 0.9350
Epoch 40/100 - Train Loss: 0.0216 - Val Loss: 0.1929 - Val Accuracy: 0.9350
Epoch 50/100 - Train Loss: 0.0215 - Val Loss: 0.1924 - Val Accuracy: 0.9325
Epoch 60/100 - Train Loss: 0.0214 - Val Loss: 0.1921 - Val Accuracy: 0.9350
Epoch 70/100 - Train Loss: 0.0214 - Val Loss: 0.1920 - Val Accuracy: 0.9350
Epoch 80/100 - Train Loss: 0.0213 - Val Loss: 0.1918 - Val Accuracy: 0.9350
Epoch 90/100 - Train Loss: 0.0213 - Val Loss: 0.1917 - Val Accuracy: 0.9350
Epoch 100/100 - Train Loss: 0.0213 - Val Loss: 0.1915 - Val Accuracy: 0.9350
test_accuracy: 93.5000 %
```

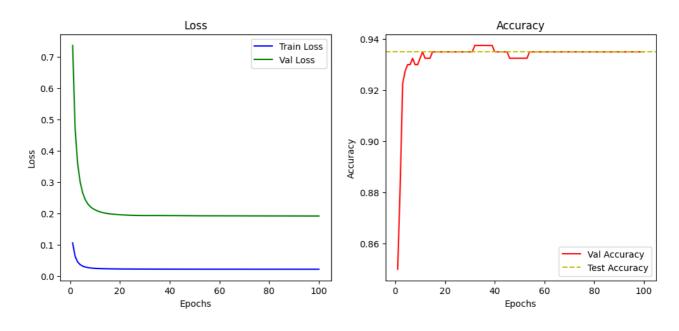


图 1

2. 使用16层隐藏层,激活函数选择Relu的结果如下

Epoch 10/100 - Train Loss: 0.0226 - Val Loss: 0.1911 - Val Accuracy: 0.9350 Epoch 20/100 - Train Loss: 0.0217 - Val Loss: 0.1801 - Val Accuracy: 0.9375 Epoch 30/100 - Train Loss: 0.0215 - Val Loss: 0.1772 - Val Accuracy: 0.9400 Epoch 40/100 - Train Loss: 0.0214 - Val Loss: 0.1760 - Val Accuracy: 0.9400 Epoch 50/100 - Train Loss: 0.0213 - Val Loss: 0.1756 - Val Accuracy: 0.9375 Epoch 60/100 - Train Loss: 0.0212 - Val Loss: 0.1749 - Val Accuracy: 0.9375 Epoch 70/100 - Train Loss: 0.0212 - Val Loss: 0.1746 - Val Accuracy: 0.9375 Epoch 80/100 - Train Loss: 0.0211 - Val Loss: 0.1746 - Val Accuracy: 0.9375 Epoch 90/100 - Train Loss: 0.0211 - Val Loss: 0.1748 - Val Accuracy: 0.9375 Epoch 100/100 - Train Loss: 0.0210 - Val Loss: 0.1748 - Val Accuracy: 0.9375 test_accuracy: 93.7500 %

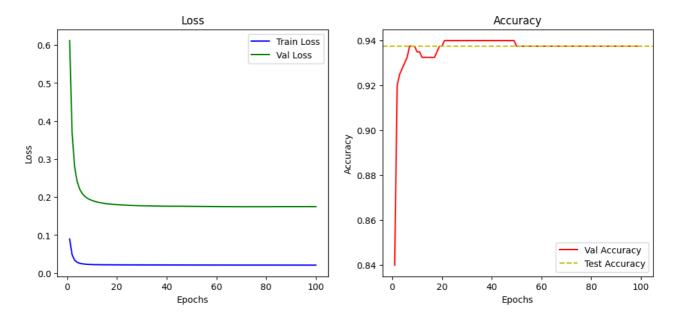


图 2

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Epoch 10/100 - Train Loss: 0.0415 - Val Loss: 0.4033 - Val Accuracy: 0.8975
Epoch 20/100 - Train Loss: 0.0257 - Val Loss: 0.2774 - Val Accuracy: 0.9225
Epoch 30/100 - Train Loss: 0.0232 - Val Loss: 0.2567 - Val Accuracy: 0.9200
Epoch 40/100 - Train Loss: 0.0225 - Val Loss: 0.2497 - Val Accuracy: 0.9225
Epoch 50/100 - Train Loss: 0.0221 - Val Loss: 0.2457 - Val Accuracy: 0.9225
Epoch 60/100 - Train Loss: 0.0219 - Val Loss: 0.2426 - Val Accuracy: 0.9225
Epoch 70/100 - Train Loss: 0.0218 - Val Loss: 0.2398 - Val Accuracy: 0.9250
Epoch 80/100 - Train Loss: 0.0216 - Val Loss: 0.2373 - Val Accuracy: 0.9275
Epoch 90/100 - Train Loss: 0.0215 - Val Loss: 0.2348 - Val Accuracy: 0.9275
Epoch 100/100 - Train Loss: 0.0214 - Val Loss: 0.2324 - Val Accuracy: 0.9300
test_accuracy: 93.0000 %

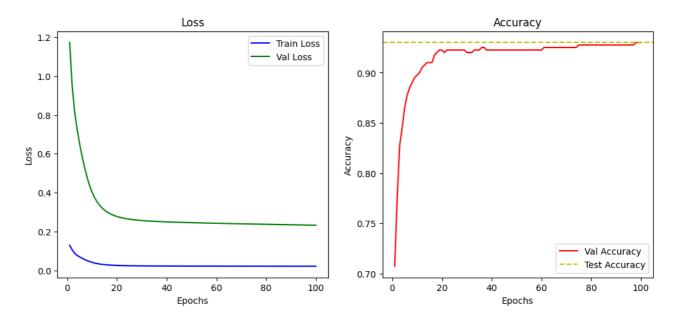


图 3

4. 使用16层隐藏层,激活函数选择Sigmoid的结果如下

Epoch 10/100 - Train Loss: 0.0315 - Val Loss: 0.3357 - Val Accuracy: 0.9150
Epoch 20/100 - Train Loss: 0.0238 - Val Loss: 0.2778 - Val Accuracy: 0.9150
Epoch 30/100 - Train Loss: 0.0223 - Val Loss: 0.2669 - Val Accuracy: 0.9150
Epoch 40/100 - Train Loss: 0.0217 - Val Loss: 0.2618 - Val Accuracy: 0.9125
Epoch 50/100 - Train Loss: 0.0214 - Val Loss: 0.2582 - Val Accuracy: 0.9150
Epoch 60/100 - Train Loss: 0.0212 - Val Loss: 0.2555 - Val Accuracy: 0.9150
Epoch 70/100 - Train Loss: 0.0210 - Val Loss: 0.2533 - Val Accuracy: 0.9175
Epoch 80/100 - Train Loss: 0.0209 - Val Loss: 0.2515 - Val Accuracy: 0.9150
Epoch 90/100 - Train Loss: 0.0208 - Val Loss: 0.2499 - Val Accuracy: 0.9150
Epoch 100/100 - Train Loss: 0.0207 - Val Loss: 0.2485 - Val Accuracy: 0.9175
test_accuracy: 91.7500 %

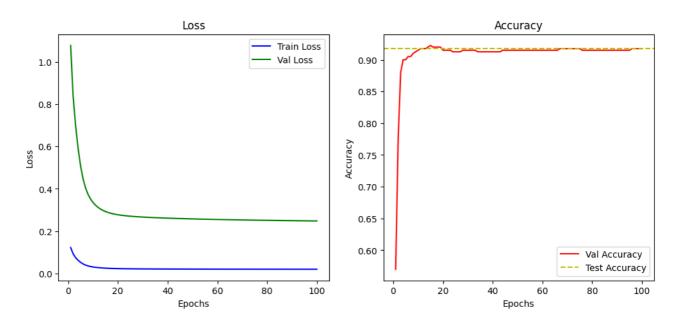


图 4

讨论

- 由对比可见,使用更复杂的网络并不一定能获得更好的准确率。
- 但是分析图四可见,在第10轮就达到了91.5%的准确率,然后基本保持不变。可见对于第四种组合,能够较快的收敛。使用较复杂的网络反而收敛的较快,这是我比较疑惑的地方。
- 使用第三种组合,准确率一直在上升,由此可见,这种组合收敛较慢。
- 每种组合的损失都会在10轮左右下降到一个比较稳定的值,而后就基本保持不变。说明我选择的损失函数CrossEntropyLoss()效果还不错。
- 除此之外,我发现使用8层隐藏层得到的准确率曲线较平滑,而是用16层隐藏层得到的准确率曲线呈现阶梯状。我猜想使用16层隐藏层每一轮训练使得梯度降低的较快,以至于接下来几轮训练的效果都不太明显,以至于出现了阶梯状。而是用8层隐藏层每一轮训练得到的梯度下降差不多,准确率曲线就会较平滑。