

# 本科生实验报告

# 机器学习技术综合训练

实验 1: 使用神经网络识别手写数字

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## 实验 1: 使用神经网络识别手写数字

### 1.1 问题描述

试根据神经网络的授课内容,结合多分类问题的四个反向传播方程,实现一个手写数字识别程序,使其可以对 MNIST 数据集图片中的手写数字进行识别。其中,神经网络包含两个隐层,第一个隐层的神经元个数为 192,第二个隐层的神经元个数为 30

损失函数采用交叉熵损失函数,激活函数使用 Sigmoid 激活函数

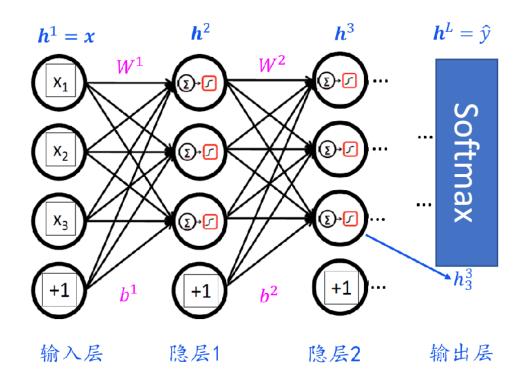
## 1.2 实验原理

神经网络算法介绍

18 end

```
Algorithm 1: Neutral Network in SGD
```

```
输入: 训练集合 x_i, i=1,\dots,n
 1 for i in 1 : n/m do
            (1). 向前传播
             for 1 in 2, \dots, L do
                   \mathbf{z}^{\mathbf{x},\mathbf{l}} \coloneqq \left(W^{l-1}\right)^T h^{x,l-1} + b^{x,l-1}
                   \mathbf{h}^{\mathbf{x},\mathbf{l}} \coloneqq \sigma\left(z^{x,L}\right)
             end
           (2). 输出误差: \delta^{x,L}
 \mathbf{s} \qquad \boldsymbol{\delta}^{\mathbf{x},\mathbf{L}} = (h^{x,L} - y^x)
       (3). 反向传播误差: \delta^{x,l}
             for 1 in L - 1 : 2 do
                   \delta^{\mathbf{x},\mathbf{l}} = \sigma'\left(z^{x,l}\right) \odot \left(W^l \delta^{x,l+1}\right)
             end
            (4). 梯度下降
             for 1 in 2:L do
                    \mathbf{W}^{l-1} := W^{l-1} - \frac{\eta}{m} \sum_{x} \frac{\partial E}{\partial W^{l-1}} = W^{l-1} - \frac{\eta}{m} \sum_{x} h^{x,l-1} \left( \delta^{x,l} \right)^{T}
                    \mathbf{b}^{l-1} \coloneqq b^{l-1} - \frac{\eta}{m} \sum_{x} \frac{\partial U}{\partial b^{l-1}} = b^{l-1} - \frac{\eta}{m} \sum_{x} \delta^{x,l}
16
             end
```



1.3 框架代码解读

题目已给定的代码有三个部分

- main.py 主函数,限定了神经网络的层数和每层的节点数,并修改工作路径以方便数据读入和输出
- mnist\_loader.py 由相对路径加载 mnist 数据集并读取为方便 network.py 处理的 形式
- network.py 神经网络的主干,上面提到的算法由 update\_mini\_batch 实现,剩余 部分主要负责提供 loss 函数的求值、求导和绘图输出
  - update\_mini\_batch 函数以最小批的方法更新参数,调用 backprop (对应算法描述中的前三步)得到反向传播误差,并应用到梯度下降上(对应算法描述中的第四步)

## 1.4 对于 TODO 的解答

一共有  $4 \uparrow TODO$ ,都是参照算法描述和代码采用的数据类型调用对应的 numpy 实现,没有什么难度

#### 1.4.1 计算 BP1

该部分对应算法描述中的第2步,已给出实现

```
# TODO: 此处为BP1的计算,已经给出参考代码如下三行
delta = (self.cost).delta(zs[-1], activations[-1], y)
nabla_b[-1] = delta
nabla_w[-1] = np.dot(activations[-2], delta.transpose())
```

#### 1.4.2 计算 $\delta$

该部分对应算法描述中的第 3 步,注意 np.dot 和 \* 的区别和递推关系中等号两边 矩阵的上标

```
# TODO: 完成 delta 的计算,对应PPT中的BP2
delta = sp * np.dot(self.weights[-l + 1], delta)
```

#### 1.4.3 计算 BP3

该部分对应算法描述中的第 4 步中计算  $\frac{\partial E}{\partial W^{l-1}} = h^{x,l-1} \left(\delta^{x,l}\right)^T$  和  $\frac{\partial E}{\partial b^{l-1}} = \delta^{x,l}$ ,注意递推关系中的先后次序和等号两边矩阵的上标

```
# TODO: 完成 Partial E / Partial b 和 Partial E / Partial w, 对应PPT中的BP3
和BP4

nabla_b[-1] = delta

nabla_w[-1] = np.dot(activations[-1 - 1], delta.transpose())
```

### 1.4.4 计算 mini batch 的 nabla b 和 nabla w

该部分对应算法描述中的第 4 步,注意 nabla\_b 和 nabla\_w 的数据类型为列表,不可以直接相加

```
1# TODO: 完成一个mini_batch中 nabla_b 和 nabla_w 的计算2nabla_b = [n_b + d_n_b for n_b, d_n_b in zip(nabla_b, delta_nabla_b)]3nabla_w = [n_w + d_n_w for n_w, d_n_w in zip(nabla_w, delta_nabla_w)]
```

### 1.5 结果展示

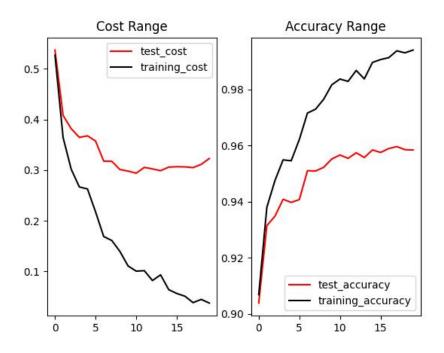


图 1: result

Listing 1: 程序输出

```
Epoch 0 training complete
 1
   Cost on training data: 0.4864990500995526
   Accuracy on training data: 45648.0 / 50000
 3
   Cost on evaluation data: 0.5171438222818319
 4
   Accuracy on evaluation data: 9082.0 / 10000
 5
 6
   Epoch 1 training complete
7
   Cost on training data: 0.32762718180961936
8
   Accuracy on training data: 47155.0 / 50000
   Cost on evaluation data: 0.38299592178724307
10
   Accuracy on evaluation data: 9341.0 / 10000
11
12
   Epoch 2 training complete
13
   Cost on training data: 0.3086852753678556
14
   Accuracy on training data: 47275.0 / 50000
15
   Cost on evaluation data: 0.399539299143793
16
   Accuracy on evaluation data: 9345.0 / 10000
17
18
```

```
Epoch 3 training complete
19
   Cost on training data: 0.24972857496684978
20
   Accuracy on training data: 47834.0 / 50000
21
   Cost on evaluation data: 0.3585633140924723
22
   Accuracy on evaluation data: 9397.0 / 10000
23
24
   Epoch 4 training complete
25
   Cost on training data: 0.22723462892513618
26
   Accuracy on training data: 47965.0 / 50000
27
   Cost on evaluation data: 0.35218769857094817
28
   Accuracy on evaluation data: 9446.0 / 10000
29
30
   Epoch 5 training complete
31
   Cost on training data: 0.16733928319261923
32
33
   Accuracy on training data: 48490.0 / 50000
   Cost on evaluation data: 0.2993975071402639
34
   Accuracy on evaluation data: 9539.0 / 10000
35
36
   Epoch 6 training complete
37
   Cost on training data: 0.13263494238211188
38
   Accuracy on training data: 48820.0 / 50000
39
   Cost on evaluation data: 0.2848000556186426
40
   Accuracy on evaluation data: 9534.0 / 10000
41
42
   Epoch 7 training complete
43
   Cost on training data: 0.12413167242826531
44
   Accuracy on training data: 48926.0 / 50000
45
   Cost on evaluation data: 0.280473272887883
46
   Accuracy on evaluation data: 9561.0 / 10000
47
48
   Epoch 8 training complete
49
50
   Cost on training data: 0.10126083659446701
   Accuracy on training data: 49146.0 / 50000
51
   Cost on evaluation data: 0.2909174340393559
52
   Accuracy on evaluation data: 9552.0 / 10000
53
54
```

```
Epoch 9 training complete
55
   Cost on training data: 0.07950659594425181
56
   Accuracy on training data: 49331.0 / 50000
57
   Cost on evaluation data: 0.28240616318534334
58
   Accuracy on evaluation data: 9589.0 / 10000
59
60
   Epoch 10 training complete
61
   Cost on training data: 0.07638307314555179
62
   Accuracy on training data: 49361.0 / 50000
63
   Cost on evaluation data: 0.2893474599986748
64
   Accuracy on evaluation data: 9598.0 / 10000
65
66
   Epoch 11 training complete
67
   Cost on training data: 0.07294671256650084
68
   Accuracy on training data: 49367.0 / 50000
69
   Cost on evaluation data: 0.30407742756999756
70
   Accuracy on evaluation data: 9567.0 / 10000
71
72
   Epoch 12 training complete
73
   Cost on training data: 0.07402528046767626
74
   Accuracy on training data: 49372.0 / 50000
75
   Cost on evaluation data: 0.3076031010374318
76
   Accuracy on evaluation data: 9555.0 / 10000
77
78
   Epoch 13 training complete
79
   Cost on training data: 0.047381324957852036
80
   Accuracy on training data: 49610.0 / 50000
81
   Cost on evaluation data: 0.2991721910718146
82
   Accuracy on evaluation data: 9613.0 / 10000
83
84
   Epoch 14 training complete
85
86
   Cost on training data: 0.039001067374808
   Accuracy on training data: 49705.0 / 50000
87
   Cost on evaluation data: 0.28855569143429394
88
   Accuracy on evaluation data: 9589.0 / 10000
89
90
```

Epoch 15 training complete 91 Cost on training data: 0.02830440419954917 92 Accuracy on training data: 49800.0 / 50000 93 Cost on evaluation data: 0.28475461901351434 94 Accuracy on evaluation data: 9619.0 / 10000 95 96 Epoch 16 training complete 97 Cost on training data: 0.02440925100559722 98 Accuracy on training data: 49846.0 / 50000 99 Cost on evaluation data: 0.28767409965738916 100 Accuracy on evaluation data: 9626.0 / 10000 101 102 103 Epoch 17 training complete Cost on training data: 0.01862185593626093 104 Accuracy on training data: 49883.0 / 50000 105 Cost on evaluation data: 0.29037706748223857 106 Accuracy on evaluation data: 9645.0 / 10000 107 108 Epoch 18 training complete 109 110 Cost on training data: 0.013902421587521778 Accuracy on training data: 49932.0 / 50000 111 112 Cost on evaluation data: 0.28878484204108124 113 Accuracy on evaluation data: 9639.0 / 10000 114 115 Epoch 19 training complete 116 Cost on training data: 0.014437471805492598 Accuracy on training data: 49916.0 / 50000 117 Cost on evaluation data: 0.29415119177326415 118 Accuracy on evaluation data: 9635.0 / 10000 119