

# MLP Report

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## Overview

In the forward propagation, given input  $\mathbf{x} = [x_1, x_2, \dots, x_n]$ , the output of Softmax layer is  $\mathbf{y} = [y_1, y_2, \dots, y_n]$ , where we have:

$$y_i = \frac{e^{x_i}}{\sum_{j=1}^n e^{x_j}}$$

In the backward propagation, we've known  $\frac{\partial E}{\partial y_i}$ , and we want to calculate  $\frac{\partial E}{\partial x_i}$ . So we should use the chain rule:

$$\frac{\partial E}{\partial x_i} = \sum_{j=1}^n \frac{\partial E}{\partial y_j} \times \frac{\partial y_j}{\partial x_i}$$

If  $j \neq i$ , then we have:

$$\frac{\partial y_j}{\partial x_i} = \frac{\partial \frac{e^{x_j}}{\sum_{j=1}^n e^{x_j}}}{\partial x_i} = e^{x_i} \times -\frac{e^{x_j}}{(\sum_{j=1}^n e^{x_j})^2} = -y_i \times y_j$$

If  $j = i$ , then we have:

$$\frac{\partial y_j}{\partial x_i} = \frac{\partial \frac{e^{x_i}}{\sum_{j=1}^n e^{x_j}}}{\partial x_i} = \frac{(\sum_{j=1}^n e^{x_j}) \times e^{x_i} - e^{x_i} \times e^{x_i}}{(\sum_{j=1}^n e^{x_j})^2} = y_i \times (1 - y_i)$$

So we can simply use the result of  $j \neq i$ , then add  $y_i$  to the answer. Finally we have:

$$\frac{\partial E}{\partial x_i} = \sum_{j=1}^n \frac{\partial E}{\partial y_j} \times -y_i \times y_j + \frac{\partial E}{\partial y_i} \times y_i$$

This is the forward and backward propagation equation for Softmax layer.

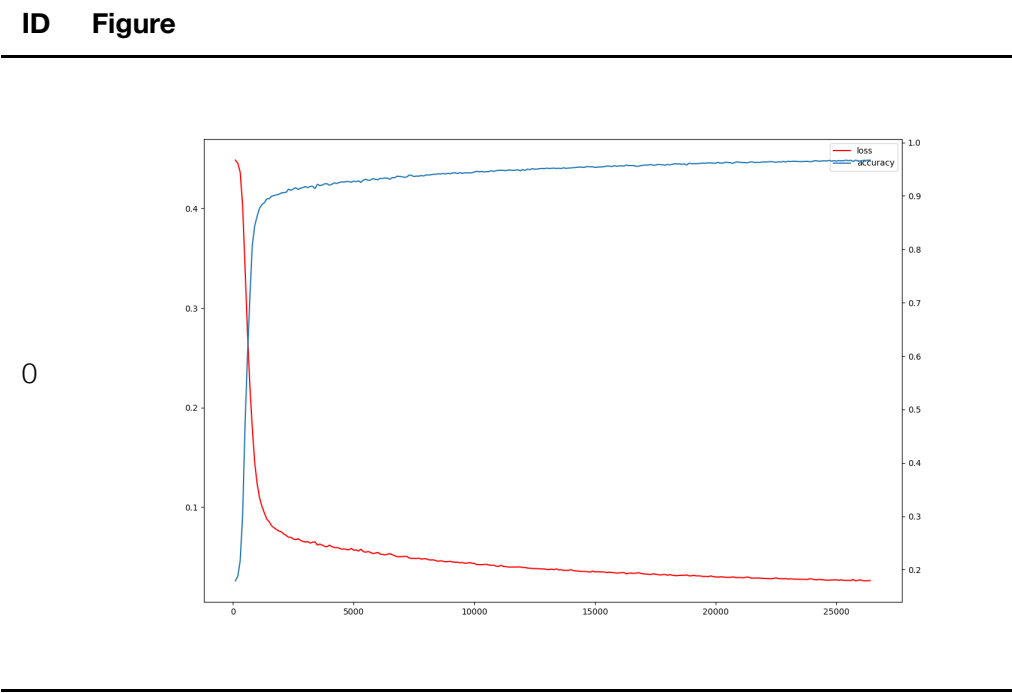
## Experiments

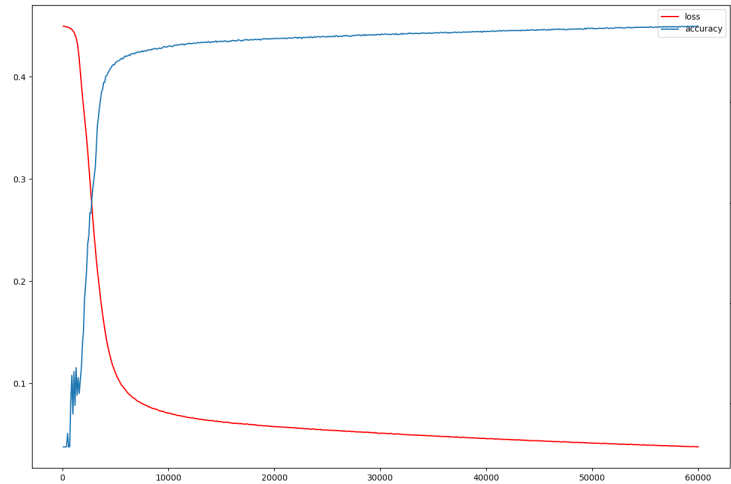
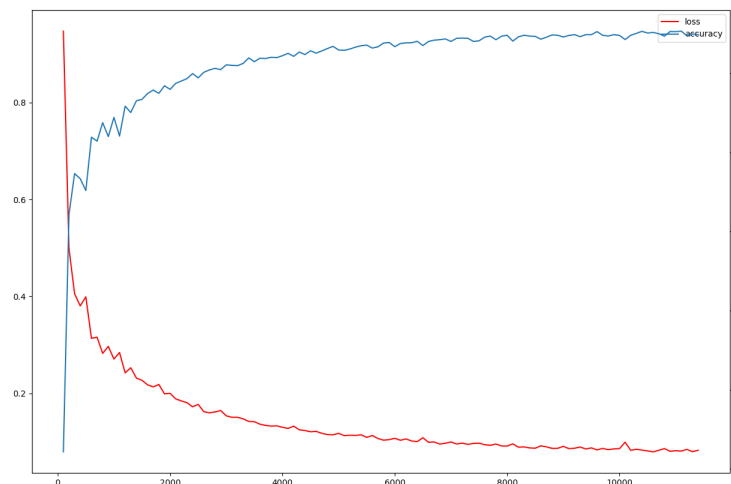
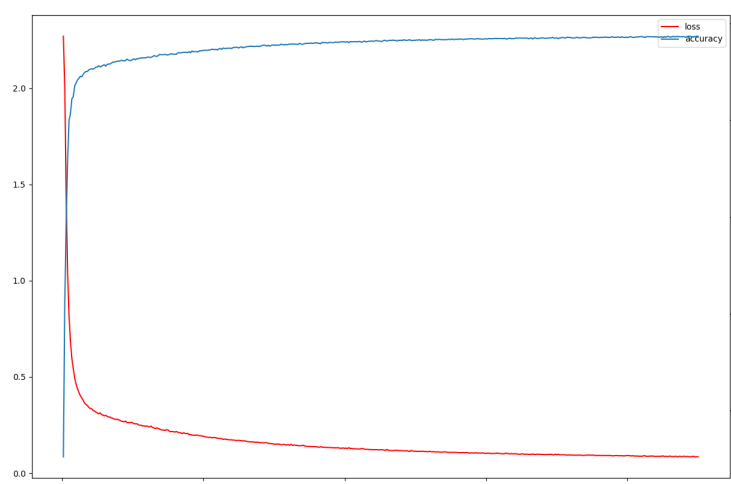
Basic Results:

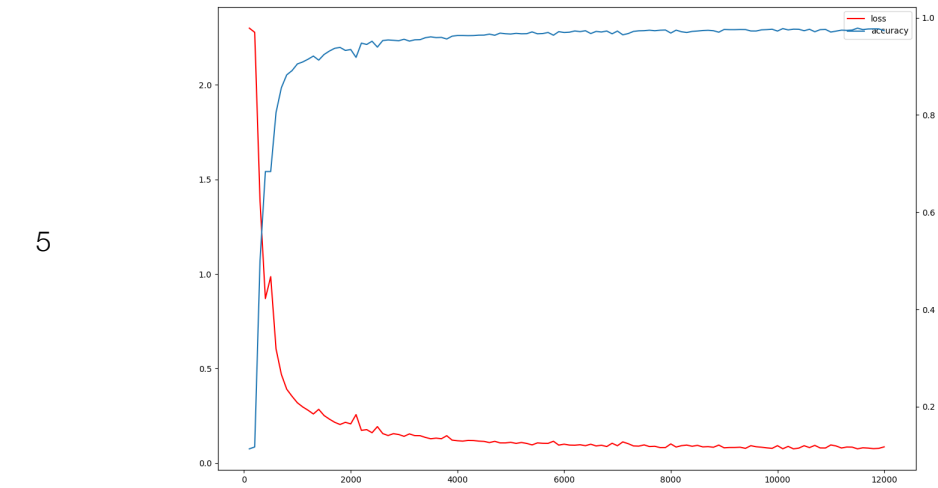
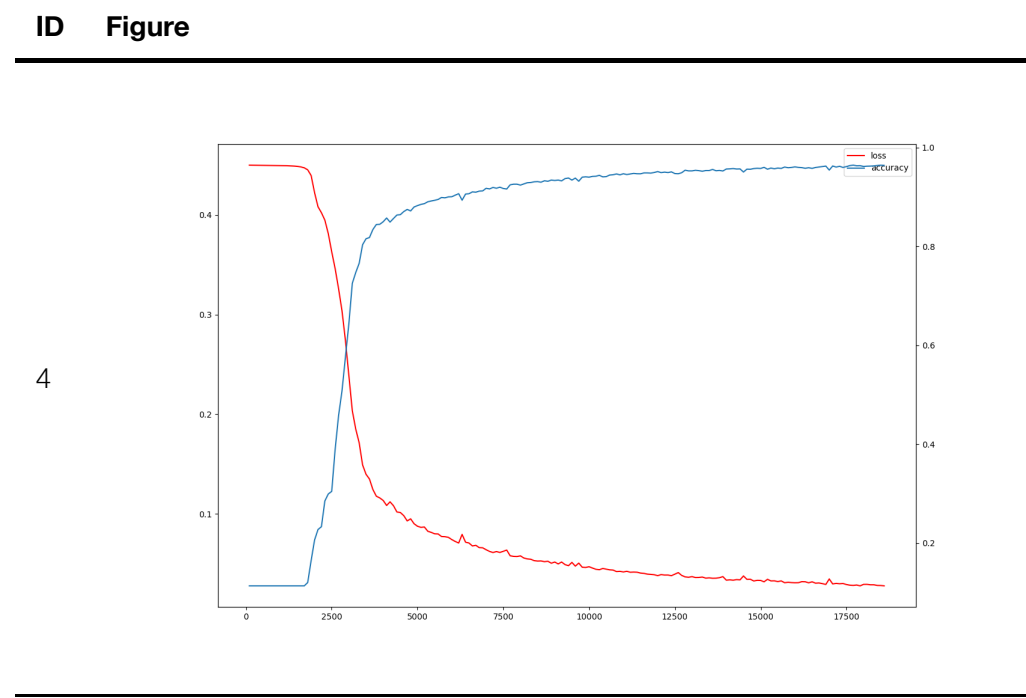
ID	Network Structure	Loss Function	Learning Rate	Train Time(s)	Best Accuracy(%)
0	Linear(784 × 128)->Relu->Linear(128 × 10)->Softmax	EuclideanLoss	0.1	179	96.7
1	Linear(784 × 128)->Sigmoid->Linear(128 × 10)->Softmax	EuclideanLoss	0.1	431	95.1

ID	Network Structure	Loss Function	Learning Rate	Train Time(s)	Best Accuracy(%)
2	Linear(784 × 128)->Relu->Linear(128 × 10)->Softmax	CrossEntropyLoss	0.1	84.3	97.6
3	Linear(784 × 128)->Sigmoid->Linear(128 × 10)->Softmax	CrossEntropyLoss	0.1	326	97.4
4	Linear(784 × 128)->Relu->Linear(128 × 10)->Softmax	EuclideanLoss	0.1	162	96.5
5	Linear(784 × 128)->Relu->Linear(128 × 10)->Softmax	CrossEntropyLoss	0.1	101	98.1

Loss-Accuracy Figures:



ID	Figure
1	
2	
3	



For higher resolution images, please visit directory [codes/images](#).