

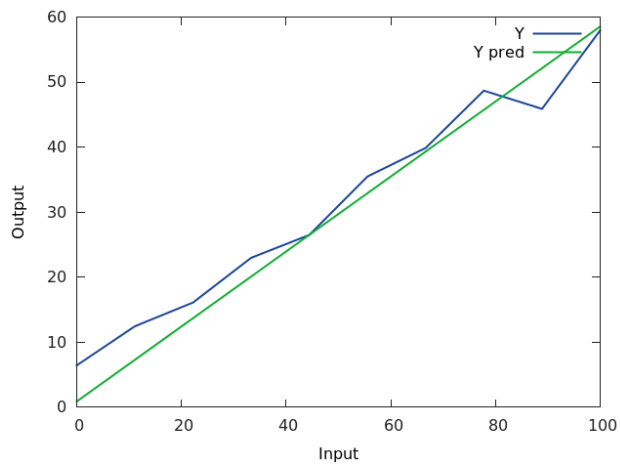
CS 8803 DL Assignment 1

Jingdao Chen

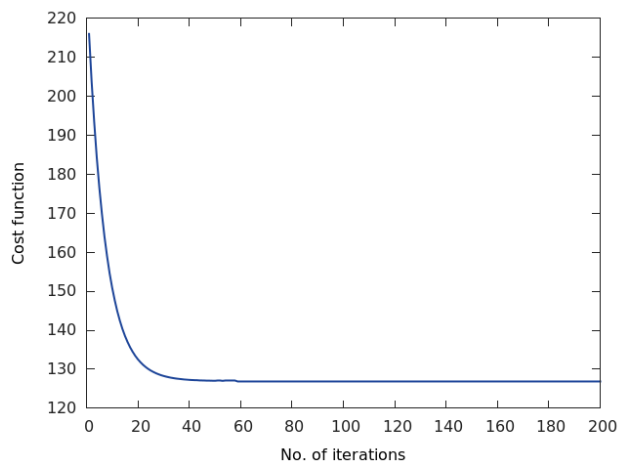
Jan 26

1 Problem 1

The plot of the estimated linear function is shown below:



The plot of cost function vs. no. of iterations is shown below:



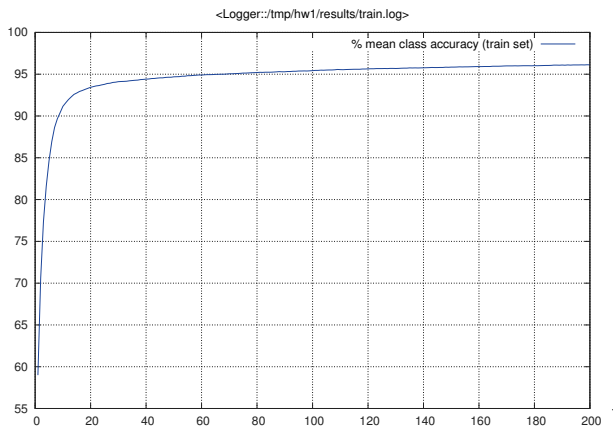
2 Problem 2

2.1 Linear Layer

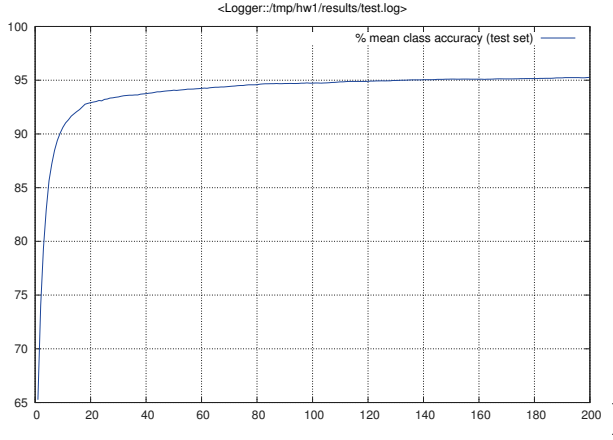
Let the input be x_0 and the output be x_1 . The linear layer is calculated as $x_1 = w_1^T x_0 + b_1$.

$$\begin{aligned}\frac{\partial E}{\partial w_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial w_1} \\ &= \frac{\partial E}{\partial x_1} x_0 \\ \frac{\partial E}{\partial b_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial b_1} \\ &= \frac{\partial E}{\partial x_1} \mathbf{1}\end{aligned}\tag{1}$$

The plot of training accuracy vs. no. of epoch for is shown below:



The plot of testing accuracy vs. no. of epoch for is shown below:



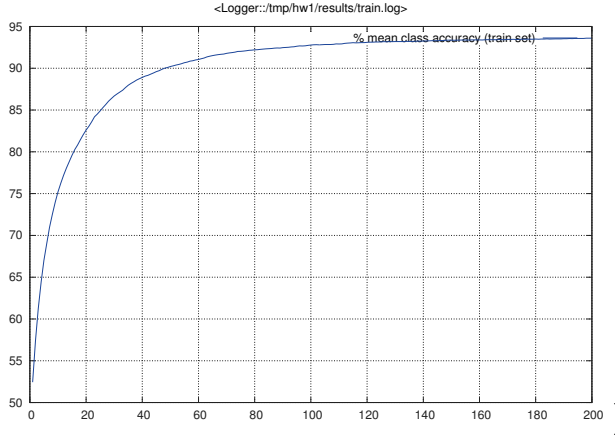
2.2 Linear-Sigmoid Layer

Let the input be x_0 and the output be x_2 . The linear layer is calculated as $x_1 = w_1^T x_0 + b_1$.

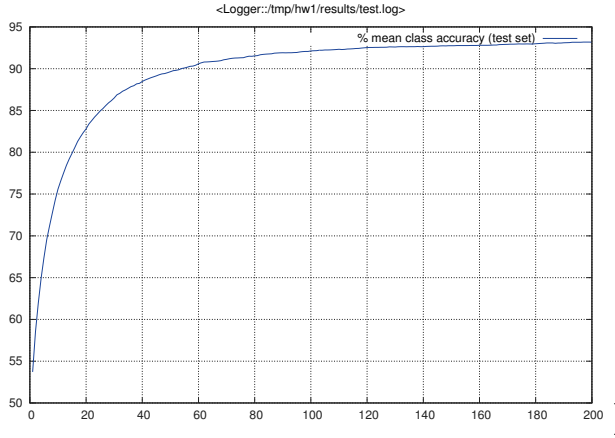
The sigmoid layer is calculated as $x_2 = \sigma(x_1)$.

$$\begin{aligned}
 \frac{\partial E}{\partial x_1} &= \frac{\partial E}{\partial x_2} \frac{\partial x_2}{\partial x_1} \\
 &= \frac{\partial E}{\partial x_2} (x_2)(1 - x_2) \\
 \frac{\partial E}{\partial w_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial w_1} \\
 &= \frac{\partial E}{\partial x_1} x_0 \\
 \frac{\partial E}{\partial b_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial b_1} \\
 &= \frac{\partial E}{\partial x_1} \mathbf{1}
 \end{aligned} \tag{2}$$

The plot of training accuracy vs. no. of epoch for is shown below:



The plot of testing accuracy vs. no. of epoch for is shown below:



2.3 Linear-Sigmoid-Linear Layer

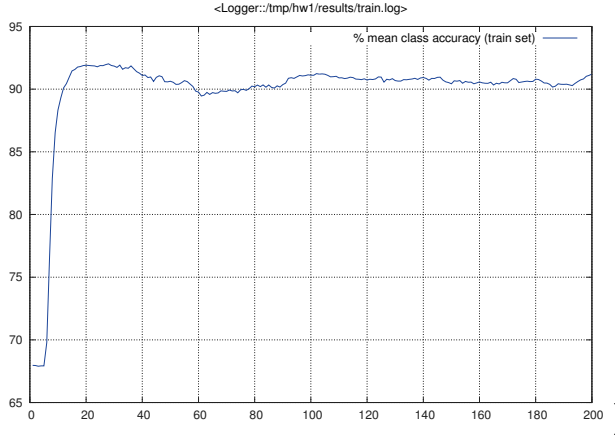
Let the input be x_0 and the output be x_3 . The first linear layer is calculated as $x_1 = w_1^T x_0 + b_1$.

The sigmoid layer is calculated as $x_2 = \sigma(x_1)$. The second linear layer is calculated as

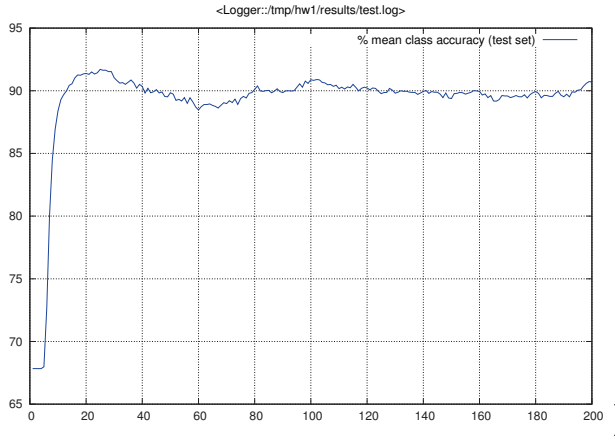
$$x_3 = w_2^T x_2 + b_2.$$

$$\begin{aligned}
\frac{\partial E}{\partial w_2} &= \frac{\partial E}{\partial x_3} \frac{\partial x_3}{\partial w_2} \\
&= \frac{\partial E}{\partial x_3} x_2 \\
\frac{\partial E}{\partial b_2} &= \frac{\partial E}{\partial x_3} \frac{\partial x_3}{\partial b_2} \\
&= \frac{\partial E}{\partial x_3} \mathbf{1} \\
\frac{\partial E}{\partial x_2} &= \frac{\partial E}{\partial x_3} \frac{\partial x_3}{\partial x_2} \\
&= \frac{\partial E}{\partial x_3} w_2 \\
\frac{\partial E}{\partial x_1} &= \frac{\partial E}{\partial x_2} \frac{\partial x_2}{\partial x_1} \\
&= \frac{\partial E}{\partial x_2} (x_2)(1 - x_2) \\
\frac{\partial E}{\partial w_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial w_1} \\
&= \frac{\partial E}{\partial x_1} x_0 \\
\frac{\partial E}{\partial b_1} &= \frac{\partial E}{\partial x_1} \frac{\partial x_1}{\partial b_1} \\
&= \frac{\partial E}{\partial x_1} \mathbf{1}
\end{aligned} \tag{3}$$

The plot of training accuracy vs. no. of epoch for is shown below:



The plot of testing accuracy vs. no. of epoch for is shown below:



3 Gradient check

One-sided finite difference approximation is calculated by:

$$\nabla J = \frac{J(\theta + \epsilon) - J(\theta)}{\epsilon} \quad (4)$$

Two-sided finite difference approximation is calculated by:

$$\nabla J = \frac{J(\theta + \epsilon) - J(\theta - \epsilon)}{2\epsilon} \quad (5)$$

The two-sided version is more accurate because taking the linear approximation of $J(\theta)$

on both sides of θ evens out the error.