Deep Learning 2021 - Homework 1

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Introduction

Logistic regression is a simple statistical model that uses a logistic function to model a binary variable. In this assignment, we were tasked to implement this model as a classifier for a simple 2-class subset of the MNIST digit dataset. As only 2 classes of these digits (3&6) are used, the classification results can be considered as binary where 3 and 6 are represented by 0 and 1.

Implementation

1. Initialization

This model is implemented using python. The given set of inputs are a list of images (784 pixels) and each image has its own label (either 3 or 6). The given images are split into train and test sets to be used in training and testing stages. After the required preprocessing, the images are one-hot encoded. Therefore, each pixel would have a binary value and is regarded as an input. In addition to these 784 inputs, we would add a bias input with the value of 1; hence, the shape of the model's input would be 785. Accordingly, we would randomly initialize a weight for each of these inputs.

2. Training

The training section includes two passes: forward and backward pass. During the forward pass, the model predicts a label by calculating the sigmoid of the dot product between the input vector x and the weights vector w as follows:

$$output = rac{1}{1 + e^{-(x \cdot w)}}$$

During the backward pass, the cross-entropy loss is used to find the gradient and accordingly update the weights as follows:

Cross-entropy loss
$$E(\theta) = -\frac{1}{N} \sum_{n=1}^{N} (t^{(n)} lnh(x^{(n)}) + (1 - t^{(n)}) ln(1 - h(x^{(n)})))$$

$$\operatorname{Gradient} \nabla E(\theta) = \frac{1}{N} \sum_{n} x^{(n)} (h(x^{(n)}) - t^{(n)})$$

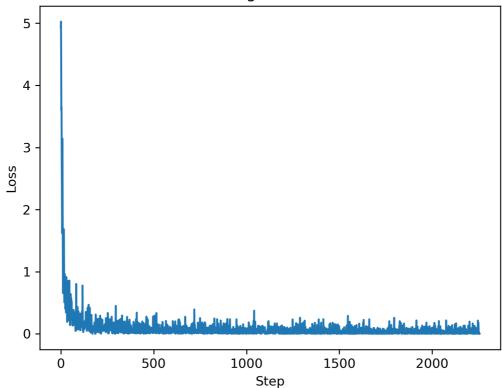
$$w_{new} = w_{old} - \text{learning rate } * \nabla E(\theta)$$

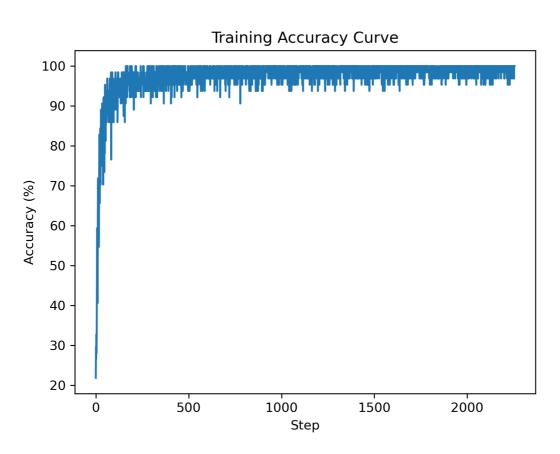
The used training parameters are provided respectively below:

learning rate = 0.1 number of epochs = 12 batch size = 64

The obtained training loss and accuracy curves are provided in the following figures.

Training Loss Curve





In addition, the recorded training and testing values are provided respectively below:

Training Accuracy = 97.36605164007092%

Testing Accuracy = 98.5416666666667%