

# Assignment 2 - DD2424 - One Layer Network

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## 1 Exercise 1

For this assignment, I decided to create a python object that would represent our classifier. I made this choice because it would simplify the creation of each different classifier and would allow me to easily test different configurations.

*How I checked my analytic gradient computations?*

After doing the math to compute the gradient and checked it with the given formula on the slides. I decided to implement the gradient computation in the python object. I then compared the results of the analytic gradient with the numerical gradient and the **PyTorch** gradient. I use the **relative error** to compare the gradients for every weights and biases. I get a relative error of  $10^{-14}$  for torch vs analytical. However, I get a relative error of  $10^{-4}$  for analytical vs numerical and torch vs numerical. The last relative error can be explained because of the numerical approximation of the gradient.

## 2 Exercise 2

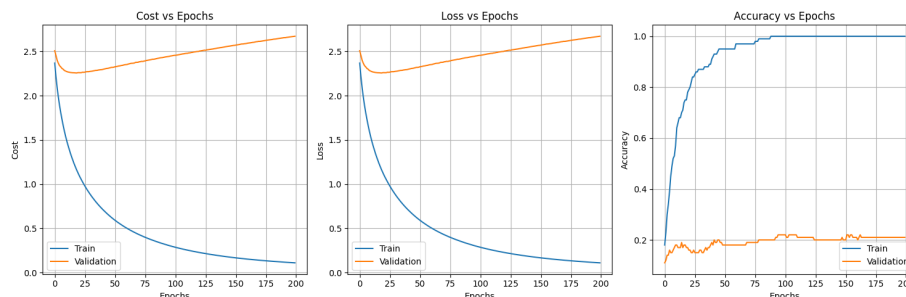


Figure 1: Small training with no regularization. It definitely overfits the data.

Here, just to see if the classifier works, I trained it on a small dataset with no regularization. As we can see in figure 1, the classifier overfits the data. The training loss goes to zero, but the validation loss increases. This is a clear sign of overfitting. I get these accuracies:

- Training accuracy: 100%
- Validation accuracy: 21%
- Test accuracy: 14%

## 3 Exercise 3

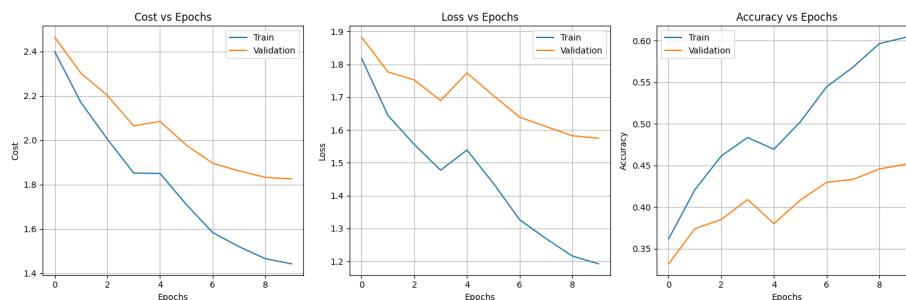


Figure 2: **Training curves (cost, loss, accuracy) for one cycle of training.** The hyper-parameter settings of the training algorithm are  $\eta_{\min} = 1e-5$ ,  $\eta_{\max} = 1e-1$ ,  $\lambda = .01$ ,  $n_s = 500$  and  $\text{batch size} = 100$ .

Now that the classifier works. I trained the classifier on the CIFAR-10 dataset with cyclical learning rates. Now, the learning rate is not constant but varies between a minimum and a maximum value.

## 4 Exercise 4

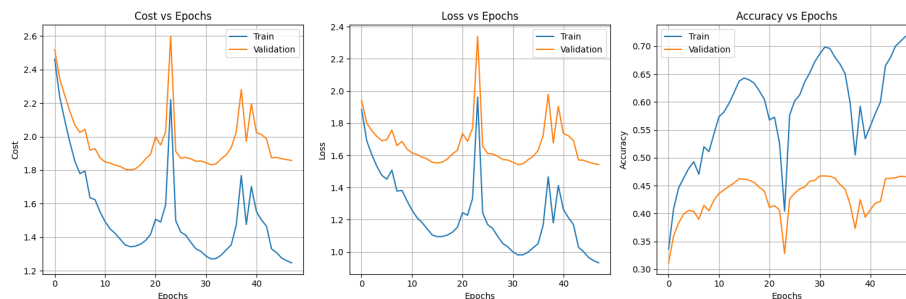


Figure 3: **Training curves (cost, loss, accuracy) for three cycles of training.** The hyper-parameter settings of the training algorithm are  $\eta_{\min} = 1e-5$ ,  $\eta_{\max} = 1e-1$ ,  $\lambda = .01$ ,  $n_s = 800$  and  $\text{batch size} = 100$ .

Now that the cyclical learning rate works, I trained the classifier on the CIFAR-10 dataset with three cycles of training. As we can see in figure 3, the training loss decreases and the validation loss decreases as well and we can see the cycles. I obtain these accuracies:

- Training accuracy: 71.78%
- Validation accuracy: 46.61%
- Test accuracy: 48.09%

### 4.1 Coarse to fine lambda search