

1 Optimize the Performance of the Network

In this part, I implement 3 optimization method.

- (a) Early stopping: Save the model with best validation accuracy, and stop training if there is no improvement on validation set for more than 10 epoches.
- (b) He Initialization: Set the standard deviation as $\sqrt{2/n_{in}}$
- (c) Do a more exhaustive search to find good value with He Initialization.
- (d) Decay the learning rate by 0.1 after every 8 epoches.

For the following experiments, these parameters are same.

lambda=0.0023292248102687557, eta=0.017453577972249945, momentum=0.95, n_batch=100

Optimization method: He Initialization + Exhaustive Random Search

n_epoch=30

Accuracy on test set: 0.5151

Optimization method: Early Stopping + Exhaustive Random Search

N_no_improvement=10

Accuracy on test set: 0.5179

Optimization method: Learning Rate Decay by 0.1 + Exhaustive Random Search

n_epoch=30, Decay_cycle=8

Accuracy on test set: 0.5200

Optimization method: All Methods

Decay_cycle=8, N_no_improvement=10

Accuracy on test set: 0.5241

Optimization method: None

Accuracy on test set: 0.5127

It indicates that the Learning Rate Decay improves the network most, and their combination gives the most gain among all experiments.

2 Leaky ReLU Activation

The activation function I use here is leaky ReLU activation function.

$$\varphi(x) = \begin{cases} x(x \geq 0) \\ 0.01x(x \leq 0) \end{cases}$$

I tests the leaky ReLU function on the experiment with no optimization method and the experiment with all optimization method mentioned above. The results are listed below.

Optimization method: All Methods + Leaky ReLU Activation

Decay_cycle=8, N_no_improvement=10

Accuracy on test set: 0.5246

Optimization method: None + Leaky ReLU Activation

Accuracy on test set: 0.5210

It shows that leaky ReLU activation actually leads to some improvement for the accuracy on test set compared to simple ReLU activation.