# **MNIST Digits Classification with PyTorch**

#### Introduction

In this homework, I compare the MLP and CNN models with and without Batch Normalization.

# **Hyperparameter Selection**

In theory, to select the right hyperparameters, a full exhaustive search that iterates through every combination should be done with a **validation dataset**. However, this is done mainly on an ad hoc basis, where the parameters work well enough in combination.

Table1: Hyperparameters used for all models

| Loss          | CrossEntropy |
|---------------|--------------|
| Learning rate | 0.01         |
| Batch size    | 100          |
| Weight decay  | 0            |
| Momentum      | 0.9          |
| Max epoch     | 50           |

# **Results Summary**

Table2: Network details

| Models        | CNN with BN    | CNN without BN | MLP with BN    | MLP without BN |
|---------------|----------------|----------------|----------------|----------------|
| Architecture  | Conv-BN-ReLU-  | Conv-ReLU-     | Linear-BN-     | Linear-ReLU-   |
|               | AvgPool-Conv-  | AvgPool-Conv-  | ReLU-Linear-   | Linear-ReLU-   |
|               | BN-ReLU-       | ReLU-AvgPool-  | BN-ReLU-Linear | Linear         |
|               | AvgPool-       | Reshape-Linear |                |                |
|               | Reshape-Linear |                |                |                |
| >98% test acc | 9 epochs       | 10 epochs      | 3 epochs       | 5 epochs       |
| Time per      | 15s            | 13s            | 13s            | 13s            |
| epoch         |                |                |                |                |
| Time to reach | ~130s          | ~130s          | ~40s           | ~65s           |
| >98% test acc |                |                |                |                |
| Training loss | 10-2           | 10-2           | 10-3           | 10-3           |
| Test loss     | 10-3           | 10-3           | 10-3           | 10-3           |
| Training acc  | 98.50%         | 98.87%         | 100%           | 100%           |
| Test acc      | 98.51%         | 98.46%         | 98.30%         | 98.27%         |

#### **Comments and Discussion**

**Training time & Convergence:** 

CNN and MLP take about the same time per epoch. MLP converges (defined as >% test acc) quicker. Adding BatchNormalization layers does not affect run time appreciably.

#### No. of Parameters:

MLP has more parameters, CNN does parameter sharing and tying. BatchNormalization does not affect no. of parameters.

#### **Accuracy:**

CNN achieves slightly better acc. BatchNormalization does not affect this.

#### Overfitting:

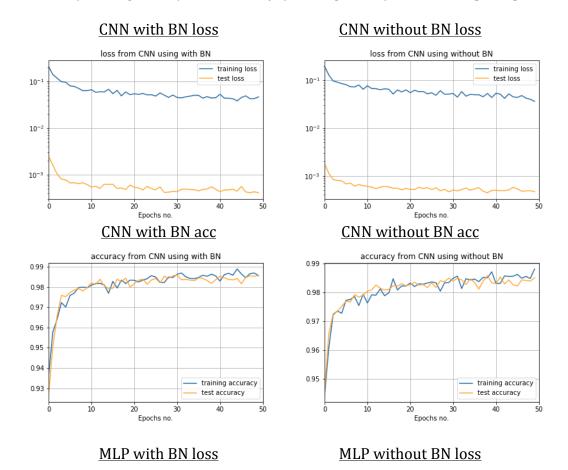
Overfitting plagues MLP but not so much for CNN. BatchNormalization does not affect this.

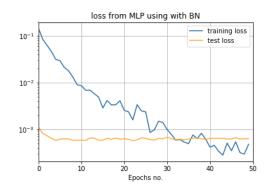
#### Effects of BN:

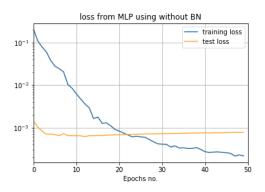
The use of Batch Normalization is to be more **robust** to bad parameter initialization. Here, we don't see much difference with or without BN. We would perhaps see the effects of BN if initialization is badly done.

#### **Plots**

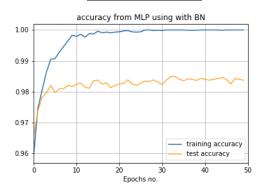
The loss (training & test) and accuracy (training & test) of each setup are plotted.



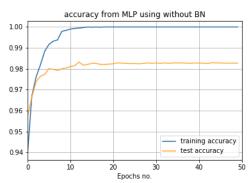








## MLP without BN acc



### **Reference:**

- \* [Deep Learning](http://www.deeplearningbook.org/)
- \* [CS231n](http://cs231n.github.io/neural-networks-2/)
- \* [Ioffe, S., Szegedy, C. (2015). Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift] (https://arxiv.org/abs/1502.03167)