

# Programming Assignment 1

Neural Network design

CSE303: Introduction to Deep Learning

# Objective

## 1. 3-layer Neural Network for Classification

1. without the deep learning framework (only python)
2. using a deep learning framework (e.g. pytorch, tensorflow)

## 2. 3-layer Convolution Neural Network for Classification

1. without the deep learning framework (only python)
2. using a deep learning framework (e.g. pytorch, tensorflow)

# Overall steps

1. Prepare the training and test datasets (MNIST)
2. (For NN) Build a 3-layer Neural Network
  1. Implement a sub-modules (Linear layer, ReLU)
  2. Implement functions (SoftMax, CE loss)
  3. Build a 3-layer NN
3. (For NN) Implement training pipeline & train NN
4. (For NN) Test the 3-layer NN, draw figures  
& Compare results from 1-(1) only python and 1-(2) using frameworks
5. (For CNN) Build a 3-layer CNN & train CNN
  1. Design a Conv layer, Pooling layer & their backpropagation
  2. Build a 3-layer CNN by replacing Linear layer to Conv layer
  3. Train 3-layer CNN
6. (For CNN) Test the 3-layer CNN, draw figures, then compare to NN  
& Compare results from 2-(1) only python and 2-(2) using frameworks

# 1. Prepare training/test dataset

## 1. Download MNIST datasets & data loader (see uploaded file)

1. Download link: <http://yann.lecun.com/exdb/mnist/>
  1. [train-images-idx3-ubyte.gz](#): training set images (9912422 bytes)  
[train-labels-idx1-ubyte.gz](#): training set labels (28881 bytes)  
[t10k-images-idx3-ubyte.gz](#): test set images (1648877 bytes)  
[t10k-labels-idx1-ubyte.gz](#): test set labels (4542 bytes)

## 2. Prepare the datasets for training

1. Ex) normalization



## 2. Design 3-layer Neural Network(NN)

### 1. Design a sub-modules & their backpropagation

1. Linear layer
2. ReLU

### 2. Design functions & their derivatives

1. SoftMax
2. Cross-entropy loss

### 3. Design 3-layer Neural Network(NN)

1. Sequence: Input – Linear-ReLU – Linear-ReLU – Linear-SoftMax  
(The input and output size of NN: input 28x28, output 10)

### **3. Implement training pipeline, train NN**

#### **1. Implement stochastic gradient descent (SGD)**

#### **2. Training pipeline**

1. Initialize the model parameters
2. Implement and do forward propagation
3. Implement and compute the cross-entropy loss
4. Implement and do backward propagation
5. Implement and update model parameter using SGD

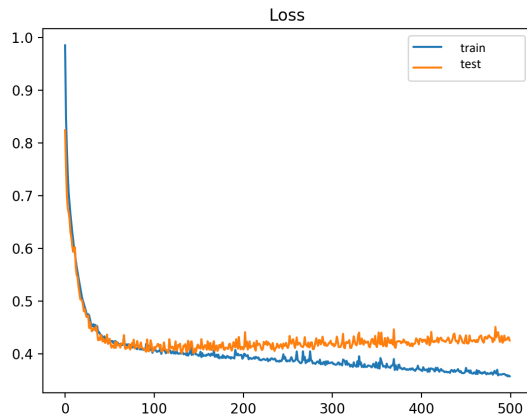
#### **3. Train a 3-layer NN**

## 4. Test the CNN, draw figures

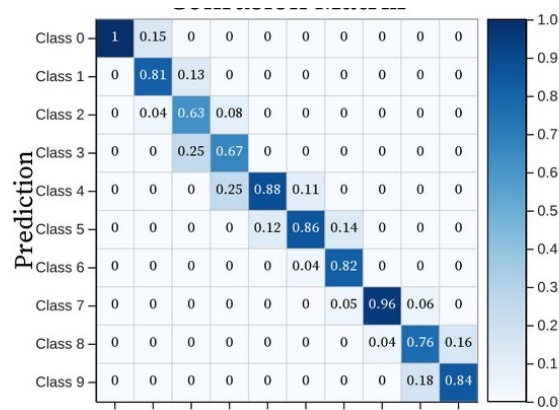
Draw all the output of NN, then write them in a report

1. Show training Loss graph (Train & test set)
2. Show 10x10 confusion matrix (the probability matrix of classification)
3. Show top 3 scored images with probability (for each class)

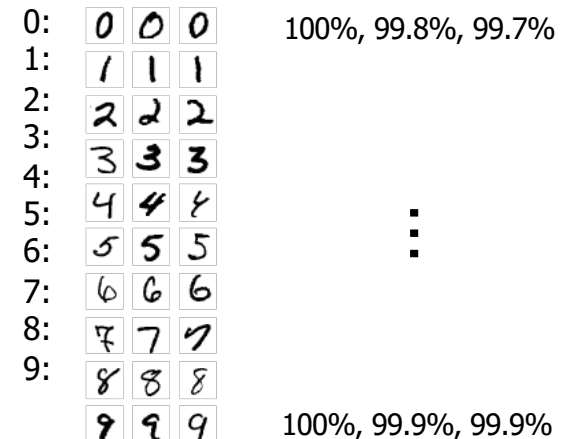
### 1. Loss graph



### 2. Confusion matrix



### 3. Top-3 images with probability



## 5. Build a CNN by replacing Linear layer to Conv layer

### 1. Design a sub-modules & their backpropagation

1. Conv layer
2. Max Pooling

### 2. Design CNN

1. Sequence: Input – Conv-ReLU-MaxPooling – Conv-ReLU-MaxPooling – Linear-SoftMax  
(The input and output size of CNN: input 28x28, output 10)

### 3. Train CNN

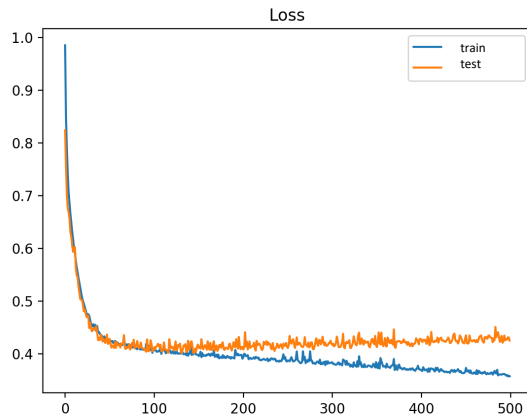


# 6. Test the CNN, draw figures

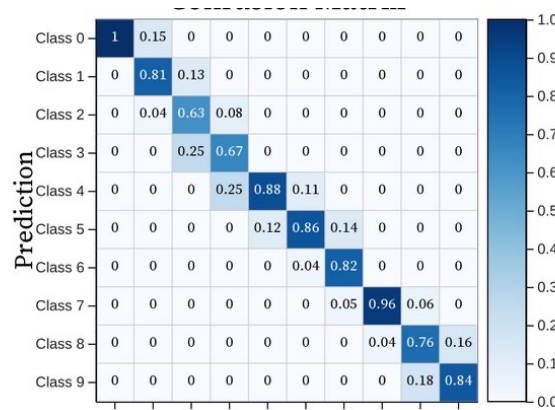
Draw all the same figures and compare with the output of NN

1. Show training Loss graph (Train & test set)
2. Show 10x10 confusion matrix (the probability matrix of classification)
3. Show top 3 scored images with probability (for each class)

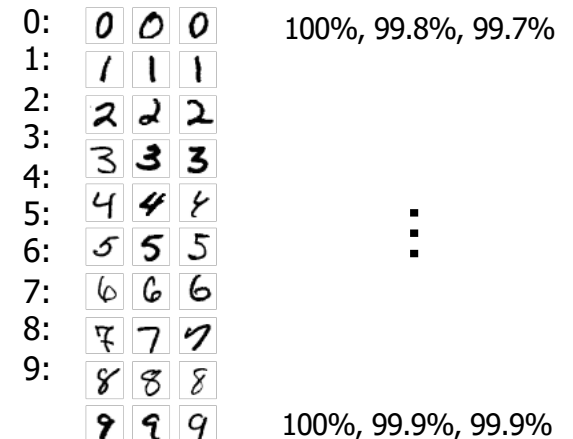
1. Loss graph



2. Confusion matrix



3. Top-3 images with probability



# Submission

**Due : Oct 25, 11:59PM**

**To : lms.dgist.ac.kr**

## **1. Submit zip file including (1) Source code, (2) PDF file(report)**

1. File name: PA1\_studentID\_name.zip (PA1\_202412345\_이경민.zip)
2. Training & Test datasets should be excluded (only code and report)

## **2. Report should include the results and results comparisons:**

1. Results from all networks (1) NN with python, (2) NN with DL framework, (3) CNN with python, and (4) CNN with DL framework should be attached.
2. Results mean:
  - (a) Training Loss graph
  - (b) 10x10 Confusion Matrix
  - (c) Top 3 score images (all classes)

\* Report does not need to include your understanding (Results are important)

## **3. Final credit: 30% of total grade**

– TA: 이경민 (kyoungmin@dgist.ac.kr)

# Notice

**Due : Oct 25, 11:59PM**

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## 1. Library

1. For the PA without the deep learning framework, you cannot use any libraies from Tensorflow, Pytorch, etc, but **you can use Numpy or other libraries**.

## 2. Delayed submission

1. 25% score will be **degraded** every 1-day delay & after 3 days delayed, you will get 10% of the total score & after 1 week delayed, 0%.  
\* 100% → 75% (1day) → 50% (2days) → 25% (3days) → 10% (3~7days) → 0% (> 7days)

## 3. Plagiarism

1. **No grade** for copied codes (from friends and the internet)
2. You can refer to sources from the internet, but do not copy and paste.

## 4. Partial credit

1. Even though you are not successfully designing the network and obtained reasonable results, please send your code.
2. **There will be partial credit** for each module implementation.