

HW-3 : Neural Network

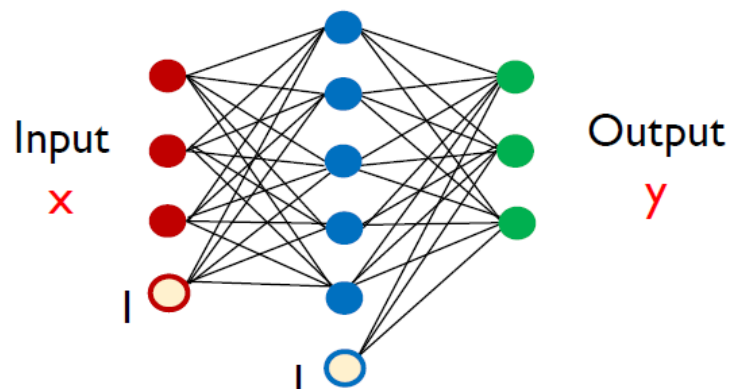
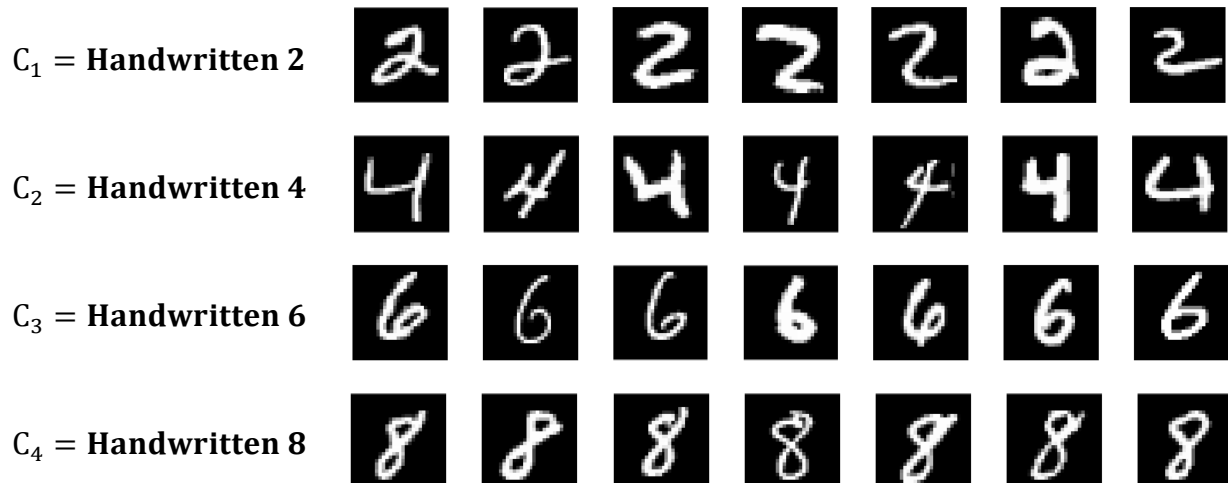
Deadlines: **2016.05.10-23:59:59**

In this homework, you are asked to build neural network models for the classification problem of the hand-written numbers x . Before doing so, the following layers need to be implemented as a Matlab function respectively:

- Inner-product layer
- Activation layer
- Softmax layer

Besides, it is required to use different types of the activation function for the tasks:

- Sigmoid function
- Rectified function



◆ Data

Input data are the extracted information from the picture of the hand-written numbers (MNIST Database). There are two data files offered: **Training_data_hw3.mat** file and one **Test_data1_hw3.mat** file.

Training_data_hw3.mat:

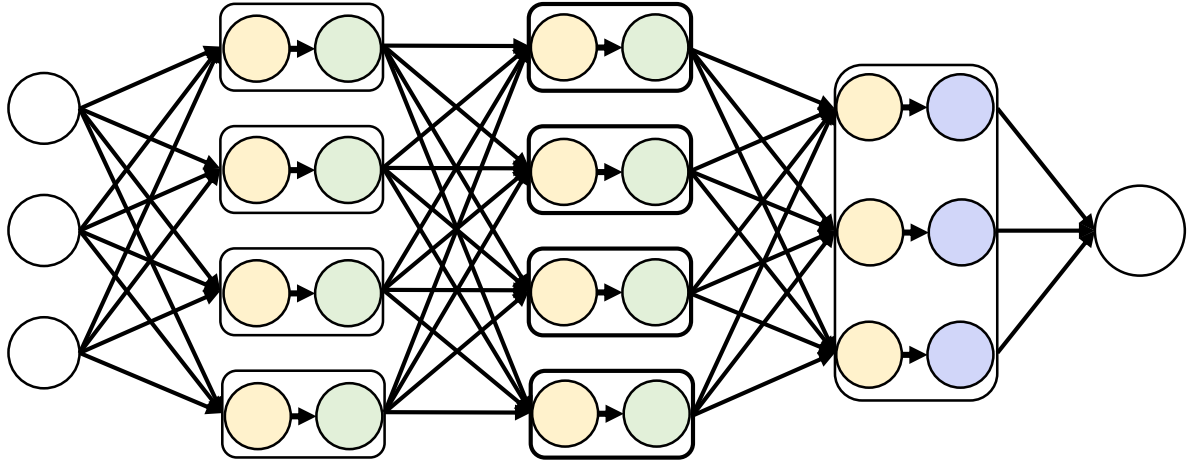
- **X_train** is a 4000x784 matrix. Every row corresponds to a 28x28 gray-scale image.
- **T_train** is a 4000x4 matrix, which records the target values of the training samples.

Test_data2_hw3.mat:

- **X_test** is a 2000x784 matrix. Every row corresponds to a 28x28 gray-scale image.

◆ Models

A simple neural network can be decomposed into the input layer, hidden layer(s) and output layer. A hidden layer is usually composed of an **inner-product layer** and an **activation layer**. Likewise, an output layer normally contains an **inner-product layer** and a **softmax layer**.



Let the m -by-1 vector \mathbf{x} and n -by-1 vector $\mathbf{y}(\mathbf{x})$ be the input and output of a layer respectively, the forward-propagation and back-propagation can be inferred as bellow:

● Inner-product layer

✧ Forward-propagation

$$\mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} \mathbf{w}_1^T \mathbf{x} + b_1 \\ \vdots \\ \mathbf{w}_n^T \mathbf{x} + b_n \end{bmatrix} = [\mathbf{w}_1 \dots \mathbf{w}_n] \cdot \mathbf{x} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \mathbf{W}^T \mathbf{x} + \mathbf{b}$$

✧ **Back-propagation**

$$\nabla_x E = \begin{bmatrix} \frac{\partial E}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial x_m} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} + \dots + \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_m} + \dots + \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_m} \end{bmatrix} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \dots & \frac{\partial y_n}{\partial x_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_1}{\partial x_m} & \dots & \frac{\partial y_n}{\partial x_m} \end{bmatrix} \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} = \mathbf{A} \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix}$$

$$\nabla_w E = \begin{bmatrix} \frac{\partial E}{\partial w_{11}} & \dots & \frac{\partial E}{\partial w_{n1}} \\ \vdots & \ddots & \vdots \\ \frac{\partial E}{\partial w_{1m}} & \dots & \frac{\partial E}{\partial w_{nm}} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial w_{11}} & \dots & \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial w_{n1}} \\ \vdots & \ddots & \vdots \\ \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial w_{1m}} & \dots & \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial w_{nm}} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{B}$$

$$\nabla_b E = \begin{bmatrix} \frac{\partial E}{\partial b_1} \\ \vdots \\ \frac{\partial E}{\partial b_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial b_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial b_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{C}$$

Please find the matrices \mathbf{A} , \mathbf{B} and \mathbf{C} by yourself.

● **Activation layer – Sigmoid function**

✧ **Forward-propagation**

$$\mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} 1/(1 + e^{-x_1}) \\ \vdots \\ 1/(1 + e^{-x_n}) \end{bmatrix}$$

✧ **Back-propagation**

$$\nabla_x E = \begin{bmatrix} \frac{\partial E}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{D}$$

Please find the matrix \mathbf{D} by yourself.

● Activation layer – Rectified function

✧ Forward-propagation

$$\mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} [x_1 > 0] \cdot x_1 \\ \vdots \\ [x_n > 0] \cdot x_n \end{bmatrix}$$

✧ Back-propagation

$$\nabla_{\mathbf{x}} E = \begin{bmatrix} \frac{\partial E}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{E}$$

Please find the matrix \mathbf{E} by yourself.

● Softmax layer

✧ Forward-propagation

$$\mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \frac{1}{\sum_{i=1}^n e^{x_i}} \begin{bmatrix} e^{x_1} \\ \vdots \\ e^{x_n} \end{bmatrix}$$

✧ Back-propagation

$$\nabla_{\mathbf{x}} E(\mathbf{y}, \mathbf{t}) = \begin{bmatrix} \frac{\partial E}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_n} \end{bmatrix} = \begin{bmatrix} y_1 - t_1 \\ \vdots \\ y_n - t_n \end{bmatrix} = \mathbf{y} - \mathbf{t}$$

There is only the inner-product layer containing parameters which need to be optimized.
Please follow the **gradient descent** approach to iteratively update the parameters.

$$\mathbf{W}^{new} = \mathbf{W}^{old} - \eta \nabla_{\mathbf{W}} E$$

$$\mathbf{b}^{new} = \mathbf{b}^{old} - \eta \nabla_{\mathbf{b}} E$$

◆ Tasks

1. Layers as Matlab Functions

Implement the forward-propagation and back-propagation of every layer into Matlab functions **(LayerName)_ForProp(.)** and **(LayerName)_BackProp(.)**. You are suggested to modify the templates attached in the HW package to accomplish the task.

2. One-hidden-layer Neural Network

Based on the Matlab functions created in the task 1, you are able to build **one-hidden-layer NNs**. Please use such models to deal with the MNIST classification problem. (Choose the **sigmoid function** as the activation function.) Predict the 2000x4 target matrix **T_test** given the input data **X_test** and save it in the file **OneHidNN_T_test.mat**.

3. Two-hidden-layer Neural Network

One step further, you are asked to build **two-hidden-layer NNs**. Please use such models to deal with the MNIST classification problem. (Choose the **rectified function** as the activation function.) Predict the 2000x4 target matrix **T_test** given the input data **X_test** and save it in the file **TwoHidNN_T_test.mat**.

What should be uploaded?

- ☐ Your Matlab source code with comments.
- ☐ The **OneHidNN_T_test.mat** file which contains **T_test**.
- ☐ The **TwoHidNN_T_test.mat** file which contains **T_test**.
- ☐ The **ReadMe.txt** file which describes how to run your program.
- ☐ Your **report** in the format of .pdf or .doc.

Reminders:

- ☐ There won't be a need for demonstration.
- ☐ Please make sure your source code can be compiled by **Matlab**.
- ☐ **DO NOT COPY!!!** (懶人包、考古題亦同)