

## HW-3 : Neural Network

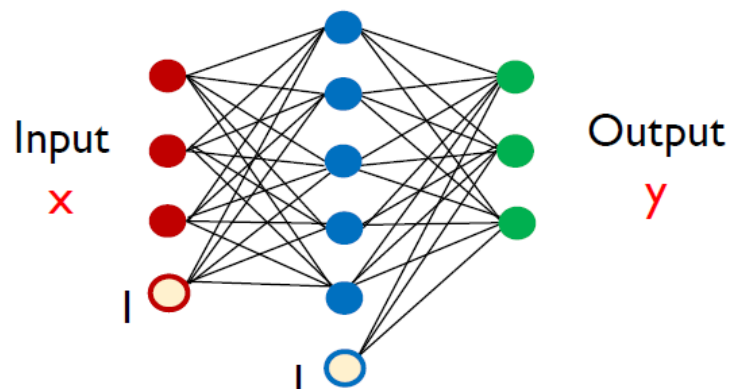
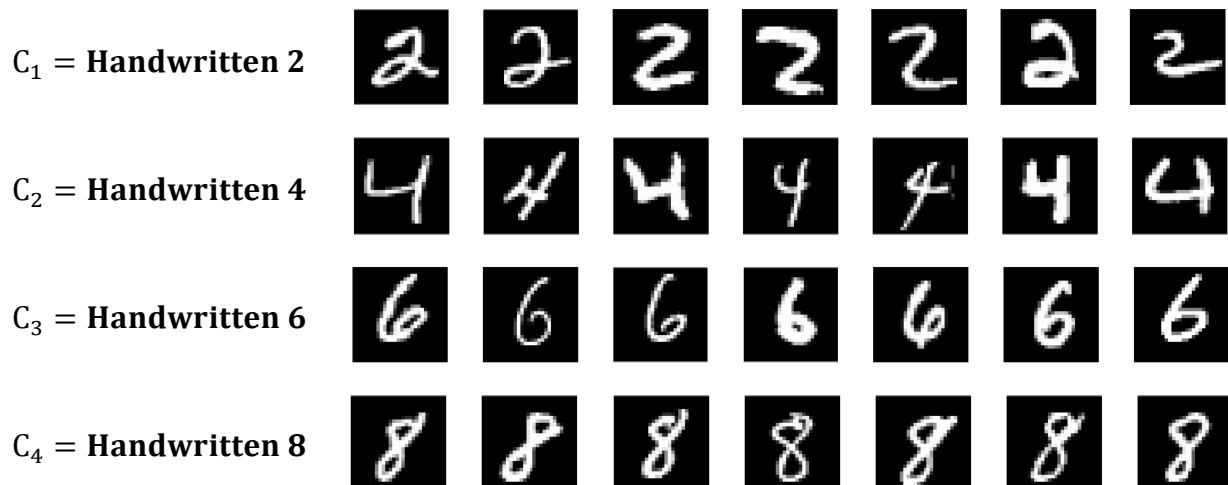
Deadlines: **2016.05.10-23:59:59**

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:

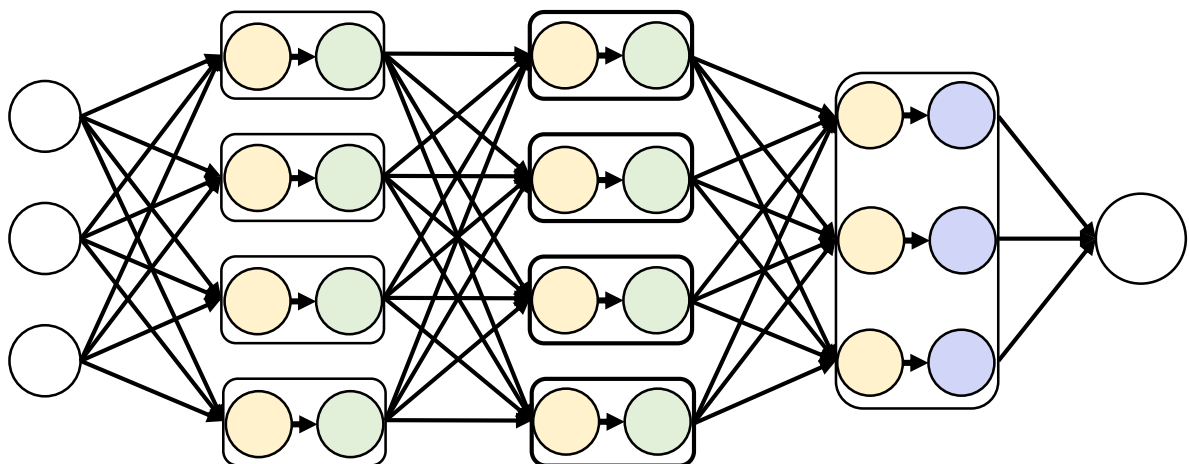
- **X\_train** is a 4000x784 matrix. Every row corresponds to a 28x28 gray-scale image.
- **T\_train** is a 4000x4 matrix, which records the target values of the training samples.

### 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

$$\begin{aligned}
 & \begin{matrix} x_1 \\ x_2 \end{matrix} \begin{matrix} \textcircled{1} \\ \textcircled{2} \end{matrix} \begin{matrix} y_1 \\ y_2 \end{matrix} & \begin{matrix} y_1 = x_1 w_1 + x_2 w_2 + b \\ y_2 = x_1 w_3 + x_2 w_4 + b \end{matrix} \\
 & \mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} \mathbf{w}_1^T \mathbf{x} + b_1 \\ \vdots \\ \mathbf{w}_n^T \mathbf{x} + b_n \end{bmatrix} = [\mathbf{w}_1 \dots \mathbf{w}_n] \cdot \mathbf{x} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \mathbf{W}^T \mathbf{x} + \mathbf{b} \\
 & \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \mathbf{w}_1 = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}, \mathbf{w}_2 = \begin{bmatrix} w_3 \\ w_4 \end{bmatrix} \Rightarrow \underbrace{\begin{bmatrix} w_1 & w_2 \\ w_3 & w_4 \end{bmatrix}}_{\mathbf{W}^T} \cdot \underbrace{\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}_{\mathbf{x}} + \underbrace{\begin{bmatrix} b_1 \\ b_2 \end{bmatrix}}_{\mathbf{b}}
 \end{aligned}$$

$\mathbf{W}^T = \begin{bmatrix} w_1^T \\ w_2^T \end{bmatrix}, \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$

$$x_1, x_2 \rightarrow y_1, y_2$$

對 loss 來說,  $y_1, y_2$  都有影響  $\rightarrow$  要 sum 起來

$$y_1 = w_{11}x_1 + w_{12}x_2$$

$$y_2 = w_{21}x_1 + w_{22}x_2$$

if  $x$  有 2 個  $y$  有 2 個

$$A = \begin{bmatrix} w_a & w_b \\ w_c & w_d \end{bmatrix} = W^T$$

### Back-propagation

這因的  $y$  是經過  $w, b$  的  $z$

$$\nabla_x E = \begin{bmatrix} \frac{\partial E}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial x_m} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} + \dots + \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_1} \\ \vdots \\ \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_m} + \dots + \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial x_m} \end{bmatrix} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \dots & \frac{\partial y_n}{\partial x_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_1}{\partial x_m} & \dots & \frac{\partial y_n}{\partial x_m} \end{bmatrix} \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} = \mathbf{A} \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \quad (2.1)$$

$$\nabla_w E = \begin{bmatrix} \frac{\partial E}{\partial w_{11a}} & \dots & \frac{\partial E}{\partial w_{n1b}} \\ \vdots & \ddots & \vdots \\ \frac{\partial E}{\partial w_{1mc}} & \dots & \frac{\partial E}{\partial w_{nm}} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial w_{1a}} & \dots & \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial w_{1a}} \\ \vdots & \ddots & \vdots \\ \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial w_{1m}} & \dots & \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial w_{1m}} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{B} \quad (2.2)$$

$$\nabla_b E = \begin{bmatrix} \frac{\partial E}{\partial b_1} \\ \vdots \\ \frac{\partial E}{\partial b_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial b_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \frac{\partial y_n}{\partial b_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial E}{\partial y_1} \\ \vdots \\ \frac{\partial E}{\partial y_n} \end{bmatrix} \mathbf{C} \quad (2.3)$$

$$\mathbf{B} = \begin{bmatrix} x_1 & x_2 \end{bmatrix}$$

$$\mathbf{C} = \begin{bmatrix} 1 & 1 \end{bmatrix}$$

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

## ● Activation layer – Sigmoid function

### Forward-propagation

$$z_1 \quad \sigma(z_1) = \frac{1}{1 + \exp(-z_1)}$$

$$z_2 \quad \sigma(z_2) = \frac{1}{1 + \exp(-z_2)}$$

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

$$y_1 = \frac{1}{1 + \exp(-x_1)}$$

### Back-propagation

這因的  $y$  是經過 sigmoid 的值  $a$  or  $g$

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

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

$$\mathbf{D} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & 0 \\ 0 & \frac{\partial y_2}{\partial x_2} \end{bmatrix} = \begin{bmatrix} \sigma(x_1)(1 - \sigma(x_1)) & 0 \\ 0 & \sigma(x_2)(1 - \sigma(x_2)) \end{bmatrix}$$

## ● Activation layer – Rectified function (操作)

✧ Forward-propagation

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

✧ Back-propagation

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

Please find the matrix **E** by yourself.

## ● Softmax layer

✧ Forward-propagation

$$\begin{matrix} x_1 & e^{x_1} & y_1 = \frac{e^{x_1}}{\text{sum}} \\ x_2 & \Rightarrow e^{x_2} & y_2 = \frac{e^{x_2}}{\text{sum}} \\ x_3 & e^{x_3} & y_3 = \frac{e^{x_3}}{\text{sum}} \end{matrix} \quad \text{loss} = (y_1 - t_1)^2 + (y_2 - t_2)^2 + (y_3 - t_3)^2$$

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

✧ Back-propagation

不知道啥只有一项

大就是 [1 0 0] 的导数

$$\frac{\partial E}{\partial x_1} = \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} + \frac{\partial E}{\partial y_2} \frac{\partial y_2}{\partial x_1} + \frac{\partial E}{\partial y_3} \frac{\partial y_3}{\partial x_1}$$

$$\frac{\partial y_1}{\partial x_1} = y_1 - e^{x_1} / (e^{x_1} + e^{x_2} + e^{x_3})^2$$

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

Handwritten notes for the first three rows of the gradient matrix:

- $(-\hat{y}_1 \times \frac{1}{x_1}) (y_1 - e^{x_1} (\text{sum})^{-2})$
- $(-\hat{y}_2 \times \frac{1}{x_2}) (y_2 - e^{x_2} (\text{sum})^{-2})$
- $(-\hat{y}_3 \times \frac{1}{x_3}) (y_3 - e^{x_3} (\text{sum})^{-2})$

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.

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

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

## ◆ 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**.

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## 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)