



*Philadelphia University*  
*Faculty of Engineering*

**Marking Scheme**

Exam Paper

BSc CE

**Neural Networks and Fuzzy Logic (630514)**

First Exam

First semester

Date: 19/11/2015

Section 1

Weighting 20% of the module total

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# Marking Scheme

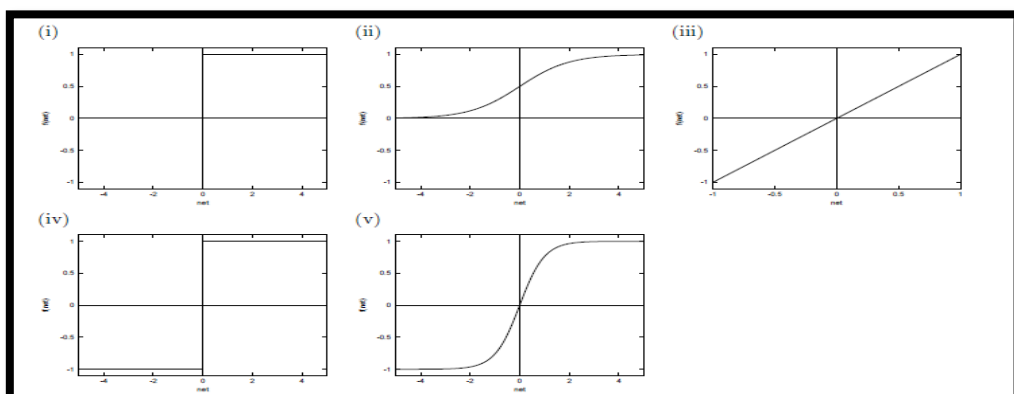
## Neural Networks and Fuzzy Logic (630514)

The presented exam questions are organized to overcome course material through 4 questions. The *all questions* are compulsory requested to be answered.

### Marking Assignments

**Question 1** This question is attributed with 8 marks if answered properly; the answers are as following:

- 1) Which of the following statements are true for typical neurons in the human brain?
- a) The neurons are connected to each other by axons, synapses and dendrites.
  - b) When the potential is bigger than a threshold, the neuron fires a pulse through the axon
  - c) Electrical potential is summed in the neuron.
  - d) **All of the above answers.**
- 2) The network that involves **backward links** from output to the input and hidden layers is called as \_\_\_\_.
- a) Self-organizing maps
  - b) **Recurrent neural network**
  - c) Multi layered perceptron
  - d) Perceptrons
- 3) Why is the **XOR** problem exceptionally interesting to neural network researchers?
- a) Because it can be expressed in a way that allows you to use a neural network.
  - b) Because it is binary operation that cannot be solved using neural networks.
  - c) Because it can be solved by a single layer perceptron.
  - d) **Because it is the simplest linearly inseparable problem that exists.**
- 4) In **supervised** learning:
- a) The algorithms are known but not the inputs
  - b) **Both the inputs and the desired outputs are known**
  - c) Only input stimuli are shown to the network
  - d) None of the above
- 5) A single-layer perceptron has **5** input units and **4** output units. How many weights does this network have?
- a) 5
  - b) 9
  - c) **20**
  - d) 24
- 6) A perceptron has input weights  $w_1 = 3$  and  $w_2 = 1$  and a threshold value  $T = 0.4$ . What output does it give for the input  $x_1 = 1, x_2 = 2$ ?
- a)  $3 * 1 + 1 * 2 = 5$ .
  - b)  **$3 * 1 + 1 * 2 = 5$ . This is more than the threshold, so output +1.**
  - c)  $3 * 1 + 1 * 2 = 5$ . Now subtract the threshold:  $5 - 0.4 = 4.6$ .
  - d)  $3 * 1 + 1 * 2 = 5$ . This is more than the threshold, so output 0.
- 7) Identify each of the following activation functions.



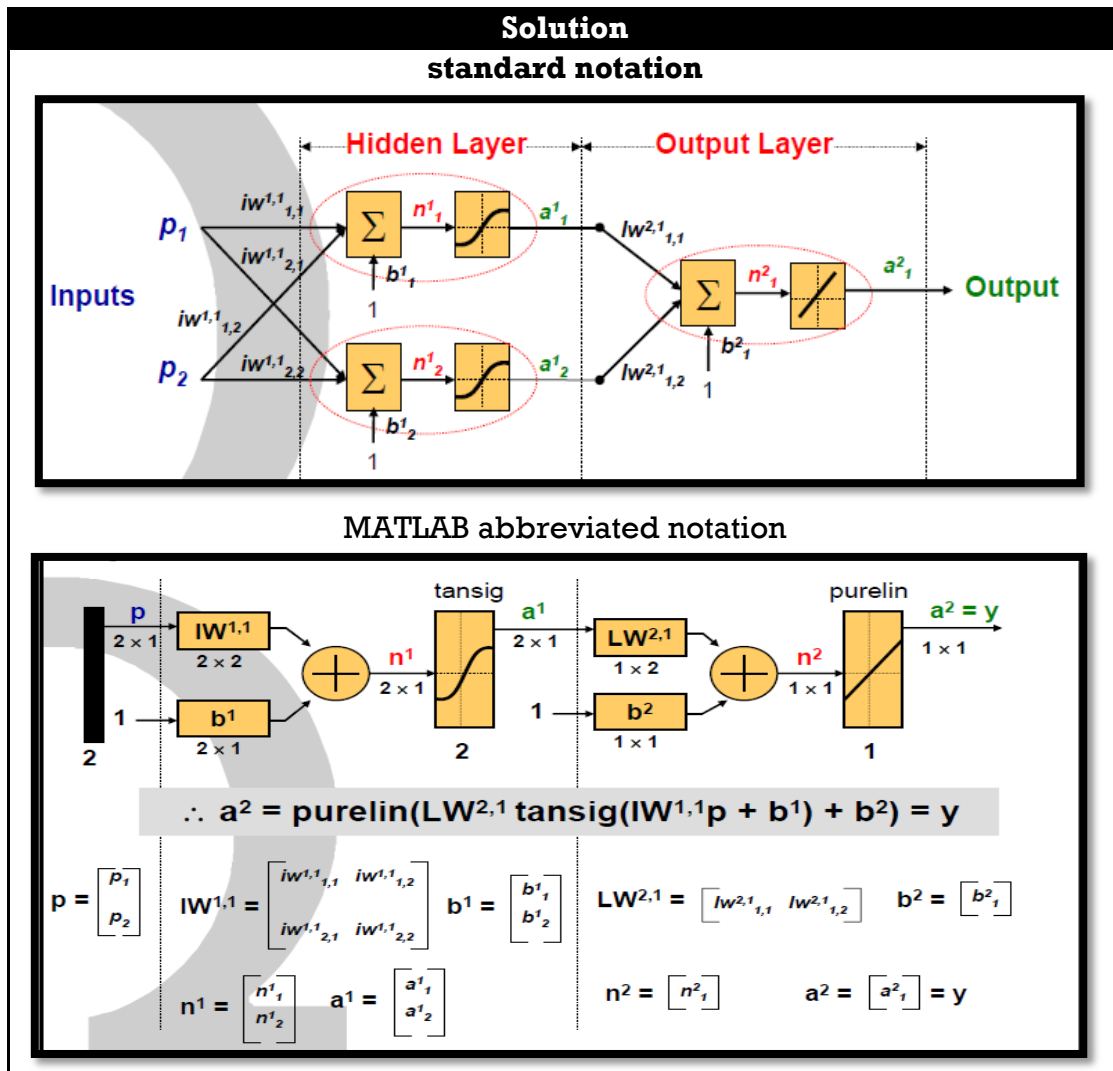
a)	(i) Unipolar step [hardlim], (ii) Unipolar sigmoid [logsig], (iii) Linear [purelin], (iv) Bipolar step [hardlims], (v) Bipolar sigmoid [tansig].
b)	(i) Bipolar step [hardlims], (ii) Unipolar sigmoid [logsig], (iii) Linear [purelin], (iv) Unipolar step [hardlim], (v) Bipolar sigmoid [tansig].
c)	(i) Bipolar step [hardlims], (ii) Bipolar sigmoid [tansig], (iii) Linear [purelin], (iv) Unipolar step [hardlim], (v) Unipolar sigmoid [logsig].
d)	(i) Unipolar step [hardlim], (ii) Bipolar sigmoid [tansig], (iii) Linear [purelin], (iv) Bipolar step [hardlims], (v) Unipolar sigmoid [logsig].

8) What does the following MATLAB function do?

`net = newff (minmax (p), [4, 2], {'tansig', 'logsig'});`

a)	Initialize a multi-layer network with non-linear activation functions and two hidden layers – the first hidden layer has 4 units and the second one has 2 units.
b)	Initialize a multi-layer network with sigmoid activation functions, 4 hidden units and 2 recurrent connections back to the input layer.
c)	Initialize a single-layer network with 4 input units, 2 output units and linear activation functions.
d)	Initialize a multi-layer network with 4 hidden units, 2 output units and sigmoid activation functions.

**Question 2** This question is attributed with 3 marks if answered properly; the answers are as following:



**Question 3** This question is attributed with 3 marks if answered properly; the answers are as following:

a)

(1.5 marks)

**Solution**

- (i) They have the ability to learn by example
- (ii) They are more faults tolerant because they are always able to respond and small changes in input do not normally cause a change in output.
- (iii) Because of their parallel architecture, they are more suited for real time operation due to their high 'computational' rates.

b)

(1.5 marks)

**Solution**

[1]. For the hard limit transfer function:

$$a = \text{hardlim}(1.6) = 1.0$$

[2]. For the linear transfer function:

$$a = \text{purelin}(1.6) = 1.6$$

[3]. For the log-sigmoid transfer function:

$$a = \text{logsig}(1.6) = \frac{1}{1 + e^{-1.6}} = 0.8320$$

**Question 4** This question is attributed with 6 marks if answered properly; the answers are as following:

**Solution**

**% Load the data points into Workspace**

```
Points = [ -3 -2 -1.5 -1 -1 0 0 1; ...
          -0.5 -1.2 0.7 3 -3.5 2 -2.5 0.7];
```

```
Group = [ 0 0 1 1 0 1 0 1];
```

**% 1 mark**

**% Assign training inputs and targets**

```
P = Points; % inputs
```

```
T = Group; % targets
```

**% Construct a two-input, single-output perceptron**

```
net = newp (minmax (P), 1);
```

**% 0.5 mark**

**% Train the perceptron network with training inputs (p) and targets (t)**

```
net = train (net, P, T);
```

**% 0.5 mark**

**% Simulate the perceptron network with same inputs again**

```
a = sim (net, P);
```

**% 0.5 mark**

```
%>> a =
```

```
%0 0 1 1 0 1 0 1 % correct classification
```

```
%>> T =
```

```
%0 0 1 1 0 1 0 1
```

**% Querying the perceptron with inputs it never seen before**

```
P9 = [-2; -3];
```

```
P10 = [0.5; 4];
```

```
a_P9 = sim (net, P9)
```

**% 0.5 mark**

```
%>> a_P9 =
```

```
%0
```

```
a_P10 = sim (net, P10);
```

**% 0.5 mark**

```
%>> a_P10 =
```

**%1; ∴ the perceptron classifies P<sub>9</sub> and P<sub>10</sub> correctly.**

