

Roll No. ....

**24489**

**B.Tech. 7th Sem. (Computer  
Science Engineering)  
Examination-May, 2013**

**NEURAL NETWORKS**

**Paper CSE-407-F**

**Time : 3 hours**

**Max. Marks : 100**

Before answering the questions, candidates should ensure that they have been supplied the correct and complete question paper. No complaint in this regard will be entertained after the examination.

**Note : Question No. 1 is compulsory. Attempt five questions in total selecting one question from each Section.**

1. (a) Differentiate supervised and unsupervised learning. 5 × 4  
(b) Explain the learning factors.  
(c) What are the separability limitations ?  
(d) Compare perception, delta and winner take II learning rules.

**SECTION - A**

2. Explain in detail the structure of biological neurons relevant to ANN. 20

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3. (a) Use Mc Culloch-Pitts neuron to design logic networks of AND and OR logic function. 8
- (b) Implement the perceptron rule training of the network using  $f(\text{net}) = \text{sgn}(\text{net})$ ,  $c=1$  and the following data specifying the initial weights  $w$  and two training pairs. 12

$$W' = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{pmatrix} x_1 = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix}, d_1 = -1 \end{pmatrix}, \begin{pmatrix} x_2 = \begin{bmatrix} 0 \\ -1 \\ -1 \end{bmatrix}, d_2 = 1 \end{pmatrix}$$

Repeat the training sequence  $(x_1, d_1)$ ,  $(x_2, d_2)$  until two correct responses in a row are achieved. List the net values obtained during training.

#### SECTION - B

- 4 (a) Explain the single layer continuous perception training algorithm for the linearly separable classification. 10
- (b) Implement the single discrete perception training algorithm for  $c = 1$  that provide the following classification to four patterns.

$X_1 = [1 \ 1]$ ,  $x_3 = [3 \ 1]$ ,  $d_1 = d_3 = 1$  : class 1

$X_2 = [0.5 \ 1]$ ,  $x_4 = [-2 \ 1]$ ,  $d_2 = d_4 = -1$  : class 2

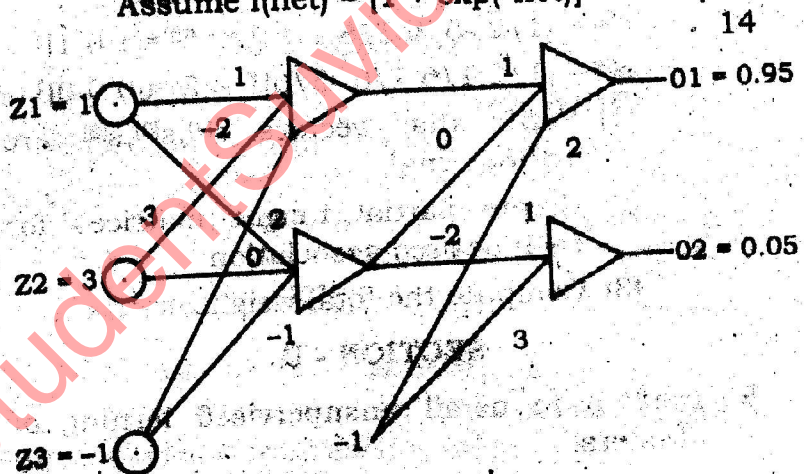
Perform the training task starting with initial weight vector  $w = [-2.5 \ 1.75]^T$ . 10

- 5 (a) Explain error back propagation training algorithm. 6

(b) For the n/w shown below analyse a single feedforward and back propagation step for the initialized n/w by doing the following :

- Find weight matrices  $w$  and  $v$
- Calculate  $net_j$ ,  $y$ ,  $net_k$ ,  $0$
- Calculate slopes  $f'(net_j)$  and  $f'(net_k)$
- Compute error signals  $\delta_0$  and  $\delta_j$
- Compute  $\Delta v$  and  $\Delta w$
- Find undated weights

Assume  $f(net) = [1 + \exp(-net)]^{-1}$  and  $\eta = 1$



### SECTION - C

6 The weight matrix  $W$  for a n/w with bipolar discrete binary neurons is given as

|    |    |    |    |    |
|----|----|----|----|----|
| 0  | 1  | -1 | -1 | -3 |
| 1  | 0  | 1  | 1  | 1  |
| -1 | 1  | 0  | 3  | 1  |
| -1 | 1  | 3  | 0  | 1  |
| 3  | -1 | 1  | 1  | 0  |



Assume threshold and external i/p of neurons are zero. Compare the values of energy for  $v = [-1 \ 1 \ 1 \ 1 \ 1]^T$  and  $v = [-1 \ -1 \ 1 \ -1 \ -1]^T$

- 7 (a) Explain the association encoding and decoding and stability consideration for bidirectional associative memory. 10
- (b) The linear associator has to associate the following pair of vectors : 10
- $s^{(1)} = [1/2 \ 1/2 \ 1/2 \ 1/2]^T \rightarrow f^{(1)} = [0 \ 1 \ 0]^T$
- $s^{(2)} = [1/2 \ -5/6 \ 1/6 \ 1/6]^T \rightarrow f^{(2)} = [1 \ 0 \ 1]^T$
- $s^{(3)} = [1/2 \ 1/6 \ 1/6 \ -5/6]^T \rightarrow f^{(3)} = [0 \ 0 \ 0]^T$
- (1) Verify that vectors  $s^{(1)}, s^{(2)}, s^{(3)}$  are orthonormal.
- (2) Create partial weight matrices for each desired association.
- (3) Compute the total weight matrix.

#### SECTION - D

- 8 Explain in detail unsupervised learning of clusters. 20
- 9 Perform the first learning cycle using the following normalized pattern set : 20
- $\{x_1, x_2, x_3, x_4\} = \{1 < 45^\circ, 1 < 135^\circ, 1 < 90^\circ, 1 < 180^\circ\}$  and  $\alpha = 0.5$

Using the winner take all training rule for two cluster neurons. Draw the resulting separating hyperplanes. Initial weights are to be assumed  $w_1^0 = [1 \ 0]^T$  and  $w_2^0 = [0 \ -1]^T$ .