

**Course No.:** CSE 2102

**Course Title:** Sessional based on CSE 2101

**Experiment No.** 4

**Name of the Experiment:**

- a. Design and implementation of Kohonen Self-organizing Neural Networks algorithm.
- b. Design and implementation of Hopfield Neural Networks algorithm.

**Course Outcomes:** CO1

**Learning Domain with Level:** Cognitive (Applying, Analyzing, Evaluating & Creating)

4 (a)

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### *Kohonen Network Algorithm*

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#### 1. Initialise network

Define  $w_{ij}(t)$  ( $0 \leq i \leq n-1$ ) to be the weight from input  $i$  to node  $j$  at time  $t$ . Initialise weights from the  $n$  inputs to the nodes to small random values. Set the initial radius of the neighbourhood around node  $j$ ,  $N_j(0)$ , to be large.

#### 2. Present input

Present input  $x_0(t), x_1(t), x_2(t), \dots, x_{n-1}(t)$ , where  $x_i(t)$  is the input to node  $i$  at time  $t$ .

#### 3. Calculate distances

Compute the distance  $d_j$  between the input and each output node  $j$ , given by

$$d_j = \sum_{i=0}^{n-1} (x_i(t) - w_{ij}(t))^2$$

#### 4. Select minimum distance

Designate the output node with minimum  $d_j$  to be  $j^*$ .

#### 5. Update weights

Update weights for node  $j^*$  and its neighbours, defined by the neighbourhood size  $N_{j^*}(t)$ . New weights are

$$w_{ij}(t+1) = w_{ij}(t) + \eta(t)(x_i(t) - w_{ij}(t))$$

For  $j$  in  $N_{j^*}(t)$ ,  $0 \leq i \leq n-1$

The term  $\eta(t)$  is a gain term ( $0 < \eta(t) < 1$ ) that decreases in time, so slowing the weight adaption. Notice that the neighbourhood  $N_{j*}(t)$  decreases in size as time goes on, thus localising the area of maximum activity.

6. Repeat by going to 2.

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Your task is to:

- **Apply and Design** a program for your dataset
- **Analyze** the program
- **Evaluate** the correctness of the program and the accuracy of the algorithm.

4 (b)

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### *Hopfield Network Algorithm*

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1. Assign connection weights

$$w_{ij} = \begin{cases} \sum_{s=0}^{M-1} x_i^s x_j^s & i \neq j \\ 0 & i = j, 0 \leq i, j \leq M-1 \end{cases}$$

where  $w_{ij}$  is the connection weight between node  $i$  and node  $j$ , and  $x_i^s$  is element  $i$  of the exemplar pattern for class  $s$ , and is either  $+1$  or  $-1$ . There are  $M$  patterns, from 0 to  $M-1$ , in total. The thresholds of the units are zero.

2. Initialise with unknown pattern

$$\mu_i(0) = x_i \quad 0 \leq i \leq N-1$$

where  $\mu_i(t)$  is the output of node  $i$  at time  $t$ .

3. Iterate until convergence

$$\mu_i(t+1) = f_h \left[ \sum_{j=0}^{N-1} w_{ij} \mu_j(t) \right] \quad 0 \leq j \leq N-1$$

The function  $f_h$  is the hard-limiting non-linearity, the step function, as in figure 3.3. Repeat the iteration until the outputs from the nodes remain unchanged.

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Your task is to:

- **Apply and Design** a program for your dataset
- **Analyze** the program
- **Evaluate** the correctness of the program and the accuracy of the algorithm.