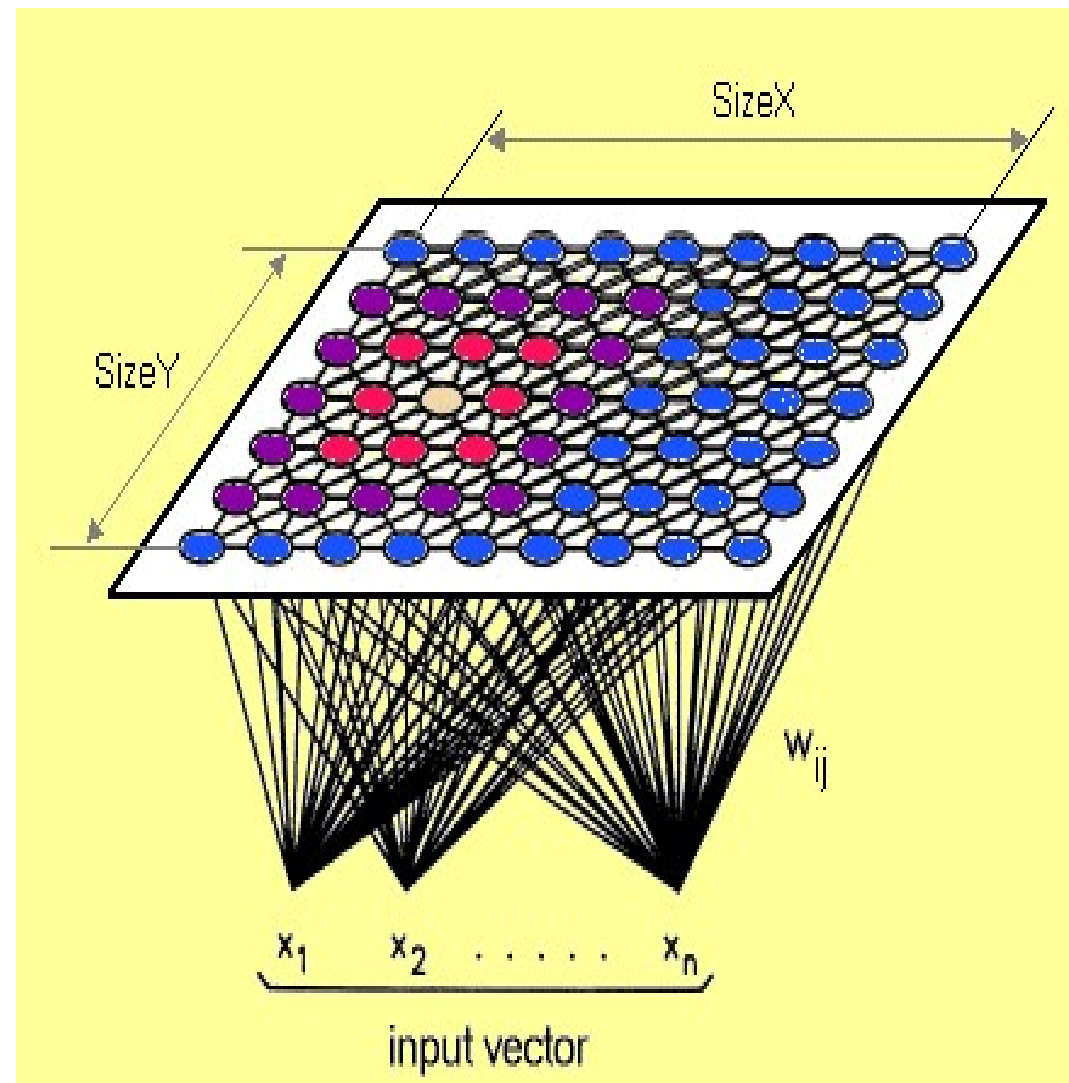


Self Organizing Map (or  
Kohonen Map or SOM)

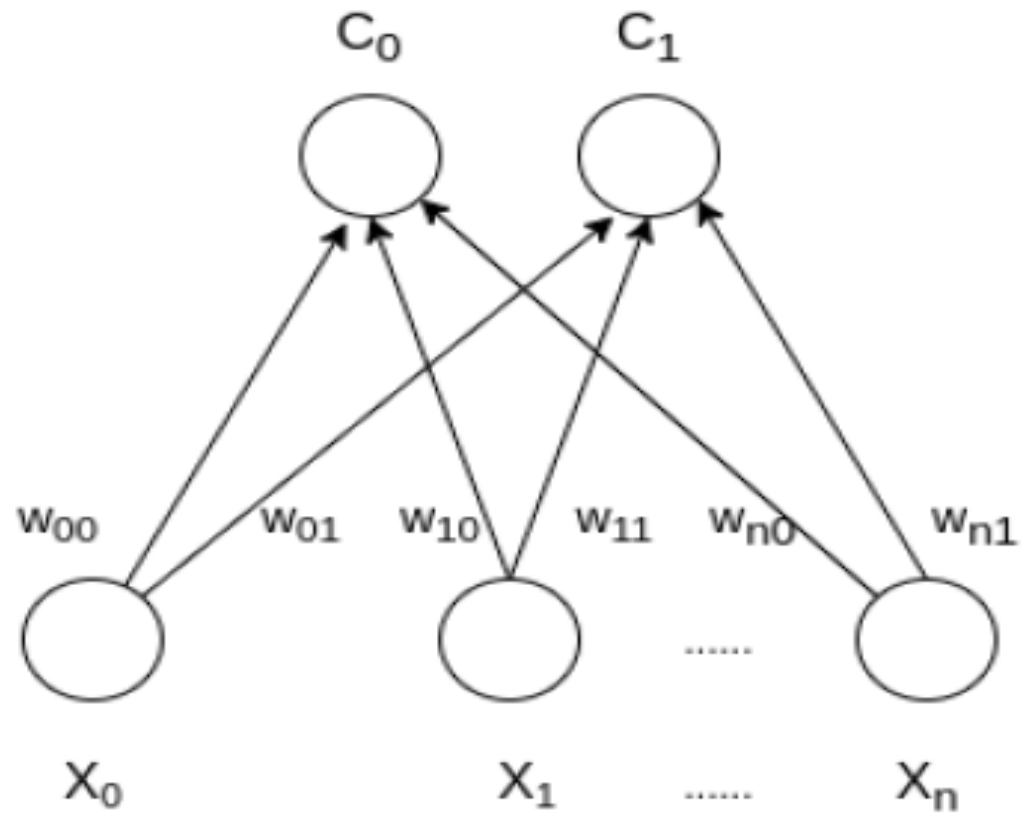
## Self Organizing Map (or Kohonen Map or SOM)

- Type of ANN, inspired by biological models of neural systems from the 1970s.
- Follows an unsupervised learning approach and trained its network through a competitive learning algorithm.
- Used for clustering and mapping (or dimensionality reduction) techniques to map multidimensional data onto lower-dimensional for easy interpretation.
- Has two layers, one is the Input layer and the other one is the Output layer.



## Architecture:

- Two clusters with 'n' input features for any sample:



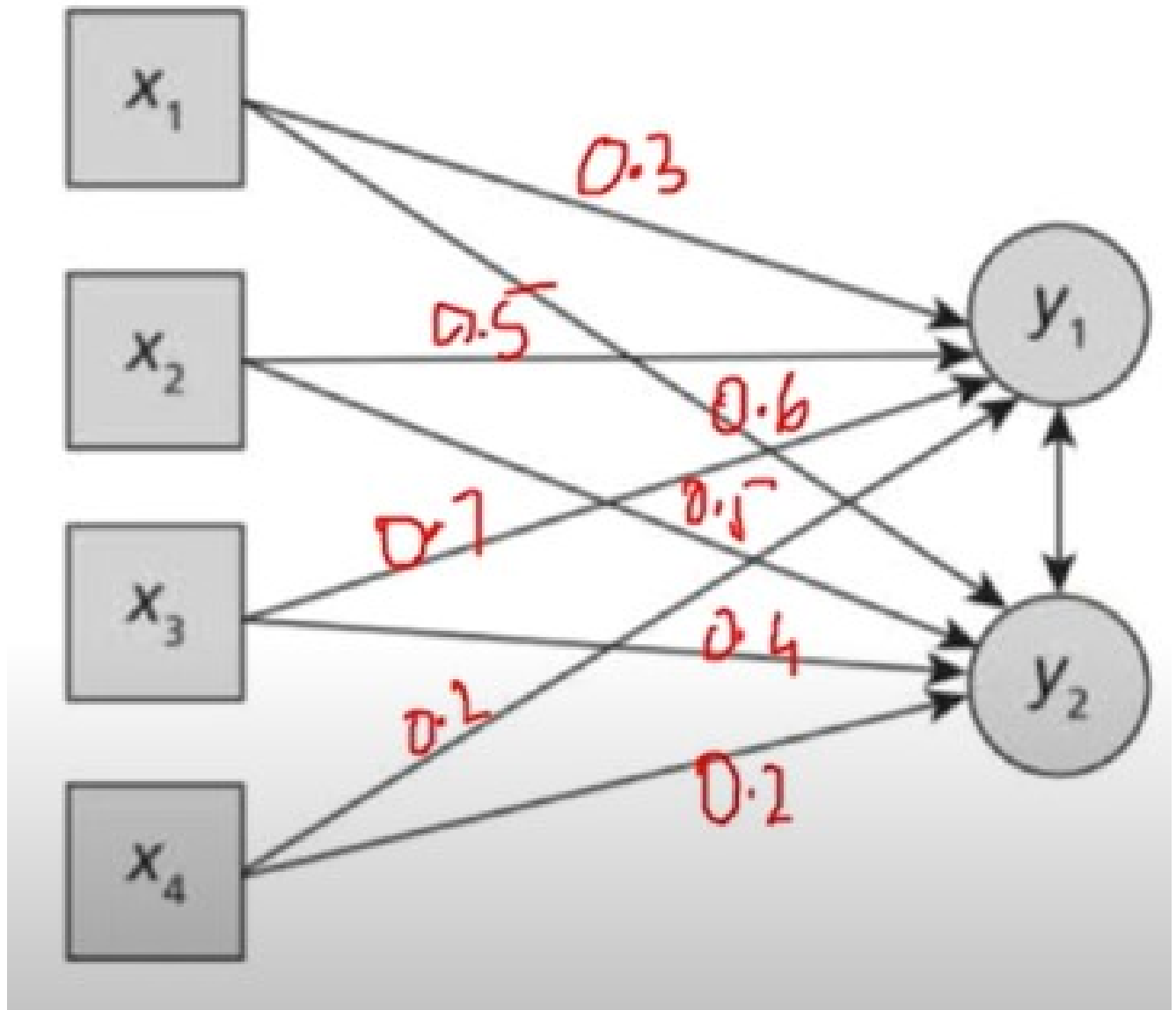
## How do SOM works?

- Let's say an input data of size  $(m, n)$  where  $m$  is the number of training examples and  $n$  is the number of features in each example.
- First, it initializes the weights of size  $(n, C)$  where  $C$  is the number of clusters.
- Then iterating over the input data, for each training example, it updates the winning vector (weight vector with the shortest distance (e.g Euclidean distance) from training example).
- Weight updation rule is given by:
$$\mathbf{w}_{ij} = \mathbf{w}_{ij}(\text{old}) + \alpha(t) * (\mathbf{x}_{ik} - \mathbf{w}_{ij}(\text{old}))$$
  - where  $\alpha$  is a learning rate at time  $t$ ,
  - $j$  denotes the winning vector,
  - $i$  denotes the  $i$ th feature of training example and
  - $k$  denotes the  $k$ th training example from the input data.
- After training, trained weights are used for clustering new examples. A new example falls in the cluster of winning vectors.

## Example:

X1: (1, 0, 1, 0)      X2: (1, 0, 0, 0).  
X3: (1, 1, 1, 1)      X4: (0, 1, 1, 0)

- Output Units: Unit 1, Unit 2
- Learning rate  $\eta(t) = 0.6$
- Initial Weight matrix
- $\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} = \begin{bmatrix} 0.3 & 0.5 & 0.7 & 0.2 \\ 0.6 & 0.5 & 0.4 & 0.2 \end{bmatrix}$



**Example: Euclidean Distance:**  $d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$ .

### Iteration 1:

Training Sample  $x_1$ : (1, 0, 1, 0)

Weight matrix:

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.3 & 0.5 & 0.7 & 0.2 \\ 0.6 & 0.7 & 0.4 & 0.3 \end{bmatrix}$$

Compute Euclidean distance between  $x_1$ : (1, 0, 1, 0) and Unit 1 weights.

$$d^2 = (0.3 - 1)^2 + (0.5 - 0)^2 + (0.7 - 1)^2 + (0.2 - 0)^2 = 0.87$$

Compute Euclidean distance between  $x_1$ : (1, 0, 1, 0) and Unit 2 weights.

$$d^2 = (0.6 - 1)^2 + (0.7 - 0)^2 + (0.4 - 1)^2 + (0.3 - 0)^2 = 1.1$$

**Unit 1 wins**

## Example:

$$\underline{w_j(t+1)} = \underline{w_j(t)} + \underline{\eta(t)}(\underline{x_s} - \underline{w_j(t)})$$

Update the weights of the winning unit.

$$\begin{aligned}\text{New Unit 1 weights} &= \underline{[0.3 \ 0.5 \ 0.7 \ 0.2]} + \underline{0.6} (\underline{[1 \ 0 \ 1 \ 0]} - \underline{[0.3 \ 0.5 \ 0.7 \ 0.2]}) \\ &= [0.3 \ 0.5 \ 0.7 \ 0.2] + 0.6 [0.7 \ -0.5 \ 0.3 \ -0.2] \\ &= [0.3 \ 0.5 \ 0.7 \ 0.2] + [0.42 \ -0.30 \ 0.18 \ -0.12] \\ &= [0.72 \ 0.2 \ 0.88 \ 0.08]\end{aligned}$$

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.72 & 0.2 & 0.88 & 0.08 \\ 0.6 & 0.7 & 0.4 & 0.3 \end{bmatrix}$$

## Example:

### Iteration 2:

Training Sample  $x_2$ : (1, 0, 0, 0)

Weight matrix:

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.72 & 0.2 & 0.88 & 0.08 \\ 0.6 & 0.7 & 0.4 & 0.3 \end{bmatrix}$$

Compute Euclidean distance between  $x_2$ : (1, 0, 0, 0) and Unit 1 weights.

$$d^2 = (0.72 - 1)^2 + (0.2 - 0)^2 + (0.88 - 0)^2 + (0.08 - 0)^2 = 0.74$$

Compute Euclidean distance between  $x_2$ : (1, 0, 0, 0) and Unit 2 weights.

$$d^2 = (0.6 - 1)^2 + (0.7 - 0)^2 + (0.4 - 0)^2 + (0.3 - 0)^2 = 0.9$$

**Unit 1 wins**



## Example:

$$\underline{w_j(t+1) = w_j(t) + \eta(t)(x_s - w_j(t))}$$

Update the weights of the winning unit:

$$\begin{aligned}\text{New Unit 1 weights} &= [0.72 \ 0.2 \ 0.88 \ 0.08] + 0.6 ([1 \ 0 \ 0 \ 0] - [0.72 \ 0.2 \ 0.88 \ 0.08]) \\ &= [0.72 \ 0.2 \ 0.88 \ 0.08] + 0.6 [0.28 \ -0.2 \ -0.88 \ -0.08] \\ &= [0.72 \ 0.2 \ 0.88 \ 0.08] + [0.17 \ -0.12 \ -0.53 \ -0.05] \\ &= [0.89 \ 0.08 \ 0.35 \ 0.03]\end{aligned}$$

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.89 & 0.08 & 0.35 & 0.03 \\ 0.6 & 0.7 & 0.4 & 0.3 \end{bmatrix}$$

## Example:

### Iteration 3:

Training Sample  $x_3$ : (1, 1, 1, 1)

Weight matrix:

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.89 & 0.08 & 0.35 & 0.03 \\ 0.6 & 0.7 & 0.4 & 0.3 \end{bmatrix}$$

Compute Euclidean distance between  $x_3$ : (1, 1, 1, 1) and Unit 1 weights.

$$\begin{aligned} d^2 &= (0.89 - 1)^2 + (0.08 - 1)^2 + (0.35 - 1)^2 + (0.03 - 1)^2 \\ &= 2.2 \end{aligned}$$

Compute Euclidean distance between  $x_3$ : (1, 1, 1, 1) and Unit 2 weights.

$$\begin{aligned} d^2 &= (0.6 - 1)^2 + (0.7 - 1)^2 + (0.4 - 1)^2 + (0.3 - 1)^2 \\ &= 1.1 \end{aligned}$$

**Unit 2 wins**

## Example:

$$\underline{w_j(t+1) = w_j(t) + \eta(t)(x_s - w_j(t))}$$

Update the weights of the winning unit:

$$\begin{aligned}\text{New Unit 2 weights} &= [0.6 \ 0.7 \ 0.4 \ 0.3] + 0.6 ([1 \ 1 \ 1 \ 1] - [0.6 \ 0.7 \ 0.4 \ 0.3]) \\ &= [0.6 \ 0.7 \ 0.4 \ 0.3] + 0.6 [0.4 \ 0.3 \ 0.6 \ 0.7] \\ &= [0.6 \ 0.7 \ 0.4 \ 0.3] + [0.24 \ 0.18 \ 0.36 \ 0.42] = [0.84 \ 0.88 \ 0.76 \ 0.72]\end{aligned}$$

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.89 & 0.08 & 0.35 & 0.03 \\ 0.84 & 0.88 & 0.76 & 0.72 \end{bmatrix}$$

## Example:

### Iteration 4:

Training Sample  $x_4$ : (0, 1, 1, 0)

Weight matrix:

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.89 & 0.08 & 0.35 & 0.03 \\ 0.84 & 0.88 & 0.76 & 0.72 \end{bmatrix}$$

Compute Euclidean distance between  $x_4$ : (0, 1, 1, 0) and Unit 1 weights.

$$\begin{aligned} d^2 &= (0.89 - 0)^2 + (0.08 - 1)^2 + (0.35 - 1)^2 + (0.03 - 0)^2 \\ &= 2.06 \end{aligned}$$

Compute Euclidean distance between  $x_4$ : (0, 1, 1, 0) and Unit 2 weights.

$$\begin{aligned} d^2 &= (0.84 - 0)^2 + (0.88 - 1)^2 + (0.76 - 1)^2 + (0.72 - 0)^2 \\ &= 1.3 \end{aligned}$$

## Example:

$$\underline{w_j(t+1)} = w_j(t) + \eta(t)(x_s - w_j(t))$$

Unit 2 wins

Update the weights of the winning unit:

$$\begin{aligned}\text{New Unit 2 weights} &= \underline{[0.84 \ 0.88 \ 0.76 \ 0.72]} + \underline{0.6} (\underline{[0 \ 1 \ 1 \ 0]} - \underline{[0.84 \ 0.88 \ 0.76 \ 0.72]}) \\ &= [0.84 \ 0.88 \ 0.76 \ 0.72] + 0.6 [-0.84 \ 0.12 \ 0.24 \ -0.72] \\ &= [0.84 \ 0.88 \ 0.76 \ 0.72] + [-0.5 \ 0.07 \ 0.14 \ -0.43] = \underline{[0.34 \ 0.95 \ 0.9 \ 0.29]}\end{aligned}$$

$$\begin{bmatrix} \text{Unit 1} \\ \text{Unit 2} \end{bmatrix} : \begin{bmatrix} 0.89 & 0.08 & 0.35 & 0.03 \\ 0.34 & 0.95 & 0.9 & 0.29 \end{bmatrix}$$

## Example:

Best mapping units for each of the sample taken are:

$x_1$ : (1, 0, 1, 0) → Unit 1 ✓

$x_2$ : (1, 0, 0, 0) → Unit 1 ✓

$x_3$ : (1, 1, 1, 1) → Unit 2 ✓

$x_4$ : (0, 1, 1, 0) → Unit 2 ✓

} Epoch ✓

This process is continued for many epochs until the feature map does not change.