

AUTOASSOCIATIVE NET

Used to recall a pattern by a its noisy or incomplete version.(pattern completion/pattern recovery)

- *For an auto associative net, the **training input and target output are identical i.e.** $t(p) = s(p)$*
- *Process of training is often called **STORING** the vectors, may be binary or bipolar **[Storage Phase]***
- *A stored vector can be **retrieved** from distorted or partial (noisy) input, if input is sufficiently similar to it.**[Recall Phase]***

Associative network can recognize as "known " vectors those vectors which are similar to stored vectors, but differ slightly from it.

Differences could be two forms:

1. MISTAKES
2. MISSING data

Bipolar (1/-1) representation of inputs and targets

Missing Data: this is represented by a 0.

Mistaken Data: A mistake in +1 is represented as -1 and vice versa.

Binary(0/1) representation of inputs and targets

Missing Data: Cannot be represented.

Mistaken Data: A mistake of 1 is represented as 0 and vice versa.

In general a net can handle more missing components than mistaken/wrong components.

An Associative net to store **one vector** : recognizing the stored vector

Vector $s = (1,1,1,-1)$ is stored with weight matrix $[s^T s]$

$$W = \begin{bmatrix} 1 \\ 1 \\ 1 \\ -1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ -1 & -1 & -1 & 1 \end{bmatrix}$$

Diagonal elements not forced to zero

Bipolar activation function with threshold as zero

$$(1,1,1,-1) \cdot W = (4,4,4,-4) \rightarrow (1,1,1,-1)$$

Testing an autoassociative network with **one mistake** in (1,1,1,-1)

$$(1,1,1,-1) \rightarrow (-1,1,1,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,-1,1,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,1,-1,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,1,1,1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

RECOGNIZES !!

$$W = \begin{bmatrix} 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ -1 & -1 & -1 & 1 \end{bmatrix}$$

TESTING AN AUTOASSOCIATIVE NETWORK WITH **TWO MISTAKES** IN (1,1,1,-1)

$$(1,1,1,-1) \rightarrow (-1,-1,1,-1).W=(0,0,0,0) \rightarrow (1,1,1,1)$$

Does not recognize

Testing an Autoassociative Network with **two “missing”** entries in input vector

$$(1,1,1,-1) \rightarrow (0,0,1,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (0,1,0,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (0,1,1,0).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,0,0,-1).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,0,1,0).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,1,0,0).W=(2,2,2,-2) \rightarrow (1,1,1,-1)$$

Successfully Retrieves

(1,1,1,-1)- Stored Vector **Two Missing Data**

$$(1,1,1,-1) \rightarrow (0,0,1,-1).W_0=(2,2,1,-1) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (0,1,0,-1).W_0=(2,1,2,-1) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (0,1,1,0).W_0=(2,1,1,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,0,0,-1).W_0=(1,2,2,-1) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,0,1,0).W_0=(1,2,1,-2) \rightarrow (1,1,1,-1)$$

$$(1,1,1,-1) \rightarrow (1,1,0,0).W_0=(1,1,2,-2) \rightarrow (1,1,1,-1)$$

RECOGNIZES IN ALL CASES

AN AUTOASSOCIATIVE NETWORK WITH NO SELF-CONNECTIONS :ZEROING OUT DIAGONAL OF ORIGINAL W

$$W = \begin{bmatrix} 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ 1 & 1 & 1 & -1 \\ -1 & -1 & -1 & 1 \end{bmatrix}$$

$$W_0 = \begin{bmatrix} 0 & 1 & 1 & -1 \\ 1 & 0 & 1 & -1 \\ 1 & 1 & 0 & -1 \\ -1 & -1 & -1 & 0 \end{bmatrix}$$

Testing an Autoassociative network with **two mistakes** in (1,1,1,-1)

$$(1,1,1,-1) \rightarrow (-1,-1,1,-1).W=(0,0,0,0) \rightarrow (0,0,0,0)$$

$$(1,1,1,-1) \rightarrow (-1,-1,1,-1).W_0=(1,1,1,1) \rightarrow (1,1,1,1)$$

Still does not recognize

STORING SEVERAL VECTORS IN AN AUTO-ASSOCIATIVE NETWORK

Non-orthogonal vectors cannot be stored in an auto-associative network in reliable manner. They may not always be identified.

More patterns can be stored if they are not similar to each other (e.g., **orthogonal**)

Two vectors x and y are orthogonal if

$$x y^T = 0 \text{ i.e., } \sum x_i y_i = 0$$

Example: Storing more than one vector in an auto-associative net

- More than one vector can be stored in an auto-associative net by simply adding the weights needed for each vector.

Theorem: an $n \times n$ network is able to store up to $n-1$ mutually orthogonal (M.O.) bipolar vectors, but not n such vectors.

The requirement of **orthogonality** places serious limitations on the Hebbian Learning Rule

Example: Storing 2 non-orthogonal vectors in an auto-associative net

Trying to store two non-orthogonal vectors (1,-1,-1,1)
and (1,1,-1,1) gives a weight matrix which cannot
distinguish between two vectors

$$\begin{bmatrix} 0 & -1 & -1 & 1 \\ -1 & 0 & 1 & -1 \\ -1 & 1 & 0 & -1 \\ 1 & -1 & -1 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 1 & -1 & 1 \\ 1 & 0 & -1 & 1 \\ -1 & -1 & 0 & -1 \\ 1 & 1 & -1 & 0 \end{bmatrix}$$
$$= \begin{bmatrix} 0 & 0 & -2 & 2 \\ 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & -2 \\ 2 & 0 & -2 & 0 \end{bmatrix}.$$

TESTING

$$[1, 1, -1, 1] \begin{bmatrix} 0 & 0 & -2 & 2 \\ 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & -2 \\ 2 & 0 & -2 & 0 \end{bmatrix} = [4, 0, -4, 4] = [1, -1, -1, 1]$$

$$[1, -1, -1, 1] \begin{bmatrix} 0 & 0 & -2 & 2 \\ 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & -2 \\ 2 & 0 & -2 & 0 \end{bmatrix} = [4, 0, -4, 4] = [1, -1, -1, 1]$$

Same result for both, cannot distinguish between the two vectors

Example : store (1 1 -1 -1) and (-1 1 1 -1)
[orthogonal] in an auto-associative net.

- to store (1 1 -1 -1), we need a weight matrix \mathbf{W}_1 and
to store (-1 1 1 -1), we need a weight matrix \mathbf{W}_2 .

$$\mathbf{W}_1 = \begin{matrix} 0 & 1 & -1 & -1 \\ 1 & 0 & -1 & -1 \\ -1 & -1 & 0 & 1 \\ -1 & -1 & 1 & 0 \end{matrix}$$

$$\mathbf{W}_2 = \begin{matrix} 0 & -1 & -1 & 1 \\ -1 & 0 & 1 & -1 \\ -1 & 1 & 0 & -1 \\ 1 & -1 & -1 & 0 \end{matrix}$$

To store (1 1 -1 -1) and (-1 1 1 -1), we simply add \mathbf{W}_1 and \mathbf{W}_2 to obtain the new weight matrix \mathbf{W} . This is because the two vectors are orthogonal.

Storing 3 mutually orthogonal vectors in an auto-associative net

Let $W_1 + W_2$ be the weight matrix to store the orthogonal vectors

$(1, 1, -1, -1)$ and $(-1, 1, 1, -1)$

W_3 be the weight matrix that stores $(-1, 1, -1, 1)$

X_3 is orthogonal to both X_1 and X_2

$W_1 + W_2$

W_3

$W_1 + W_2 + W_3$

$$\begin{bmatrix} 0 & 0 & -2 & 0 \\ 0 & 0 & 0 & -2 \\ -2 & 0 & 0 & 0 \\ 0 & -2 & 0 & 0 \end{bmatrix} + \begin{bmatrix} 0 & -1 & 1 & -1 \\ -1 & 0 & -1 & 1 \\ 1 & -1 & 0 & -1 \\ -1 & 1 & -1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -1 & -1 & -1 \\ -1 & 0 & -1 & -1 \\ -1 & -1 & 0 & -1 \\ -1 & -1 & -1 & 0 \end{bmatrix}$$

which correctly classifies each of the three vectors on which it was trained.

Storing 4 mutually orthogonal vectors in an auto-associative net

Attempting to store a fourth vector, $(1, 1, 1, 1)$, with weight matrix W_4 orthogonal to each of the foregoing three,

$$\begin{array}{c} \mathbf{W}_1 + \mathbf{W}_2 + \mathbf{W}_3 \end{array} \quad \begin{array}{c} \mathbf{W}_4 \end{array} \quad \begin{array}{c} \mathbf{W}^* \end{array}$$
$$\begin{bmatrix} 0 & -1 & -1 & -1 \\ -1 & 0 & -1 & -1 \\ -1 & -1 & 0 & -1 \\ -1 & -1 & -1 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

which cannot recognize any vector

Autoassociative Memory

pattern
and target same
6x5 matrix



$\mathbf{p}_1, \mathbf{t}_1$

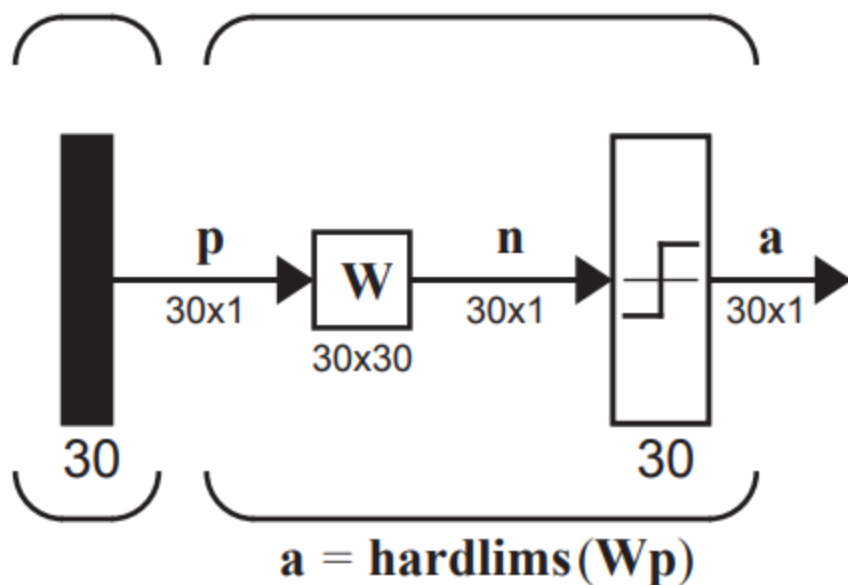


$\mathbf{p}_2, \mathbf{t}_2$



$\mathbf{p}_3, \mathbf{t}_3$

Inputs Sym. Hard Limit Layer



$$\mathbf{W} = \mathbf{p}_1 \mathbf{p}_1^T + \mathbf{p}_2 \mathbf{p}_2^T + \mathbf{p}_3 \mathbf{p}_3^T$$



$\mathbf{p}_1, \mathbf{t}_1$

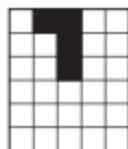
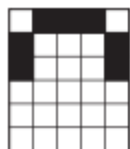


$\mathbf{p}_2, \mathbf{t}_2$

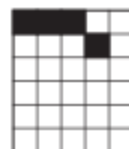
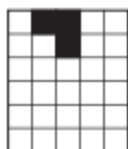
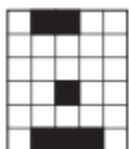
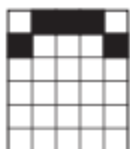


$\mathbf{p}_3, \mathbf{t}_3$

50% Occluded



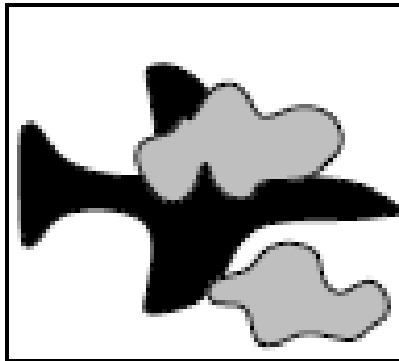
67% Occluded



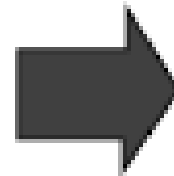
Noisy Patterns (7 pixels)



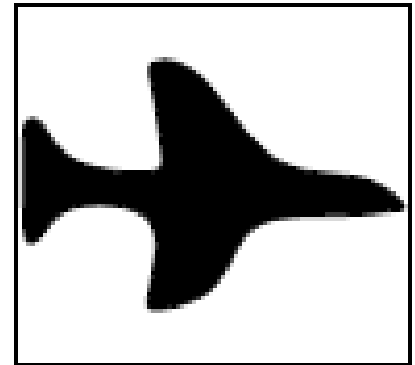
Airplane partially
occluded by clouds



Associative
Memory



Retrieved airplane



END