

# Hebb's learning postulate:

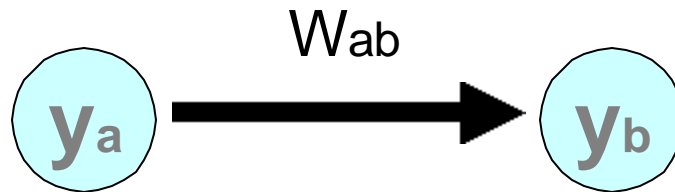
Whenever  
neuron A  
fires,



and neuron B  
fires soon  
afterwards ...

the synaptic efficacy,  $W_{ab}$ , increases.

# A Hebbian learning rule:



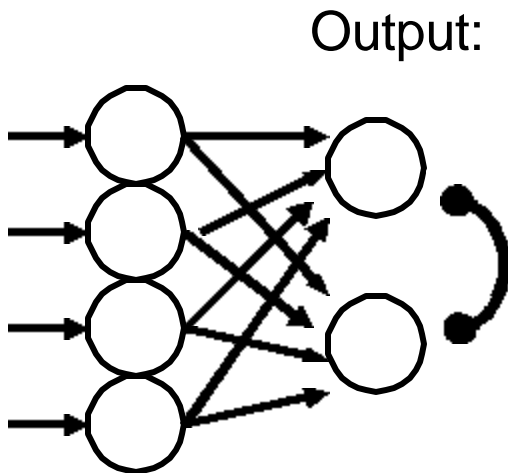
$$W_{ab} = W_{ab} + \mathbf{h} y_a y_b$$

# “Self-organizing” models based on Hebbian learning

## I. Competitive learning

Input  
patterns:

$A_2$	$A_1$
0	1
0	0
0	1
1	0

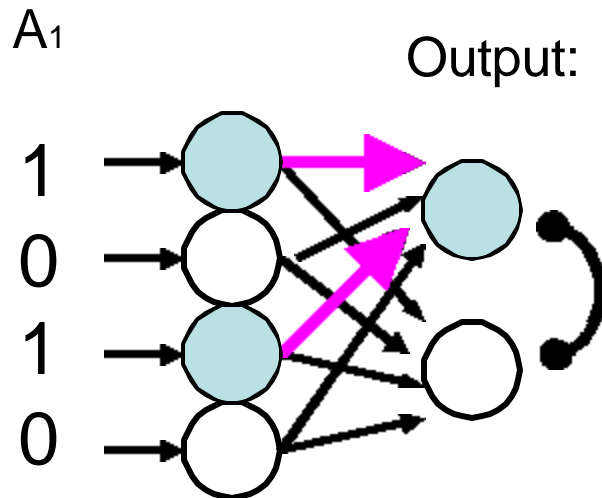


- Lateral inhibition
- Units compete to respond to pattern
- Winning unit does Hebbian learning

# “Self-organizing” models based on Hebbian learning

## I. Competitive learning

Input  
patterns:



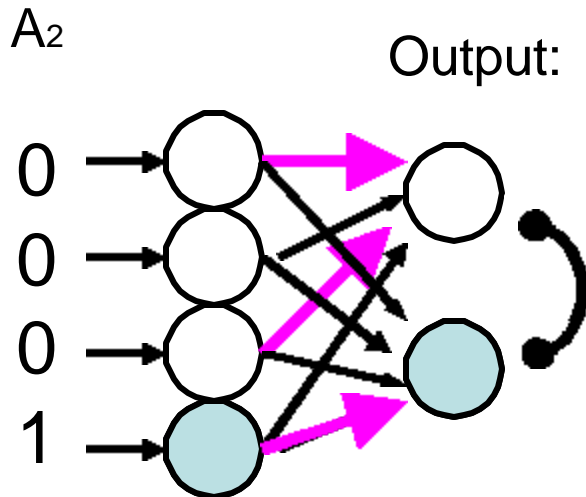
First learn pattern A<sub>1</sub>:

- The first output unit wins the competition
- Its connections to active inputs are strengthened

# “Self-organizing” models based on Hebbian learning

## I. Competitive learning

Input patterns:



Then learn pattern  $A_2$ :

- Second output unit wins the competition
- Its connections to active inputs are strengthened

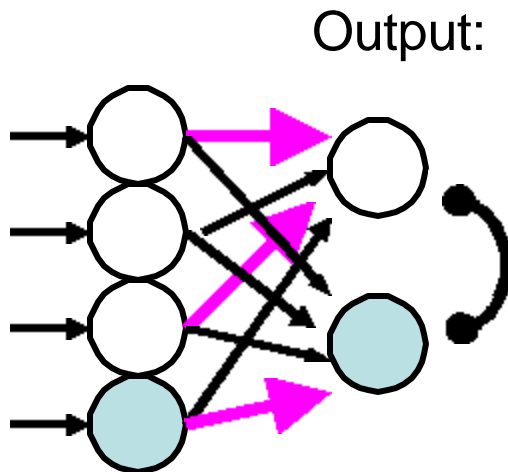
# “Self-organizing” models based on Hebbian learning

## I. Competitive learning

Input  
patterns:

A<sub>2</sub> A<sub>1</sub>

0 1  
0 0  
0 1  
1 0



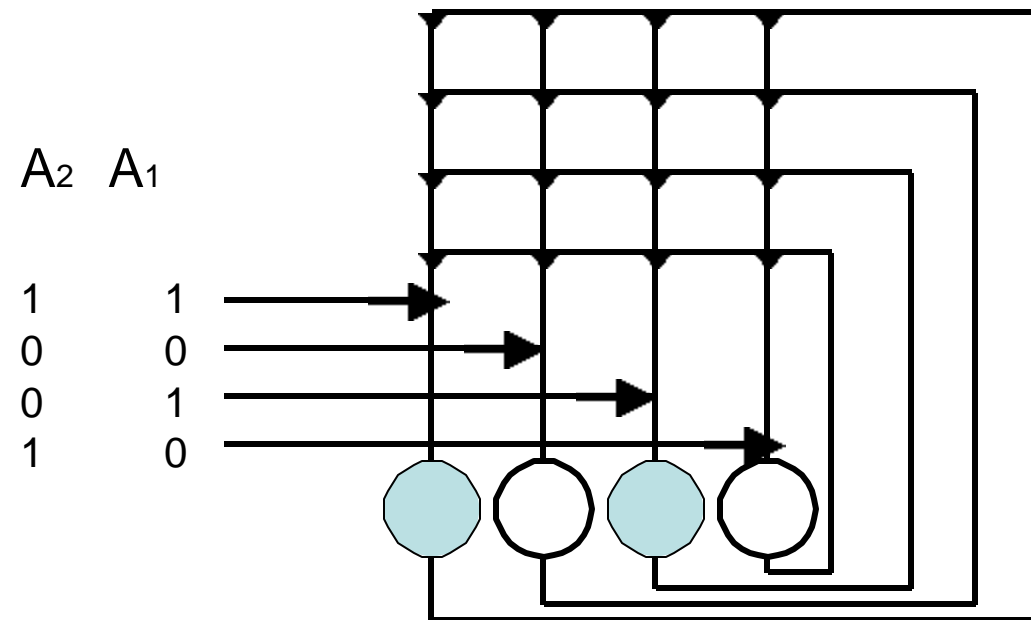
After learning:

- Each output unit responds to a different pattern (or subset of patterns)

# “Self-organizing” models based on Hebbian learning

## II. Auto-associator

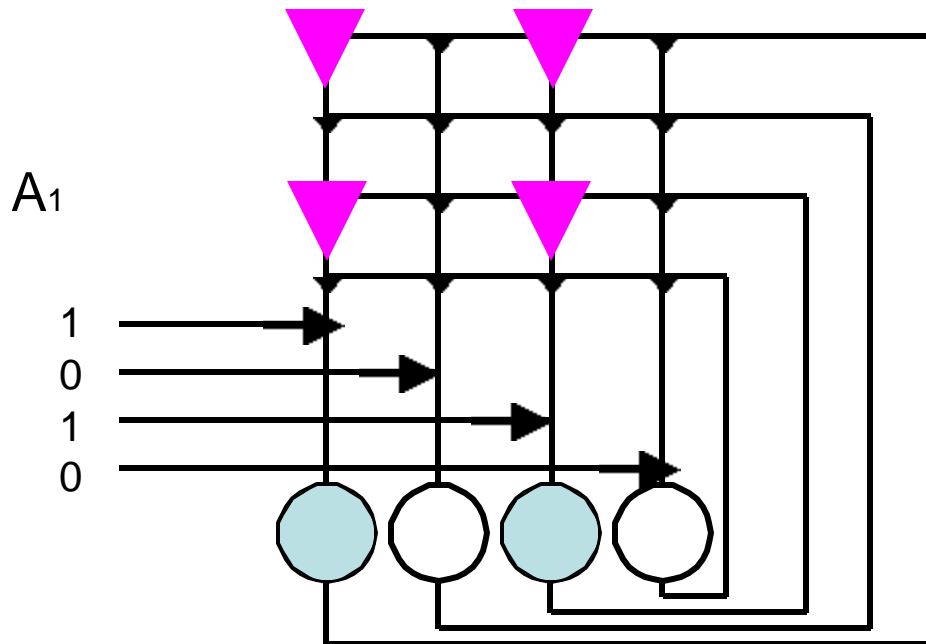
During learning:



- patterns  $A_1 \dots A_n$  are memorized via Hebbian learning

# “Self-organizing” models based on Hebbian learning

## II. Auto-associator



First present pattern  $A_1$

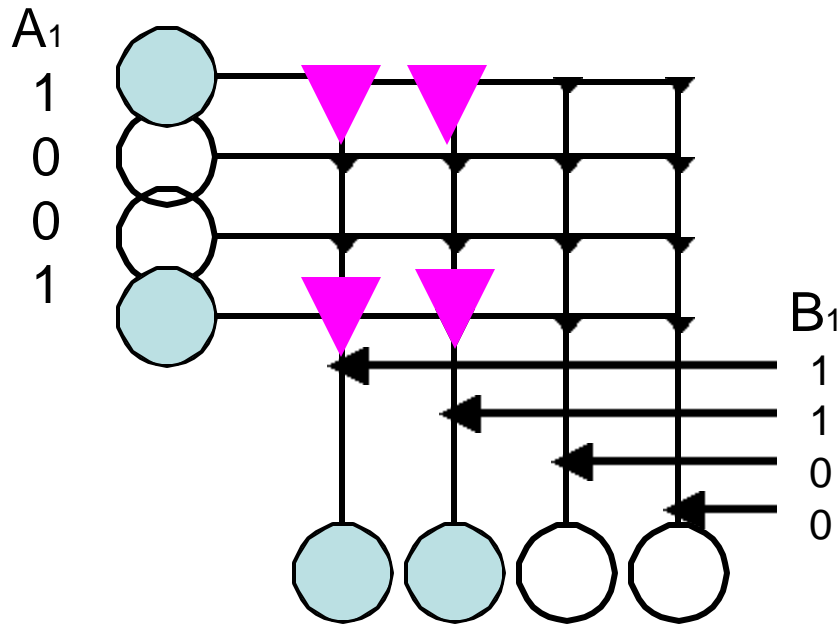
After learning pattern  $A_1$ :

- Some synapses have been strengthened
- Elements of pattern have been associated together



# “Self-organizing” models based on Hebbian learning

## III. Hetero-associator



### During learning:

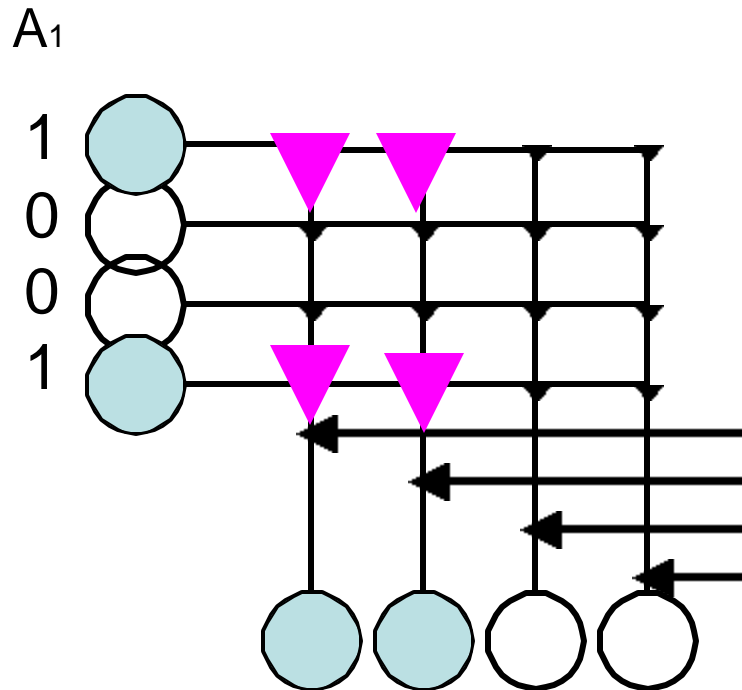
- Present pairs of input, output patterns A,B
- Apply Hebbian learning

### After learning:

- elements of the patterns A,B have been associated together

# “Self-organizing” models based on Hebbian learning

## III. Hetero-associator

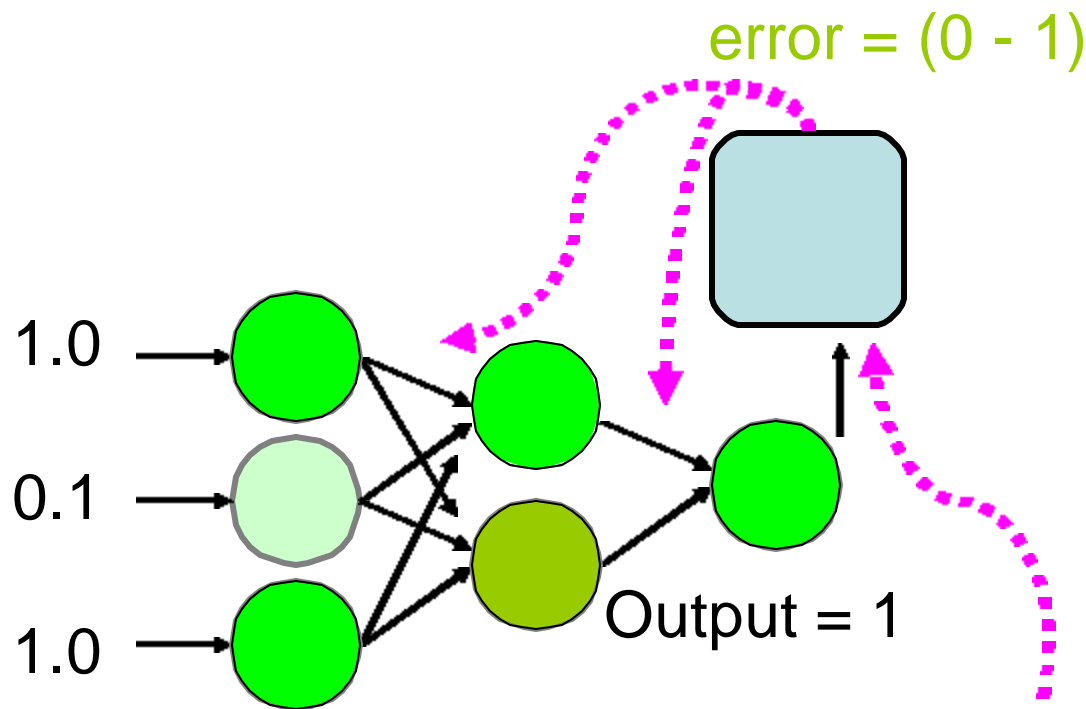


After learning, test with  
 $A_1$  alone

Then activate  $B$  units by  
propagating activity  
through synapses

- correct recall of  $B_1$

# “Supervised” learning based on error back-propagation



- very powerful learning procedure
- biologically implausible

Input  
pattern

“Teaching Signal”:  
Desired output = 0

# Reinforcement learning

