



**DEPARTMENT OF COMPUTER SCIENCE & ENGINEERING
(ARTIFICIAL INTELLIGENCE & MACHINE LEARNING)**

Hebbian Rule

- This rule was proposed by Donald Hebb in 1949.
- If neuron i is near enough to excite neuron j and repeatedly participate in its activation, the synaptic connection between these neurons is strengthened and neuron j becomes more sensitive to stimuli from neuron i .
- Hebb's law can be represented in the form of 2 rules:
 - i) If 2 neurons on either side of connection are activated synchronously, then the weight of that connection is increased.
 - ii) If 2 neurons on either ~~of~~ side of connection are activated asynchronously, then the weight of that connection is decreased.
- If 2 neurons don't have any relationship, then the weight will not increase.



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- The weight update in Hebb rule is given by,

$$\underline{w_i(\text{new}) = w_i(\text{old}) + x_i y}$$

- The Hebb rule is more suitable for bipolar data, than binary data.

~~Flowchart~~ :

Training algorithm:

step 0: initialize weights = 0

step 1: step 2 ~~to~~ have to be performed for each training input vector, and target o/p pair i.e. $s:t$

step 2: weight adjustments and bias adjustments are as follows :

$$w_i(\text{new}) = w_i(\text{old}) + x_i y$$

$$b(\text{new}) = b(\text{old}) + y$$

Here, the change in weight can be expressed as

$$\Delta w = x y$$

$$\therefore \underline{w(\text{new}) = w(\text{old}) + \Delta w}$$



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Flowchart :

