

Perceptron Algorithms

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ML 2 BDA3321

Neurons

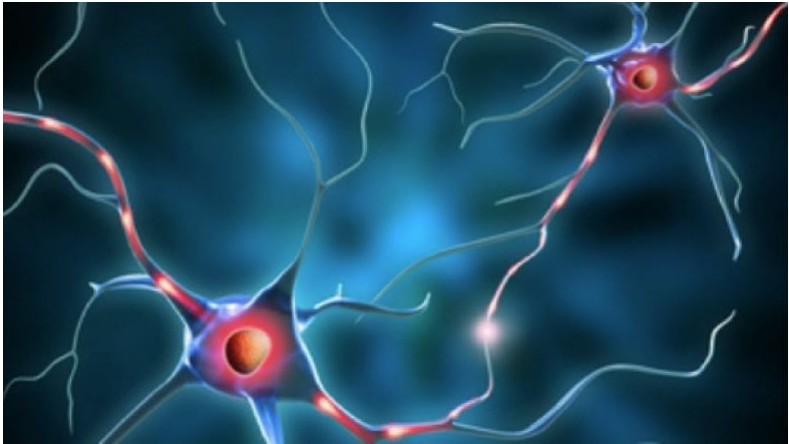


Figure: <https://readbiology.com/types-of-neurons/>

Neurons

McCulloch and Pitts Neurons

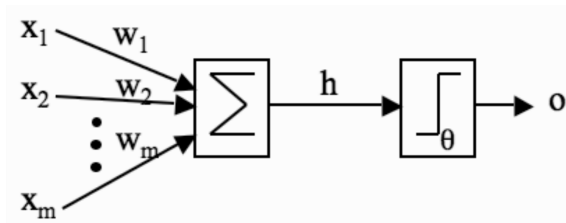


Figure: Neuron, [Mar14]

1. Weighted Inputs
2. Adder
3. Activation Function

McCulloch and Pitt Neurons

Components

Weighted Inputs

For each input node, x_i a weight w_i is assigned.

Adder

$$h = \sum_{i=1}^m x_i w_i$$

Activation Function and Output

A threshold function decided if the neuron fires.

$$o = g(h) = \begin{cases} 1 & \text{if } h > \theta \\ 0 & \text{if } h \leq \theta \end{cases}$$

Neural Networks

Perceptron

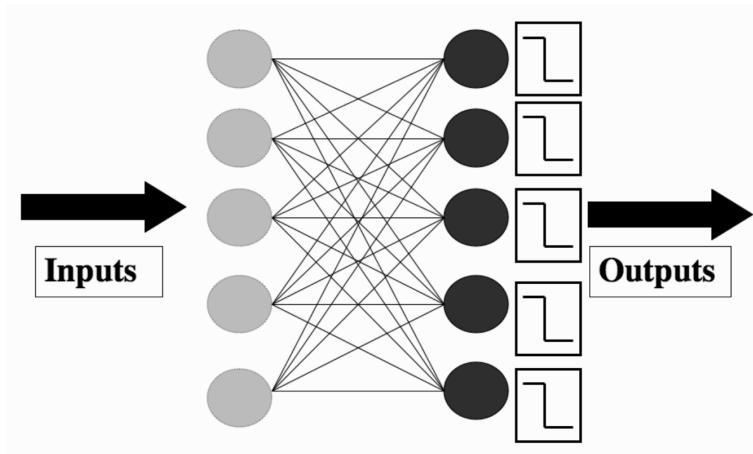


Figure: Perceptron, [Mar14]

Perceptron

- ▶ We label weights for neurons as w_{ij} ; where i is the index running over number of inputs, and j index runs over number of neurons.
- ▶ To check if a neuron fires or not, we use adder and activation function as defined on slide 4.
- ▶ The output is a vector of 0s and 1s.
- ▶ Compare this pattern to the target.

Perceptron

Training the perceptron

Learning Rate

$$w_{ij} \leftarrow w_{ij} - \eta(y_j - t_j) \cdot x_i$$

Bias Input

Add a fixed non zero input weight.

Perceptron

The Perceptron Algorithm

Initialize:

- Set all of the weights to small random numbers.

Training:

- For T iterations or until all outputs are correct

For each input vector

Compute the activation of each neuron j using

$$y_j = g(\sum_{i=0}^m w_{ij}x_i) = \begin{cases} 1 & \text{if } \sum_{i=1}^m w_{ij}x_i > 0 \\ 0 & \text{if } \sum_{i=1}^m w_{ij}x_i \leq 0 \end{cases}$$

Update each weight individually using

$$w_{ij} \leftarrow w_{ij} - \eta(y_j - t_j) \cdot x_i.$$

Recall:

Compute the activation of each neuron j using:

$$y_j = g(\sum_{i=0}^m w_{ij}x_i) = \begin{cases} 1 & \text{if } w_{ij}x_i > 0 \\ 0 & \text{if } w_{ij}x_i \leq 0 \end{cases}$$

References I

[Mar14] Stephen Marsland. *Machine Learning, an algorithmic perspective*. CRC Press, 2014.