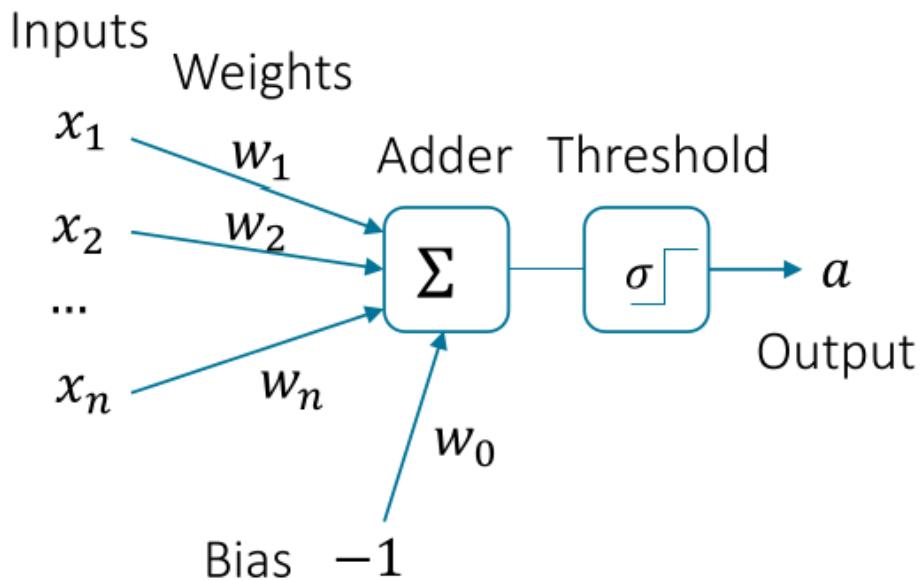




1. NEURON MODEL

Given n numerical features x_j , $j = \{1, \dots, n\}$ $a(x, w) = \sigma(\langle w, x \rangle) = \sigma(\sum_{j=1}^n w_j x_j - w_0)$, where $w_0, w_1, \dots, w_n \in \mathbb{R}$ are feature weights, $\sigma(z)$ is an activation function, for example, $\text{sign}(z) = \begin{cases} +1, z \leq 0 \\ -1, z < 0 \end{cases}$, $\text{sigmoid}(z) = \frac{1}{1+e^{-z}}$, etc.



Linear neuron model
(McCulloch and Pitts, 1943)

2. PERCEPTRON MODEL

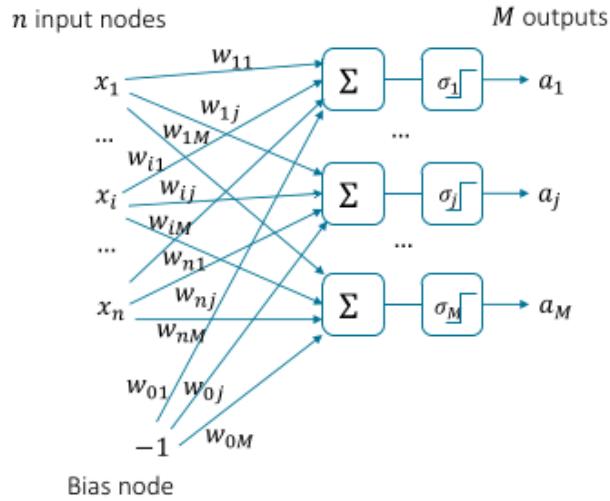
DEFINITION 1.

The perceptron is a collection of neurons together with a set of inputs and weights to associate the inputs to the neurons.

- Weights $w_{ij}, i = \{1, \dots, n\}, j = \{1, \dots, M\}$
- The rule for updating a weight w_{ij}

$$w_{ij} = w_{ij} - \mu(a_j - y_j) \cdot x_i \quad (1)$$

, where y_j is the target value, $0.1 \leq \mu \leq 0.4$ is the learning rate.



DEFINITION 2: The perceptron learning algorithm.

Initialization

Set all the weights w_{ij} to small positive and negative random values

Training phase

for T iterations or until all the outputs are correct

for each input vector

 compute the activation of each neuron using *activation function* σ

$$a_j = \sigma(\sum_{i=0}^M w_{ij}x_i) = \begin{cases} 1 & \text{if } \sum_{i=0}^M w_{ij}x_i > 0 \\ 0 & \text{if } \sum_{i=0}^M w_{ij}x_i \leq 0 \end{cases}$$

 update each of the weights individually using

$$w_{ij} := w_{ij} - \mu(a_j - y_j) \cdot x_i$$

Recall phase

compute the activation of each neuron as $a_j = \sigma(\sum_{i=0}^M w_{ij}x_i) = \begin{cases} 1 & \text{if } \sum_{i=0}^M w_{ij}x_i > 0 \\ 0 & \text{if } \sum_{i=0}^M w_{ij}x_i \leq 0 \end{cases}$