

function BACK-PROP-LEARNING(*examples*, *network*) **returns** a neural network

inputs: *examples*, a set of examples, each with input vector \mathbf{x} and output vector \mathbf{y}
network, a multilayer network with L layers, weights $w_{i,j}$, activation function g

local variables: Δ , a vector of errors, indexed by network node

for each weight $w_{i,j}$ **in** *network* **do**
 $w_{i,j} \leftarrow$ a small random number

repeat

for each example (\mathbf{x}, \mathbf{y}) **in** *examples* **do**
 */ * Propagate the inputs forward to compute the outputs */*
 for each node i **in the input layer do**
 $a_i \leftarrow x_i$
 for $\ell = 2$ **to** L **do**
 for each node j **in layer** ℓ **do**
 $in_j \leftarrow \sum_i w_{i,j} a_i$
 $a_j \leftarrow g(in_j)$
 */ * Propagate deltas backward from output layer to input layer */*
 for each node j **in the output layer do**
 $\Delta[j] \leftarrow y_j - a_j \quad (= -\partial \text{Loss} / \partial in_j)$
 for $\ell = L - 1$ **to** 1 **do**
 for each node i **in layer** ℓ **do**
 $\Delta[i] \leftarrow g(in_i)(1 - g(in_i)) \sum_j w_{i,j} \Delta[j]$
 */ * Update every weight in network using deltas */*
 for each weight $w_{i,j}$ **in network do**
 $w_{i,j} \leftarrow w_{i,j} + \alpha \times a_i \times \Delta[j]$

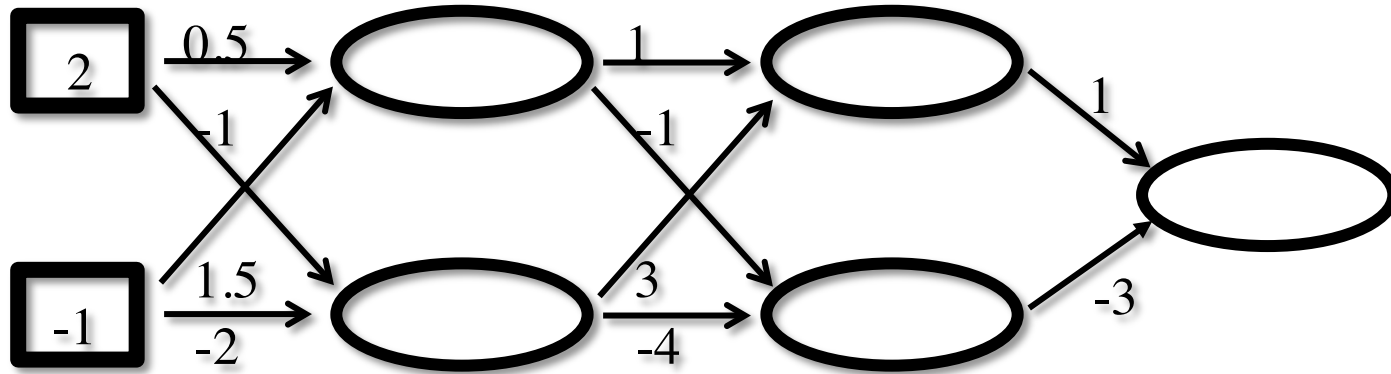
until some stopping criterion is satisfied

return *network*

$\text{Logistic}(\cdot) \equiv g(\cdot)$
 (pour simplifier notation)

Exemple

- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$

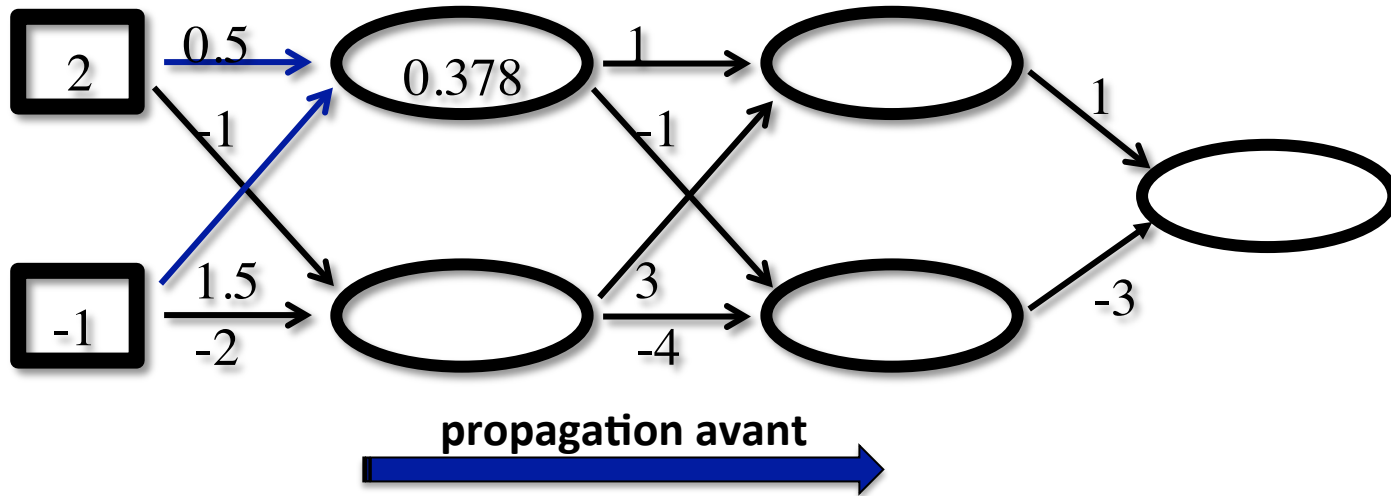


propagation avant

$$a_k = g \left(\sum_j w_{j,k} a_j \right)$$

Exemple

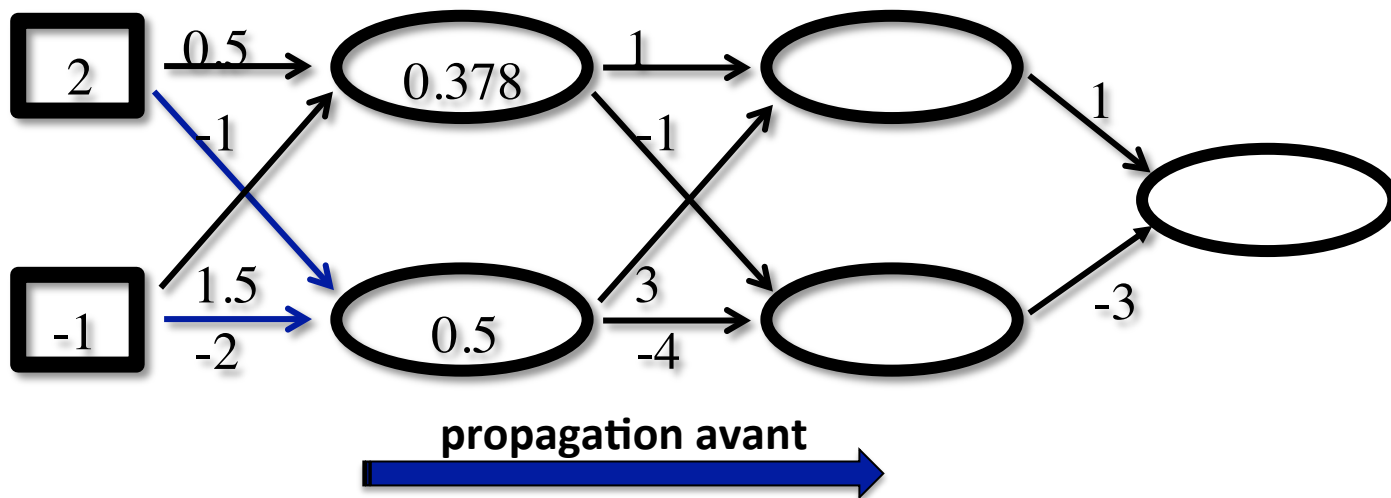
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\text{Logistic}(0.5 * 2 + 1.5 * -1) = \text{Logistic}(-0.5) = 0.378$$

Exemple

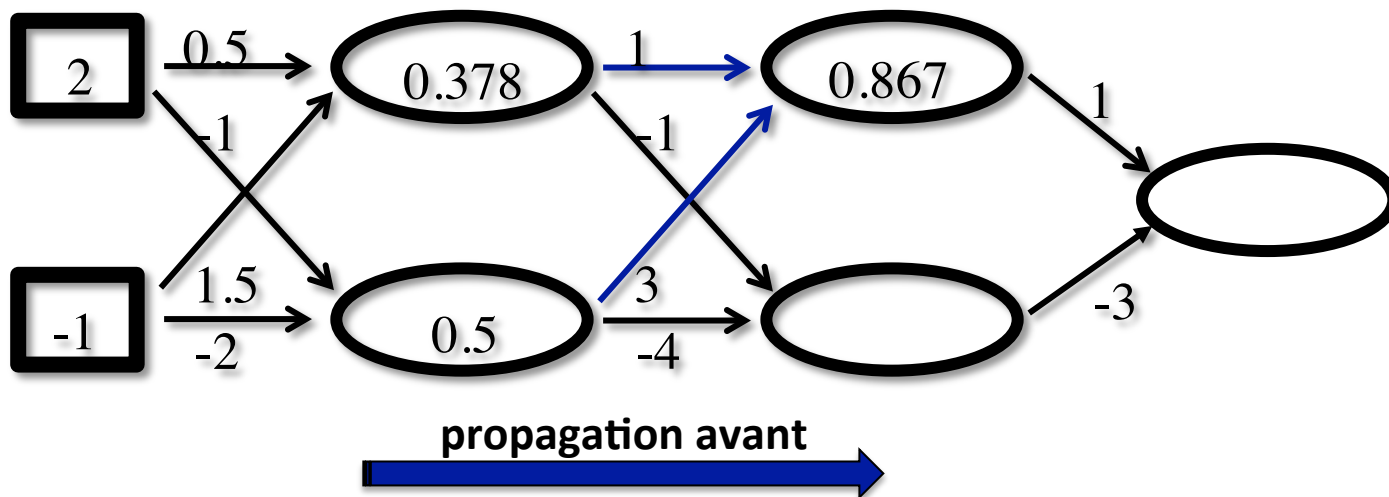
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\text{Logistic}(-1 * 2 + -2 * -1) = \text{Logistic}(0) = 0.5$$

Exemple

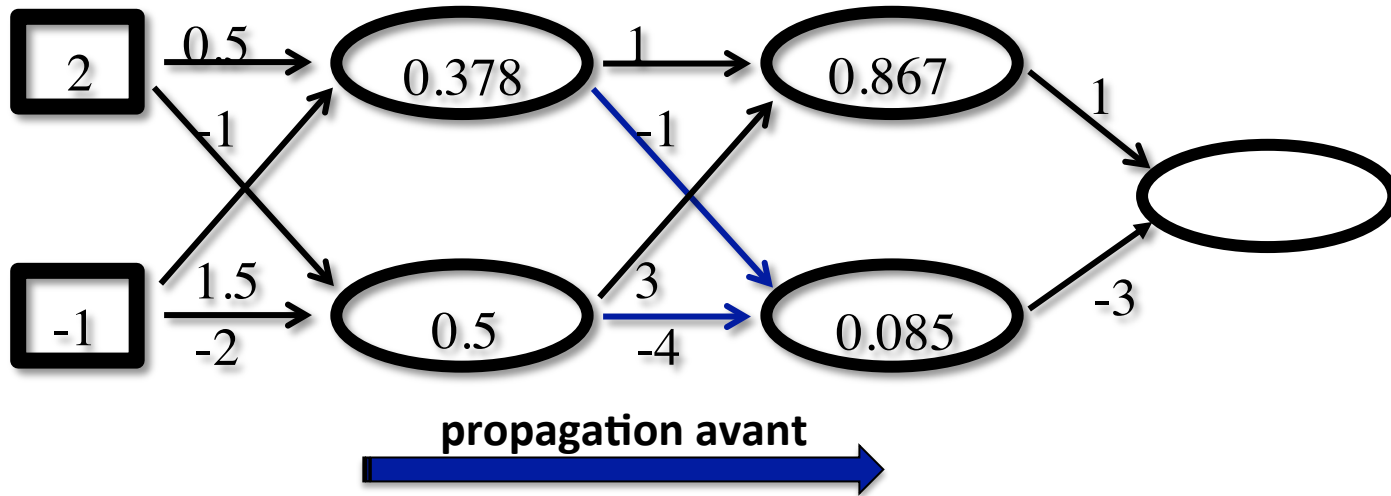
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\text{Logistic}(1 * 0.378 + 3 * 0.5) = \text{Logistic}(1.878) = 0.867$$

Exemple

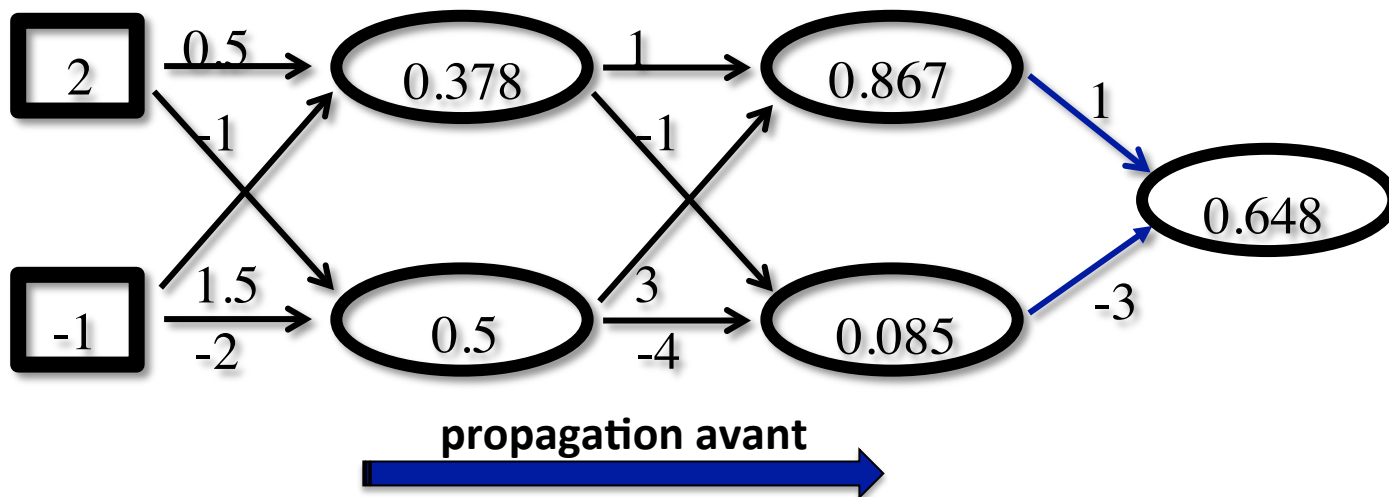
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\text{Logistic}(-1 * 0.378 + -4 * 0.5) = \text{Logistic}(-2.378) = 0.085$$

Exemple

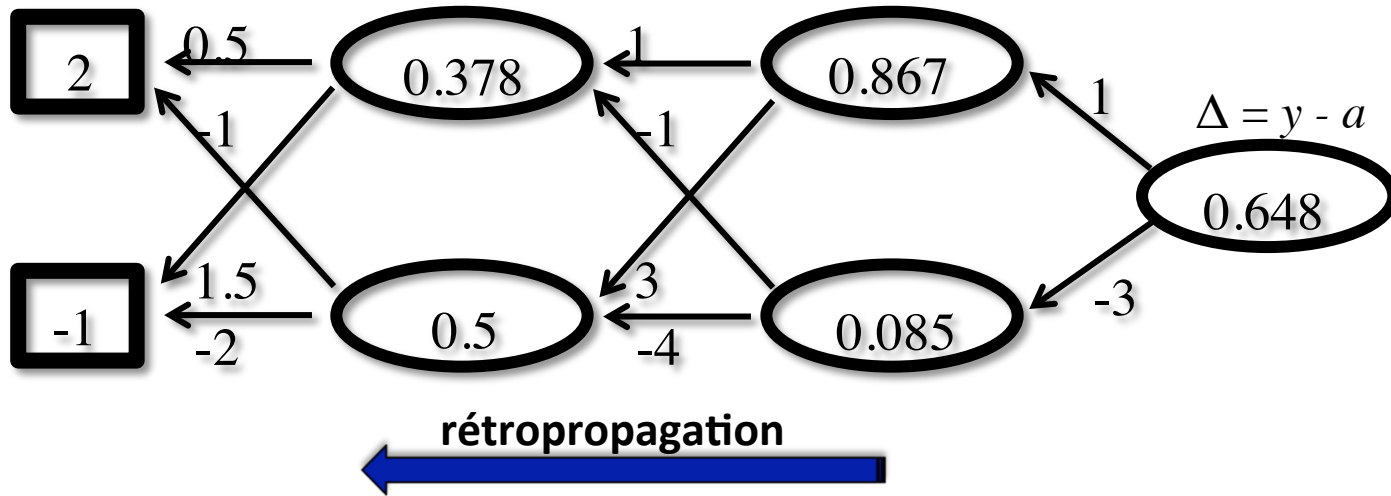
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\text{Logistic}(1 * 0.867 + -3 * 0.085) = \text{Logistic}(0.612) = 0.648$$

Exemple

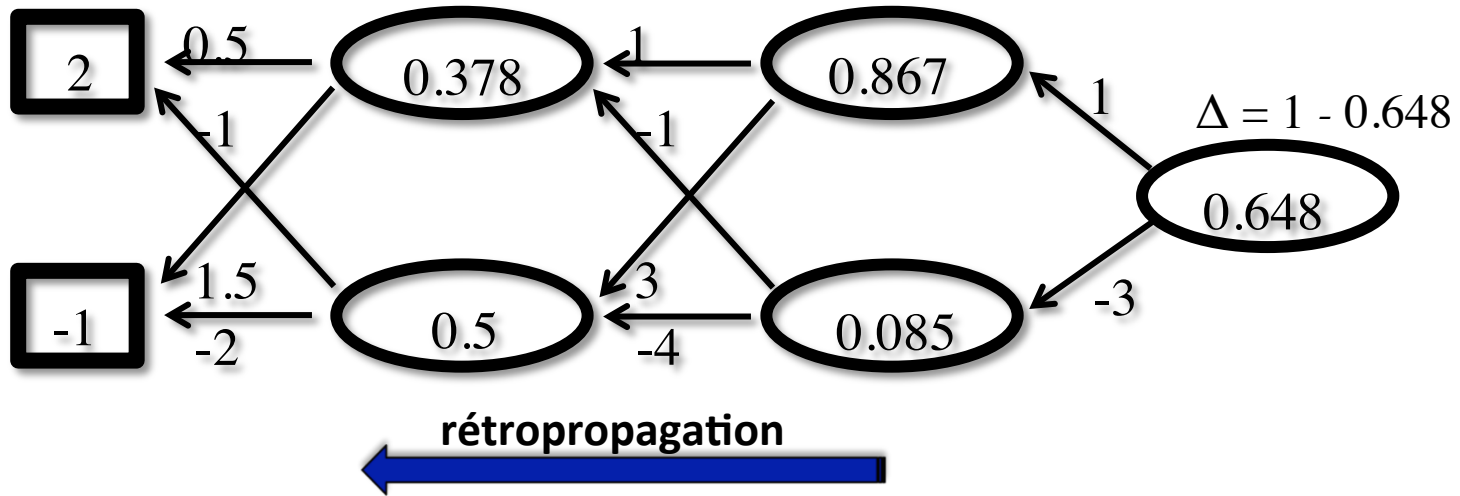
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta[j] = g(in_j)(1 - g(in_j)) \sum_k w_{j,k} \Delta[k]$$

Exemple

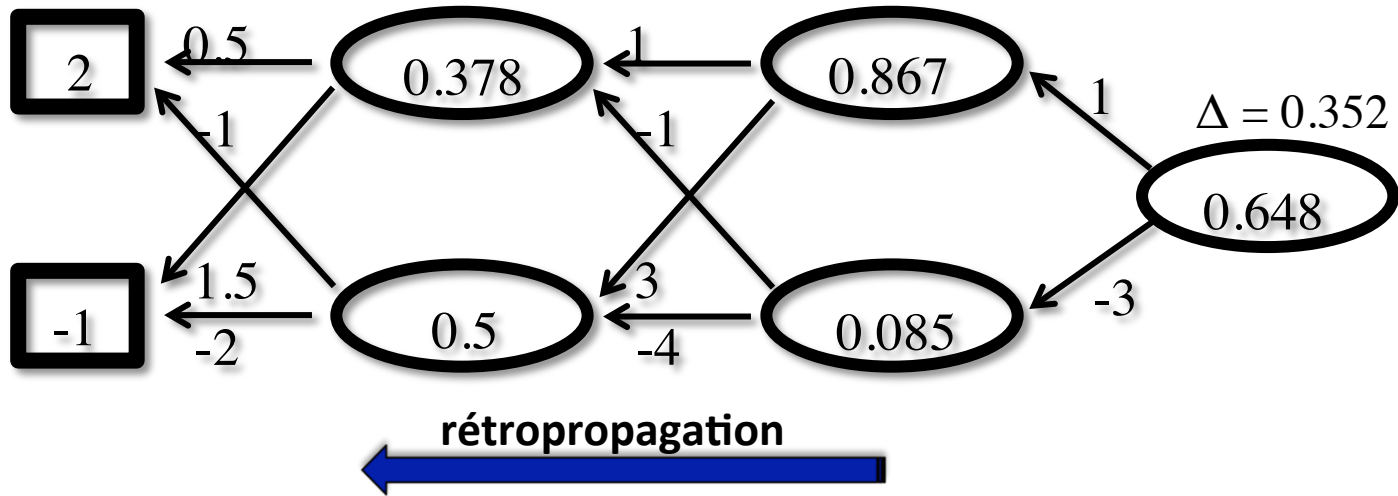
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta[j] = g(in_j)(1 - g(in_j)) \sum_k w_{j,k} \Delta[k]$$

Exemple

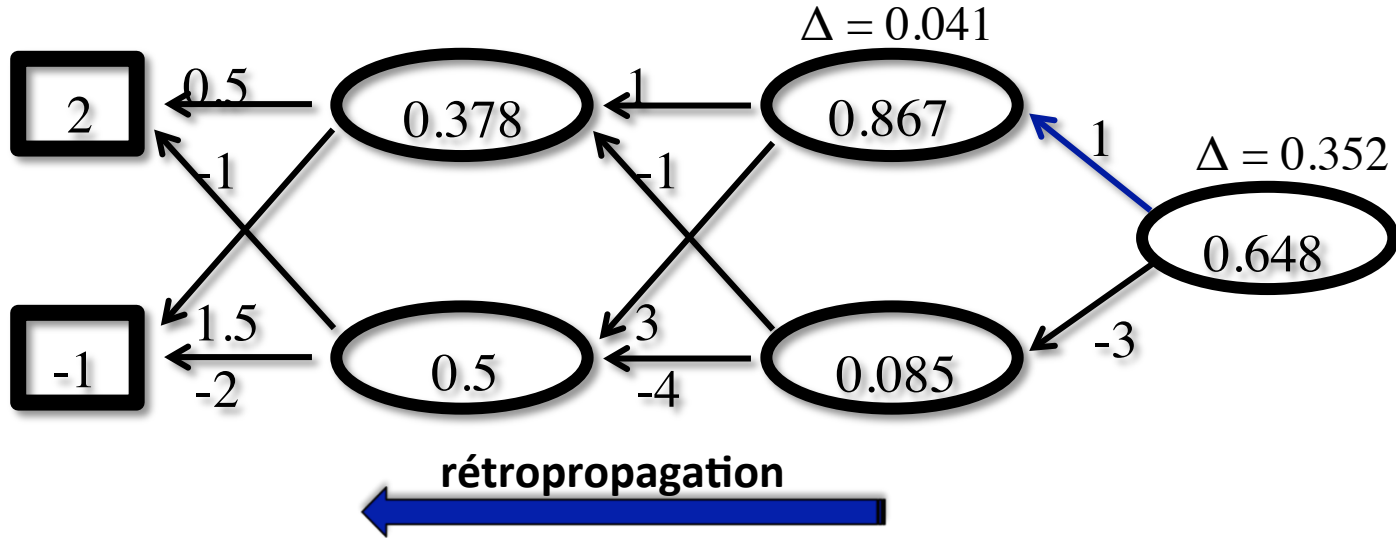
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta[j] = g(in_j)(1 - g(in_j)) \sum_k w_{j,k} \Delta[k]$$

Exemple

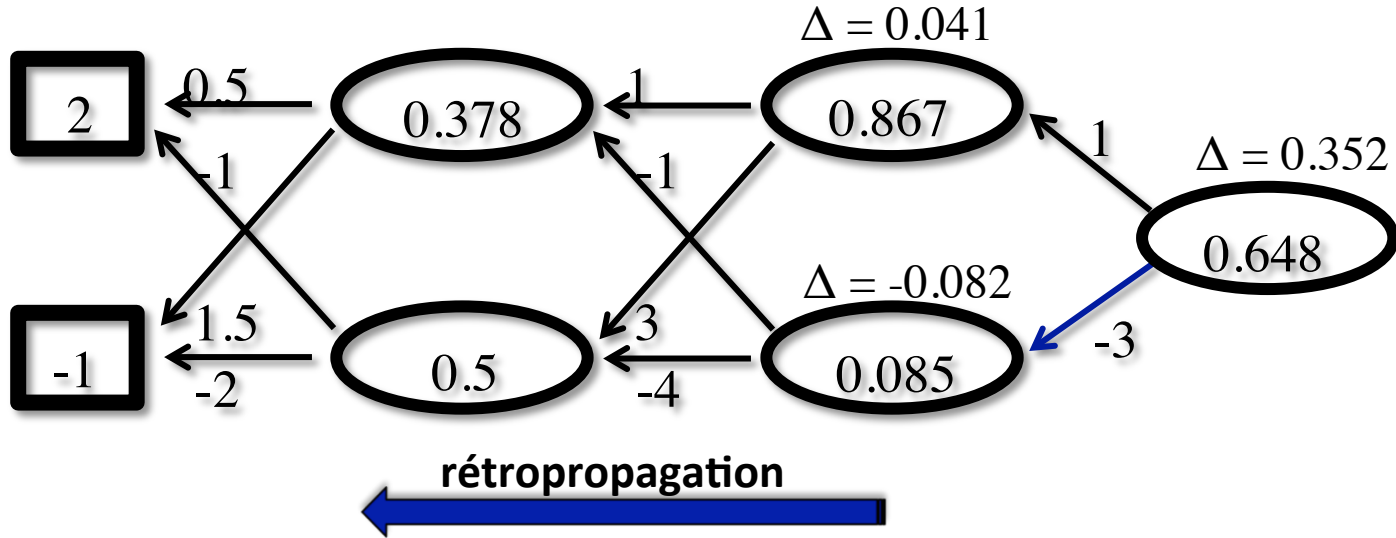
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta = 0.867 * (1 - 0.867) * 1 * 0.352 = 0.041$$

Exemple

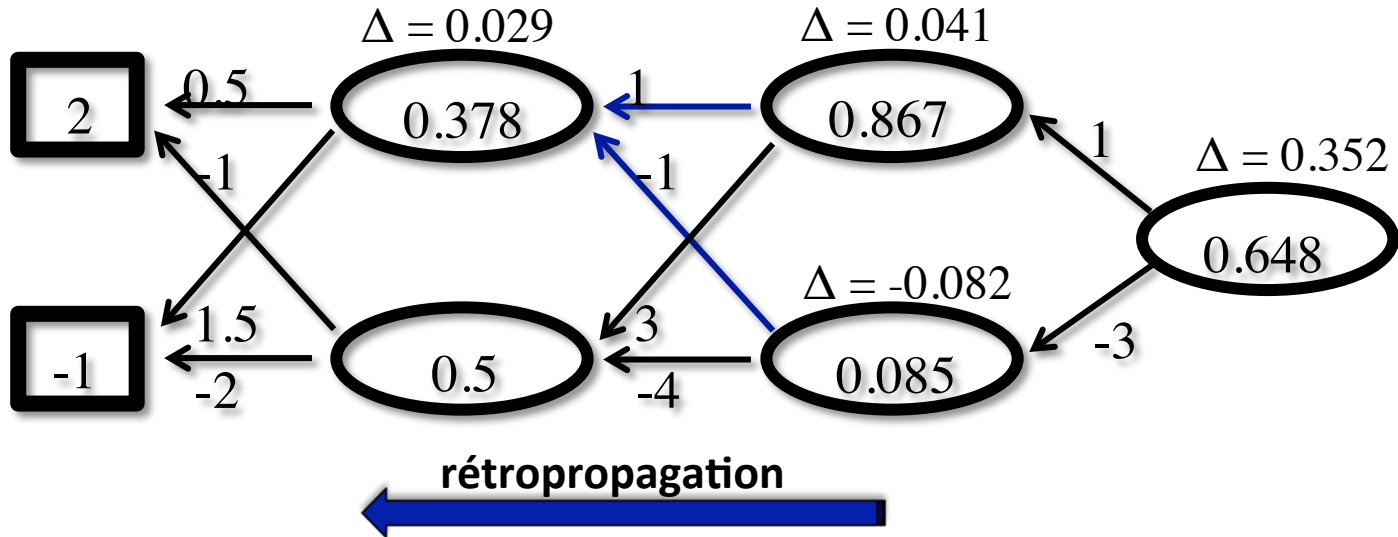
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta = 0.085 * (1 - 0.085) * -3 * 0.352 = -0.082$$

Exemple

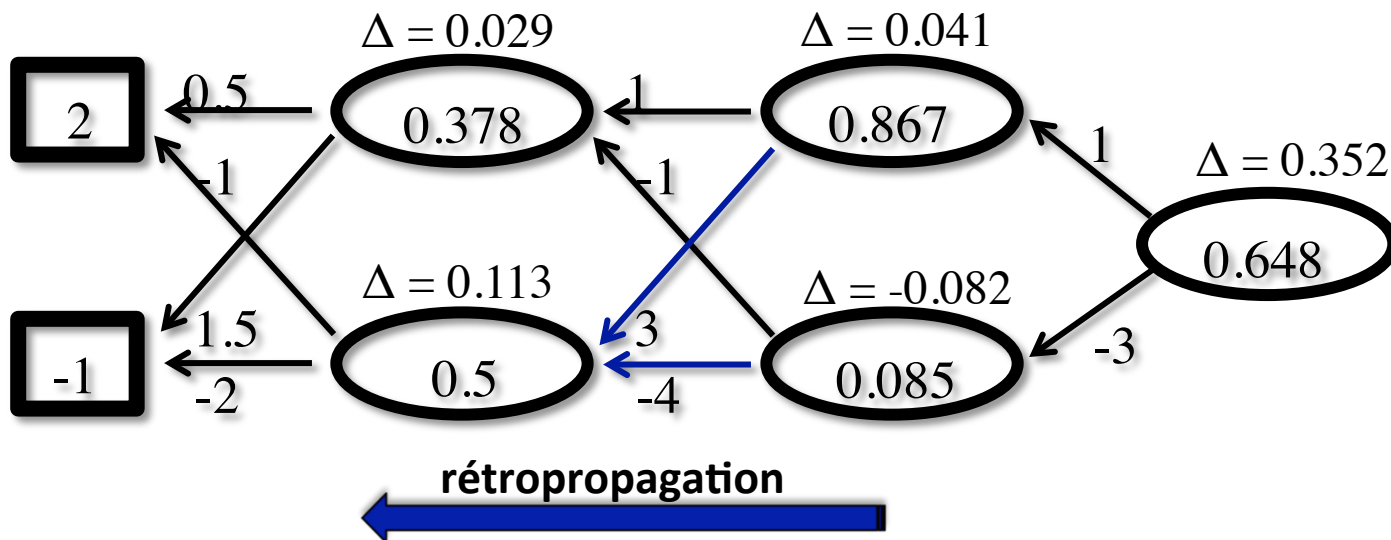
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta = 0.378 * (1 - 0.378) * (1 * 0.041 + -1 * -0.082) = 0.029$$

Exemple

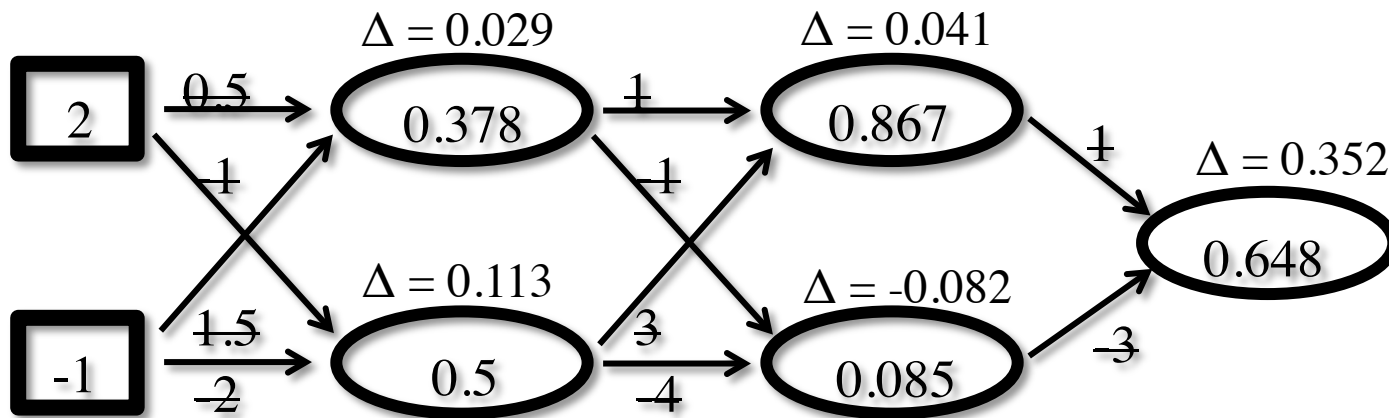
- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



$$\Delta = 0.5 * (1-0.5) * (3 * 0.041 + -4 * -0.082) = 0.113$$

Exemple

- Exemple: $\mathbf{x} = [2, -1]$, $y = 1$



mise à jour ($\alpha=0.1$)

$$w_{1,3} \leftarrow 0.5 + 0.1 * 2 * 0.029 = 0.506$$

$$w_{1,4} \leftarrow -1 + 0.1 * 2 * 0.113 = -0.977$$

$$w_{2,3} \leftarrow 1.5 + 0.1 * -1 * 0.029 = 1.497$$

$$w_{2,4} \leftarrow -2 + 0.1 * -1 * 0.113 = -2.011$$

$$w_{3,5} \leftarrow 1 + 0.1 * 0.378 * 0.041 = 1.002$$

$$w_{3,6} \leftarrow -1 + 0.1 * 0.378 * -0.082 = -1.003$$

$$w_{4,5} \leftarrow 3 + 0.1 * 0.5 * 0.041 = 3.002$$

$$w_{4,6} \leftarrow -4 + 0.1 * 0.5 * -0.082 = -4.004$$

$$w_{5,7} \leftarrow 1 + 0.1 * 0.867 * 0.352 = 1.031$$

$$w_{6,7} \leftarrow -3 + 0.1 * 0.085 * 0.352 = -2.997$$