## Useful information about neural nets

- Neural nets are numerical classifiers with binary (0/1) output.
- The neuron is a primitive circuit element.
- Forward propagation computes the overall output of a neural net.
- Graphical intuition:

inputs  $\rightarrow$  First layer (draw lines)  $\rightarrow$  Remaining layers (perform logic functions)  $\rightarrow$  output (0/1)

- Backward propagation trains a neural net to imitate a given (unknown) function.
  - Threshold trick
  - Helper functions
  - · Hill climbing (gradient ascent)
  - Quick formulas to replace the chain rule

## **Helper functions**

$$Stairstep_{T}(x) = \begin{cases} 1 & \text{if } x \ge T \\ 0 & \text{if } x < T \end{cases}$$

$$Sigmoid_{S,M}(x) = \frac{1}{1 + e^{-S(x-M)}}$$

Performance = Accuracy (out\*, out) = 
$$-\frac{1}{2}$$
 (out\* - out)<sup>2</sup>

## Quick formulas for backward propagation

$$W_{A \rightarrow B, new} = W_{A \rightarrow B, old} + \Delta W_{A \rightarrow B}$$

$$\Delta w_{A \rightarrow B} = r \times out_A \times \delta_B$$

$$\delta_{\text{B}} = \textit{out}_{\text{B}} \, (1 - \textit{out}_{\text{B}}) \, (\textit{out*} \, \text{-} \, \textit{out}) \,$$
 , if neuron B is in final layer

$$\delta_{\mathit{B}} = out_{\mathit{B}} (1 - out_{\mathit{B}}) \sum_{\mathit{outgoing}\,\mathit{C}_{i}} w_{\mathit{B} \rightarrow \mathit{C}_{i}} \delta_{\mathit{C}_{i}} \ \, \text{, if neuron B is } \underline{not} \text{ in final layer}$$

## A single step of backward propagation involves:

- 1. Compute output of each neuron using forward propagation and stairstep function
- 2. Compute  $\delta_B$  for final layer
- 3. Compute  $\delta_B$  for earlier layers
- 4. Compute updates for weights
- 5. Update all weights