

Computational Psycholinguistics

Neural Networks Review



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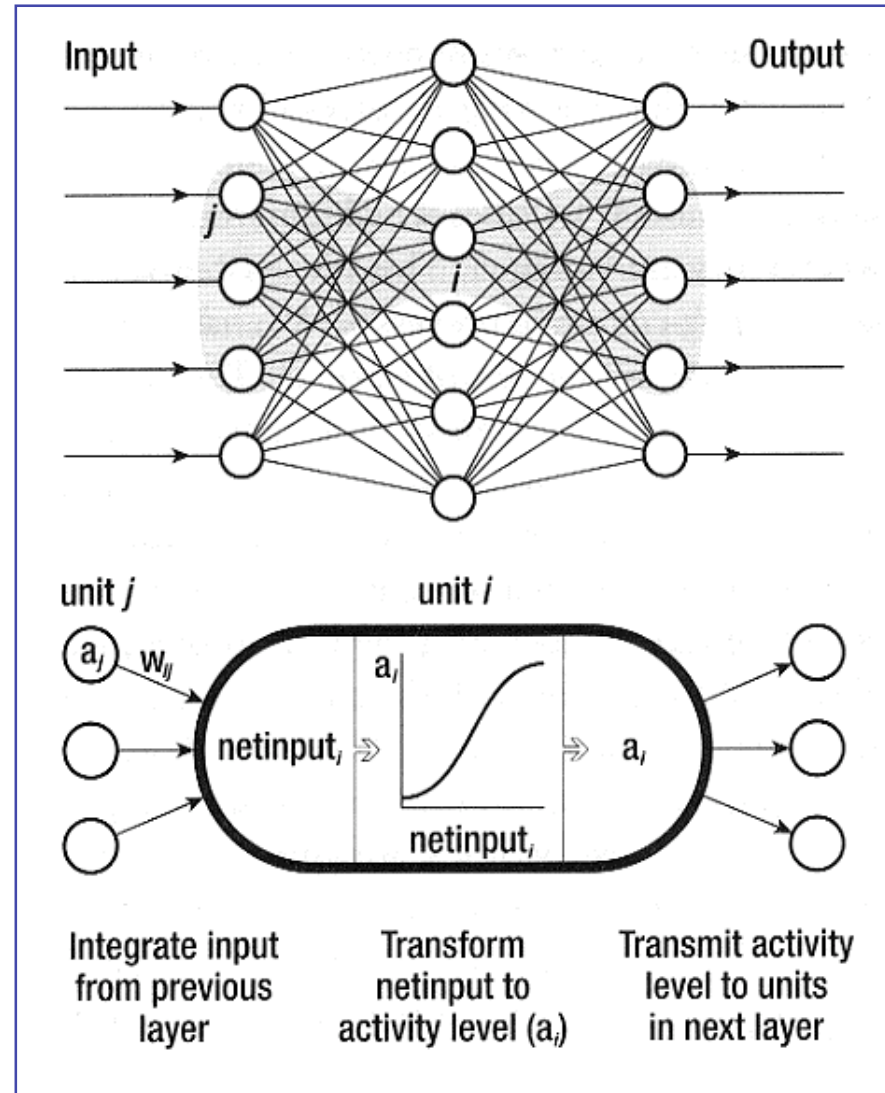
Neural network architecture

- The **activation** of a unit i is represented by the symbol a_i .
- The extent to which unit j influences unit i is determined by the **weight** w_{ij}
- The **input** from unit j to unit i is the product: $a_j * w_{ij}$
- For a node i in the network:

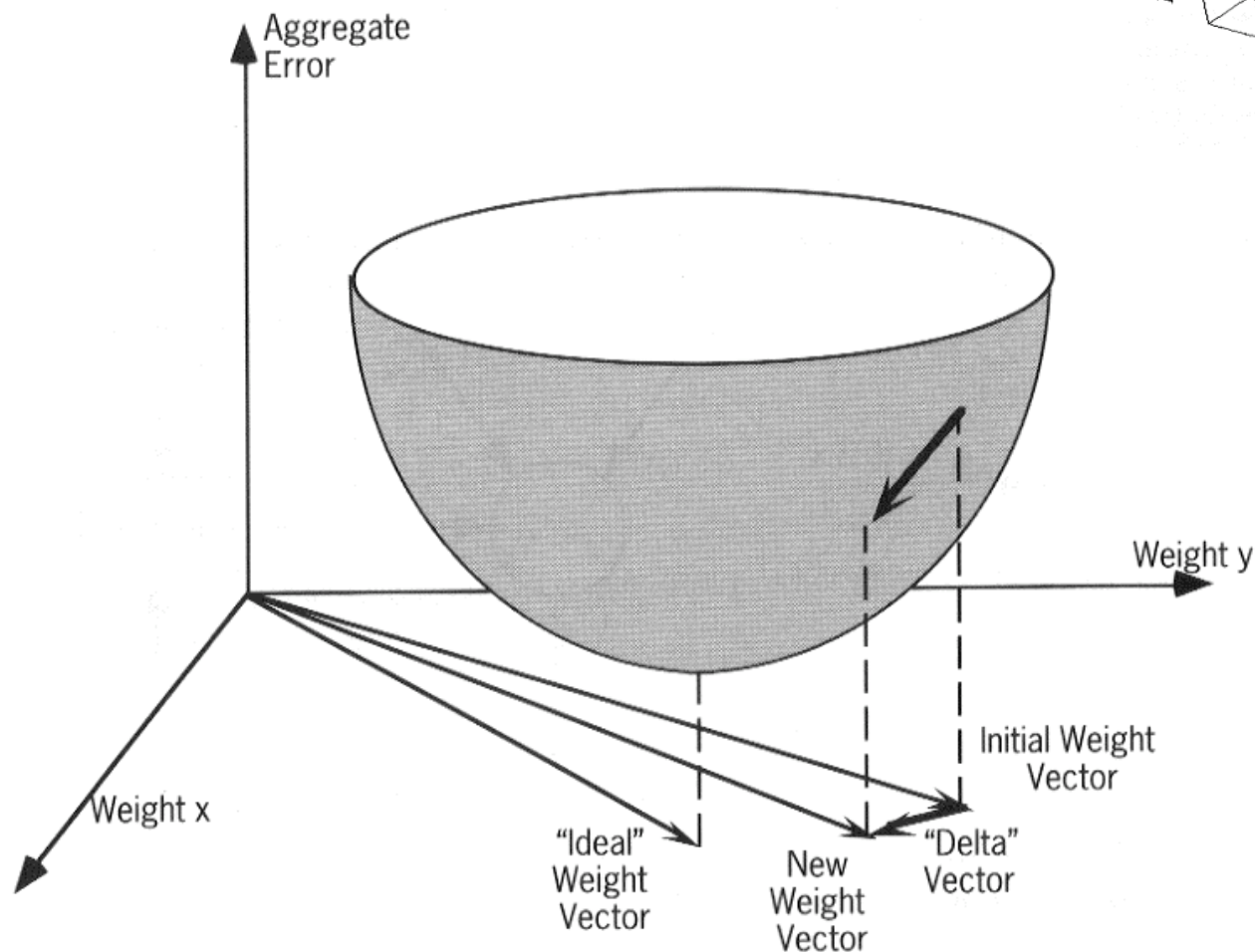
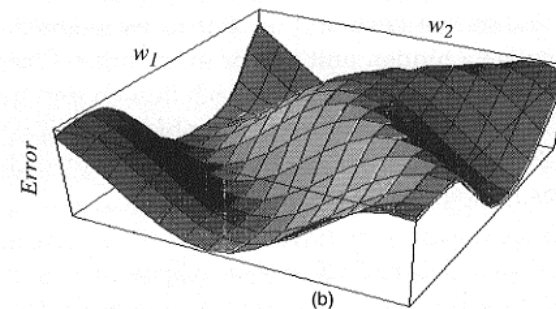
$$netinput_i = \sum_j w_{ij} a_j$$

- The output activation of node i is determined by the activation function, e.g. the logistic:

$$a_i = \sigma(netinput_i) = \frac{1}{1 + e^{-net_i}}$$



Visualising the error “surface”



Learning with the Sigmoid activation function

■ Networks with linear activation functions:

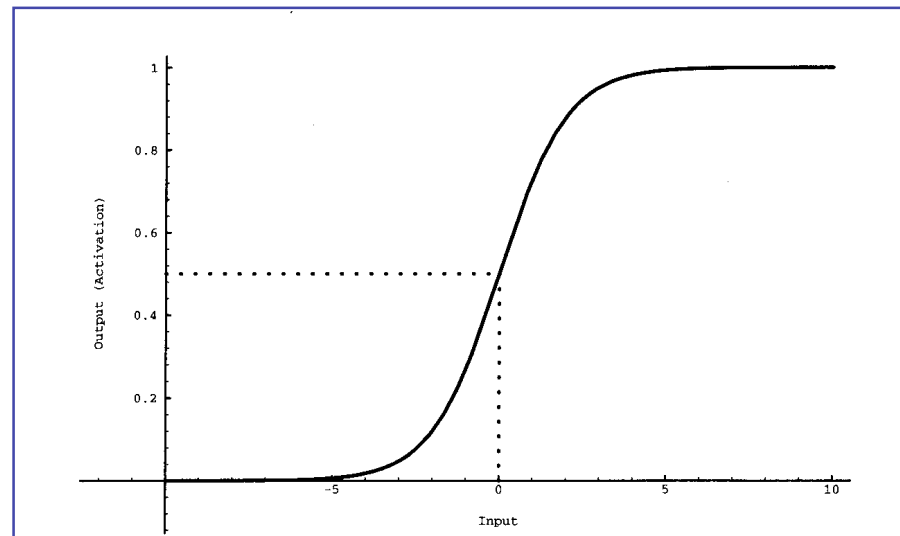
- have mathematically well-defined learning capacities (linear algebra)
- they are known to be limited in the kinds of problems they can solve

■ The logistic, or sigmoid, function is:

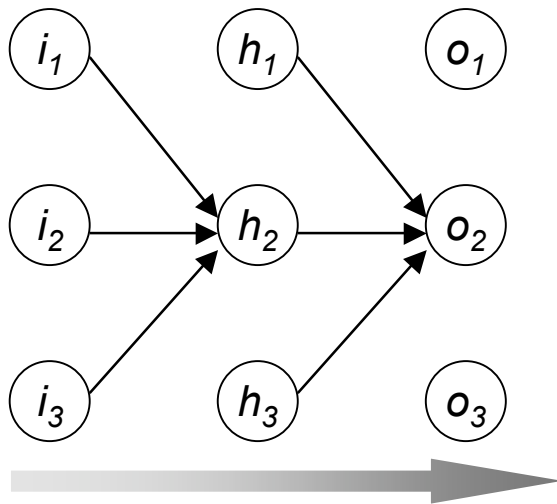
- Nonlinear: more powerful
- More neurologically plausible
- Less well-understood, more difficult to analyse mathematically

■ Recall:

$$a_i = \sigma(\text{net}_i) = \frac{1}{1 + e^{-\text{net}_i}}$$



Backpropagation of Error



(a) Forward propagation of activity :

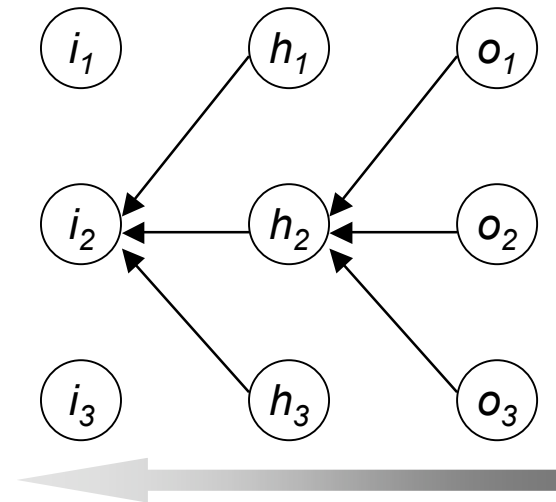
$$\text{net}_{out} = \sum w_{oh} \cdot a_{hidden}$$

$$a_{out} = f(\text{net}_{out})$$

(b) Backward propagation of error :

$$\text{err}_{hidden} = \sum w_{oh} \cdot \delta_{out}$$

$$\delta_{hidden} = f'(\text{net}_{hidden}) \cdot \text{err}_{hidden}$$



Example of Backpropagation

- Consider the following network, containing a single hidden node
- Calculate the weight changes for both layers of the network, assuming learning rate $\varepsilon = 0.3$, inputs 0.4 and 0.7, and targets 0.8 and 0.2

The generalised Delta rule :

$$\Delta w_{ij} = \varepsilon \delta_i a_j$$

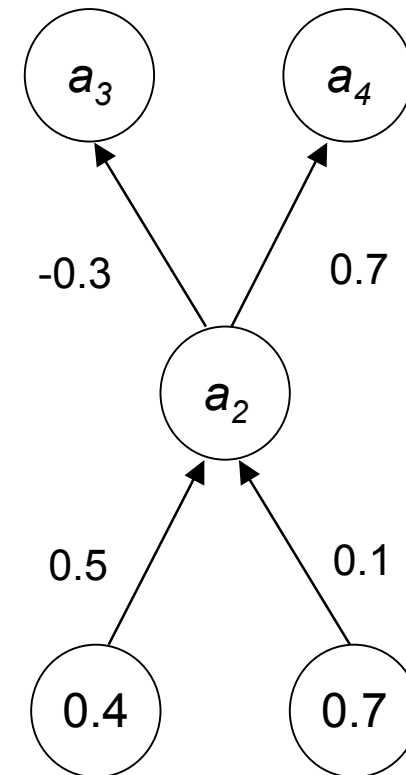
For output nodes :

$$\delta_k = \sigma'(net_k)(t_k - a_k) = a_k(1 - a_k)(t_k - a_k)$$

For hidden nodes :

$$\delta_i = \sigma'(net_i) \sum_k \delta_k w_{ki} = a_i(1 - a_i) \sum_k \delta_k w_{ki}$$

because $\sigma'(net_i) = a_i(1 - a_i)$



Forward and Backpropagation

