### Sheet 4

### November 15, 2015

# Assignment 19

### Assignment 20

Given an array N with n elements and a function called random that.

#### Randomize elements

- 1: for i = 0 such that i < n do
- $i \leftarrow i + 1$
- 3:  $swap \leftarrow random(0, n)$
- 4:  $temp \leftarrow N[swap]$
- 5:  $N[swap] \leftarrow N[i]$
- 6:  $N[i] \leftarrow temp$
- 7: end for

## Assignment 21

**Backpropagation** of errors (BP) is an algorithm used to modify the weights in a Technical Neural Network (TNN) in order to improve the learning of the network.

After the computation of the output values, we calculate the difference between our teacher value  ${}^{p}\hat{y}_{m}$  and each output unit  ${}^{p}y_{m}$ . We obtain the sum of all and apply to it an error function:

$${}^{p}E = \frac{1}{2} \sum_{m=1}^{M} ({}^{p}\hat{y}_{m} - {}^{p}y_{m})^{2}$$

Now, is possible to calculate  $\delta$  values. This calculation is different for the neurons located in the output layer:

$$\delta_m = (\hat{y}_m - y_m) \cdot f'(net_m)$$

and the neurons located in the hidden layers:

$$\delta_h = (\sum_{k=1}^k \delta_k w_{hk}) \cdot f'(net_h)$$

We must be warned that  $\delta_k w_{hk}$  are referring to values in the next layer. The way we calculate the weight changes is using this formula:

$$\Delta w_{ij} = \eta \cdot \delta w_{ij} \cdot out_i$$

We iterate in this process until we reach the input layer. Finally, we update the current weights  $w_{ij}$  adding to them the changed weights  $\Delta w_{ij}$ :

$$\Delta w_{ij} = w_{ij} + \Delta w_{ij}$$

# Assignment 22

# Assignment 23