

## 4.3 NN

**Exercise 1**

Describe the topology for a multilayer network solving the problem of parity check with three coordinate vectors (3-3-1 Network).

Output  $z=1$  (odd number of +1 in  $x$ )

Output  $z=-1$  (odd number of -1)

Give the number of weights and propose some laws based on the hard limiting function:  $\text{sign}(x)$  to build the complete function.

**Exercise 2**

Show that the  $\Delta \mathbf{w}[n]$  weight adaptation increment in the Levenberg-Marquadt algorithm is equivalent to  $\Delta \mathbf{w}[n]$  that minimises the error function:

$J[n+1] \approx J[n] + \nabla J_{\mathbf{w}} \Delta \mathbf{w}[n] + \frac{1}{2} \Delta \mathbf{w}[n]^T \mathbf{H} \Delta \mathbf{w}[n] + \lambda (\|\Delta \mathbf{w}[n]\|^2 - K)$ , where  $\lambda$  is a positive Lagrange Multiplier and  $K$  is a constant to control the trust-region where the approximation for  $J[n+1]$  is valid.

**Exercise 3 (Optativo)**

En el diseño de redes neuronales basadas en funciones base radiales, demuestre que al minimizar la función de error  $\frac{1}{2} \text{Trace}((\Phi^T(\mathbf{x})\mathbf{W}^T - \mathbf{T}^T)(\mathbf{W}\Phi(\mathbf{x}) - \mathbf{T}))$  respecto a la matriz

$\mathbf{W} \in \mathbb{R}^{c \times h}$ , la solución obtenida es:  $\mathbf{W}^T = (\Phi(\mathbf{x})\Phi^T(\mathbf{x}))^{-1} \Phi(\mathbf{x})\mathbf{T}^T$

Nota: Recuerde que  $\nabla_{\mathbf{R}} (\text{Trace}(\mathbf{A}\mathbf{R}\mathbf{B})) = \mathbf{A}^T \mathbf{B}^T$  with  $\mathbf{A} \in \mathbb{R}^{a \times b}$ ;  $\mathbf{R} \in \mathbb{R}^{b \times c}$ ;  $\mathbf{B} \in \mathbb{R}^{c \times a}$

$$\nabla_{\mathbf{R}} (\text{Trace}(\mathbf{A}\mathbf{R}\mathbf{B})) = \left( \nabla_{\mathbf{R}^T} (\text{Trace}(\mathbf{A}\mathbf{R}\mathbf{B})^T) \right)^T$$

$$\nabla_{\mathbf{R}^T} (\text{Trace}(\mathbf{A}\mathbf{R}^T \mathbf{B})) = 2\mathbf{A}^T \mathbf{B}^T \mathbf{R}^T \text{ with } \mathbf{A} \in \mathbb{R}^{a \times b}; \mathbf{R} \in \mathbb{R}^{c \times b}; \mathbf{B} \in \mathbb{R}^{b \times a}$$