

5.16 In case of multiple outputs, the error function is given by:

$$E = \frac{1}{2} \sum_{n=1}^N \|\mathbf{y}_n - \mathbf{t}_n\|^2 = \frac{1}{2} \sum_{n=1}^N (\mathbf{y}_n - \mathbf{t}_n)^T (\mathbf{y}_n - \mathbf{t}_n)$$

The first derivative is given by:

$$\nabla E = \sum_{n=1}^N \nabla \mathbf{y}_n^T (\mathbf{y}_n - \mathbf{t}_n)$$

Assuming K outputs, \mathbf{y}_n has K elements, and assuming \mathbf{w} has W elements, $\nabla \mathbf{y}_n$ will be a matrix of size $K * W$.

Let $\mathbf{B}_n = \nabla \mathbf{y}_n$.

The second derivative is given by:

$$\begin{aligned} \nabla \nabla E &= \left(\sum_{n=1}^N \nabla \mathbf{y}_n^T \nabla \mathbf{y}_n \right) + \left(\sum_{n=1}^N (\mathbf{y}_n - \mathbf{t}_n)^T \nabla \nabla \mathbf{y}_n \right) \\ &= \left(\sum_{n=1}^N \mathbf{B}_n^T \mathbf{B}_n \right) + \left(\sum_{n=1}^N (\mathbf{y}_n - \mathbf{t}_n)^T \nabla \nabla \mathbf{y}_n \right) \end{aligned}$$

Neglecting the second term as done in the book, we get:

$$\Rightarrow \mathbf{H} \simeq \sum_{n=1}^N \mathbf{B}_n^T \mathbf{B}_n$$