

Exercises

Paul Scemama

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1 Chapter 1

1.1

Note: although the exercise asks for an arbitrary w_i , I will do it for the entire \mathbf{w} , and then show that each row of the resultant solution vector corresponds to the necessary scalar equation mentioned in the book. The reason being is that I believe it is more instructive.

Get derivative of (1.2) with respect to \mathbf{w} . $\frac{\partial E}{\partial \mathbf{w}}$ is the derivative of a function with respect to a vector, and thus the output will be a vector.

$$\frac{\partial E}{\partial \mathbf{w}} = \frac{\partial E}{\partial y(x_n, \mathbf{w})} \cdot \frac{\partial y(x_n, \mathbf{w})}{\partial \mathbf{w}} \quad (\text{Chain rule})$$

We will first compute $\frac{\partial E}{\partial y(x_n, \mathbf{w})}$.

$$\frac{\partial E}{\partial y(x_n, \mathbf{w})} = \sum_{n=1}^N \{y(x_n, \mathbf{w}) - t_n\}$$

And $\frac{\partial y(x_n, \mathbf{w})}{\partial \mathbf{w}}$.

$$\frac{\partial y(x_n, \mathbf{w})}{\partial \mathbf{w}} = [1 \ x_n^1 \ x_n^2 \ x_n^3 \dots x_n^M]$$

So,

$$\frac{\partial E}{\partial \mathbf{w}} = \underbrace{\sum_{n=1}^N \{y(x_n, \mathbf{w}) - t_n\}}_{\text{scalar}} \cdot \underbrace{\mathbf{x}_n}_{\text{vector}}$$

where $\mathbf{x}_n = [1 \ x_n^1 \ x_n^2 \ x_n^3 \dots x_n^M]$. Rewriting the above equation we get,

$$\begin{aligned} \frac{\partial E}{\partial \mathbf{w}} &= \sum_{n=1}^N \{y(x_n, \mathbf{w}) - t_n\} \cdot \mathbf{x}_n \\ &= \begin{bmatrix} \sum_{n=1}^N \{y(x_n, \mathbf{w}) - t_n\} \cdot x_n^0 \\ \dots \\ \sum_{n=1}^N \{y(x_n, \mathbf{w}) - t_n\} \cdot x_n^M \end{bmatrix} = \begin{bmatrix} \sum_{n=1}^N \{ \sum_{j=0}^M w_j x_n^j - t_n \} \cdot x_n^0 \\ \dots \\ \sum_{n=1}^N \{ \sum_{j=0}^M w_j x_n^j - t_n \} \cdot x_n^M \end{bmatrix} = \begin{bmatrix} \sum_{n=1}^N \{ \sum_{j=0}^M w_j x_n^j x_n^0 - t_n x_n^0 \} \\ \dots \\ \sum_{n=1}^N \{ \sum_{j=0}^M w_j x_n^j x_n^M - t_n x_n^M \} \end{bmatrix} \end{aligned}$$

So each row can be written as

$$\sum_{n=1}^N \left\{ \sum_{j=0}^M w_j x_n^{(j+i)} - t_n x_n^i \right\}$$

where i indicates the row.

Set derivative equal to 0 and rearrange

$$\sum_{n=1}^N \sum_{j=0}^M w_j x_n^{(i+j)} = \sum_{n=1}^N t_n (x_n)^i \quad (1)$$

$$\sum_{j=0}^M \sum_{n=1}^N x_n^{(i+j)} w_j = \sum_{n=1}^N t_n (x_n)^i \quad (2)$$

$$\sum_{j=0}^M A_{ij} w_j = T_i \quad (3)$$

So to get each element of the vector that contains w_i 's that minimize E , we solve (3) for $i = 0 \dots M$. This equates to what Bishop wanted us to show.