

Correction to the derivation of the LS classifier

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Given $\{t_n, x_n\}_{n=1}^N$ as training data, the likelihood is

$$p(t_{1:N}; \tilde{w}, \tilde{x}_{1:N}) = \prod_{n=1}^N \mathcal{N}(t_n; \tilde{w}^T \tilde{x}_n, I). \quad (1)$$

Rather than maximizing the likelihood we will consider the equivalent problem of minimizing the negative log-likelihood, where

$$\begin{aligned} -\log p(t_{1:N}; \tilde{w}, \tilde{x}_{1:N}) &\propto \frac{1}{2} \sum_{n=1}^N (t_n - \tilde{w}_n^T \tilde{x}_n)^T (t_n - \tilde{w}_n^T \tilde{x}_n) \\ &= \frac{1}{2} \sum_{n=1}^N (t_n^T - \tilde{x}_n^T \tilde{w}_n) (t_n^T - \tilde{x}_n^T \tilde{w}_n)^T. \end{aligned} \quad (2)$$

Introducing the matrices

$$T = \begin{pmatrix} t_1^T \\ t_2^T \\ \vdots \\ t_N^T \end{pmatrix} \in \mathbb{R}^{N \times K}, \quad \tilde{X} = \begin{pmatrix} \tilde{x}_1^T \\ \tilde{x}_2^T \\ \vdots \\ \tilde{x}_N^T \end{pmatrix} \in \mathbb{R}^{N \times (D+1)} \quad (3)$$

we can now write (2) according to

$$\begin{aligned} -\log p(t_{1:N}; \tilde{w}, \tilde{x}_{1:N}) &\propto \frac{1}{2} \mathbf{Tr}(T - \tilde{X}\tilde{W})(T - \tilde{X}\tilde{W})^T \\ &= \frac{1}{2} \mathbf{Tr}(T - \tilde{X}\tilde{W})^T (T - \tilde{X}\tilde{W}), \end{aligned} \quad (4)$$

where we made use of the fact $\mathbf{Tr}(AB) = \mathbf{Tr}(BA)$ in order to establish the last equality.