



Why

It is natural to embed a dataset.

Definition

Let $(x : \Omega \rightarrow \mathbf{R}^d, A \in \mathbf{R}^{n \times d}, e : \Omega \rightarrow \mathbf{R}^n)$ be a probabilistic linear model over the probability space $(\Omega, \mathcal{A}, \mathbf{P})$. Let $\phi : \mathbf{R}^d \rightarrow \mathbf{R}^{d'}$ be a feature embedding. Then (x, A, e, ϕ) is an *embedded probabilistic linear model*. We are modeling the function $h : \Omega \rightarrow (\mathbf{R}^d \rightarrow \mathbf{R})$ as linear in the features

$$h_\omega(a) = \phi(a)^\top x(\omega).$$

Correspondence to linear model

Denote the data matrix of the embedded feature vectors by $\phi(A)$. Then, of course, the embedded linear model (x, A, e, ϕ) corresponds to the linear model $(x, \phi(A), e)$.

Normal case

In the normal case, the parameter posterior $g_{x|y}(\cdot, \gamma)$ is a normal density with mean

$$\Sigma_x \phi(A)^\top \left(\phi(A) \Sigma_x \phi(A)^\top + \Sigma_e \right)^{-1} \gamma$$

and covariance

$$\left(\Sigma_x^{-1} + \phi(A)^\top \Sigma_e^{-1} \phi(A) \right)^{-1}.$$

The predictive density for a point $a \in \mathbf{R}^d$ is normal with mean

$$(\phi(a)^\top \Sigma_x \phi(A)^\top + \Sigma_{fe}) \left(\phi(a)^\top \Sigma_x a + \Sigma_f \right)^{-1} \gamma.$$

