



Why

We can cast various common probabilistic regression models into the probabilistic errors linear model by mentioning the input space and feature maps. This unifies our analysis.

Definition

A *line fit model* has input space \mathbf{R} and output space \mathbf{R} . We use a regression function $\phi : \mathbf{R} \rightarrow \mathbf{R}^2$ defined by $\phi(t) = (1, t)^\top$.

We think of $t \in T \subset \mathbf{R}$ as a “dose level” (T is an interval). Given dose levels t_1, \dots, t_ℓ and repetitions n_1, \dots, n_ℓ we obtain the design matrix. Here the regression function generates a line segment embedded in the plane \mathbf{R}^2 . We call the parameters the *intercept parameter* and *slope parameter*.

A *parabola fit model* has input space \mathbf{R} and output space \mathbf{R} . We use a regression function $\phi : \mathbf{R} \rightarrow \mathbf{R}^3$ defined by $\phi(t) = (1, t, t^2)^\top$. Here the regression space is a segment of a parabola embedded in space \mathbf{R}^3 (since $t \in T$ an interval).

These two are instance of *polynomial fit models* of degree $d \geq 1$, in which the regression function becomes $\phi : \mathbf{R} \rightarrow \mathbf{R}^{d+1}$ defined by $\phi(t) = (1, t, t^2, \dots, t^d)^\top$. In this case, the regression range $\phi(T)$ is a one-dimensional curve embedded in \mathbf{R}^{d+1} . In cases in which it is clear that the input space is a single real variable t , a linear model for a line fit (parabola fit, polynomial fit of degree d) is called a *first-degree model* (*second-degree model*, *dth degree model*).

***m*-way models**

We can generalize to *m*-way *d*th degree polynomial fit models in which the input space is $X \subset \mathbf{R}^m$ and the regression function $\phi : \mathbf{R}^m \rightarrow \mathbf{R}^k$ (k is $d+m$ choose d) is the vector of all monomials of degree d in m variables.

For example, a two-way third-degree model has a regression function

$$\phi(t_1, t_2) = \begin{bmatrix} 1 & t_1 & t_2 & t_1^2 & t_1 t_2 & t_2^2 & t_1^3 & t_1^2 t_2 & t_1 t_2^2 & t_2^3 \end{bmatrix}^\top.$$

Or consider a three way second-degree model with regression function

$$\phi(t_1, t_2, t_3) = \begin{bmatrix} 1 & t_1 & t_2 & t_3 & t_1^2 & t_1 t_2 & t_1 t_3 & t_2^2 & t_2 t_3 & t_3^2 \end{bmatrix}^\top.$$

Both models will result in parameter vectors of size ten. We call these models *saturated* because they have every possible *d*th degree power or cross product of variables. In generally, a *m*-way *d*th degree model has $d+m$ choose d mean parameters.

In contrast to saturated models we can talk about *nonsaturated* models. For example, a nonsaturated two-way second-degree model has $\phi : \mathbf{R}^2 \rightarrow \mathbf{R}^4$ where $\phi(t_1, t_2) = (1, t_1, t_2, t_1^2)^\top$.

