

Probability Model For A Multi-Predictor Linear Regression Model

Nan Wu
nanw@udel.edu

1 Linear Regression Model

The linear regression model is given as

$$y = \alpha + X\beta + \epsilon \quad (1)$$

where $y \in \mathbb{R}^N$, $X \in \mathbb{R}^{N \times P}$, $\alpha \in \mathbb{R}$, $\beta \in \mathbb{R}^P$, $\epsilon \in \mathbb{R}^N$, and $\epsilon \sim \text{Normal}(0, \sigma)$.

In this model, X is the predictor matrix, y is the outcome, α is the intercept, β is the coefficient vector associated with predictors, ϵ is the residual.

2 Probability Model

Stan allows us to use improper priors if we don't have any prior knowledge about the parameters. We can therefore start with a simple model by assuming improper priors for α , β and σ :

$$\begin{aligned} \alpha &\sim \text{Uniform}(-\infty, \infty) \\ \beta_p &\sim \text{Uniform}(-\infty, \infty), p = 1, \dots, P \\ \sigma &\sim \text{Uniform}(0, \infty) \end{aligned}$$

Putting it all together, the probability model for the multi-predictor linear regression model is:

$$\begin{aligned} y_n &\sim \text{Normal}(\mu_n, \sigma) \\ \mu_n &= \alpha + \sum_{p=1}^P X_{n,p} \beta_p \\ \alpha &\sim \text{Uniform}(-\infty, \infty) \\ \beta_p &\sim \text{Uniform}(-\infty, \infty), p = 1, \dots, P \\ \sigma &\sim \text{Uniform}(0, \infty) \end{aligned}$$