

Likelihood computation of regression for the time order determination

Based on the normality assumption, we have $\mathbf{y}_{it} \sim \mathcal{N}(\mathbf{X}_t \mathbf{b}_i, \sigma_{MRi}^2)$, \mathbf{X}_t is the row of \mathbf{X} at time t . Therefore the likelihood of the models in Equations 12 and 13 for the set of 12 i.i.d. samples \mathcal{D} is:

$$\mathcal{L}(\mathbf{b}_i, \sigma_{MRi}^2 | \mathcal{D}) = (2\pi\sigma_{MRi}^2)^{-\frac{12}{2}} \prod_{0 \leq t \leq 32} \exp\left(-\frac{(\mathbf{y}_{it} - \mathbf{X}_t \mathbf{b}_i)^2}{2\sigma_{MRi}^2}\right).$$

The corresponding log-likelihood is:

$$\log \mathcal{L}(\mathbf{b}_i, \sigma_{MRi}^2 | \mathcal{D}) = - \sum_{0 \leq t \leq 32} \left(\frac{(\mathbf{y}_{it} - \mathbf{X}_t \mathbf{b}_i)^2}{2\sigma_{MRi}^2} \right) - \frac{12}{2} \log(\sigma_{MRi}^2) - \frac{12}{2} \log(2\pi).$$

Parameters \mathbf{b}_i and σ_{MRi}^2 are estimated by ordinary least squares:

$$\begin{aligned} \hat{\mathbf{b}}_i &= (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y}_i \\ \hat{\sigma}_{MRi}^2 &= \frac{(\mathbf{y}_i - \mathbf{X} \hat{\mathbf{b}}_i)^T (\mathbf{y}_i - \mathbf{X} \hat{\mathbf{b}}_i)}{12}. \end{aligned}$$

Using the estimated parameters $\hat{\mathbf{b}}_i$ and $\hat{\sigma}_{MRi}^2$, we can formulate the log-likelihood such as:

$$\log \mathcal{L}(\mathbf{b}_i, \sigma_{MRi}^2 | \mathcal{D}) = -\frac{12}{2} \log\left(\frac{(\mathbf{y}_i - \mathbf{X} \hat{\mathbf{b}}_i)^T (\mathbf{y}_i - \mathbf{X} \hat{\mathbf{b}}_i)}{12}\right) + C,$$

where C is a constant term and is the same for every gene.