

How lmer Works

Joint dist:

$$p(y, u; \theta, \beta, \sigma^2) \propto (\sigma^2)^{-(N+Q)/2} \exp \left\{ -\frac{1}{2\sigma^2} \left\| \begin{bmatrix} y \\ 0 \end{bmatrix} - \begin{bmatrix} Z\Lambda(\theta) & X \\ I & 0 \end{bmatrix} \begin{bmatrix} u \\ \beta \end{bmatrix} \right\|^2 \right\}$$

1. Calculate joint mode of u, β : $\tilde{u}, \tilde{\beta}$
 - 1.1 Compute decomposition of augmented design, L, R_{ZX}, R_X
 - 1.2 $(A^\top A)^{-1} A^\top y'$
2. Integrate u - marginal likelihood
3. Profile β
4. Profile σ^2
5. Loop over deviance

Marginal likelihood

$$p(y; \theta, \beta, \sigma^2) \propto (\sigma^2)^{-N/2} |L(\theta)|^{-1} \exp \left\{ -\frac{1}{2\sigma^2} \left\| R_X(\beta - \tilde{\beta}) \right\|^2 \right\} \times \\ \exp \left\{ -\frac{1}{2\sigma^2} \left\| \begin{bmatrix} y \\ 0 \end{bmatrix} - \begin{bmatrix} Z\Lambda(\theta) & X \\ I & 0 \end{bmatrix} \begin{bmatrix} \tilde{u} \\ \tilde{\beta} \end{bmatrix} \right\|^2 \right\}$$

- ▶ Joint mode in β maximizes likelihood
- ▶ Joint mode gives PWRSS, maximizes σ^2

Priors “For Free”

Add $\beta \sim N(0, \sigma^2 \Sigma_\beta)$, joint dist:

$$p(y, u; \theta, \beta, \sigma^2) \propto (\sigma^2)^{-(N+Q+P)/2} |\Sigma_\beta|^{-1/2}$$

$$\exp \left\{ -\frac{1}{2\sigma^2} \left\| \begin{bmatrix} y \\ 0 \\ 0 \end{bmatrix} - \begin{bmatrix} Z\Lambda(\theta) & X \\ 0 & \Sigma_\beta^{-1/2} \\ I & 0 \end{bmatrix} \begin{bmatrix} u \\ \beta \end{bmatrix} \right\|^2 \right\}$$

Changes:

1. PWRSS includes $\tilde{\beta}^\top \Sigma_\beta^{-1} \tilde{\beta}$ (falls out of joint mode)
2. $R_X^\top R_X = X^\top X - R_{ZX}^\top R_{ZX} + \Sigma_\beta^{-1}$

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$\beta \sim N(0, \Sigma_\beta)$ much, much harder

More “For Free”

- ▶ $\sigma^2 \sim \Gamma^{-1}$ - adjust degrees of freedom, add constant to PWRSS
- ▶ $\sigma^2 \sim \Gamma$, $\sigma \sim \Gamma^{-1}$ yield quadratic maximizer for σ^2
- ▶ $\sigma^2 = \sigma_0^2$, skip profiling step and plug in

PWRSS less useful/concrete

θ Priors

Any prior on $\Sigma(\theta)$ can be tacked on as “penalty”

What of $\tilde{\Sigma} = \sigma^2 \Sigma(\theta)$?

- ▶ Inv-Wishart yields functional form equivalent to Inv-Gamma on σ^2
- ▶ Wishart and $\tilde{\Sigma}^{1/2} \sim \text{Inv-Wishart}$ yield quadratic in σ^2 as before

GLMMs

- ▶ σ^2 not (yet) relevant
- ▶ β not profiled, numerically optimized

All priors easy

Wishlist

Hooks into/pause optimization at:

- ▶ For given θ , just do decomp & let me tweak R_X
- ▶ For given θ , decomp, just do joint mode
- ▶ Less ambitiously, way to penalize deviance
- ▶ Something like an interface

For last point, can already wrap deviance

Or, just throw everything into numeric optimizer