

# THINKING ABOUT MIXED-EFFECTS MODELS

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# OUTLINE

- 1 THE DEEP END
  - Model Statement
  - MLE
  - ReML

# BASIC MODEL STATEMENT

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\mathbf{b} + \boldsymbol{\epsilon}$$

$$\mathbf{b} \sim \mathcal{N}(\mathbf{0}, \mathbf{D})$$

$$\boldsymbol{\epsilon} \sim \mathcal{N}(\mathbf{0}, \mathbf{R})$$

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## COVARIANCE MATRICES

- $\mathbf{D}$  describes the random effects covariance.
- $\mathbf{R}$  allocates the residuals covariance.

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$$\text{Var}(\mathbf{Y} \mid \mathbf{X}, \boldsymbol{\beta}) = \mathbf{Z}\mathbf{D}\mathbf{Z}' + \mathbf{R} = \mathbf{V}$$

## LOG LIKELIHOOD

$$\mathcal{L}(\boldsymbol{\beta}, \mathbf{V} \mid \mathbf{Y}, \mathbf{X}) = -\frac{1}{2} \ln(|\mathbf{V}|) - \frac{n}{2} \ln(2\pi) - \frac{1}{2} (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{V}^{-1} (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})$$



PROFILE  $\beta$  OUT

$$\hat{\beta} = (\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{Y}$$

$\beta$  IS GONE!

$$\mathcal{L}(\beta, \mathbf{V} \mid \mathbf{Y}, \mathbf{X}) = f(\mathbf{V}, \mathbf{Y}, \mathbf{X}, \mathbf{Z}, \mathbf{D}, \mathbf{R})$$

# $\beta$ IS GONE!

Estimate  $\hat{V}$  by maximization and then  $\hat{\beta}$  by substitution.

# REML

Maximum likelihood estimators of covariance parameters are usually negatively biased.

# REML

Briefly, ReML involves applying ML, but replacing

- $\mathbf{Y}$  with  $\mathbf{KY}$ ;
- $\mathbf{X}$  with  $\mathbf{0}$ ;
- $\mathbf{Z}$  with  $\mathbf{K}'\mathbf{Z}$ ; and
- $\mathbf{V}$  with  $\mathbf{K}'\mathbf{VK}$

where  $\mathbf{K}$  is such that  $\mathbf{K}'\mathbf{X} = \mathbf{0}$ .