MA50260 Statistical Modelling

Lecture 16: Mixed Effects Models - Nested and Crossed Designs

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Mixed Effect Models

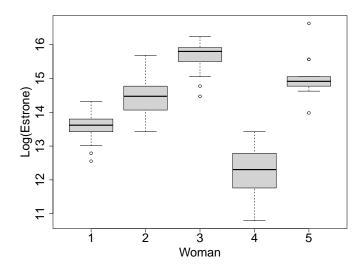
A mixed effects (normal) linear model is written as

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

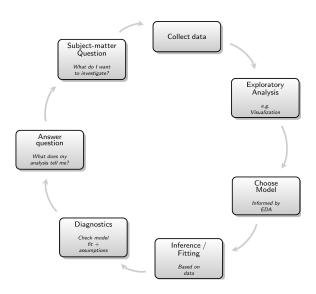
where

- **Y** is the vector of responses
- **X** is the design matrix
- ightharpoonup are the regression coefficients
- **Z** is the matrix of covariates associated to the *q* (unknown) random effects **b**.
- ightharpoonup is the vector of residuals

Example - Analysis of Estrone Levels



Summary of Last Lecture



Nested and Crossed Effects

In several applications, we have multiple random effects.

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Nested and Crossed Effects

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How can we extend the introduced linear mixed model?

We generally distinguish between two types of random effects:

- Nested effects: levels of one factor are completely contained within the levels of another factor
- Crossed effects: levels of a factor vary across the levels of another

Example - Setup (I)

Consider a study with

- N machines
- ▶ P types of moulds that can be used on any machine
- n components being produced for each mould and each machine



Example - Setup (II)

Let $Y_{i,j,s}$ be the measurement for the *s*-th component from the *j*-th mould for the *i*-th machine.

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We are interested in the influence of machine and mould on the measurements:

- treat machine effect as random effect
- treat mould effect as random effect

Example - Nested Random Effects

Suppose that a different set of moulds is used on each machine.

We then model

$$Y_{i,j,s} = \mu + b_{1,i} + b_{12,ij} + \epsilon_{i,j,s},$$

where

- ▶ $b_{1,i} \sim \text{Normal}(0, \sigma_M^2)$
- ▶ $b_{12,ij} \sim \text{Normal}(0, \sigma_P^2)$
- $ightharpoonup \epsilon_{i,j,s} \sim \operatorname{Normal}\left(0,\sigma_{\epsilon}^{2}\right)$

Note, the mould factor appears **only** within a particular level of the machine factor.

Example - Crossed Random Effects

Assume that the same moulds are used on each machine.

This is modelled as

$$Y_{i,j,s} = \mu + b_{1,i} + b_{2,j} + \epsilon_{i,j,s},$$

where

- ▶ $b_{1,i} \sim \text{Normal}(0, \sigma_M^2)$
- ▶ $b_{2,j} \sim \text{Normal}(0, \sigma_P^2)$
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Note, mould effects are no longer specific to machines, but each mould effect b_j remains the same for all machine effects b_i .

Measurements of NO_X across Bath at 4:00, 10:00, 16:00 and 22:00.

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- Location
- ► Time of day

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General Notation

For nested models, we can write

$$\mathbf{Y} = \mathbf{X}\underline{\beta} + \mathbf{Z}_1\mathbf{b}_1 + \mathbf{Z}_{12}\mathbf{b}_{12} + \underline{\epsilon},$$

where we use the double subscript to indicate that the random effects are nested.

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On the other hand, for crossed random effects, we can write

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We can also write these models in the general form

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

The Story so far (I)

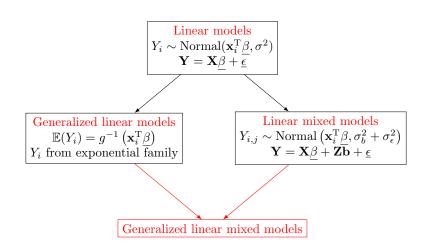
In the past weeks, we considered a range of models, including:

- Normal linear regression
- Logistic regression (Binomial GLM)
- Poisson regression
- Gamma/Exponential regression
- Ordinal regression
- Linear mixed models

The Story so far (II)

Generalized linear models $\mathbb{E}(Y_i) = g^{-1} \left(\mathbf{x}_i^{\mathrm{T}} \underline{\beta} \right)$ Y_i from exponential family

The Story so far (III)



Decide which model to use to analyse the following responses:

1. Size of a car insurance claim

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- 5. Wingspan of an albatross

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- 4. Student performance across units
 - → Linear mixed model
- 5. Wingspan of an albatross
 - → Normal linear regression