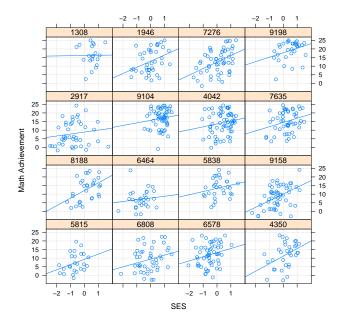
Multilevel Regression

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Sept 14, 2016



Consider we have reaction time data from J subjects,

$$\{x_{j1}, x_{j2}, x_{j3} \dots x_{jn_j}\}_{j=1}^{J}$$
.

► A simple multilevel model for this data might be:

$$\begin{split} &x_{j\,i} \sim N(\mu_j, \sigma^2), \quad \text{for } i \in \{1\dots n_j\}, \\ &\mu_j \sim N(\theta, \tau^2), \quad \text{for } j \in \{1\dots J\}. \end{split}$$

▶ In words, each x_{ji} is drawn from a Gaussian with mean μ_j and variance σ^2 , and each μ_j is drawn from a Gaussian with mean θ and variance τ^2 .

• We can re-write $x_{ji} \sim N(\mu_j, \sigma^2)$ as

$$x_{ji} = \mu_j + \varepsilon_{ji}, \quad \varepsilon_{ji} \sim N(0, \sigma^2).$$

• We can re-write $\mu_j \sim N(\theta, \tau^2)$ as

$$\mu_j = \theta + \eta_j, \quad \eta_j \sim N(0, \tau^2).$$

► The multilevel model can be re-written

$$x_{\text{ji}} = \theta + \eta_{\text{j}} + \varepsilon_{\text{ji}} \quad \varepsilon_{\text{ji}} \sim N(0, \sigma^2), \\ \eta_{\text{j}} \sim N(0, \tau^2).$$

▶ This is often termed a *random-effects* model.

- ▶ In the model just described, there are three unknowns: θ , σ^2 and τ^2 .
- Model estimation (fitting) estimates values for these variables.
- The variable θ denotes the global average reaction time.
- ► The variable σ^2 denotes the variance within any given subject.
- ► The variable τ^2 denotes the variance across subjects.

- ▶ In the model just described, θ tells us the global average.
- The variance τ^2 tells us how much any given subject's average varies about θ.
- ► For example, 95% and 99% of the averages for individual subjects, will be in the ranges

$$\theta \pm 1.96 \times \tau$$
, $\theta \pm 2.56 \times \tau$,

respectively.

Likewise, 95% and 99% of any given subject's reaction times, i.e. x_{ji} , will be in the ranges

$$\theta + \eta_i \pm 1.96 \times \sigma$$
, $\theta + \eta_i \pm 2.56 \times \sigma$.



► A model, such as the previous one, would be specified as follows in the lmer program in R:

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lmer(latency ~ 1 + (1|subject))
where latency is reaction time, and subject is a categorical
variable that indicates the identity of the subject.
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▶ In other words, here we are saying that reaction time is modelled as a variation around a global average plus an individual average.

- ► Let's say we want to measure the mpg of a given model of car (e.g. a Porsche 911).
- ▶ Because any one car could vary from others of the same model, we have K different examples of this model of car.
- ▶ Likewise, because any one driver could affect the recorded mpg of the car he drives, we have J different drivers.
- We get each of the J drivers to drive each of the K cars, and record the mpg as

 $y_{jk} = mpg$ for driver j, car k.

A multilevel model for this mpg experiment could be

$$\begin{split} y_{jk} &\sim N(\mu_j + \nu_k, \sigma^2), \\ \mu_j &\sim N(\varphi, \tau^2) \\ \nu_k &\sim N(\psi, \upsilon^2) \end{split}$$

which would work out as

$$y_{jk} = \underbrace{\theta}_{\Phi + \Psi} + \eta_j + \zeta_k + \varepsilon_{jk},$$

with

$$\eta_{\text{j}} \sim N(0,\tau^2), \ \zeta_k \sim N(0,\upsilon^2), \ \varepsilon_{\text{j}k} \sim N(0,\sigma^2).$$

▶ In this example, we have three sources of variation

$$y_{jk} = \theta + \underbrace{\eta_j}_{\text{within driver}} + \underbrace{\zeta_k}_{\text{within car}} + \underbrace{\varepsilon_{jk}}_{\text{within trial}},$$

where τ^2 gives the within driver variance, υ^2 gives the within car variation, and σ^2 gives within trial variation.

- The variable θ provides the average mpg for the car model (i.e. the Porsche 911)
- ▶ Your mileage may vary: The variables τ^2 , υ^2 and σ^2 provide measures of the relative variation across in mpg drivers, cars and trials, respectively.

► The mpg model would be specified as follows in the lmer: lmer(mpg ~ 1 + (1|driver) + (1|car))

where mpg is continuous variable, and driver and car are categorical variables that indicate the identity of the driver and car, respectively.

- ▶ In this problem, we have J schools. Within each school, we have n_i students.
- ▶ For student i in school j, their ses score is x_{ji} and their mathematical achievemnt score is y_{ji} .
- A multilevel model for this data is

$$\begin{split} y_{ji} &\sim N(\alpha_j + \beta_j x_{ji}, \sigma^2), \\ \alpha_j &\sim N(\alpha, \tau_\alpha^2), \\ \beta_j &\sim N(b, \tau_b^2). \end{split}$$

► The model

$$y_{ji} \sim N(\alpha_j + \beta_j x_{ji}, \sigma^2),$$

$$\alpha_j \sim N(\alpha, \tau_\alpha^2),$$

$$\beta_j \sim N(b, \tau_b^2),$$

can be re-written

$$y_{ji} = a + bx_{ji} + \eta_j + \zeta_j x_{ji} + \epsilon_{ji},$$

where

$$\eta_j \sim N(0,\tau_\alpha^2), \; \zeta_j \sim N(0,\tau_b^2), \; \varepsilon_j \sim N(0,\sigma^2).$$

We can look at this model as

$$y_{\mathtt{j}\mathtt{i}} = \underbrace{\alpha + b x_{\mathtt{j}\mathtt{i}}}_{\mathtt{general \, model}} + \underbrace{\eta_{\mathtt{j}} + \zeta_{\mathtt{j}} x_{\mathtt{j}\mathtt{i}}}_{\mathtt{school-level \, model}} + \varepsilon_{\mathtt{j}\mathtt{i}}.$$

▶ For any given school, the model can be viewed as

$$y_{ji} = \underbrace{(a + \eta_j)}_{\alpha_j} + \underbrace{(b + \zeta_j)}_{\beta_j} x_{ji} + \epsilon_{ji}.$$

► In other words, any given school's regression model is a variation on a general regression model.

- In the model just described, a and b are the general regression coefficients.
- ► The variance τ_a^2 tells us how much variation in the intercept term there is across schools. The variance τ_b^2 tells us how much variation in the slope term there is across schools.
- ► For example, 95% and 99% of the intercepts for individual schools will be in the ranges

$$a \pm 1.96 \times \tau_a$$
, $a \pm 2.56 \times \tau_a$,

respectively. Likewise, 95% and 99% of the slope terms for schools will be in the ranges

$$b \pm 1.96 \times \tau_h$$
, $b \pm 2.56 \times \tau_h$.



► The ses regression model would be specified as follows in the lmer:

lmer(math \sim 1 + ses + (1|school) + (0 + ses|school)) where math is continuous variable, and school is a categorical variable that indicates the identity of the school¹

¹Writing (1 + ses|school) would imply correlated slopes and intercepts. ■ ▶ ■ ✓ ९ €

Why multilevel models?

- When data occurs in groups, multilevel models should always be considered.
- Multilevel models provide a macro/micro perspective on the data: The show the general pattern across all groups, and show how each individual group varies about this general pattern.
- ▶ By appropriately indentifying different sources of variation, the general patterns in the data are more accurately inferred.
- ▶ Even estimates for an individual group are improved by a process of *strength-sharing*. In other words, knowing the general pattern in the data facilitates the inference of the patterns for individual