Multilevel Model

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Multilevel Model

Multilevel model is all about shrinkage.

Suppose we have a hierarchical structure:

$$y_i \sim N(\alpha_{j[i]} + \beta x_i, \sigma^2), \text{ for } i = 1, \dots, n,$$

 $\alpha_j \sim N(\mu_\alpha, \sigma^2_\alpha) \text{ for } j = 1, \dots, J$

- Exactly the same as what we have done in the section of random effects model,
 - For any unit *i*, $var(\epsilon_i^{all}) = \sigma^2 + \sigma_\alpha^2$
 - ▶ For any units i and k in the same group j, $cov(\epsilon_i^{\mathit{all}}, \epsilon_k^{\mathit{all}}) = \sigma_\alpha^2$
 - For any units *i* and *k* in different groups, their covariance is 0
- ▶ The shrinkage:

estimate of
$$\alpha_j = \frac{\frac{n_j}{\sigma^2}}{\frac{n_j}{\sigma^2} + \frac{1}{\sigma_\alpha^2}}\underbrace{\frac{(\bar{y}_j - \beta \bar{x}_j)}{\sigma^2 + \beta \bar{x}_j}}_{\text{no-pooling estimate for group } j} + \frac{\frac{1}{\sigma_\alpha^2}}{\frac{n_j}{\sigma^2} + \frac{1}{\sigma_\alpha^2}}_{\text{regression prediction } \alpha_j}\underbrace{\mu_\alpha}_{\text{regression prediction } \alpha_j}$$

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Several implications we can draw from the estimate of group coefficients α_j :

- $\sigma_{\alpha}^2 \downarrow$ causes more pooling, $\sigma_{\alpha}^2 = 0$ means complete pooling. As the distribution of group-level parameter is more accurate, each group can borrow more information from groups that are closer than when σ_{α}^2 is large.
- ▶ When $\sigma^2 \uparrow$ and $\sigma^2_{\alpha} \downarrow$, effective sample size is n, as the influence of intra-group correlation goes down
- ▶ When $\sigma^2 \downarrow$ and $\sigma^2_\alpha \uparrow$, effective sample size is j
- ▶ When $n_j \uparrow$, the estimate is pooled closer to the non-pooling estimate for its own group.



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