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Mixed-effects model.

$$y_{ij} = \alpha + \beta x_{ij} + \delta_i + \varepsilon_{ij}$$

$$\delta_i \sim N(0, s^2)$$

$$\varepsilon_{ij} \sim N(0, \sigma^2)$$

We estimate $\alpha, \beta, s^2, \sigma^2$ by MLE.

$$p(y_{ij} | \alpha, \beta, \delta_i, \sigma^2) = N(\alpha + \beta x_{ij} + \delta_i, \sigma^2)$$

$$p(\delta_i | s^2) = N(0, s^2)$$

$$\begin{aligned} p(y_{ij} | \alpha, \beta, s^2, \sigma^2) &= \int_{-\infty}^{+\infty} p(y_{ij} | \alpha, \beta, \delta_i, \sigma^2) p(\delta_i | s^2) d\delta_i \\ &= \int_{-\infty}^{+\infty} p(y_{ij}, \delta_i | \alpha, \beta, \sigma^2, s^2) d\delta_i \end{aligned}$$

$$\rightarrow p(y_{ij} | \alpha, \beta, \delta_i, \sigma^2)$$

$$= E_{\delta_i} p(y_{ij} | \alpha, \beta, \delta_i, \sigma^2)$$

$$= \frac{1}{N} \sum_{i=1}^N p(y_{ij} | \alpha, \beta, \delta_i^{(n)}, \sigma^2)$$

A trick to make MLE valid:

when sampling ~~$\delta_i^{(n)}$~~ $\delta_i^{(n)} \sim N(0, s^2)$,

~~or~~ we let $\delta_i^{(n)} = s \cdot z^{(n)}$, $z^{(n)} \sim N(0, 1)$.

$$p(y_{ij} | \alpha, \beta, \delta_i, \sigma^2)$$

$$= \frac{1}{N} \sum_{n=1}^N p(y_{ij} | \alpha, \beta, \delta_i, s \cdot z^{(n)}, \sigma^2) \quad (*)$$

we optimize $(*)$ and find $\hat{\alpha}, \hat{\beta}, \hat{s}^2, \hat{\sigma}^2$.

where $p(y_{ij} | \alpha, \beta, s \cdot z^{(n)}, \sigma^2)$

$$= \phi(y_{ij} | \alpha + \beta x_{ij} + s \cdot z^{(n)}, \sigma^2).$$