

Whiteboard Lecture

Longitudinal Data Analysis:

Whiteboard Interlude II

We now describe the essential properties of the *random intercept model*

$$y_{ij} = U_i + \beta_0 + \beta_1 x_{ij} + \varepsilon_{ij}$$

$$U_i \stackrel{\text{ind.}}{\sim} N(0, \sigma_U^2), \quad \varepsilon_{ij} \stackrel{\text{ind.}}{\sim} N(0, \sigma_\varepsilon^2)$$

and the U_i and ε_{ij} are independent of each other.

For $j \neq j'$, $\text{Cov}(y_{ij}, y_{ij'})$ is the covariance between different measurements on the same subject:

$$\begin{aligned} \text{Cov}(y_{ij}, y_{ij'}) &= \text{Cov}(U_i + \beta_0 + \beta_1 x_{ij} + \varepsilon_{ij}, \\ &\quad U_i + \beta_0 + \beta_1 x_{ij'} + \varepsilon_{ij'}) \\ &= \text{Cov}(U_i + \varepsilon_{ij}, U_i + \varepsilon_{ij'}) \\ &= \text{Cov}(U_i, U_i) + \text{Cov}(U_i, \varepsilon_{ij'}) + \text{Cov}(U_i, \varepsilon_{ij}) \\ &\quad + \text{Cov}(\varepsilon_{ij}, \varepsilon_{ij'}) \\ &= \sigma_U^2 + 0 + 0 + 0 \\ &= \sigma_U^2 \end{aligned}$$

For $j = j'$, $\text{Cov}(y_{ij}, y_{ij'}) = \text{Var}(y_{ij})$ is the variance of the j th measurement on subject i , and has expression:

$$\begin{aligned}\text{Cov}(y_{ij}, y_{ij}) &= \text{Var}(U_i + \beta_0 + \beta_1 x_{ij} + \varepsilon_{ij}) \\ &= \text{Var}(U_i + \varepsilon_{ij}) \\ &= \text{Var}(U_i) + \text{Var}(\varepsilon_{ij}) \\ &= \sigma_U^2 + \sigma_\varepsilon^2\end{aligned}$$

For $i \neq i'$ we get

$$\text{Cov}(y_{ij}, y_{i'j'}) = 0.$$

This says that observations of different individuals are uncorrelated (e.g. Anne's blood pressure is not correlated with Bill's blood pressure).

Consider the sample sizes corresponding to the example:

$$m = 3, \quad n_1 = 2, \quad n_2 = 3, \quad n_3 = 2.$$

The covariance matrix of

$$\begin{bmatrix} y_{11} \\ y_{12} \\ y_{21} \\ y_{22} \\ y_{23} \\ y_{31} \\ y_{32} \end{bmatrix}$$

is

$$\left[\begin{array}{cc|ccc|cc} \sigma_U^2 + \sigma_\varepsilon^2 & \sigma_U^2 & 0 & 0 & 0 & 0 & 0 \\ \sigma_U^2 & \sigma_U^2 + \sigma_\varepsilon^2 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & \sigma_U^2 + \sigma_\varepsilon^2 & \sigma_U^2 & \sigma_U^2 & 0 & 0 \\ 0 & 0 & \sigma_U^2 & \sigma_U^2 + \sigma_\varepsilon^2 & \sigma_U^2 & 0 & 0 \\ 0 & 0 & \sigma_U^2 & \sigma_U^2 & \sigma_U^2 + \sigma_\varepsilon^2 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & \sigma_U^2 + \sigma_\varepsilon^2 & \sigma_U^2 \\ 0 & 0 & 0 & 0 & 0 & \sigma_U^2 & \sigma_U^2 + \sigma_\varepsilon^2 \end{array} \right]$$

The correlation matrix is then

$$\left[\begin{array}{cc|ccc|cc} 1 & \rho & 0 & 0 & 0 & 0 & 0 \\ \rho & 1 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 1 & \rho & \rho & 0 & 0 \\ 0 & 0 & \rho & 1 & \rho & 0 & 0 \\ 0 & 0 & \rho & \rho & 1 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 1 & \rho \\ 0 & 0 & 0 & 0 & 0 & \rho & 1 \end{array} \right] \quad \text{where} \quad \rho = \frac{\sigma_U^2}{\sigma_U^2 + \sigma_\varepsilon^2}.$$

Note the blocking structure corresponding to within-subject correlation. The first block is for subject 1 (Anne), the second is for subject 2 (Bill) and the third is for subject 3 (Clare).

Remarks

1. The random intercept U_i invokes correlation between measurements on same subject.
2. A shortcoming of the random intercept model is that the correlation is the same for each subject; e.g. Anne's ρ is the same as Bill's ρ .
3. Another shortcoming is that the within-subject correlation is constant over time; e.g. the correlation between Anne's blood pressure measurements 2 days apart is the same as those taken 10 days apart.

