

(Bailey) Sec. 4.4 models for block designs

(Q: how to model  $Z_w$ : the effect of plot structure)

- Fixed effect model

$$Z_w = \underbrace{\zeta_{B(w)}}_{\substack{\text{fixed parameters} \\ (\text{c.f. treatment effect})}} + \underbrace{\epsilon_w}_{\substack{\text{random variable} \\ E(\epsilon_w) = 0 \\ \text{cov}(\epsilon_w, \epsilon_{w'}) = 0}}$$

Q: when to use fixed effect?

- blocks = population
- conditional on conclusion the blocks
- block difference not change, say over time.

Then,  $E(Z_w) = \zeta_{B(w)}$

$$\text{cov}(Z_w, Z_{w'}) = \begin{cases} 0 & \text{if } w \neq w' \\ \sigma^2 & \text{if } w = w' \end{cases}$$

$$\begin{aligned} Y &= \zeta_{T(w)} + Z_w \\ \Rightarrow E(Y) &= \zeta + \zeta \\ \text{cov}(Y) &= \sigma^2 I \end{aligned}$$

3-12

- random effect model

$$\begin{aligned} Z_w &= \underbrace{\zeta_{B(w)}}_{\substack{\text{random} \\ \text{variable}}} + \underbrace{\epsilon_w}_{\substack{\text{random} \\ \text{variable}}} \\ &= \zeta_0 + (\zeta_{B(w)} - \zeta_0) + \epsilon_w \\ &- \zeta_0, \zeta_{B(w)} - \zeta_0, \epsilon_w: \text{uncorrelated,} \\ &\text{mean zero, variance } \sigma_0^2, \sigma_B^2, \sigma_e^2, \text{ respectively.} \end{aligned}$$

Q: when to use random effect?

- blocks << population
- conclusion would like to be applied on the population
- block difference change, say over time --

Q: do we need to restrict sum of  $\zeta_{B(w)} - \zeta_0 = 0$ ?

(if  $\zeta_0$ : average of  $\zeta_{B(w)}$ , yes; if  $\zeta_0$ : average of whole population, no)

$$E(Z_w) = 0$$

$$\alpha = \beta$$

$$\text{cov}(Z_\alpha, Z_\beta) = \sigma_0^2 + \sigma_B^2 + \sigma_e^2$$

$$\alpha \neq \beta$$

$$B(\alpha) = B(\beta)$$

$$B(\alpha) \neq B(\beta)$$

$$\begin{aligned} &= \sigma_0^2 + \sigma_B^2 \\ &= \sigma_0^2 = \rho_2 \sigma^2 \end{aligned}$$

$$\begin{bmatrix} \sigma_0^2 & \rho_1 \sigma^2 & \rho_2 \sigma^2 \\ \rho_1 \sigma^2 & \sigma_0^2 & \rho_2 \sigma^2 \\ \rho_2 \sigma^2 & \rho_2 \sigma^2 & \sigma_0^2 \end{bmatrix}$$

3-13

$$\Rightarrow E(\mathbb{Y}) = \mathbb{C}$$

$\text{cov}(\mathbb{Y})$  = the matrix in previous slide

- Let  $J_B$  be the  $N \times N$  matrix whose  $(\alpha, \beta)$  entry is
$$\begin{cases} 1, & \text{if } B(\alpha) = B(\beta) \\ 0, & \text{o.w.} \end{cases}$$

$$J_B = \begin{bmatrix} \begin{smallmatrix} 1 & 1 \\ 1 & 1 \end{smallmatrix} & \begin{smallmatrix} 0 \\ \vdots \\ 0 \end{smallmatrix} \\ \begin{smallmatrix} 0 \\ \vdots \\ 0 \end{smallmatrix} & \begin{smallmatrix} \mathbb{I} \\ \vdots \\ \mathbb{I} \end{smallmatrix} \end{bmatrix}$$

$$\begin{bmatrix} 1 & \cdots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \cdots & 1 \end{bmatrix}_{N \times N}$$

then  $\text{cov}(\mathbb{Y}) = \sigma^2 \mathbb{I} + \rho_1 \sigma^2 (J_B - \mathbb{I}) + \rho_2 \sigma^2 (J - J_B)$

$$= \sigma^2 \left[ (1 - \rho_1) \mathbb{I} + (\rho_1 - \rho_2) J_B + \rho_2 J \right]$$

- An alternative approach:  $\zeta_0$  regarded as fixed effect.

$$\Rightarrow \rho_2 = 0 \text{ & } \zeta_0 \text{ goes into } E(\mathbb{Y})$$

3-14

### (Reading material 02, Sec 1.3) Randomization model

- randomness comes from the restricted randomization procedure in plan.
- $b$  blocks, labeled by  $\xi_p = 1, 2, \dots, b$
- $k$  plots in block  $\xi_p$ , labeled by  $\pi(\xi_p) = 1, 2, \dots, k$
- randomization of blocks:

choose a permutation of  $\{1, \dots, b\}$

after the permutation, relabel blocks by  $j = 1, 2, \dots, b$

- randomization of plots in block  $\xi_p$

(foreach block) choose a permutation of  $\{1, 2, \dots, k\}$

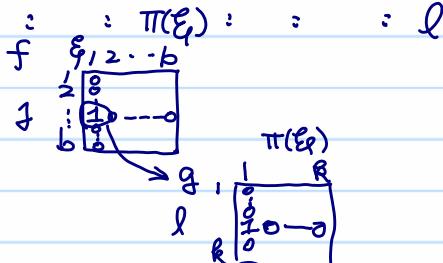
after permutation, relabel the plots by  $l = 1, 2, \dots, k$

- let  $f_{j\xi_p} = \begin{cases} 1, & \text{if block originally labeled by } \xi_p \text{ is relabel by } j \\ 0, & \text{o.w.} \end{cases}$

random variable  $\mathcal{G}_{j\xi_p} = \begin{cases} 1, & \text{if plot } \pi(\xi_p) = l \text{ is relabel by } j \\ 0, & \text{o.w.} \end{cases}$

$$(\star 1) \quad \sum_{\xi_p=1}^b f_{j\xi_p} = 1, \quad \sum_{\xi_p=1}^b \mathcal{G}_{j\xi_p} = 1$$

$$\sum_{j=1}^b f_{j\xi_p} = 1, \quad \sum_{j=1}^b \mathcal{G}_{j\xi_p} = 1$$



3-15

$$(*2) \quad E(f_{j|\epsilon}) = \frac{1}{b} \quad E(f_{j|\epsilon} f_{j'|\epsilon'}) = \begin{cases} \frac{1}{b} & , \begin{array}{l} j=j', \epsilon=\epsilon' \\ j=j', \epsilon \neq \epsilon' \\ j \neq j', \epsilon=\epsilon' \\ j \neq j', \epsilon \neq \epsilon' \end{array} \\ 0 & \end{cases}$$

$$E(g_{l|\pi(\epsilon)}) = \frac{1}{R} \quad E(g_{l|\pi(\epsilon)} g_{l'|\pi(\epsilon')}) = \begin{cases} \frac{1}{R} & , \begin{array}{l} l=l', \epsilon=\epsilon' \\ l=l', \epsilon \neq \epsilon' \\ l \neq l', \epsilon=\epsilon' \\ l \neq l', \epsilon \neq \epsilon' \end{array} \\ 0 & \end{cases}$$

because of the independence of randomization in different blocks

$$\therefore E(g_{l|\pi(\epsilon)} g_{l'|\pi(\epsilon')}) = E(g_{l|\pi(\epsilon)}) E(g_{l'|\pi(\epsilon')}) = \frac{1}{R^2}, \epsilon \neq \epsilon'$$

(\*3)  $\because$  randomization of blocks is independent of the plot randomization within block

$$E(f_{j|\epsilon} g_{l|\pi(\epsilon')}) = E(f_{j|\epsilon}) E(g_{l|\pi(\epsilon')}) = \frac{1}{b} \cdot \frac{1}{R}$$

for  $\epsilon=\epsilon'$  or  $\epsilon \neq \epsilon'$

- Let the response contributed by the plot originally labeled by  $\pi(\epsilon)$  be denoted by  $M_{\pi(\epsilon)}$ .

Let the block of the plot receives label  $j$  after block randomization and the plot receives label  $l$  after plot randomization within block.

3-16

$$m_{l(j)} = \sum_{\epsilon=1}^b \sum_{\pi(\epsilon)=l}^R f_{j|\epsilon} g_{l|\pi(\epsilon)} \cdot M_{\pi(\epsilon)}$$

- Introduce the identity:

$$M_{\pi(\epsilon)} = M_{\cdot(\cdot)} + [M_{\cdot(\epsilon)} - M_{\cdot(\cdot)}] + [M_{\pi(\epsilon)} - M_{\cdot(\epsilon)}]$$

where  $M_{\cdot(\epsilon)} = \frac{1}{R} \sum_{\pi(\epsilon)=1}^R M_{\pi(\epsilon)}$   $\leftarrow$  average of plot contribution within block  $\epsilon$

$M_{\cdot(\cdot)} = \frac{1}{b} \sum_{\epsilon=1}^b M_{\cdot(\epsilon)}$   $\leftarrow$  average of all plots contribution ( $bR$  plots)

- Then,

$$m_{l(j)} = \underbrace{\sum_{\epsilon=1}^b \sum_{\pi(\epsilon)=l}^R f_{j|\epsilon} g_{l|\pi(\epsilon)} M_{\cdot(\cdot)}}_{M_{\cdot(\cdot)} \equiv \mu} + \underbrace{\sum_{\epsilon=1}^b \sum_{\pi(\epsilon)=l}^R f_{j|\epsilon} g_{l|\pi(\epsilon)} (M_{\cdot(\epsilon)} - M_{\cdot(\cdot)})}_{= \sum_{\epsilon=1}^b f_{j|\epsilon} (M_{\cdot(\epsilon)} - M_{\cdot(\cdot)}) \equiv \beta_j} \\ + \underbrace{\sum_{\epsilon=1}^b \sum_{\pi(\epsilon)=l}^R f_{j|\epsilon} g_{l|\pi(\epsilon)} (M_{\pi(\epsilon)} - M_{\cdot(\epsilon)})}_{\equiv \gamma_{l(j)}}$$

$$\Rightarrow m_{l(j)} = \mu + \beta_j + \gamma_{l(j)}$$

where  $\mu$  is a constant parameter and

3-17

$\beta_j$ : random variable  $\Rightarrow$  block random effect  
 ↳ because of the randomization of blocks

Note that  $\sum_{j=1}^b \beta_j = 0$  ( $\Rightarrow$  sum of random effects = 0)

$\gamma_{l(j)}$ : unit (plot) error: random variable

because of the block randomization & plot randomization

- It is easy to get (from (1), (2), (3))

$$(1) E(\beta_j) = 0 \quad \& \quad E(\gamma_{l(j)}) = 0$$

$$(2) \text{cov}(\beta_j, \beta_{j'}) = \left\{ \begin{array}{l} b \sum_{\varepsilon=1}^b (\mu_{\cdot(\varepsilon)} - \mu_{\cdot(\varepsilon)})^2, \quad j=j' \\ -\frac{1}{b(b-1)} \sum_{\varepsilon=1}^b (\quad)^2, \quad j \neq j' \end{array} \right. , \quad \text{c.f. random effect model} \quad \& \quad \text{the restriction} \quad \sum \beta_j = 0$$

$$(3) \text{cov}(\gamma_{l(j)}, \gamma_{l(j')}) = \left\{ \begin{array}{l} \frac{1}{bK} \sum_{\varepsilon=1}^b \sum_{l=1}^K (\mu_{\pi(\varepsilon)} - \mu_{\cdot(\varepsilon)})^2, \quad j=j', \quad l=l' \\ 0, \quad j=j', \quad l \neq l' \\ -\frac{1}{K(b(K-1))} \sum_{\varepsilon=1}^b \sum_{l=1}^K (\quad)^2, \quad j=j', \quad l \neq l' \end{array} \right. , \quad \text{c.f. random effect model} \quad \& \quad \text{the restriction} \quad \sum \beta_j = 0$$

$$(4) \text{cov}(\beta_j, \gamma_{l(j')}) = 0$$

3-18

- From (1), (2), (3), (4), we can get

$$E(m_{l(j)}) = \mu$$

c.f. the matrix in LNp.3-13

$$\text{cov}(m_{l(j)}, m_{l'(j')}) = \left\{ \begin{array}{l} = (\delta_{jj'} - \frac{1}{b}) \sigma_B^2 + (\delta_{jj'} (\delta_{ll'} - \frac{1}{K})) \cdot \sigma_{\varepsilon}^2 \\ \sigma^2, \quad \text{if } j=j', \quad l=l' \\ \rho_1 \sigma^2, \quad j=j', \quad l \neq l' \\ \rho_2 \sigma^2, \quad j \neq j' \end{array} \right.$$

$$\sigma^2 = \frac{b-1}{b} \sigma_B^2 + \frac{K-1}{K} \sigma_{\varepsilon}^2$$

$$\rho_1 = \left[ \frac{b-1}{b} \sigma_B^2 - \frac{1}{K} \sigma_{\varepsilon}^2 \right] / \sigma^2$$

$$\rho_2 = -\frac{1}{b} \sigma_B^2 / \sigma^2$$

dimension = 2

$$\begin{bmatrix} \sigma^2 & 0 \\ 0 & \sigma^2 \end{bmatrix} \leftarrow \begin{array}{l} \text{fixed effect} \\ \text{random effect} \end{array}$$

\*sum of block effect  $\neq 0$

$$\begin{bmatrix} \square & 0 \\ 0 & \square \end{bmatrix} \leftarrow \begin{array}{l} \text{fixed intercept effect} \\ \text{random intercept effect} \end{array}$$

— random intercept effect

$$\begin{bmatrix} \square & \beta_B \\ \beta_B & \square \end{bmatrix} \leftarrow \begin{array}{l} \text{sum of block effect} = 0 \\ \text{fixed} \\ \text{random} \end{array}$$

- add a technical error  $\epsilon_{l(j)}$

$$Z_{l(j)} = m_{l(j)} + \epsilon_{l(j)}$$

$$E(Z_{l(j)}) = \mu$$

$$\text{dimension} = 3 \leftarrow \text{cov}(Z_{l(j)}) = (\delta_{jj'} - \frac{1}{b}) \sigma_B^2 + [\delta_{jj'} (\delta_{ll'} - \frac{1}{K})] \cdot \sigma_{\varepsilon}^2 + \delta_{jj'} \cdot \delta_{ll'} \cdot \sigma_{\varepsilon}^2$$

3-19