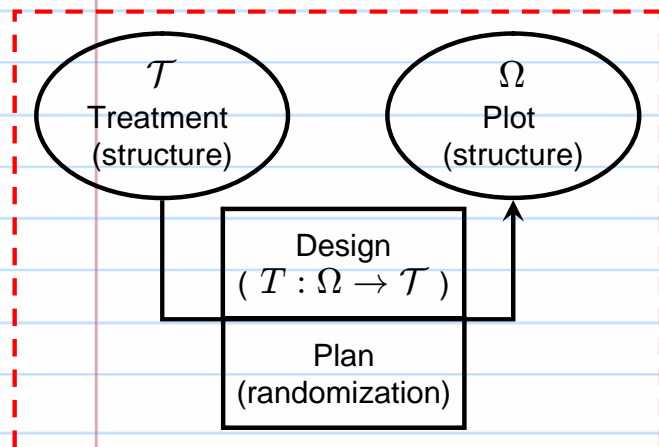


Bailey, Sec.1.5



$$\text{Model} \quad \left(\begin{array}{l} Y_w = \\ \tau_{T(w)} + Z_w \end{array} \right)$$

- The response on plot w is Y_w : a random variable, whose value denoted by y_w
- collected data vector $\mathcal{Y} = \{y_w | w \in \Omega\}$ is a realization of the r.v. \mathcal{Y}

- For plot w , assume treatment-plot additivity (no interaction effects between treatments & plots)

$$Y_w = \tau_{T(w)} + Z_w$$

- $\tau_{T(w)}$: (1) a constant (2) depend on the treatment $T(w)$ (3) can be thought of the contribution of treatment $T(w)$
- Z_w : (1) random variable (2) depend on w (3) can be thought of the contribution of plot w

2.34

- Z_w is a r.v. $\Rightarrow Z_w$ must have probability space \Rightarrow the set of occasions & uncontrolled conditions.
- Z_α & Z_β are different random variables for different plots α & $\beta \Rightarrow Z_\alpha$ & Z_β must have a joint distribution, including a correlation (may be zero)
- Q: What should be the joint dist. of $\{Z_w | w \in \Omega\} \equiv Z_\Omega$

1. simple textbook model

$$Z_w \text{ i.i.d. } N(0, \sigma^2) \Rightarrow Z_\Omega \sim N(0, \sigma^2 I) \rightarrow \text{often seen in completely randomized design}$$

2. fixed effect model

$$Z_w \text{ are indep. Normal with variance } \sigma^2 \text{ \& mean } \mu_w \Rightarrow Z_\Omega \sim N(\mu_\Omega, \sigma^2 I) \rightarrow \text{often seen in block design.}$$

3. random effect model

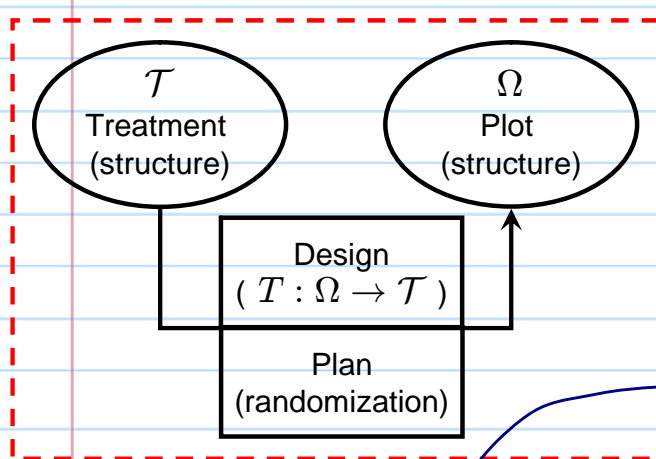
$$Z_w \text{ have identical (marginal) distribution} \rightarrow \text{often seen in block design, split-plot design.}$$

4. randomization model $\Rightarrow Z_\Omega \sim N(0, \Sigma)$

$$Z_w \text{ have identical (marginal) distribution} \rightarrow \text{non-parametric methods.}$$

$$\text{COV}(Z_\alpha, Z_\beta) \text{ depends on the method of randomization.}$$

2.35



Analysis: estimation & testing for function of \mathcal{T} 's

linear combinations

depend on the treatment structure

Example: ($E(Y) = XB$)

1. treatment structure: no structure
e.g. 4 treatments.

$$\begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \tau_i \\ \tau_j \\ \tau_k \\ \tau_l \end{array} \quad \begin{array}{c} \beta_0 = \tau_i, \beta_1 = \tau_j, \beta_2 = \tau_k, \beta_3 = \tau_l \\ X = \begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \beta_0 \ \beta_1 \ \beta_2 \ \beta_3 \\ \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \end{array}
 \end{array}$$

$$\begin{aligned} \beta_0 &= (\tau_i + \tau_j + \tau_k + \tau_l) / 4 = \bar{\tau} \\ \beta_1 &= \tau_j - \bar{\tau} \\ \beta_2 &= \tau_k - \bar{\tau} \\ \beta_3 &= \tau_l - \bar{\tau} \end{aligned} \quad X = \begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \beta_0 \ \beta_1 \ \beta_2 \ \beta_3 \\ \begin{bmatrix} 1 & -1 & -1 & -1 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \end{array}$$

2.36

2. several treatments v.s. a control

$$\begin{array}{c} \text{control} \leftarrow i \\ j \\ k \\ l \end{array} \begin{array}{c} \tau_i \\ \tau_j \\ \tau_k \\ \tau_l \end{array} \quad \begin{array}{c} \beta_0 = \tau_i \\ \beta_1 = \tau_j - \tau_i \\ \beta_2 = \tau_k - \tau_i \\ \beta_3 = \tau_l - \tau_i \end{array} \quad X = \begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \beta_0 \ \beta_1 \ \beta_2 \ \beta_3 \\ \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \end{array}$$

3. factorial structure

$$\begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \tau_i \\ \tau_j \\ \tau_k \\ \tau_l \end{array} \begin{array}{c|c|c} A & B & AB \\ \hline + & + & + \\ + & - & - \\ - & + & - \\ - & - & + \end{array} \quad \begin{array}{c} \beta_0 = (\tau_i + \tau_j + \tau_k + \tau_l) / 4 \\ \beta_A = [(\tau_i + \tau_j) - (\tau_k + \tau_l)] / 2 \\ \beta_B = [(\tau_i + \tau_k) - (\tau_j + \tau_l)] / 2 \\ \beta_{AB} = [(\tau_i + \tau_l) - (\tau_j + \tau_k)] / 2 \end{array} \quad X = \begin{array}{c} i \\ j \\ k \\ l \end{array} \begin{array}{c} \beta_0 \ \beta_A \ \beta_B \ \beta_{AB} \\ \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \end{array}$$

Note: all the X-matrices have same column space, i.e., the space spanned by the columns of X, which will be referred to as treatment space.

(Bailey) the data analysts should give

- minimum variance unbiased estimator of τ_i 's & linear combinations such as $\tau_i - \tau_j$
- estimate the variance of these estimators
- inference about present/absent of effects \Rightarrow test $H_0: \beta_i = 0$.

2.37