

Discussion section 8

Math 181B

1 Review

1.1 Randomized Block Design

From the lecture slides, we have

Suppose we get b blocks and k treatments.

Let Y_{ij} be the observation in the i^{th} block assigned to treatment j .

	Treatment 1	Treatment 2	...	Treatment k	Block effects	
					Block sample means	
Block 1	Y_{11}	Y_{12}		Y_{1k}	$\bar{Y}_{1.}$	β_1
...						
Block b	Y_{b1}	Y_{b2}		Y_{bk}	$\bar{Y}_{b.}$	β_b
Treatment sample means	$\bar{Y}_{.1}$	$\bar{Y}_{.2}$		$\bar{Y}_{.k}$	$\bar{Y}_{..}$	
Average treatment effects	μ_1	μ_2		μ_k		

$$\bar{Y}_{.j} = \frac{1}{b} \sum_{i=1}^b Y_{ij} \qquad \bar{Y}_{i.} = \frac{1}{k} \sum_{j=1}^k Y_{ij} \qquad \bar{Y}_{..} = \frac{1}{bk} \sum_{j=1}^k \sum_{i=1}^b Y_{ij}$$

Figure 1: The illustration of the RBD

We assume

$$Y_{ij} = \mu_j + \beta_i + \varepsilon_{ij}$$

where ε_{ij} are *i.i.d.* normally distributed with mean zero and variance σ^2 , for $i = 1, 2, \dots, b$ and $j = 1, 2, \dots, k$.

We define

- Total sum of squares: $SSTOT = \sum_{j=1}^k \sum_{i=1}^b (Y_{ij} - \bar{Y}_{..})^2$.
- Treatment sum of squares: $SSTR = \sum_{j=1}^k \sum_{i=1}^b (\bar{Y}_{.j} - \bar{Y}_{..})^2 = \sum_{j=1}^k b (\bar{Y}_{.j} - \bar{Y}_{..})^2$.
- Block sum of squares: $SSB = \sum_{i=1}^b \sum_{j=1}^k (\bar{Y}_{i.} - \bar{Y}_{..})^2 = \sum_{i=1}^b k (\bar{Y}_{i.} - \bar{Y}_{..})^2$.

- Error sum of squares: $SSE = \sum_{j=1}^k \sum_{i=1}^b (Y_{ij} - \bar{Y}_{.j} - \bar{Y}_{i.} + \bar{Y}_{..})^2$.

Some facts:

- $SSTOT = SSTR + SSB + SSE$.
- $SSTR$, SSB , and SSE are independent random variables.
- When $H_0 : \mu_1 = \mu_2 = \dots = \mu_k$ is true, $SSTR / \sigma^2$ has a chi square distribution with $k - 1$ degrees of freedom.
- When $H_0 : \beta_1 = \beta_2 = \dots = \beta_b$ is true, SSB / σ^2 has a chi square distribution with $b - 1$ degrees of freedom.
- Regardless of whether the μ_j 's and/or the β_i 's are equal, SSE / σ^2 has a chi square distribution with $(b - 1)(k - 1)$ degrees of freedom.
- When $H_0 : \mu_1 = \mu_2 = \dots = \mu_k$ is true,

$$F = \frac{SSTR / (k - 1)}{SSE / (b - 1)(k - 1)}$$

has an F distribution with $k - 1$ and $(b - 1)(k - 1)$ degrees of freedom.

- When $H_0 : \beta_1 = \beta_2 = \dots = \beta_b$ is true,

$$F = \frac{SSB / (b - 1)}{SSE / (b - 1)(k - 1)}$$

has an F distribution with $b - 1$ and $(b - 1)(k - 1)$ degrees of freedom.