

# MA50259: Statistical Design of Investigations

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Lecture 5: Blocking

# Blocking

- ▶ **Objective:** Reduce the variance of the experimental error ( $\sigma^2$ ) and increase the power for detecting treatment factor effects so that results **generalise to whole population**
- ▶ Choose the experimental units for a study to be as homogeneous as possible. Sometimes difficult!
- ▶ Heterogeneous experimental units are grouped into homogeneous subgroups before they are randomly assigned to treatment factor levels
- ▶ The act of grouping the experimental units together in homogeneous groups is called blocking.
- ▶ In a **randomized block design**, a group of heterogeneous experimental units is used so that the conclusions can be more general

## Examples of blocking

- ▶ Plots of land in agricultural experiments are usually **blocked by proximity** because plots in close proximity normally have similar soil characteristics
- ▶ When experimental units are animals, the **grouping (blocking) of genetically similar animals**, such as littermates, often reduces variability within groups
- ▶ When experimental units are trials, or points in time where treatments will be applied, they are often **blocked by time** since many lurking variables may change over time and trials in close temporal proximity are more alike

## Randomized complete block design (RCB) with one treatment factor

- ▶ Treatment factor has  $t$  levels
- ▶  $b$  blocks (or subgroups of homogeneous experimental units)
- ▶ Each block contains exactly  $t$  experimental units for a total of  $t \times b$  experimental units
- ▶ The  $t$  experimental units within each block are as similar as possible
- ▶ The groups of experimental units vary enough from block to block to allow general conclusions to be drawn
- ▶ The randomization of experimental units to treatment factor levels is performed within each block.

## Comparison between CRD and RCB $t = 3$ , $b = 4$

```
levels<-c("level 1","level 2","level 3")
fac <- levels %>% rep(each = 4) %>% sample(12) %>% factor()
blocks <- factor( rep(c("block 1", "block 2", "block 3", "block 4"), each=3))
CRD <- data.frame( units=1:12,block=blocks,treatmentCRD=fac)
block1 <- sample(levels,3); block2 <- sample(levels,3)
block3 <- sample(levels,3); block4 <- sample(levels,3)
t<-c(block1,block2,block3,block4) %>% factor()
RCB<-data.frame(block = blocks, treatmentRCB = t)
cbind(CRD,RCB)
```

	units	block	treatmentCRD	block	treatmentRCB
1	1	block 1	level 2	block 1	level 3
2	2	block 1	level 3	block 1	level 2
3	3	block 1	level 3	block 1	level 1
4	4	block 2	level 1	block 2	level 3
5	5	block 2	level 2	block 2	level 2
6	6	block 2	level 2	block 2	level 1
7	7	block 3	level 1	block 3	level 2
8	8	block 3	level 2	block 3	level 1
9	9	block 3	level 1	block 3	level 3
10	10	block 4	level 3	block 4	level 2
11	11	block 4	level 3	block 4	level 1
12	12	block 4	level 1	block 4	level 3

# Statistical model

$$y_{ij} = \mu + b_i + \tau_j + \epsilon_{ij}$$

where

- ▶  $b_i$  is the effect of block  $i \in \{1, \dots, b\}$
- ▶  $\tau_1, \dots, \tau_t$  are the treatment effects
- ▶ All  $\{\epsilon_{ij}\}$  are  $N(0, \sigma^2)$  and independent
- ▶ Note: There is no interaction between block and treatment
- ▶ Only  $t \times b$  experimental units, there would be zero degrees of freedom for the error term  $ssE$  if a block by treatment interaction term were included!