# Assignment 1

Your Name

2025-08-25

## Goal of this assignment

The goals of this assignment are to:

- Review a few basic concepts, and
- Make sure that you can knit an Rmd document with the basic features that will be needed throughout
  this course.

#### 1.

Using your own words, describe a randomized complete block design (RCBD) in 3-5 sentences. Then write out a statistical model (using mathematical notation) that could be fitted to data generated by an RCBD.

An RCBD is a type of experiment design where there were groups of similar experimental units (i.e., the blocks) and all treatments fit once in each block. The capacity to run an experiment in an RCBD is independent of the treatment structure. To design an RCBD, we would randomize all treatments within each block. The "block effect" is typically considered a nuisance parameter, but they account for known variability in the data and may reduce the residual variance  $\sigma_{\varepsilon}^2$ . We assume that there is no block × treatment interaction.

A statistical model that could model data generated by an RCBD is

$$y_{ij} = \mu + T_i + b_j + \varepsilon_{ij}, \varepsilon_{ij} \sim N(0, \sigma^2),$$

where:

- $y_{ij}$  is the observation of the *i*th treatment in the *k*th block,
- $\mu$  is the overall mean,
- $T_i$  is the effect of the jth treatment,
- $b_j$  is the effect of the kth block,
- $\varepsilon_{ij}$  is the residual of the *i*th treatment in the *j*th block,
- $\sigma^2$  is the residual variance.

A more broad statistical model that could model data generated by an RCBD is

$$y_{ij} \sim P(\mu_{ij}, \phi), g(\mu_{ij}) = \eta_{ij} = \eta_0 + T_i + b_j,$$

where:

- $y_{ij}$  is the observation of the ith treatment in the kth block, that arises from probability distribution P,
- $\mu_{ij}$  is the mean of said observation and distribution,
- $\psi$  is the dispersion parameter of the distribution P,
- $g(\cdot)$  is the link function that connects the linear predictor to the mean, so that the value for the mean is always within its support,
- $\eta_0$  is the overall mean of the linear predictor,

- $T_i$  is the effect of the jth treatment,
- $b_j$  is the effect of the kth block.

### 2.

#### Edit the R code below:

- Silence the messages and warnings: your submitted pdf or html should include the code, but not the warnings/messages seen below.
- Print a summary of model1.
- Find and print the value for  $\hat{\sigma}^2$  estimated in model1.

```
library(tidyverse)
library(agridat)
data("omer.sorghum")
df <- omer.sorghum %>% filter(env == "E3")
model1 <- lm(yield ~ 1 + gen + rep, data = df)</pre>
model1
##
## Call:
## lm(formula = yield ~ 1 + gen + rep, data = df)
##
## Coefficients:
   (Intercept)
                                    genG03
                                                  genG04
##
                      genG02
                                                               genG05
                                                                             genG06
##
        672.03
                      -66.52
                                    223.07
                                                  167.40
                                                                61.05
                                                                            -267.21
                      genG08
                                    genG09
                                                  genG10
##
        genG07
                                                               genG11
                                                                             genG12
##
        355.20
                      159.53
                                    231.30
                                                 156.66
                                                                             -42.87
                                                               146.29
##
        genG13
                      genG14
                                                  genG16
                                                               genG17
                                                                             genG18
                                    genG15
                                                  22.81
                                                              -296.66
                                                                             448.28
##
        243.17
                        1.06
                                     64.72
##
         repR2
                       repR3
                                     repR4
##
       -150.83
                      -85.66
                                   -124.21
# different ways to get #sigma^2
sigma(model1)^2
## [1] 25627.36
summary(model1)$sigma^2
## [1] 25627.36
sum(model1$residuals^2) / model1$df.residual
## [1] 25627.36
```