

# **GENERALIZED RCBDS AND FACTORIAL RCBDS**

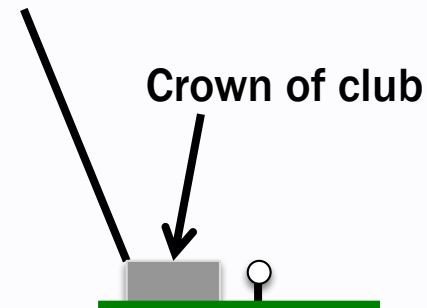
**Chapter 10**

# LEARNING OBJECTIVES

- Define generalized RCBD in terms of block size
- Describe randomization and perform in R
- Write statistical model and compare to RCBD model
- Perform treatment factorial analysis in RCBD
- Explain how some block/treatment interactions could be estimated in this scenario

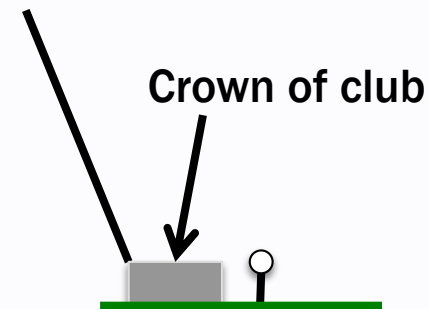
# GOLF TEE HEIGHT

- *Golf Magazine* (2006, June) experimented to determine the **ideal tee height** for driving a golf ball
- **Purpose:** recommend tee height for any reader
- **Treatment levels:**
  1. **Entire ball below crown**
  2. Half ball above crown
  3. Bottom of ball at top of club-face
- Nine golfers chosen that were representative sample of all readers



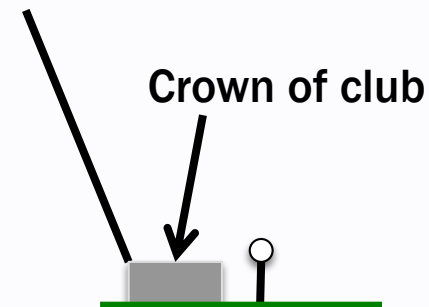
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# GOLF TEE HEIGHT

- For this to be an RCBD, each golfer would use each tee height exactly one time
- Most of the experimental error variability comes from golfer-to-golfer variability
- Do not anticipate much variability between drives from same golfer
- Ask each golfer to **replicate each treatment 5 times** in a randomized order
- **Each golfer experiment is a CRD with  $r=5$ ,  $t=3$**

# GENERALIZED RCBD

- This is an example of a **generalized RCBD**
  - Each block is an **equally-replicated CRD** with  $r > 1$
  - Block size is a multiple of  $t$
- Randomize like RCBD, we just have more replicates
- Only do this if within-block EU variability won't significantly increase by including more EUs
  - Increasing EUs was why we blocked in the first place!
- Gives estimate of pure error and we can separate out block/treatment interaction!

# GENERALIZED RCBD STATISTICAL MODEL

- Since we replicate treatments in each block we now have df for error after fitting block/trt interactions

$$Y_{hij} = \mu + \beta_h + \tau_i + \beta\tau_{hi} + E_{hij}$$

- **Group:** write what the index ranges are for  $h$ ,  $i$ , and  $j$
- Pure error estimate based on comparing each block/treatment replicate response to average response

$$\sum_{h,i,j} (Y_{hij} - \bar{Y}_{hi.})^2$$

- Pure error has  $bt(r-1)$  degrees of freedom



# GENERALIZED RCBD STATISTICAL MODEL

- The F-ratio test for block effects is still invalid
- The F-ratio test for block/treatment interaction though is **valid!**
  - $F = ms_{BT} / ms_E$  with  $(b-1)(t-1)$  numerator df
- If we have a significant interaction, then consider **consistency of treatment contrasts** across blocks
  - Plots can help guide you on this

# GENERALIZED RCBD

## GENERAL TREATMENT EFFECTS

- If interaction effect insignificant, remove it from the model and **ssBT and its df are pooled with error**
- General conclusions about treatment effects **in the presence of interaction** uses msBT in  $F$  denominator
  - Strong overall treatment effects should **overcome the block-to-block effect variability**
  - Numerator df =  $t-1$ , Denominator df =  $(b-1)(t-1)$
  - Mimics test from RCBD
- Only recommended when the blocks are representative sample of a large population
  - **Counter-example:** experiment with 2 blocks

# FACTORIAL TREATMENTS IN RCBD

- Let's revisit the RCBD scenario but with a twist
- What if the treatments have a factorial structure?
  - Single blocking factor
  - Set of treatment factors  $A, B, C, \dots$
- Called an **RCBD factorial** experiment
- Randomization is the same, the analysis approach changes

# RCBD FACTORIAL ANALYSIS STANDARD MODEL

- Expand treatment effects just like we did before

$$Y_{hij} = \mu + \beta_h + \tau_{ij} + E_{hij}$$

$$\tau_{ij} = \alpha_i + \gamma_j + \alpha\gamma_{ij}$$

$$h = 1, \dots, b$$

$$i = 1, \dots, a$$

$$j = 1, \dots, c$$

$$Y_{hij} = \mu + \beta_h + \alpha_i + \gamma_j + \alpha\gamma_{ij} + E_{hij}$$

- This is a **3 factor model** that includes:
  - All main effects
  - One two-factor interaction between the last two factors
- **We already know how to fit this model!**

# RCBD FACTORIAL ANALYSIS

## BLOCK/TREATMENT INTERACTIONS

- For RCBD we cannot estimate the block/treatment interactions because we didn't have enough DF
- Only necessary if we want to consider **any treatment contrast** can change depending on the block
  - **Necessary DF = (b-1)(ac-1)...remember ac=t, # treatments**
- Block/treatment interactions here expand to three different classes

$$\beta\alpha_{hi}$$

$$(b-1)(a-1)$$

$$\beta\gamma_{hj}$$

$$(b-1)(c-1)$$

$$\beta\alpha\gamma_{hij}$$

$$(b-1)(a-1)(c-1)$$

Test DF =

- Adding these DF gives total interaction DF **(b-1)(ac-1)**

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