# Lab 8 - Nested ANOVA

October 8 & 9, 2017 FANR 6750

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# Scenario

We subsample each experimental unit

For example

- We count larvae at multiple subplots within a plot
- We weigh multiple chicks in a brood

We're interested in treatment effects at the experimental (whole) unit level, not the subunit level

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# THE ADDITIVE MODEL

$$y_{ijk} = \mu + \alpha_i + \beta_{ij} + \varepsilon_{ijk}$$

Because we want our inferences to apply to all experimental units, not just the ones in our sample,  $\beta_{ij}$  is random.

Specifically:

$$\beta_{ij} \sim \mathsf{Normal}(0, \sigma_B^2)$$

And as always,

$$\varepsilon_{ijk} \sim \mathsf{Normal}(0, \sigma^2)$$

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#### Hypotheses

#### **Treatment effects**

 $H_0: \alpha_1 = \cdots = \alpha_a = 0$  $H_a:$  at least one inequality

#### Random variation among experimental units

 $H_0: \sigma_B^2 = 0$  $H_a: \sigma_B^2 > 0$ 

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# Incorrect analysis

## The denominator degrees-of-freedom are wrong

## Example data

#### Import data

```
gypsyData <- read.csv("gypsyData.csv")
str(gypsyData)

## 'data.frame': 36 obs. of 3 variables:
## $ larvae : num 16 16 15.8 14.2 13.9 14.2 13.5 13.4 14 13.1
## $ Treatment: Factor w/ 3 levels "Bt", "Control",..: 1 1 1 1 1
## $ Plot : int 1 1 1 1 2 2 2 2 3 3 ...</pre>
```

#### Convert Plot to a factor and then cross-tabulate

```
gypsyData$Plot <- factor(gypsyData$Plot)
table(gypsyData$Treatment, gypsyData$Plot)

##

##

1 2 3 4 5 6 7 8 9

##

Bt

4 4 4 0 0 0 0 0 0

##

Control 0 0 0 4 4 4 0 0 0

##

Dimilin 0 0 0 0 0 4 4 4</pre>
```

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## CORRECT ANALYSIS

USING aov

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#### WHAT HAPPENS IF WE ANALYZE PLOT-LEVEL MEANS?

The aggregate function is similar to tapply but it works on entire data.frames. Here we get averages for each whole plot.

```
plotData
    Treatment Plot larvae
## 1
                 1 15.50
## 2
           Bt
                 2 13.75
## 3
                 3 14.00
      Control
                 4 18.25
      Control
                 5 18.75
## 5
## 6
      Control
                 6 19.25
## 7
      Dimilin
                 7 12.50
## 8
      Dimilin
                 8 13.50
## 9
      Dimilin
                 9 13.00
```

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### Issues

#### When using using aov with Error term:

- You can't use TukeyHSD
- You don't get a direct estimate of  $\sigma_R^2$
- Doesn't handle unbalanced designs well
- But, you can use model.tables and se.contrast

#### An alternative is to use lme function in nlme package

- $\bullet$  Possible to get direct estimates of  $\sigma_B^2$  and other variance parameters
- Handles very complex models and unbalanced designs
- Possible to do multiple comparisons and contrasts using the the glht function in the multcomp package.
- But...
- Only works if there random effects
- ANOVA tables aren't as complete as aov

# F AND p VALUES ARE THE SAME AS BEFORE

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## USING THE lme FUNCTION

```
anova(lme1, Terms="Treatment")

## F-test for: Treatment

## numDF denDF F-value p-value

## 1 2 6 57.86567 1e-04
```

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#### Variance parameter estimates

The first row shows the estimates of  $\sigma_B^2$  and  $\sigma_B$ . The second row shows the estimates of  $\sigma^2$  and  $\sigma$ 

```
VarCorr(lme1)

## Plot = pdLogChol(1)

## Variance StdDev

## (Intercept) 0.3363889 0.5799904

## Residual 0.5155556 0.7180220
```

There is more random variation within whole units than among whole units (after accounting for treatment effects)

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# EXTRACT THE PLOT-LEVEL RANDOM EFFECTS

These are the  $\beta_{ij}$ 's

```
round(ranef(lme1), 2)
##
     (Intercept)
             0.78
## 1
            -0.48
## 2
            -0.30
## 3
            -0.36
## 4
## 5
             0.00
## 6
             0.36
## 7
            -0.36
## 8
             0.36
             0.00
## 9
```

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## Multiple comparisons

```
tuk <- glht(lme1, linfct=mcp(Treatment="Tukey"))</pre>
summary(tuk)
    Simultaneous Tests for General Linear Hypotheses
## Multiple Comparisons of Means: Tukey Contrasts
##
## Fit: lme.formula(fixed = larvae ~ Treatment, data = gypsyData, random = ~1 |
##
      Plot)
## Linear Hypotheses:
                         Estimate Std. Error z value Pr(>|z|)
## Control - Bt == 0
                         4.3333
                                      0.5569 7.781 < 0.001 ***
                                      0.5569 -2.544 0.0295 *
## Dimilin - Bt == 0
                          -1.4167
## Dimilin - Control == 0 -5.7500
                                      0.5569 -10.324
                                                     <0.001 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

## (Adjusted p values reported -- single-step method)

### Assignment

To determine if salinity affects adult fish reproductive performance, a researcher places one pregnant female in a tank with one of three salinity levels: low, medium, and high, or a control tank. A week after birth, two offspring (fry) are measured.

Run a nested ANOVA using aov and lme on the fishData.csv dataset. Answer the following questions:

- (1) What are the null and alternative hypotheses?
- (2) Does salinity affect fry growth?
- (3) If so, which salinity levels differ?
- (4) Is there more random variation among or within experimental units?

Upload your self-contained  $\mathbf{R}$  script to ELC at least one day before your next lab

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