

Two Way ANOVA: Special Cases

Lecture 21

Alexandra Chronopoulou



COLLEGE OF LIBERAL ARTS & SCIENCES

Department of Statistics
101 Illini Hall, MC-374
725 S. Wright St.
Champaign, IL 61820-5710

- When the treatment sample sizes are unequal, the analysis of variance for two-factor studies becomes more complex.
- The least-squares equations are no longer of a simple structure and the regular analysis of variance formulas are now inappropriate.
- Furthermore, the factor effect component sum of squares are no longer orthogonal; that is, **they do not sum up to TSS**.

- Consider the rats example from the previous lecture.
- Remove the first observation to make the data *unbalanced*.
- Use the `anova()` command for *each* models:
 - (1) `poison*treat` - i.e. poison first, treat second
 - (2) `treat*poison` - i.e. treat first, poison second
- Observe that the results change depending on the order the factors are introduced in the model.

```
anova(lm(1/time ~ treat*poison, newrats))
```

```
## Analysis of Variance Table
##
## Response: 1/time
##           Df Sum Sq Mean Sq F value    Pr(>F)
## treat       3  20.136   6.7121  29.6807 9.986e-10 ***
## poison      2  35.102  17.5510  77.6094 1.362e-13 ***
## treat:poison 6   1.980   0.3300   1.4592  0.2207
## Residuals  35   7.915   0.2261
##
```

```
anova(lm(1/time ~ poison*treat, newrats))
```

```
## Analysis of Variance Table
##
## Response: 1/time
##           Df Sum Sq Mean Sq F value    Pr(>F)
## poison      2  36.672  18.3358  81.0799 7.276e-14 ***
## treat       3  18.567   6.1889  27.3670 2.706e-09 ***
## poison:treat 6   1.980   0.3300   1.4592  0.2207
## Residuals  35   7.915   0.2261
##
```

$$Y_{ijk} = \mu_{..} + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \epsilon_{ijk}$$

- Due to the lack of orthogonality, the ANOVA F -tests are not applicable.
- We will express the ANOVA model as a regression model with indicator (dummy) variables.
- We need $a - 1$ indicator variables for factor A main effects and $b - 1$ indicator variables for factor B main effects. The interactions correspond to the cross products of the indicator variables for A and B .

$$\left\{ \begin{array}{l} H_0 : \text{Smaller model} \\ H_\alpha : \text{Larger model} \end{array} \right.$$

- The partial F -test is:

$$F = \frac{(RSS_0 - RSS_\alpha)/(df_0 - df_\alpha)}{RSS_\alpha/df_\alpha} \sim F_{df_0 - df_\alpha, df_\alpha}$$

- Reject H_0 , if F test statistic is large, that is, the variation missed by the reduced model, when being compared with the error variance, is significantly large.

Partial F -tests in R

```
ginv1 <- lm(time~-1 ~ poison*treat, data=newrats)
ginv2 <- lm(time~-1 ~ poison + treat, data=newrats)
anova(ginv2, ginv1)
```

```
## Analysis of Variance Table
##
## Model 1: time~-1 ~ poison + treat
## Model 2: time~-1 ~ poison * treat
##   Res.Df    RSS Df Sum of Sq      F Pr(>F)
## 1      41  9.8951
## 2      35  7.9151   6      1.98 1.4592 0.2207
```

```
ginv2 <- lm(time~-1 ~ poison + treat, data=newrats)
ginv3 <- lm(time~-1 ~ poison, data=newrats)
anova(ginv3, ginv2)
```

```
## Analysis of Variance Table
##
## Model 1: time~-1 ~ poison
## Model 2: time~-1 ~ poison + treat
##   Res.Df    RSS Df Sum of Sq      F      Pr(>F)
## 1      44 28.4618
## 2      41  9.8951   3    18.567 25.644 1.669e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Type III Sums of Squares

- Use the `Anova()` command with `Type="III"` specification from the `car` library.
- This type tests for the presence of an effect given that both the other effects are in the model.

```
Anova(lm(1/time ~ treat*poison, data=newrats), type="III")
```

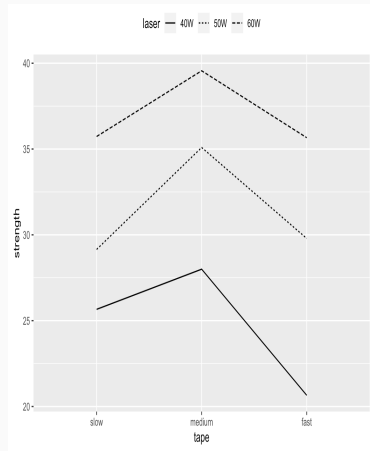
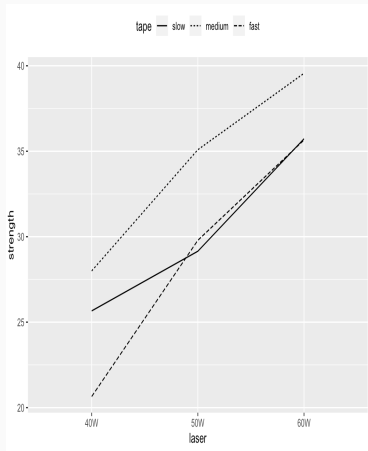
```
## Anova Table (Type III tests)
##
## Response: 1/time
##              Sum Sq Df F value    Pr(>F)
## (Intercept) 15.0605  1 66.5967 1.298e-09 ***
## treat        2.1340  3  3.1455  0.03723 *
## poison      11.7375  2 25.9514 1.225e-07 ***
## treat:poison  1.9800  6  1.4592  0.22073
## Residuals    7.9151 35
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```


- Only one observation in each cell, so we cannot fit the interaction model.
- There are no degrees of freedom left for estimating the error.
- $RSS = 0$ when the model includes main effects and interaction term.
- All F -tests are valid, but the interaction model is not a candidate model.

```
fullmodel <- lm(strength ~ laser*tape, composite)
summary(fullmodel)
```

```
##
## Call:
## lm(formula = strength ~ laser * tape, data = composite)
##
## Residuals:
## ALL 9 residuals are 0: no residual degrees of freedom!
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25.66          NA      NA      NA
## laser50W          3.49          NA      NA      NA
## laser60W         10.07          NA      NA      NA
## tapemedium        2.34          NA      NA      NA
## tapefast        -5.01          NA      NA      NA
## laser50W:tapemedium  3.60          NA      NA      NA
## laser60W:tapemedium  1.49          NA      NA      NA
## laser50W:tapefast   5.65          NA      NA      NA
## laser60W:tapefast   4.94          NA      NA      NA
##
## Residual standard error: NaN on 0 degrees of freedom
## Multiple R-squared:      1, Adjusted R-squared:      NaN
## F-statistic:      NaN on 8 and 0 DF,  p-value: NA
```

Interaction Plots



```
meffectmodel <- lm(strength ~ laser + tape, composite)
summary(meffectmodel)
```

```
##
## Call:
## lm(formula = strength ~ laser + tape, data = composite)
##
## Residuals:
##      1      2      3      4      5      6      7      8
## 1.74222 -1.34111 -0.40111  0.04556  0.56222 -0.60778 -1.78778  0.77889
##      9
## 1.00889
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   23.918      1.208   19.803 3.84e-05 ***
## laser50W       6.573      1.323    4.968 0.007661 **
## laser60W      12.213      1.323    9.231 0.000765 ***
## tapemedium     4.037      1.323    3.051 0.037991 *
## tapefast     -1.480      1.323   -1.119 0.325944
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.62 on 4 degrees of freedom
## Multiple R-squared:  0.963, Adjusted R-squared:  0.9259
## F-statistic:    26 on 4 and 4 DF, p-value: 0.004013
```

- Consider the following model that includes interactions:

$$y_{ij} = \mu + \alpha_i + \beta_j + \theta \alpha_i \beta_j + \epsilon_{ij}$$

- Here, we assume that the interactions are of *multiplicative* nature, i.e.

$$(\alpha\beta)_{ij} = \theta \alpha_i \beta_j$$

- Consider the SSA , SSB as before and :

$$SSAB^* = \frac{\left(\sum_i \sum_j (\bar{y}_{i\cdot} - \bar{y}_{..})(\bar{y}_{\cdot j} - \bar{y}_{..})y_{ij} \right)^2}{\sum_i (\bar{y}_{i\cdot} - \bar{y}_{..})^2 \sum_j (\bar{y}_{\cdot j} - \bar{y}_{..})^2}$$

- The TSS is computed as usual and is decomposed as

$$TSS = SSA + SSB + SSAB^* + SSRem^*$$

where the remainder is

$$SSRem^* = TSS - SSA - SSB - SSAB^*$$

- We want to test the following hypothesis

$$\begin{cases} H_0 : \theta = 0 & (\text{no interactions}) \\ H_\alpha : \theta \neq 0 & (\text{interactions}) \end{cases}$$

which is essentially a test for model additivity.

- The test statistic computes as

$$F^* = \frac{SSAB^*/1}{SSRem^*/(ab - a - b)}$$

Implementation of Tukey's Additivity Test in R

```
meffectmodel <- lm(strength ~ laser + tape, composite)
lasercoefs <- rep(c(0, 6.5733, 12.2133), 3)
tapecoefs <- rep(c(0, 4.0367, -1.4800), each=3)

newmod <- update(meffectmodel, .~. + I(lasercoefs*tapecoefs))
anova(newmod)
```

```
## Analysis of Variance Table
##
## Response: strength
##
```

| | Df | Sum Sq | Mean Sq | F value | Pr(>F) |
|------------------------------|----|---------|---------|---------|-------------|
| ## laser | 2 | 224.184 | 112.092 | 36.8201 | 0.007745 ** |
| ## tape | 2 | 48.919 | 24.459 | 8.0344 | 0.062401 . |
| ## I(lasercoefs * tapecoefs) | 1 | 1.370 | 1.370 | 0.4501 | 0.550340 |
| ## Residuals | 3 | 9.133 | 3.044 | | |

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```