

# **Analysis of Variance (Unbalanced Case)**

# Review: Balanced Case

- Breaking up variation of data based on source
- Terms **orthogonal** i.e., You can uniquely decompose variation
- The **order of variables** does not change the result

This is not the case for an unbalanced case. The order of the variables does change the result.

# Review: Balanced Case

This code only focuses on the two main effects, does not include the interaction effect)

```
summary(aov(Toothlength ~ Supplement + Dose, data=tooth))
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Supplement    1  205.4    205.4    14.02 0.000429 ***
## Dose          2 2426.4   1213.2    82.81 < 2e-16 ***
## Residuals    56  820.4     14.7
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1
' ' 1
```

```
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##              Df Sum Sq Mean Sq F value    Pr(>F)
## Dose          2 2426.4   1213.2    82.81 < 2e-16 ***
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## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '
1
```

# Example: ozkids data

Continuous response:

Testing to see the significant effect of predictors on days absent.

- **days**: days absent

Categorical predictors:

- **origin**: Aboriginal or not
- **sex**: male or female
- **grade**: level in school
- **type**: type of learner

```
str(ozkid)
```

```
## 'data.frame':    154 obs. of  5 variables:
## $ origin: Factor w/ 2 levels "A","N": 1 1 1 1 1 1 1 1 1 1 ...
## $ sex   : Factor w/ 2 levels "F","M": 2 2 2 2 2 2 2 2 2 2 ...
## $ grade : Factor w/ 4 levels "F0","F1","F2",...: 1 1 1 1 1 1 1 1 2 2
..
## $ type  : Factor w/ 2 levels "AL","SL": 2 2 2 1 1 1 1 1 2 2 ...
## $ days  : int   2 11 14 5 5 13 20 22 6 6 ...
```

# Unbalanced Case

- Still want to decompose variation – to test significance of variables
- Effects are not orthogonal, i.e., **Decomposition is not unique**

There is some overlap

```
table(ozkid$origin); table(ozkid$sex)  # check unbalance
```

```
##
```

```
##   A   N
```

```
## 74 80
```

```
##
```

```
##   F   M
```

```
## 84 70
```

Confirmed that the data is unbalanced.

# Unbalanced Case

```
summary(aov(days ~ origin + grade, data=ozkid))
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## origin         1   2646   2645.7    11.580 0.000856 ***
## grade          3   2020    673.4     2.947 0.034821 *
## Residuals     149  34040    228.5
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
=====
summary(aov(days ~ grade + origin, data=ozkid))
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## grade          3   2277    759.1     3.323 0.02149 *
## origin         1   2389   2388.5    10.455 0.00151 **
## Residuals     149  34040    228.5
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

When you change the order of the variables, it completely changes the results when comparing the top to the bottom code

# Types of Sums of Squares

- 3-way ANOVA model with variables A, B, and C
- Notations:
- **$SS(C|A\ B)$** -additional contribution when C is added to model containing A and B
- **$SS(C) \neq SS(C|AB)$**  if unbalanced
- **$SS(C) = SS(C|AB)$**  if balanced

"C" given (|) "A" and "B"

"Types" are used to calculate the sum of squares when you have an unbalanced case.

# Type I

- Sequential sum of squares (SS)
- Results from `aov()` or `lm()`
- **Additional variation explained by the model when that term is added to terms already in**
  - If we run `aov(Y~A+B+C)`, it displays `SS(A)`, `SS(B | A)`, and `SS(C | A,B)`

Sequential means: "A" goes in, then when "B" goes in, it is added to "A", and when "C" goes in, it is added to "A" and "B".
  - If we run `aov(Y~B+A+C)`, it displays `SS(B)`, `SS(A | B)`, and `SS(C | A,B)`
- Order does matter, like which variable is added in the model first

With Type 1, order matters, but the order will not impact the Total SS



# Type III

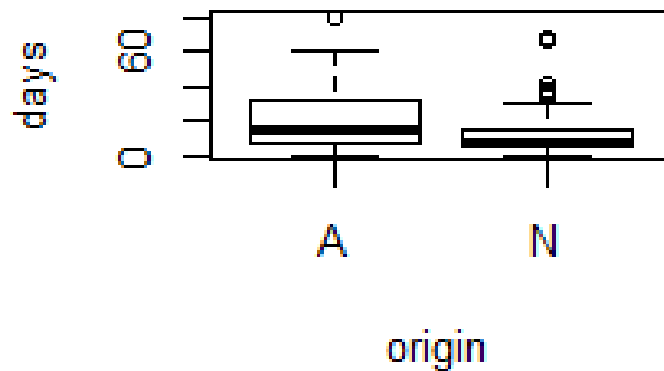
- Partial sums of squares
- Available from Anova() in package “car”
- **Explained variation that term adds when all other terms are already included**
  - If we run `aov(Y~A+B+C)`, it displays  $SS(A|B,C)$ ,  $SS(B|A,C)$ , and  $SS(C|A,B)$
  - If we run `aov(Y~B+A+C)`, it displays  $SS(B|A,C)$ ,  $SS(A|B,C)$ , and  $SS(C|A,B)$
- Order does not matter. Consistent outputs
- <https://www.r-bloggers.com/anova-%E2%80%93-type-iiiii-ss-explained/>

# Exercise: data exploration

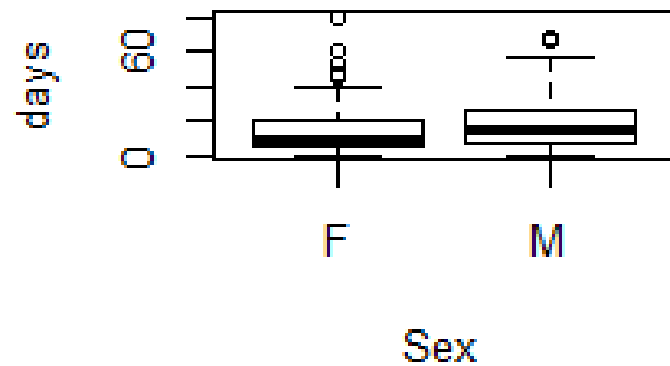
- Boxplots of days by Origin
- Or by sex, grade, or type
- Informally compare means and variations of absent day among groups

# Exercise: data exploration

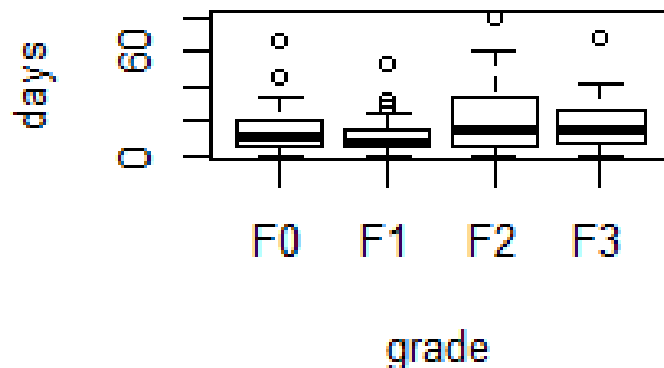
**Days by Origin**



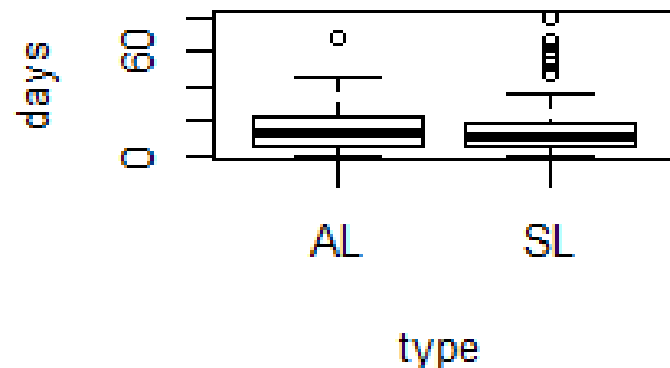
**Days by Sex**



**Days by grade**



**Days by type**



# Example: **origin** and **grade** Models

- Days absent with **origin** and **grade** predictors
- See Type I and Type III sums of squares
- Reverse order of terms (e.g. **grade** first and then **origin**)
- What stays the same, and what changes?
- Impact of reversing main effects on interaction model?

# Example: origin and grade Models

```
aov.res1= aov(days ~ grade + origin, data=ozkid) # type 1 test
```

```
Anova(aov.res1, type=3) # type 3 test
```

```
## Anova Table (Type III tests)
```

```
##
```

```
## Response: days
```

	Sum Sq	Df	F value	Pr(>F)	
## (Intercept)	8148	1	35.6655	1.648e-08	***
## grade	2020	3	2.9474	0.034821	*
## origin	2389	1	10.4550	0.001506	**
## Residuals	34040	149			

```
=====
```

```
aov.res2= aov(days ~ origin + grade, data=ozkid)
```

```
Anova(aov.res2, type=3)
```

```
## Anova Table (Type III tests)
```

```
##
```

```
## Response: days
```

	Sum Sq	Df	F value	Pr(>F)	
## (Intercept)	8148	1	35.6655	1.648e-08	***
## origin	2389	1	10.4550	0.001506	**
## grade	2020	3	2.9474	0.034821	*
## Residuals	34040	149			

# Exercise: Type III Analysis in Four-Way Main Effects Model

- Type III SS for the four-way main effects model
- Backward elimination and conclusions about main effects to keep in the model?
- Get Type I SS for each of the orderings of the terms we might want to keep
- Could we further reduce the main effects we would want to keep in the model?

# Exercise: Multiple Comparisons

- Fit model with previous main effects and all interactions between them.
- Which main effects and interactions kept?
- Do Tukey multiple comparison for the main effects and note differences.
- Significantly different groups?