# One-Way ANOVA

## R Handout

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#### Initialization

```
> library(NCStats)
> library(multcomp) # glht()
```

# Raspberry Example

## Background

A researcher is interested in the effect of irrigation on fruit production by raspberry plants. The researcher has determined that he will examine the effects of 100 ml (a maintenance amount), 200, 400, and 800 ml of water per pot. The researcher had 16 identical planting pots available and much more than that number of raspberry plant seedlings. A square table for growing the plants in a greenhouse was also available. He had enough time to let the plants mature to the point of producing fruit (i.e. berries) or not. At the end of this period, the total weight (g) of mature berries was recorded.

## Fitting the Linear Model

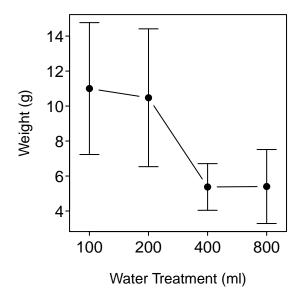
#### > summary(lm1)

#### Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 11.0000 0.9425 11.672 6.58e-08
water200 -0.5250 1.3328 -0.394 0.70057
water400 -5.6250 1.3328 -4.220 0.00119
water800 -5.6000 1.3328 -4.202 0.00123

Residual standard error: 1.885 on 12 degrees of freedom Multiple R-squared: 0.7296, Adjusted R-squared: 0.662 F-statistic: 10.79 on 3 and 12 DF,  $\,$ p-value: 0.001004

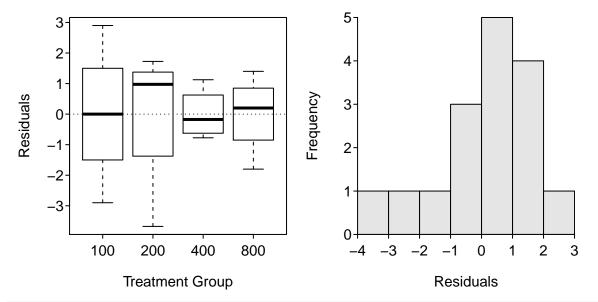
> fitPlot(lm1,xlab="Water Treatment (ml)",ylab="Weight (g)",main="")



# Checking the Assumptions

### > levenesTest(lm1)

> residPlot(lm1)



> adTest(lm1\$residuals)

Anderson-Darling normality test with x A = 0.4308, p-value = 0.2688

> outlierTest(lm1)

No Studentized residuals with Bonferonni p < 0.05 Largest |rstudent|: rstudent unadjusted p-value Bonferonni p 8 -2.836044 0.016196 0.25914

## **Multiple Comparison Tests**

```
> rasp.mc <- glht(lm1, mcp(water = "Tukey"))
> summary(rasp.mc)
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: lm(formula = weight ~ water, data = rasp)

Linear Hypotheses:

```
Estimate Std. Error t value Pr(>|t|)
                             1.333 -0.394 0.97835
200 - 100 == 0
                -0.525
400 - 100 == 0
                 -5.625
                             1.333
                                   -4.220 0.00571
800 - 100 == 0
                                   -4.202 0.00579
                 -5.600
                             1.333
400 - 200 == 0
                 -5.100
                             1.333
                                    -3.826 0.01113
800 - 200 == 0
                 -5.075
                             1.333
                                    -3.808 0.01160
800 - 400 == 0
                 0.025
                             1.333
                                     0.019
                                           1.00000
(Adjusted p values reported -- single-step method)
```

### > confint(rasp.mc)

```
Simultaneous Confidence Intervals
```

```
Multiple Comparisons of Means: Tukey Contrasts
```

```
Fit: lm(formula = weight ~ water, data = rasp)
```

```
Quantile = 2.9676
```

95% family-wise confidence level

### Linear Hypotheses:

```
Estimate lwr upr

200 - 100 == 0 -0.5250 -4.4803 3.4303

400 - 100 == 0 -5.6250 -9.5803 -1.6697

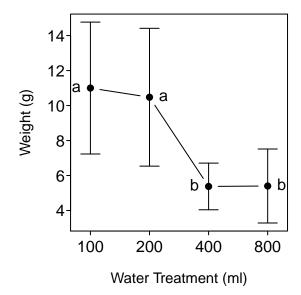
800 - 100 == 0 -5.6000 -9.5553 -1.6447

400 - 200 == 0 -5.1000 -9.0553 -1.1447

800 - 200 == 0 -5.0750 -9.0303 -1.1197

800 - 400 == 0 0.0250 -3.9303 3.9803
```

```
> fitPlot(lm1,xlab="Water Treatment (ml)",ylab="Weight (g)",main="")
> addSigLetters(lm1,lets=c("a","a","b","b"),pos=c(2,4,2,4))
```



# Benthic Infaunal Example

## Background

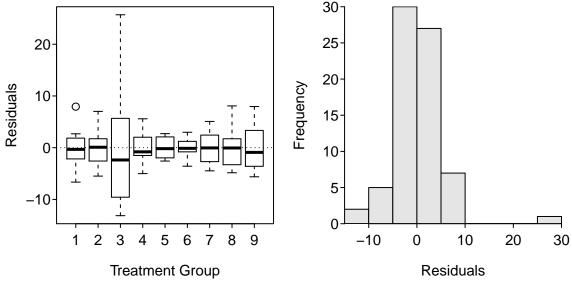
Australian researchers were interested in the effect of effluent releases on benthic organisms in the release area. To examine the effect, the researchers recorded the total abundance of benthic organisms at 8 haphazardly-

selected sublocations at each of 8 control locations (thought to have not been impacted by the effluent release) and 1 potentially impacted location. The results are recorded in BenthicInfaunal.csv. Use these data to determine if the mean abundance of benthic organisms differs between the locations (and, especially, if the impacted location differs from any of the control locations).

```
> ben <- read.csv("BenthicInfaunal.csv")
> ben$site <- factor(ben$site)
> str(ben)

'data.frame': 72 obs. of 2 variables:
$ site : Factor w/ 9 levels "1","2","3","4",..: 1 1 1 1 1 1 1 2 2 ...
$ abundance: num 14.4 20.4 21.2 17.6 29 ...
```

## Assumption Checking with Possible Transformations



> adTest(lm2\$residuals)

```
Anderson-Darling normality test with x A = 1.6389, p-value = 0.0002996
```

```
> outlierTest(lm2)
```

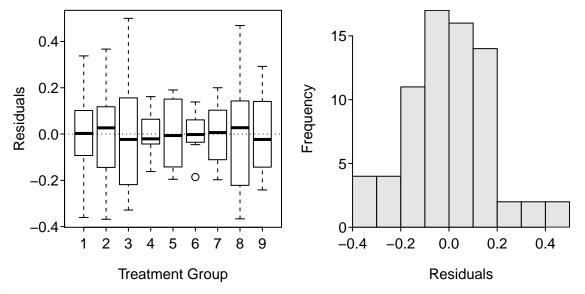
```
rstudent unadjusted p-value Bonferonni p
20 6.624666     9.5554e-09   6.8799e-07
> ## transChooser(lm2) # interactive, result not shown
> ben$logab <- log(ben$abundance)</pre>
```

```
> 1m3 <- lm(logab~site,data=ben)
```

> levenesTest(lm3)

Levene's Test for Homogeneity of Variance (center = median)
 Df F value Pr(>F)
group 8 1.5339 0.1636
 63

> residPlot(lm3)



> adTest(lm3\$residuals)

Anderson-Darling normality test with x A = 0.3323, p-value = 0.5062

> outlierTest(lm3)

No Studentized residuals with Bonferonni p < 0.05 Largest |rstudent|: rstudent unadjusted p-value Bonferonni p 20 2.928889 0.004754 0.34229

## Model Summarization

### > anova(lm3)

Analysis of Variance Table

Response: logab

site

Df Sum Sq Mean Sq F value Pr(>F) 8 8.6683 1.08353 29.066 < 2.2e-16

Residuals 63 2.3485 0.03728

- > ben.mc <- glht(lm3, mcp(site = "Dunnett"))</pre>
- > summary(ben.mc)

#### Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Dunnett Contrasts

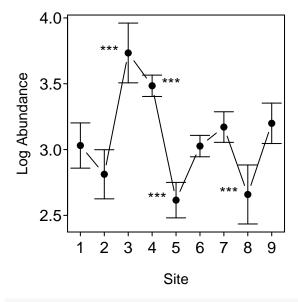
Fit: lm(formula = logab ~ site, data = ben)

### Linear Hypotheses:

```
Estimate Std. Error t value Pr(>|t|)
2 - 1 == 0 -0.218435
                       0.096537 -2.263 0.14556
3 - 1 == 0 \quad 0.703189
                       0.096537
                                  7.284 < 0.001
4 - 1 == 0 \quad 0.453836
                       0.096537
                                  4.701 < 0.001
5 - 1 == 0 -0.414859
                       0.096537
                                 -4.297 < 0.001
6 - 1 == 0 -0.004238
                       0.096537
                                  -0.044 1.00000
7 - 1 == 0 \quad 0.140280
                       0.096537
                                  1.453 0.57984
8 - 1 == 0 -0.371867
                       0.096537
                                 -3.852 0.00199
9 - 1 == 0 0.168668
                       0.096537
                                   1.747 0.37962
```

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

```
> fitPlot(lm3,ylab="Log Abundance",xlab="Site",main="")
> addSigLetters(lm3,lets=c("","","***","***","***","","***",""),pos=c(2,4,2,4,2,2,4,2,4))
```



### > confint(ben.mc)

Simultaneous Confidence Intervals

Multiple Comparisons of Means: Dunnett Contrasts

Fit: lm(formula = logab ~ site, data = ben)

Quantile = 2.7297

95% family-wise confidence level

### Linear Hypotheses:

```
Estimate lwr upr

2 - 1 == 0 -0.218435 -0.481957 0.045088

3 - 1 == 0 0.703189 0.439667 0.966711

4 - 1 == 0 0.453836 0.190314 0.717358

5 - 1 == 0 -0.414859 -0.678381 -0.151337

6 - 1 == 0 -0.004238 -0.267760 0.259284

7 - 1 == 0 0.140280 -0.123242 0.403803

8 - 1 == 0 -0.371867 -0.635389 -0.108345

9 - 1 == 0 0.168668 -0.094854 0.432190
```

## > exp(confint(ben.mc)\$confint)

```
Estimate lwr upr
2 - 1 0.8037761 0.6175641 1.0461360
3 - 1 2.0201841 1.5521649 2.6293235
4 - 1 1.5743404 1.2096105 2.0490460
5 - 1 0.6604332 0.5074296 0.8595714
6 - 1 0.9957713 0.7650794 1.2960229
7 - 1 1.1505965 0.8840360 1.4975320
8 - 1 0.6894457 0.5297208 0.8973320
9 - 1 1.1837272 0.9094913 1.5406525
attr(,"conf.level")
[1] 0.95
attr(,"calpha")
[1] 2.729909
```