Two-Way ANOVA analysis and interpretation

General Setup

Data in columns, one column for response, two columns for treatment levels.

library(car)

Fit the linear model e.g.

m1 <- lm(response~factor1*factor2, data=data)</pre>

Checking the assumptions

1. Equal variances:

plot(m1, which=1)

If spread of residual values changes along the x axis (for example a triangle of values) then consider transformation. Check equality formally:

leveneTest(response~factor1*factor2, data=data)

2. Normality of residuals

plot(m1, which=2)

If residuals do not fall close to line then consider transformation

Check normality formally

shapiro.test(m1\$residuals)

Interpreting the results

If plots and tests are OK, then proceed:

Anova (m1, type=3) Look at significance of each line in the ANOVA table, starting with the interaction Interaction (factor1:factor2) Interaction not significant (p > 0.05), but No significant results (all effects significant (p < 0.05) p > 0.05) one or more main effects significant To visualise differences. One main effect Both main effects Stop use Interaction plot: significant significant interaction.plot(dat a\$factor1, data\$factor2, To visualise differences, use data\$response, To visualise differences, use Interaction type="b", pch=19) boxplot **or** beeswarm plot: (response~factor, interaction.plot(data\$factor1, data=data). Factor is the main data\$factor2, data\$response, effect that is significant type="b", pch=19) Use TukeyHSD (aov (m1), which=c("factor1:fact or2"))

to look at mean for each combination of treatments

Examine output to determine where significant differences lie between all combinations, but really only need to interpret those which are biologically interesting

For figure, use boxplot or beeswarm for all combinations of factor levels e.g.

boxplot(response~factor1* factor2, data=data)

Use

TukeyHSD (aov (m1), which=c(factor)), specify the factor that is significant

Examine output to determine where significant differences lie between levels of significant factor

For figure, use boxplot or beeswarm for the significant factor e.g.

boxplot (response~factor1, data=data)

Use

TukeyHSD (aov (m1), which=c(factor1, factor2)) to look at both factors

Examine output to determine where significant differences lie between levels of each factor separately

For figure, use boxplot or beeswarm for both of the significant factors e.g.

boxplot(response~factor1*factor2, data=data)