

# ch19output\_beetles

February 18, 2020

```
#####
##### Example: Beetles #####
beetles<- read.table(file="/Desktop/jenn/teaching/stat445545/data/beetle.dat",header = TRUE)
#or you can read from Professor Erik Erhardt's website
#beetles <- read.table("https://statacumen.com/teach/ADA2/notes/ADA2_notes_Ch05_beetles.dat"
#                      , header = TRUE)
beetles

##   dose insecticide  t1   t2   t3   t4
## 1     1             A 0.31 0.45 0.46 0.43
## 2     1             B 0.82 1.10 0.88 0.72
## 3     1             C 0.43 0.45 0.63 0.76
## 4     1             D 0.45 0.71 0.66 0.62
## 5     2             A 0.36 0.29 0.40 0.23
## 6     2             B 0.92 0.61 0.49 1.24
## 7     2             C 0.44 0.35 0.31 0.40
## 8     2             D 0.56 1.02 0.71 0.38
## 9     3             A 0.22 0.21 0.18 0.23
## 10    3             B 0.30 0.37 0.38 0.29
## 11    3             C 0.23 0.25 0.24 0.22
## 12    3             D 0.30 0.36 0.31 0.33

# make dose a factor variable and label the levels
beetles$dose <- factor(beetles$dose, labels = c("low","medium","high"))
beetles$dose

## [1] low    low    low    low    medium medium medium medium high   high
## [11] high   high
## Levels: low medium high

library(reshape2)
beetles.long <- melt(beetles
                      , id.vars      = c("dose", "insecticide")
                      , variable.name = "number"
                      , value.name    = "hours10"
)
str(beetles.long)

## 'data.frame': 48 obs. of 4 variables:
## $ dose       : Factor w/ 3 levels "low","medium",...: 1 1 1 1 2 2 2 2 3 3 ...
## $ insecticide: Factor w/ 4 levels "A","B","C","D": 1 2 3 4 1 2 3 4 1 2 ...
## $ number     : Factor w/ 4 levels "t1","t2","t3",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ hours10    : num  0.31 0.82 0.43 0.45 0.36 0.92 0.44 0.56 0.22 0.3 ...

head(beetles.long)

##   dose insecticide number hours10
## 1   low         A     t1    0.31
## 2   low         B     t1    0.82
## 3   low         C     t1    0.43
## 4   low         D     t1    0.45
## 5 medium       A     t1    0.36
```

```

## 6 medium           B     t1    0.92
library(plyr)
# Calculate the cell means for each (dose, insecticide) combination
mean(beetles.long[, "hours10"])

## [1] 0.479375

beetles.mean <- ddply(beetles.long, .(), summarise, m = mean(hours10))
beetles.mean

##      .id      m
## 1 <NA> 0.479375

beetles.mean.d <- ddply(beetles.long, .(dose), summarise, m = mean(hours10))
beetles.mean.d

##      dose      m
## 1   low 0.617500
## 2 medium 0.544375
## 3   high 0.276250

beetles.mean.i <- ddply(beetles.long, .(insecticide), summarise, m = mean(hours10))
beetles.mean.i

##      insecticide      m
## 1             A 0.3141667
## 2             B 0.6766667
## 3             C 0.3925000
## 4             D 0.5341667

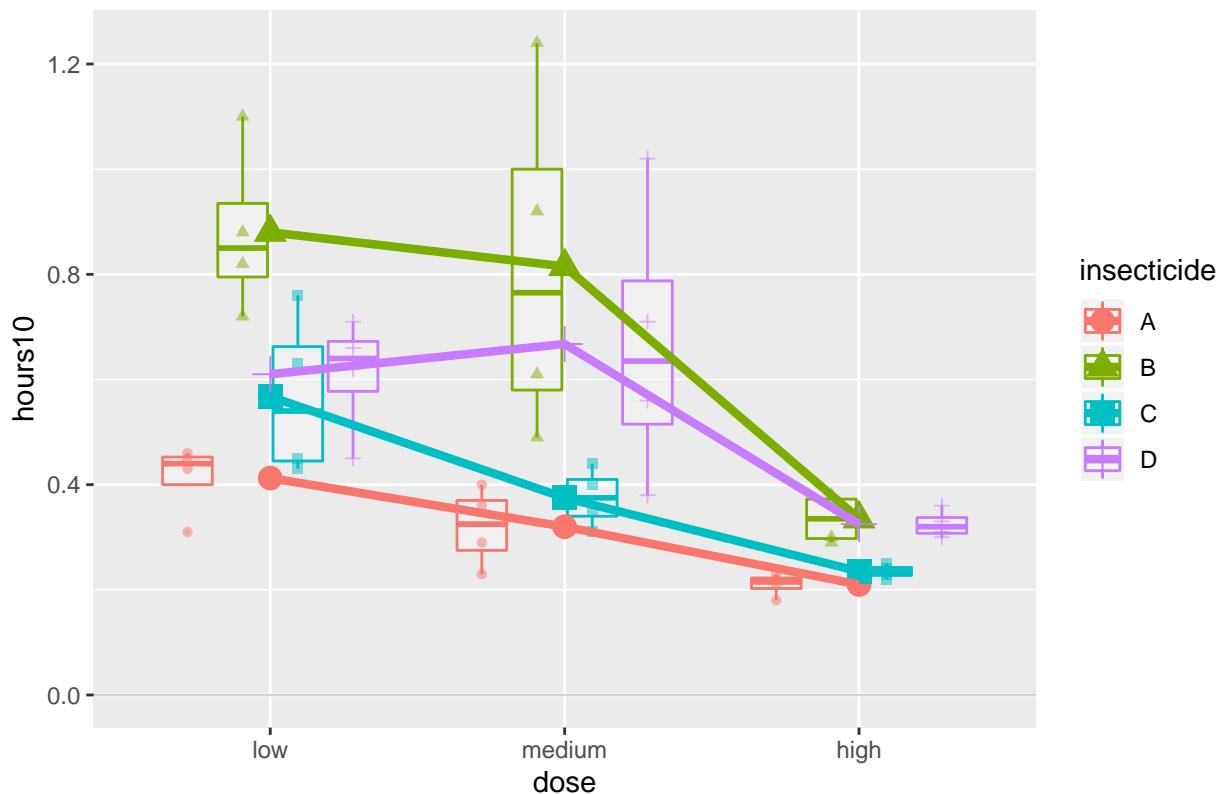
beetles.mean.di <- ddply(beetles.long, .(dose,insecticide), summarise, m = mean(hours10))
beetles.mean.di

##      dose insecticide      m
## 1   low         A 0.4125
## 2   low         B 0.8800
## 3   low         C 0.5675
## 4   low         D 0.6100
## 5 medium        A 0.3200
## 6 medium        B 0.8150
## 7 medium        C 0.3750
## 8 medium        D 0.6675
## 9   high        A 0.2100
## 10  high        B 0.3350
## 11  high        C 0.2350
## 12  high        D 0.3250

# Interaction plots, ggplot
library(ggplot2)
p <- ggplot(beetles.long, aes(x = dose, y = hours10, colour = insecticide, shape = insecticide))
p <- p + geom_hline(aes(yintercept = 0), colour = "black",
                     , linetype = "solid", size = 0.2, alpha = 0.3)
p <- p + geom_boxplot(alpha = 0.25, outlier.size=0.1)
p <- p + geom_point(alpha = 0.5, position=position_dodge(width=0.75))
p <- p + geom_point(data = beetles.mean.di, aes(y = m), size = 4)
p <- p + geom_line(data = beetles.mean.di, aes(y = m, group = insecticide), size = 1.5)
p <- p + labs(title = "Beetles interaction plot, insecticide by dose")
print(p)

```

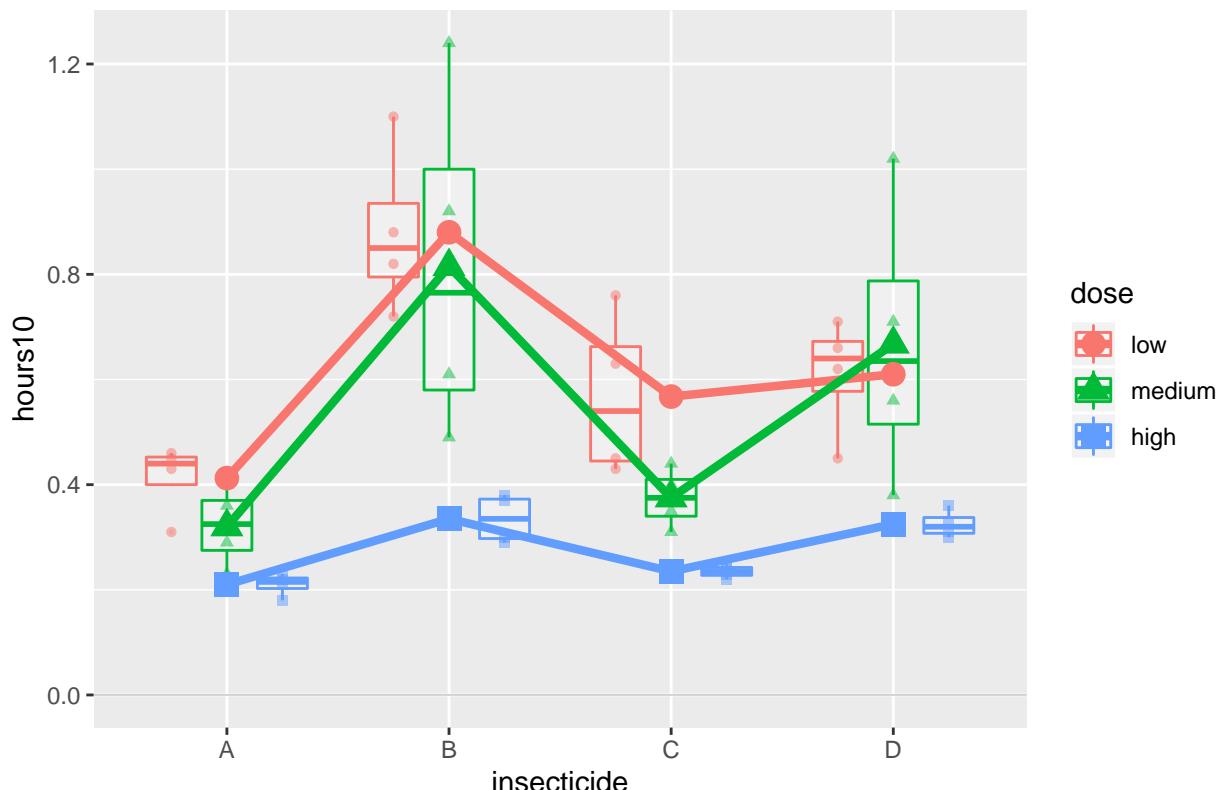
## Beetles interaction plot, insecticide by dose



```
#dev.copy(jpeg,filename "~/Desktop/jenn/teaching/ADA2/lecture notes/plots/ch05plot3.jpg")
#dev.off()
```

```
p <- ggplot(beetles.long, aes(x = insecticide, y = hours10, colour = dose, shape = dose))
p <- p + geom_hline(aes(yintercept = 0), colour = "black",
                     , linetype = "solid", size = 0.2, alpha = 0.3)
p <- p + geom_boxplot(alpha = 0.25, outlier.size=0.1)
p <- p + geom_point(alpha = 0.5, position=position_dodge(width=0.75))
p <- p + geom_point(data = beetles.mean.di, aes(y = m), size = 4)
p <- p + geom_line(data = beetles.mean.di, aes(y = m, group = dose), size = 1.5)
p <- p + labs(title = "Beetles interaction plot, dose by insecticide")
print(p)
```

## Beetles interaction plot, dose by insecticide



```
dev.copy(jpeg,filename "~/Desktop/jenn/teaching/ADA2/lecture notes/plots/ch05plot4.jpg")
```

```
## jpeg
## 3
```

```
dev.off()
```

```
## pdf
## 2
```

```
##fit ANOVA model
myfit <- aov(hours10~ dose*insecticide, data=beetles.long)
summary(myfit) #ANOVA table
```

```
##          Df Sum Sq Mean Sq F value    Pr(>F)
## dose        2 1.0330  0.5165 23.222 3.33e-07 ***
## insecticide  3 0.9212  0.3071 13.806 3.78e-06 ***
## dose:insecticide 6 0.2501  0.0417   1.874    0.112
## Residuals   36 0.8007  0.0222
## ---
```

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

```
##notice that interaction effect is not significant (p value =0.112)
```

```
#report row, column and cell means
```

```
print(model.tables(myfit,"means"),digits=3) #you can also find cell mean, marginal mean etc using myfi
```

```
## Tables of means
```

```
## Grand mean
```

```
##
```

```
## 0.479375
```

```

##  

## dose  

## dose  

##    low medium   high  

##  0.618  0.544  0.276  

##  

## insecticide  

## insecticide  

##      A      B      C      D  

## 0.314  0.677  0.393  0.534  

##  

## dose:insecticide  

##          insecticide  

## dose      A      B      C      D  

##  low     0.413  0.880  0.568  0.610  

##  medium  0.320  0.815  0.375  0.668  

##  high    0.210  0.335  0.235  0.325  

#### checking assumptions

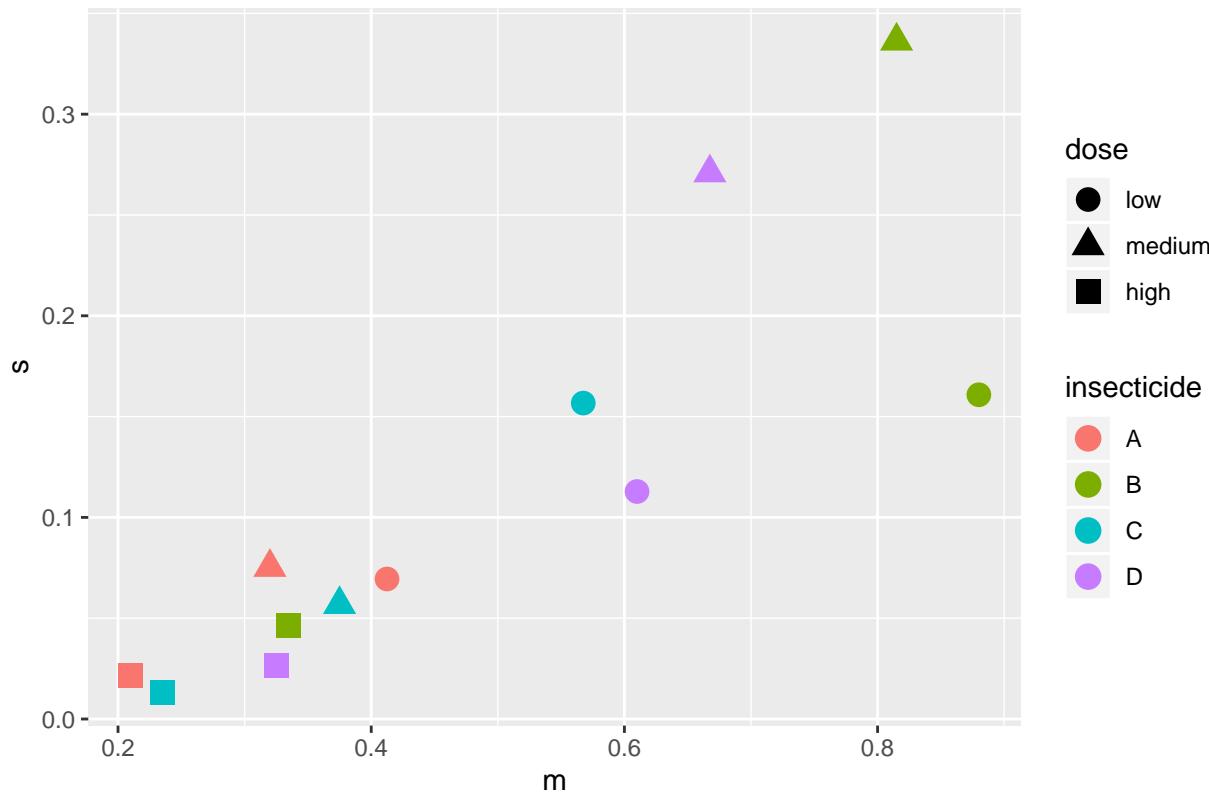
# mean vs sd plot
library(plyr)
# means and standard deviations for each dose/interaction cell
beetles.meansd.di <- ddply(beetles.long, .(dose,insecticide), summarise
                           , m = mean(hours10), s = sd(hours10))
beetles.meansd.di

##      dose insecticide      m      s
## 1    low           A 0.4125 0.06946222
## 2    low           B 0.8800 0.16083117
## 3    low           C 0.5675 0.15671099
## 4    low           D 0.6100 0.11284207
## 5  medium          A 0.3200 0.07527727
## 6  medium          B 0.8150 0.33630343
## 7  medium          C 0.3750 0.05686241
## 8  medium          D 0.6675 0.27097048
## 9   high           A 0.2100 0.02160247
## 10  high           B 0.3350 0.04654747
## 11  high           C 0.2350 0.01290994
## 12  high           D 0.3250 0.02645751

p <- ggplot(beetles.meansd.di, aes(x = m, y = s, shape = dose, colour = insecticide))
p <- p + geom_point(size=4)
p <- p + labs(title = "Beetles standard deviation vs mean")
print(p)

```

## Beetles standard deviation vs mean



```
dev.copy(jpeg,filename "~/Desktop/jenn/teaching/stat445545/notes/plots/ch05sdvsmean.jpg")
```

```
## jpeg
## 3
dev.off()

## pdf
## 2
# plot diagnostics
par(mfrow=c(2,2))

#residual vs fitted value
plot(myfit$fitted, myfit$residuals, main="Residuals vs Fitted Values")
abline(h = 0, col = "gray75")
```

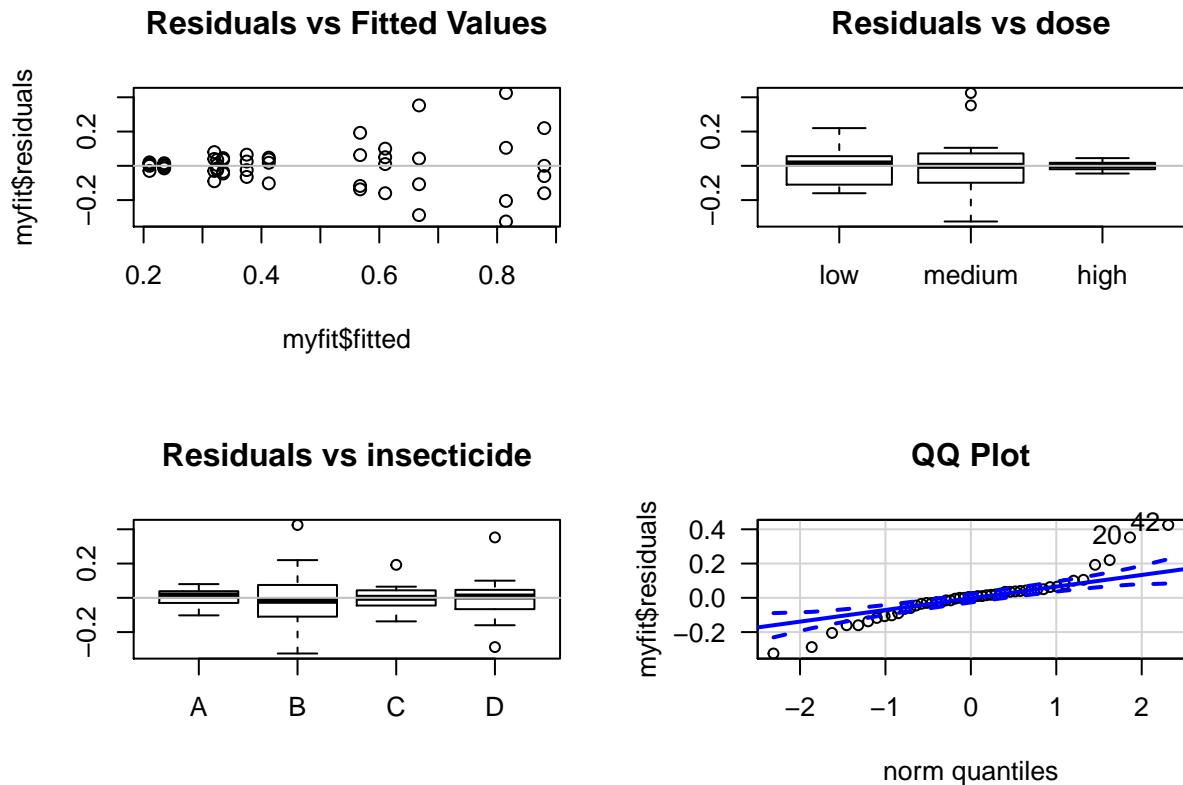
```
plot(beetles.long$dose, myfit$residuals, main="Residuals vs dose")
# horizontal line at zero
abline(h = 0, col = "gray75")
```

```
plot(beetles.long$insecticide, myfit$residuals, main="Residuals vs insecticide")
# horizontal line at zero
abline(h = 0, col = "gray75")
```

```
# Normality of Residuals
library(car)
```

```
## Loading required package: carData
```

```
qqPlot(myfit$residuals, las = 1, main="QQ Plot")
```



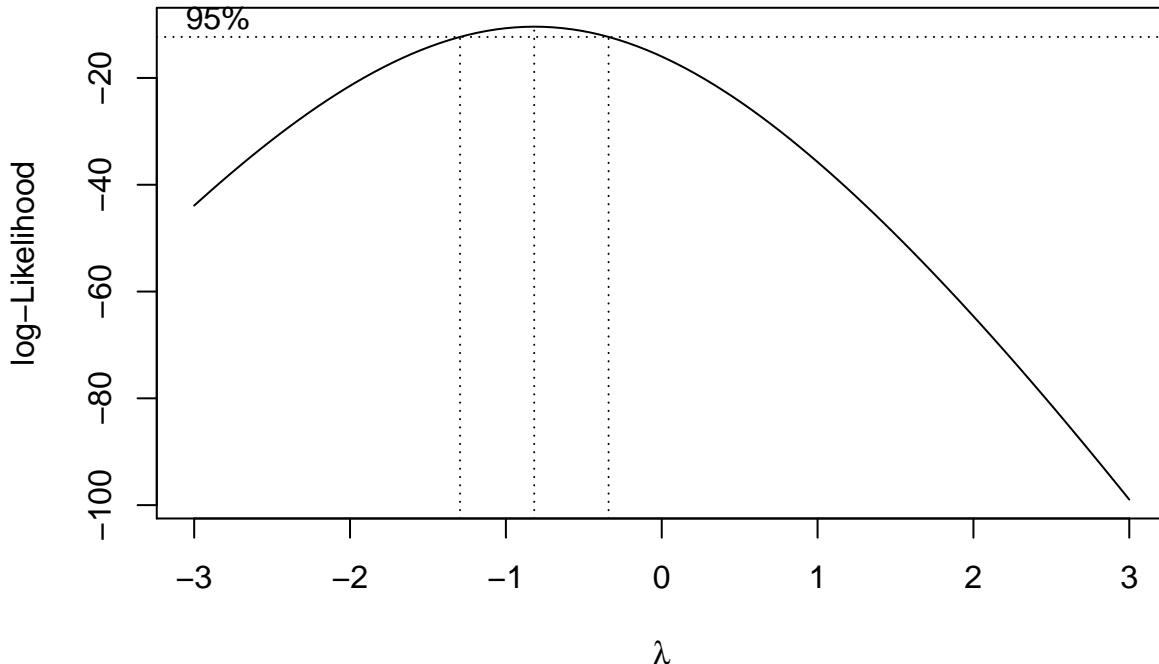
```
## [1] 42 20
```

```
dev.copy(jpeg,filename=~"/Desktop/jenn/teaching/stat445545/notes/plots/ch05diagfull.jpg")
```

```
## jpeg  
## 3
```

```
dev.off()
```

```
## pdf  
## 2  
#transformation  
par(mfrow=c(1,1))  
library(MASS)  
boxcox(myfit, lambda = seq(-3, 3, length = 10), plotit = TRUE)
```



```

dev.copy(jpeg,filename=~"/Desktop/jenn/teaching/stat445545/notes/plots/ch05boxcox.jpg")

## jpeg
## 3
dev.off()

## pdf
## 2

##transformations on $y$
myfit2 <- aov(1/hours10 ~ dose*insecticide, data = beetles.long)
summary(myfit2)

##                               Df Sum Sq Mean Sq F value    Pr(>F)
## dose                      2  34.88  17.439   72.64 2.31e-13 ***
## insecticide                3  20.41   6.805   28.34 1.38e-09 ***
## dose:insecticide          6   1.57   0.262     1.09    0.387
## Residuals                 36   8.64   0.240
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

#plot diagnostics
par(mfrow=c(2,3))

#residual vs fitted value
plot(myfit2$fitted, myfit2$residuals, main="Residuals vs Fitted Values for transformed data")
abline(h = 0, col = "gray75")

plot(myfit2$fitted, rstudent(myfit2), main="Studentized Deleted Residuals vs Fitted Values for transformed data")
abline(h = 0, col = "gray75")

plot(beetles.long$dose, myfit2$residuals, main="Residuals vs dose")
# horizontal line at zero
abline(h = 0, col = "gray75")

```

```

plot(beetles.long$insecticide, myfit2$residuals, main="Residuals vs insecticide")
# horizontal line at zero
abline(h = 0, col = "gray75")

# Normality of Residuals
library(car)
qqPlot(myfit2$residuals, las = 1, main="QQ Plot")

## [1] 41 44
dev.copy(jpeg,filename=~"/Desktop/jenn/teaching/stat445545/notes/plots/ch05diagtrans.jpg")

## jpeg
##   3
dev.off()

## pdf
##   2
##refit model without interaction term

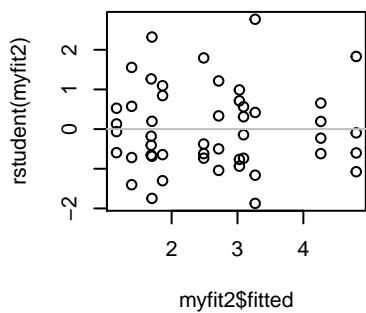
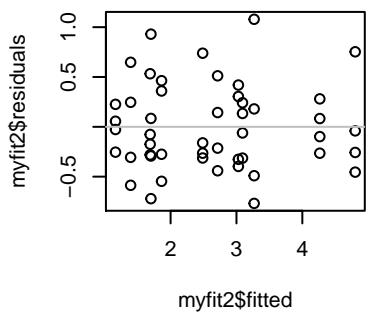
myfit3 <- aov(1/hours10 ~ dose+insecticide, data = beetles.long)
summary(myfit3)

##           Df Sum Sq Mean Sq F value    Pr(>F)
## dose        2  34.88  17.439   71.71 2.86e-14 ***
## insecticide 3   20.41    6.805   27.98 4.19e-10 ***
## Residuals   42   10.21    0.243
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

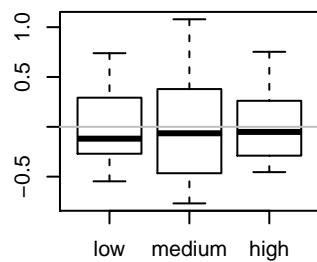
#plot diagnostics
par(mfrow=c(1,2))

```

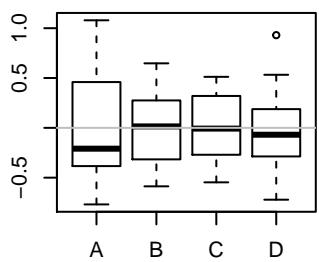
**Residuals vs Fitted Values for transformed Residuals vs Fitted Values**



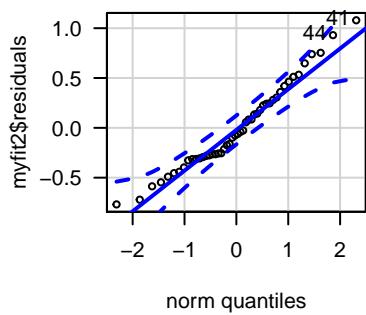
**Residuals vs dose**



**Residuals vs insecticide**



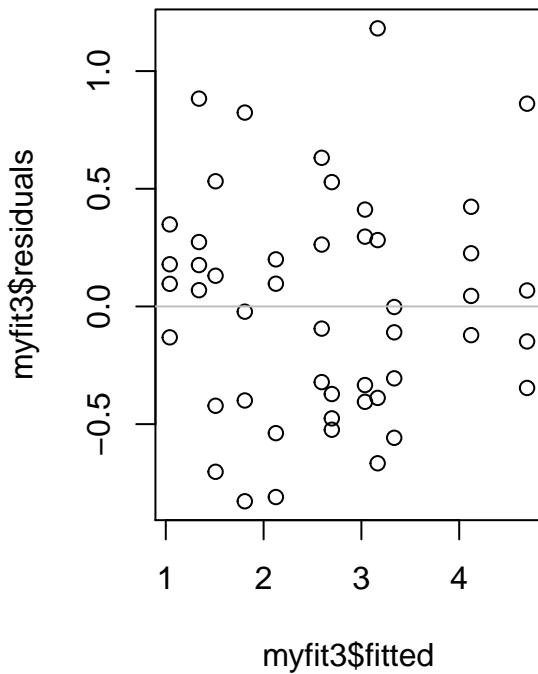
**QQ Plot**



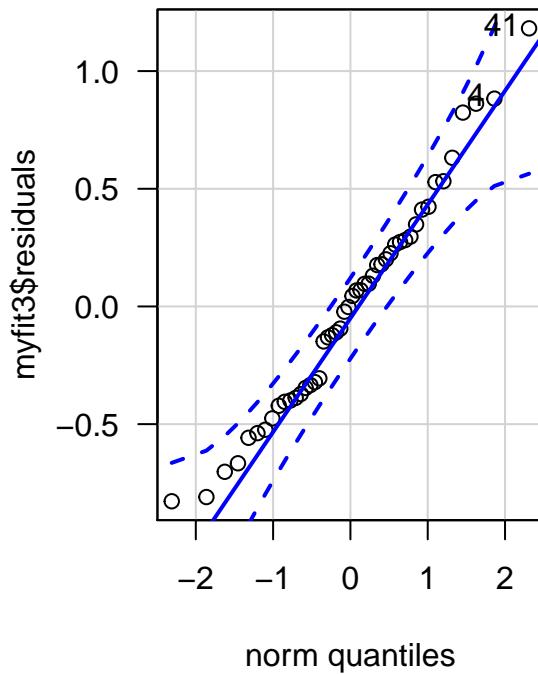
```
#residual vs fitted value
plot(myfit3$fitted, myfit3$residuals, main="Residuals vs Fitted Values for transformed reduced model")
abline(h = 0, col = "gray75")

# Normality of Residuals
library(car)
qqPlot(myfit3$residuals, las = 1, main="QQ Plot")
```

## s vs Fitted Values for transformed re



## QQ Plot



```

## [1] 41  4
dev.copy(jpeg,filename="~/Desktop/jenn/teaching/stat445545/notes/plots/ch05diagreduce.jpg")

## jpeg
##   3
dev.off()

## pdf
##   2

# Testing multiple factors is of interest here.
#multiple comparisons
#Tukey's multiple comparison for factor Dose
beetleT<-TukeyHSD(myfit3,"dose",conf.level=.95)
beetleT

## Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = 1/hours10 ~ dose + insecticide, data = beetles.long)
## 
## $dose
##             diff      lwr      upr     p adj
## medium-low  0.4686413 0.04505584 0.8922267 0.0271587
## high-low    1.9964249 1.57283950 2.4200103 0.0000000
## high-medium 1.5277837 1.10419824 1.9513691 0.0000000
plot(beetleT, sub="Tukey Honest Significant Differences")

#Tukey's multiple comparison for factor Insecticide

```

```

beetleT2<-TukeyHSD(myfit3,"insecticide",conf.level=.95)
beetleT2

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = 1/hours10 ~ dose + insecticide, data = beetles.long)
##
## $insecticide
##      diff      lwr      upr      p adj
## B-A -1.6574024 -2.1959343 -1.11887050 0.0000000
## C-A -0.5721354 -1.1106673 -0.03360355 0.0335202
## D-A -1.3583383 -1.8968702 -0.81980640 0.0000002
## C-B  1.0852669  0.5467351  1.62379883 0.0000172
## D-B  0.2990641 -0.2394678  0.83759598 0.4550931
## D-C -0.7862029 -1.3247347 -0.24767096 0.0018399

plot(beetleT2, sub="Tukey Honest Significant Differences")

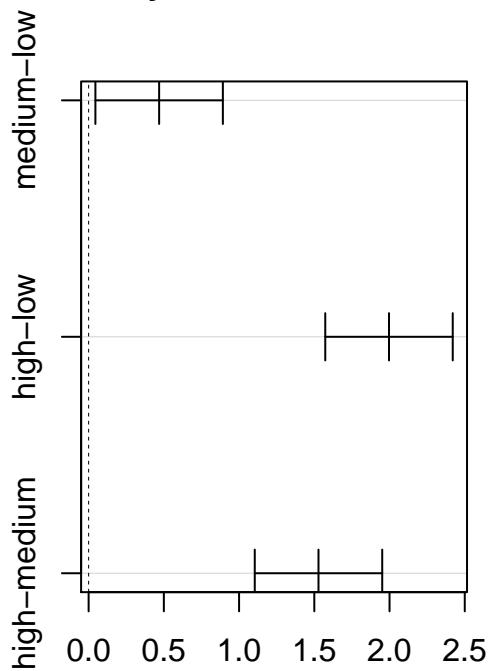
#or you can use the default to derive all the comparisons, however, if you
want to do simultaneous comparisons for both factors Dose and Insecticide,
confidence coefficient is less than 0.95, at least 90% by using Bonferroni inequality
beetleT3<-TukeyHSD(myfit3,conf.level=.95)
beetleT3

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = 1/hours10 ~ dose + insecticide, data = beetles.long)
##
## $dose
##      diff      lwr      upr      p adj
## medium-low  0.4686413 0.04505584 0.8922267 0.0271587
## high-low    1.9964249 1.57283950 2.4200103 0.0000000
## high-medium 1.5277837 1.10419824 1.9513691 0.0000000
##
## $insecticide
##      diff      lwr      upr      p adj
## B-A -1.6574024 -2.1959343 -1.11887050 0.0000000
## C-A -0.5721354 -1.1106673 -0.03360355 0.0335202
## D-A -1.3583383 -1.8968702 -0.81980640 0.0000002
## C-B  1.0852669  0.5467351  1.62379883 0.0000172
## D-B  0.2990641 -0.2394678  0.83759598 0.4550931
## D-C -0.7862029 -1.3247347 -0.24767096 0.0018399

plot(beetleT3, sub="Tukey Honest Significant Differences")

```

### 95% family-wise confidence level



Differences in mean levels of dose  
Tukey Honest Significant Differences

```
beetles.mean.d2 <- ddply(beetles.long, .(dose), summarise, m = mean(1/hours10))
beetles.mean.d2

##      dose      m
## 1    low 1.800688
## 2   medium 2.269329
## 3    high 3.797112

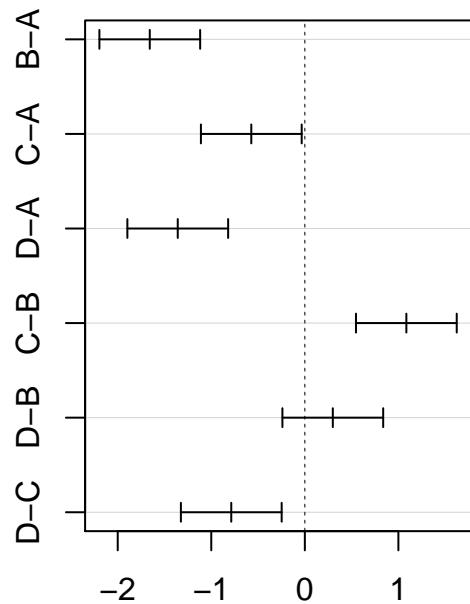
beetles.mean.i2 <- ddply(beetles.long, .(insecticide), summarise, m = mean(1/hours10))
beetles.mean.i2

##      insecticide      m
## 1             A 3.519345
## 2             B 1.861943
## 3             C 2.947210
## 4             D 2.161007

beetles.mean.di2 <- ddply(beetles.long, .(dose,insecticide), summarise, m = mean(1/hours10))
beetles.mean.di2

##      dose insecticide      m
## 1    low         A 2.486881
## 2    low         B 1.163464
## 3    low         C 1.862724
## 4    low         D 1.689682
## 5   medium        A 3.268470
## 6   medium        B 1.393392
## 7   medium        C 2.713919
## 8   medium        D 1.701534
## 9    high         A 4.802685
```

### 95% family-wise confidence level



Differences in mean levels of insecticide  
Tukey Honest Significant Differences

```

## 10   high          B 3.028973
## 11   high          C 4.264987
## 12   high          D 3.091805

#with family confidence coefficient at least 90%, we conclude that
#(1)the population mean dying rate for the high dose
#(averaged over insecticides) is significantly faster than the population mean dying rate
#for the low and medium doses (averaged over insecticides).
#(2) the mean dying rate of the two lower doses are slightly significantly different from each other.
#(3) the population mean dying rate for the Insecticide A
# (averaged over doses) is slightly significantly different from the population mean dying rate for
#the C insecticide #(averaged over doses).
#(4) Both insectides A and C has significant faster dying rate than those of insectides B and D,
#regardless of the doses.
#(5) The population mean dying rate for the Insecticide B
#(averaged over doses) is not significantly different from the population mean dying rate
#for the D insecticide (averaged over doses).

```