

## R Workshop Module 4: Some Statistical Analyses

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### Reading in Data

Start by reading the dataset 'practicedata.csv' into R. (Having trouble? See the instructions in Module 1.)

### Two-Sample *t*-tests

```
## Find group means for response variable
aggregate((practicedata$respvar)~practicedata$groupvar*practicedata$groupvar2,FUN=mean)

##   practicedata$groupvar practicedata$groupvar2 (practicedata$respvar)
## 1           Control          A          39.92996
## 2           Treatment          A          46.32736
## 3           Control          B          37.24586
## 4           Treatment          B          45.49279

## Find group standard deviations for response variable
aggregate((practicedata$respvar)~practicedata$groupvar*practicedata$groupvar2,FUN=sd)

##   practicedata$groupvar practicedata$groupvar2 (practicedata$respvar)
## 1           Control          A           8.523437
## 2           Treatment          A           7.430197
## 3           Control          B           7.160007
## 4           Treatment          B           8.608150

## Perform a t-test
t.test(respvar~groupvar,data=practicedata,
       var.equal=TRUE)

##
## Two Sample t-test
##
## data:  respvar by groupvar
## t = -4.6117, df = 98, p-value = 1.207e-05
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -10.472948 -4.171384
## sample estimates:
## mean in group Control mean in group Treatment
##           38.58791           45.91007

t.test(respvar~groupvar2,data=practicedata,
       var.equal=TRUE)
```

```
##
## Two Sample t-test
##
## data: respvar by groupvar2
## t = 1.0097, df = 98, p-value = 0.3151
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.698629  5.217300
## sample estimates:
## mean in group A mean in group B
##      43.12866      41.36932
```

## Analysis of Variance (ANOVA)

```
## Perform an ANOVA
anova.results=lm(respvar~groupvar+groupvar2+groupvar*groupvar2,data=practicedata)

anova.results

##
## Call:
## lm(formula = respvar ~ groupvar + groupvar2 + groupvar * groupvar2,
##     data = practicedata)
##
## Coefficients:
##             (Intercept)             groupvarTreatment
##                   39.930                   6.397
##             groupvar2B groupvarTreatment:groupvar2B
##                   -2.684                   1.850

anova(anova.results)

## Analysis of Variance Table
##
## Response: respvar
##           Df Sum Sq Mean Sq F value    Pr(>F)
## groupvar    1 1340.4  1340.35  21.1727 1.283e-05 ***
## groupvar2    1   77.4    77.38   1.2223  0.2717
## groupvar:groupvar2  1   21.4    21.38   0.3377  0.5625
## Residuals   96 6077.3    63.31
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

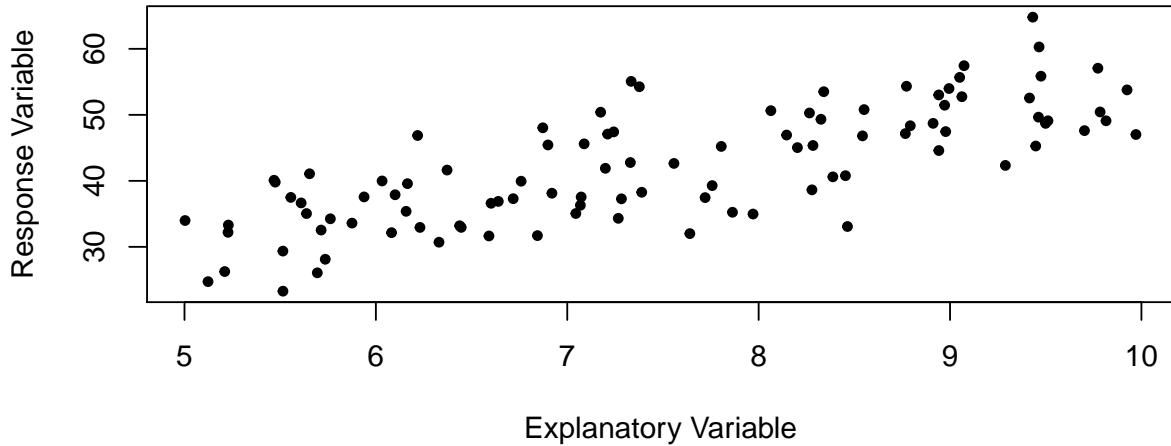
summary(anova.results)
```

```
##
## Call:
## lm(formula = respvar ~ groupvar + groupvar2 + groupvar * groupvar2,
##     data = practicedata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.6440  -5.7905   0.0287   5.7551  19.3054
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)          39.930     1.591  25.093 < 2e-16 ***
## groupvarTreatment      6.397     2.250   2.843  0.00546 **
## groupvar2B            -2.684     2.250  -1.193  0.23593
## groupvarTreatment:groupvar2B  1.850     3.183   0.581  0.56251
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7.956 on 96 degrees of freedom
## Multiple R-squared:  0.1915, Adjusted R-squared:  0.1662
## F-statistic: 7.578 on 3 and 96 DF,  p-value: 0.0001333
```

## Linear Regression

For this example, we will investigate the relationship between the variables `respvar` and `expvar` from the data set, `practicedata`. Remember than by using a '\$', we can refer to the variable as `practicedata$respvar` in the following code. To fit a linear model, we use the function `lm()` as follows.

```
## Scatter plot of the data
plot(practicedata$expvar,practicedata$respvar,xlab='Explanatory Variable',
      ylab='Response Variable',pch=20)
```

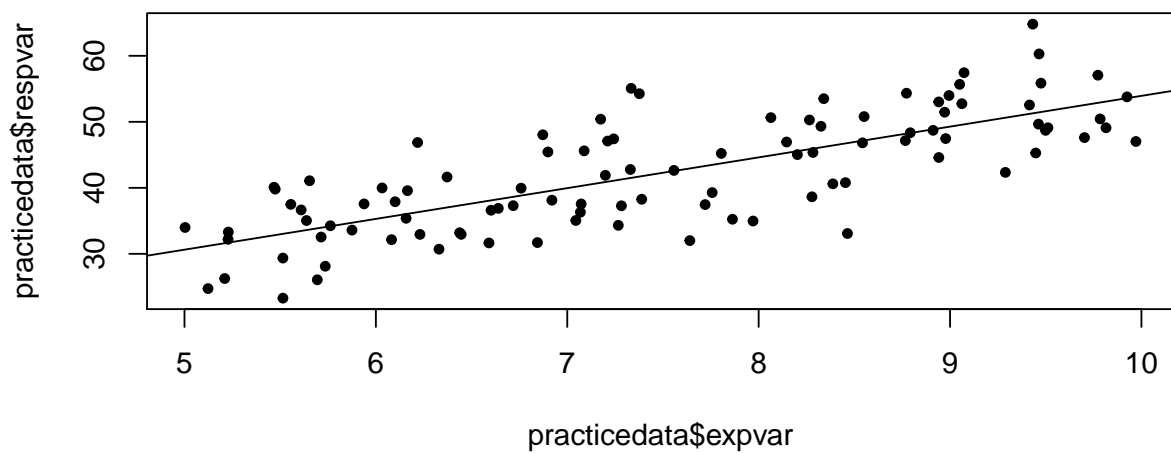


```
## Fit a linear model to the Data
fitted.model = lm(respvar~expvar,
                  data=practicedata) # Fit a linear model with
                                     # y-variable respvar and x-variable expvar
summary(fitted.model) #Summarize the linear model

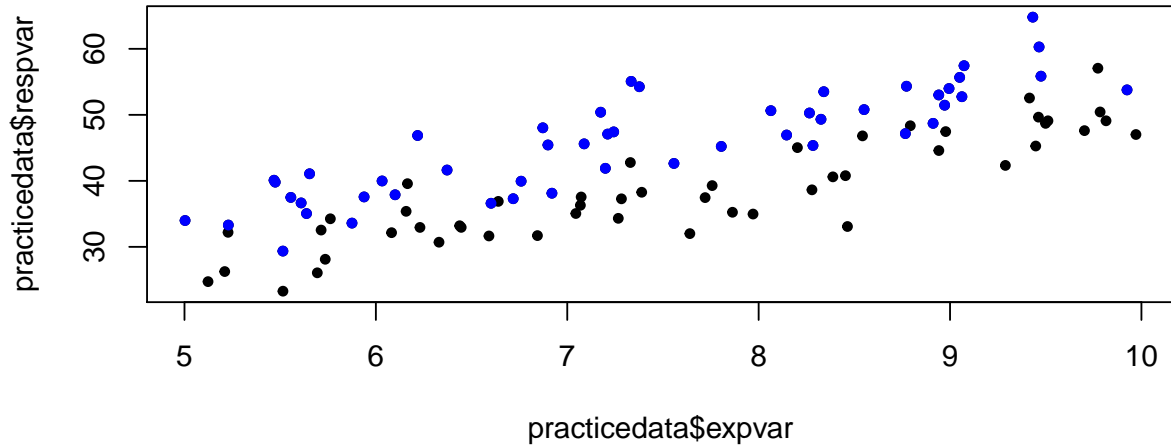
##
## Call:
## lm(formula = respvar ~ expvar, data = practicedata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13.6918  -4.1475  -0.2382   4.2102  13.5615
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   7.3299     3.0495   2.404  0.0181 *
## expvar        4.6604     0.3999  11.654 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.67 on 98 degrees of freedom
## Multiple R-squared:  0.5809, Adjusted R-squared:  0.5766
## F-statistic: 135.8 on 1 and 98 DF,  p-value: < 2.2e-16
```

```
## Check to make sure that the model looks appropriate
plot(fitted.model)
```

```
##Plot the data with the fitted regression line
plot(practicedata$expvar,practicedata$respvar,pch=20)
abline(fitted.model)
```



```
##Plotting the data and the fitted model
plot(practicedata$expvar,practicedata$respvar,pch=20)
points(practicedata$expvar[practicedata$groupvar=='Treatment'],
       practicedata$respvar[practicedata$groupvar=='Treatment'],
       col='blue',pch=20)
```



```
##You can also fit a linear model with more than one variable
fitted.group.model = lm(respvar~expvar + groupvar,
                        data=practicedata) # Fit a linear model with
                        # y-variable respvar and x-variables expvar and groupvar
summary(fitted.group.model) #Summarize the linear model

##
## Call:
## lm(formula = respvar ~ expvar + groupvar, data = practicedata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.7226 -2.4108 -0.1681  2.1640  9.4101
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.4628     2.1175   0.691   0.491
## expvar           4.8846     0.2694  18.129 <2e-16 ***
## groupvarTreatment  8.3739     0.7640  10.960 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.809 on 97 degrees of freedom
## Multiple R-squared:  0.8128, Adjusted R-squared:  0.8089
## F-statistic: 210.5 on 2 and 97 DF,  p-value: < 2.2e-16

##You can also fit a linear model with both variables and their interaction
fitted.int.model = lm(respvar~expvar + groupvar + expvar*groupvar,
```

```

data=practicedata) # Fit a linear model with
# y-variable respvar and
# x-variables expvar, groupvar and their interaction
summary(fitted.int.model) #Summarize the linear model

##
## Call:
## lm(formula = respvar ~ expvar + groupvar + expvar * groupvar,
##     data = practicedata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.6342 -2.4548 -0.0682  2.2381  9.4161
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)         2.2408     2.8744   0.780   0.438
## expvar              4.7822     0.3714  12.875 <2e-16 ***
## groupvarTreatment    6.7406     4.1313   1.632   0.106
## expvar:groupvarTreatment 0.2182     0.5423   0.402   0.688
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.826 on 96 degrees of freedom
## Multiple R-squared:  0.8131, Adjusted R-squared:  0.8072
## F-statistic: 139.2 on 3 and 96 DF,  p-value: < 2.2e-16

```