ANOVA as Regression (For Illustration)

We consider fitting the ANOVA model 4 different ways.

1. Model1: Fit the default "effects" model using the lm() function. This will be our typical approach! Look at parameter (coefficient) estimates, model matrix, ANOVA table and emmeans.
2. Model2: Fit the alternate "no intercept" or "means" model using the lm() function.
3. Model3: Fit the default model "by hand" by creating the 3 indicator variables. This model is overparameterized, for illustration only.
4. Model4: Fit the default model "by hand" using just 2 indicator variables. This is equivalent to Model1.
5. Model5: If we do not define trt (1, 2, 3) as.factor, then we fit a regression model instead of an ANOVA model. This woud not be appropriate, for illustration only.

```
library(emmeans)
InData <- read.csv("C:/hess/STAT512/RNotes/Intro and R/RegANOVA.csv")
str(InData)
```

```
## 'data.frame':    6 obs. of  5 variables:
##  $ trt: int  1 1 2 2 3 3
##  $ y  : num  6.3 5.9 4.3 4.8 3.7 3.9
##  $ x1 : int  1 1 0 0 0 0
##  $ x2 : int  0 0 1 1 0 0
##  $ x3 : int  0 0 0 0 1 1
```

```
#Important: Need to redefine trt as.factor!
InData$trt <- as.factor(InData$trt)
str(InData)
```

```
## 'data.frame':    6 obs. of  5 variables:
##  $ trt: Factor w/ 3 levels "1","2","3": 1 1 2 2 3 3
##  $ y  : num  6.3 5.9 4.3 4.8 3.7 3.9
##  $ x1 : int  1 1 0 0 0 0
##  $ x2 : int  0 0 1 1 0 0
##  $ x3 : int  0 0 0 0 1 1
```

```
aggregate(y ~ trt, FUN = mean, data = InData)
```

```
##   trt    y
## 1   1 6.10
## 2   2 4.55
## 3   3 3.80
```

Approach1: one-way ANOVA

This is the standard approach corresponding to the "Effects Model". Typical research questions are addressed using the ANOVA table and pairwise comparison of means. Note that the emmeans are the same as the simple means.

```
Model1 <- lm(y ~ trt, data = InData)
anova(Model1)
```

```
## Analysis of Variance Table
##
## Response: y
```

```
##           Df Sum Sq Mean Sq F value   Pr(>F)
## trt        2 5.5033  2.7517  36.689 0.007785 **
## Residuals  3 0.2250  0.0750
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```r
emmeans(Model1, pairwise ~ trt, adjust = "none")
```

```
## $emmeans
##  trt emmean        SE df lower.CL upper.CL
##  1     6.10 0.1936492  3 5.483722 6.716278
##  2     4.55 0.1936492  3 3.933722 5.166278
##  3     3.80 0.1936492  3 3.183722 4.416278
##
## Confidence level used: 0.95
##
## $contrasts
##  contrast estimate        SE df t.ratio p.value
##  1 - 2        1.55 0.2738613  3   5.660  0.0222
##  1 - 3        2.30 0.2738613  3   8.398  0.0073
##  2 - 3        0.75 0.2738613  3   2.739  0.1384
##
## P value adjustment: tukey method for comparing a family of 3 estimates
```

```r
model.matrix(Model1)
```

```
##   (Intercept) trt2 trt3
## 1           1    0    0
## 2           1    0    0
## 3           1    1    0
## 4           1    1    0
## 5           1    0    1
## 6           1    0    1
## attr(,"assign")
## [1] 0 1 1
## attr(,"contrasts")
## attr(,"contrasts")$trt
## [1] "contr.treatment"
```

```r
summary(Model1)
```

```
##
## Call:
## lm(formula = y ~ trt, data = InData)
##
## Residuals:
##     1     2     3     4     5     6
##  0.20 -0.20 -0.25  0.25 -0.10  0.10
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)   6.1000     0.1936  31.500 7.03e-05 ***
## trt2         -1.5500     0.2739  -5.660  0.01092 *
## trt3         -2.3000     0.2739  -8.398  0.00354 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 0.2739 on 3 degrees of freedom
## Multiple R-squared:  0.9607, Adjusted R-squared:  0.9345
## F-statistic: 36.69 on 2 and 3 DF,  p-value: 0.007785
```

Approach2: No intercept model

When we fit the model without the intercept, the parameters/coefficients correspond to the trt means! This is also called the "Means Model".

```
Model2 <- lm(y ~ trt - 1, data = InData)
model.matrix(Model2)
```

```
##   trt1 trt2 trt3
## 1    1    0    0
## 2    1    0    0
## 3    0    1    0
## 4    0    1    0
## 5    0    0    1
## 6    0    0    1
## attr(,"assign")
## [1] 1 1 1
## attr(,"contrasts")
## attr(,"contrasts")$trt
## [1] "contr.treatment"
```

```
summary(Model2)
```

```
##
## Call:
## lm(formula = y ~ trt - 1, data = InData)
##
## Residuals:
##     1     2     3     4     5     6
##  0.20 -0.20 -0.25  0.25 -0.10  0.10
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## trt1   6.1000     0.1936   31.50 7.03e-05 ***
## trt2   4.5500     0.1936   23.50 0.000169 ***
## trt3   3.8000     0.1936   19.62 0.000289 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2739 on 3 degrees of freedom
## Multiple R-squared:  0.9984, Adjusted R-squared:  0.9969
## F-statistic: 643.1 on 3 and 3 DF,  p-value: 0.0001038
```

```
anova(Model2)
```

```
## Analysis of Variance Table
##
## Response: y
##           Df  Sum Sq Mean Sq F value    Pr(>F)
## trt        3 144.705  48.235  643.13 0.0001038 ***
```

```
## Residuals  3   0.225   0.075
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Approach3: Regression with Indicator Variables

This model is *overparameterized*. That is why the NA values appear. For Illustration only!

```r
Model3 <- lm(y ~ x1 + x2 +x3, data = InData)
summary(Model3)
```

```
##
## Call:
## lm(formula = y ~ x1 + x2 + x3, data = InData)
##
## Residuals:
##     1     2     3     4     5     6
##  0.20 -0.20 -0.25  0.25 -0.10  0.10
##
## Coefficients: (1 not defined because of singularities)
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.8000     0.1936  19.623 0.000289 ***
## x1            2.3000     0.2739   8.398 0.003541 **
## x2            0.7500     0.2739   2.739 0.071422 .
## x3                NA         NA      NA       NA
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2739 on 3 degrees of freedom
## Multiple R-squared:  0.9607, Adjusted R-squared:  0.9345
## F-statistic: 36.69 on 2 and 3 DF,  p-value: 0.007785
```

Approach4: Regression with Indicator Variables

Note that x1 is not included in the model statement. This model is equivalent to the default one-way ANOVA model in R.

```r
Model4 <- lm(y ~ x2 + x3, data = InData)
model.matrix(Model4)
```

```
##   (Intercept) x2 x3
## 1           1  0  0
## 2           1  0  0
## 3           1  1  0
## 4           1  1  0
## 5           1  0  1
## 6           1  0  1
## attr(,"assign")
## [1] 0 1 2
```

```r
summary(Model4)
```

```
##
```

```
## Call:
## lm(formula = y ~ x2 + x3, data = InData)
##
## Residuals:
##     1     2     3     4     5     6
##  0.20 -0.20 -0.25  0.25 -0.10  0.10
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)   6.1000     0.1936  31.500 7.03e-05 ***
## x2           -1.5500     0.2739  -5.660  0.01092 *
## x3           -2.3000     0.2739  -8.398  0.00354 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2739 on 3 degrees of freedom
## Multiple R-squared:  0.9607, Adjusted R-squared:  0.9345
## F-statistic: 36.69 on 2 and 3 DF,  p-value: 0.007785
```

What happens if we don't define trt as a factor?

Since trt is coded as 1,2,3 it will be defined as a numerical variable by default. If we don't define trt as a factor, a regression model will be fit! This is NOT appropriate for this data.

```r
InData <- read.csv("C:/hess/STAT512/RNotes/Intro and R/RegANOVA.csv")
str(InData)
```

```
## 'data.frame':    6 obs. of  5 variables:
##  $ trt: int  1 1 2 2 3 3
##  $ y  : num  6.3 5.9 4.3 4.8 3.7 3.9
##  $ x1 : int  1 1 0 0 0 0
##  $ x2 : int  0 0 1 1 0 0
##  $ x3 : int  0 0 0 0 1 1
```

```r
Model5 <- lm(y ~ trt, data= InData)
summary(Model5)
```

```
##
## Call:
## lm(formula = y ~ trt, data = InData)
##
## Residuals:
##        1        2        3        4        5        6
##  0.33333 -0.06667 -0.51667 -0.01667  0.03333  0.23333
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)   7.1167     0.3576  19.904 3.76e-05 ***
## trt          -1.1500     0.1655  -6.948  0.00225 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.331 on 4 degrees of freedom
## Multiple R-squared:  0.9235, Adjusted R-squared:  0.9043
```

```
## F-statistic: 48.27 on 1 and 4 DF,  p-value: 0.002254
```

```r
model.matrix(Model5)
```

```
##   (Intercept) trt
## 1           1   1
## 2           1   1
## 3           1   2
## 4           1   2
## 5           1   3
## 6           1   3
## attr(,"assign")
## [1] 0 1
```