

ORIGIN = 0

## Linear Modeling & "Extra" Sums of Squares

Multiple Regression forms the basis for the extensive field called "Linear Modeling". The independent variables in any particular "linear model" can be many things, including squared  $X^2$  or higher-order terms  $X^3$  or  $X^4$  etc. as in "Polynomial Regression", crossed-factors  $X_1 \cdot X_2$  for interactions as seen in ANOVA models, or many kinds of transformed variables for curvilinear regression fits such as  $\ln(X)$ ,  $e^X$ , or the like. The only requirement is that there is only a single Dependent variable  $Y$ , and all Independent variables  $X_i$  enter into regression as "linear factors" meaning that they are only added together:  $Y = X_1 + X_2 + X_3 \dots$  - even though any Independent variable may be transformed or crossed composites of other variables.

In constructing one or more linear models, the researcher is often confronted with many possible independent variables from which to choose a "best set". Biological intuition should definitely play a role in the choice. The Partial t/F tests may be used to determine whether the addition of any particular independent variable  $X_i$  makes a difference to a model consisting of other  $X$ 's previously constructed. Statistical software typically provides output to assist in the choice of variables, although they differ in details. Commonly encountered are tables listing "extra" sums of squares for each independent variable with associated F statistics and P values. It is important to note that all "extra" sums of squares add to Sum of Squares Regression (SSR) for what is termed the "Full Model" - a *multiple regression including all independent variables*. Each "extra" sum of squares represents the decrease in SSR, or equivalently the increase in Sum of Squares Error (SSE), with subtraction of an independent variable from the multiple regression producing what is termed a "Reduced Model".

Of course, there are multiple ways to subtract an independent variable from a Full Model to produce a Reduced Model. Two of the most common are described here.

### Marginal "Extra" Sums of Squares:

In this instance, only a single independent variable is removed at a time from the Full Model (FM) to produce one or more Reduced Models (RM). Resultant comparisons between RM's and FM thus involves statistical hypotheses of the form:  $H_0: \beta_i = 0$  versus  $H_1: \beta_i \neq 0$  for each independent variable at a time, *with all other independent variables retained within the Reduced Model*. "Extra" Sums of Squares (and tests of this type) are called "Marginal" or "Type II" Regression Sums of Squares.

```
ZAR := READPRN("c:/DATA/Biostatistics/ZarEX20.1aR.txt")
```

< Zar Example 20.1a

### Prototype in R:

```
#MULTIPLE REGRESSION MODEL SS
#ZAR EXAMPLE 20.1a
ZAR=read.table("c:/DATA/Biostatistics/ZarEX20.1aR.txt")
ZAR
attach(ZAR)

Y=var5
X1=var1
X2=var2
X3=var3
X4=var4

#DOWNLOAD {CAR} PACKAGE FROM CRAN:
library(car)
#TYPE 2 SS:

FM=lm(Y~X1+X2+X3+X4)
```

**summary(FM)**  
**Anova(FM)**

**Partial t-statistics >**

Here the Anova() function in package {car} produces marginal sums of squares, F statistics, and P values that are identical to the Partial t/F tests shown in Lecture Worksheet 380 (although only t values are given by R's summary() function above). These Partial tests are "marginal" in the sense that only one independent variable is removed from FM at a time for each test and all other variables are left within the RM.

$$6.075^2 = 36.9056 < t^2 \text{ statistic for } X_1$$

^ note t<sup>2</sup> above = F here

**Partial F statistics >**

**#REDUCED MODELS:**

- RM1=lm(Y~X2+X3+X4) #subtracting X1**
- RM2=lm(Y~X1+X3+X4) #subtracting X2**
- RM3=lm(Y~X1+X2+X4) #subtracting X3**
- RM4=lm(Y~X1+X2+X3) #subtracting X4**

- Anova(RM1)**
- Anova(RM2)**
- Anova(RM3)**
- Anova(RM4)**

Note the *change* in Residual SS between FM above and Reduced Model RM<sub>1</sub> excluding independent variable X<sub>1</sub>. This difference is shown as the "Extra" SS in Anova(FM) above.

**> summary(FM)**

```
Call:
lm(formula = Y ~ X1 + X2 + X3 + X4)

Residuals:
    Min       1Q   Median       3Q      Max
-1.5070 -0.1112  0.0401  0.1550  0.9617

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  2.95828    1.36361   2.169  0.03869 *
X1          -0.12932    0.02129  -6.075 1.50e-06 ***
X2          -0.01878    0.05628  -0.334  0.74102
X3          -0.04621    0.20727  -0.223  0.82518
X4           0.20876    0.06703   3.114  0.00423 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.4238 on 28 degrees of freedom
Multiple R-squared:  0.6589,    Adjusted R-squared:  0.6102
F-statistic: 13.52 on 4 and 28 DF,  p-value: 2.948e-06
```

**> Anova(FM)**

```
Anova Table (Type II tests)

Response: Y
      Sum Sq Df F value    Pr(>F)
X1     6.6298  1  36.9063 1.496e-06 ***
X2     0.0200  1   0.1114  0.741024
X3     0.0089  1   0.0497  0.825178
X4     1.7421  1   9.6979  0.004227 **
Residuals 5.0299 28
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

**> Anova(RM1)**

```
Anova Table (Type II tests)

Response: Y
      Sum Sq Df F value    Pr(>F)
X2     1.3777  1   3.4267  0.07436 .
X3     0.5477  1   1.3622  0.25267
X4     2.3125  1   5.7516  0.02313 *
Residuals 11.6596 29
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## Serial "Extra" Sums of Squares:

"Serial" Sums of Squares, also known as "Type I" SS and associated Partial F-tests represent the *change* in SSR (or SSE) when independent variables are added or deleted in an ordered fashion. In R, the order is determined by the order given to independent variables in the formula within the `lm()` function from left to right. Thus, at each step of addition or deletion of an independent variable, the Full Model (FM) and (RM) changes in a serial fashion.

**TYPE 1 SS:**  
**`anova(FM)`**

**`FM1=lm(Y~X1+X2)`**  
**`RM1=lm(Y~X1)`**  
**`anova(FM1)`**  
**`anova(RM1)`**

^ Here serial addition of independent variables follows the order seen in the `lm()` function:  
 $X_1, X_2, X_3, X_4$ .

`anova(FM)` shows the change in SSR (or SSE) with the *addition* of  $X_1$ , then  $X_2$ , followed by  $X_3$ , and then  $X_4$ . Alternatively it shows the *subtraction* of  $X_4$  first followed by  $X_3$ ,  $X_2$  and then  $X_1$ .

**> `anova(FM)`**

Analysis of Variance Table

```
Response: Y
      Df Sum Sq Mean Sq F value    Pr(>F)
X1      1  7.8762   7.8762  43.8450 3.496e-07 ***
X2      1  0.0131   0.0131   0.0732 0.788785
X3      1  0.0859   0.0859   0.4781 0.494963
X4      1  1.7421   1.7421   9.6979 0.004227 **
Residuals 28  5.0299   0.1796
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

`anova(FM1)` shows an intermediate step in which only independent variables  $X_1$  &  $X_2$  have been added in order.

**> `anova(FM1)`**

Analysis of Variance Table

```
Response: Y
      Df Sum Sq Mean Sq F value    Pr(>F)
X1      1  7.8762   7.8762  34.4549 2.007e-06 ***
X2      1  0.0131   0.0131   0.0575  0.8121
Residuals 30  6.8579   0.2286
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

`anova(RM1)` shows an intermediate step in which only independent variable  $X_1$  has been added.

**> `anova(RM1)`**

Analysis of Variance Table

```
Response: Y
      Df Sum Sq Mean Sq F value    Pr(>F)
X1      1  7.8762   7.8762  35.535 1.371e-06 ***
Residuals 31  6.8710   0.2216
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```