# TYPE I, TYPE II, AND TYPE III SUMS OF SQUARES 

## David C. Howell <br> 1/12/2010

## The data file:

These are the data from Howell (2010) Table 16.5 except that two additional
observations were added to cell(2,4). The example in the book had the unfortunate feature that both levels of B had the same number of observations. That is no longer true, and makes it a bit easier to see the differences in the analyses.

I have added two extra columns to my SPSS data set. Agrp is just the coding of the levels of A as 1 and 2, while Bgrp is the coding of the levels of B as $1,2,3,4$. The other columns (except for dv) represent contrast coding.

| 5 | 1 | 1 | 0 | 0 | 1 | 0 | 0 | 1.00 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 7 | 1 | 1 | 0 | 0 | 1 | 0 | 0 | 1.00 |
| 9 | 1 | 1 | 0 | 0 | 1 | 0 | 0 | 1.00 |
| 8 | 1 | 1 | 0 | 0 | 1 | 0 | 0 | 1.00 |
| 2 | 1 | 0 | 1 | 0 | 0 | 1 | 0 | 1.00 |
| 5 | 1 | 0 | 1 | 0 | 0 | 1 | 0 | 1.00 |
| 7 | 1 | 0 | 1 | 0 | 0 | 1 | 0 | 1.00 |
| 3 | 1 | 0 | 1 | 0 | 0 | 1 | 0 | 1.00 |
| 9 | 1 | 0 | 1 | 0 | 0 | 1 | 0 | 1.00 |
| 8 | 1 | 0 | 0 | 1 | 0 | 0 | 1 | 1.00 |
| 11 | 1 | 0 | 0 | 1 | 0 | 0 | 1 | 1.00 |
| 12 | 1 | 0 | 0 | 1 | 0 | 0 | 1 | 1.00 |
| 14 | 1 | 0 | 0 | 1 | 0 | 0 | 1 | 1.00 |
| 11 | 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1.00 |
| 15 | 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1.00 |
| 16 | 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1.00 |
| 10 | 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1.00 |
| 9 | 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1.00 |
| 7 | -1 | 1 | 0 | 0 | -1 | 0 | 0 | 2.00 |
| 9 | -1 | 1 | 0 | 0 | -1 | 0 | 0 | 2.00 |
| 10 | -1 | 1 | 0 | 0 | -1 | 0 | 0 | 2.00 |
| 9 | -1 | 1 | 0 | 0 | -1 | 0 | 0 | 2.00 |
| 3 | -1 | 0 | 1 | 0 | 0 | -1 | 0 | 2.00 |
| 8 | -1 | 0 | 1 | 0 | 0 | -1 | 0 | 2.00 |
| 9 | -1 | 0 | 1 | 0 | 0 | -1 | 0 | 2.00 |
| 11 | -1 | 0 | 1 | 0 | 0 | -1 | 0 | 2.00 |
| 9 | -1 | 0 | 0 | 1 | 0 | 0 | -1 | 2.00 |
| 12 | -1 | 0 | 0 | 1 | 0 | 0 | -1 | 2.00 |


| 14 | -1 | 0 | 0 | 1 | 0 | 0 | -1 | 2.00 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 8 | -1 | 0 | 0 | 1 | 0 | 0 | -1 | 2.00 |
| 7 | -1 | 0 | 0 | 1 | 0 | 0 | -1 | 2.00 |
| 11 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 14 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 10 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 12 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 13 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 11 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |
| 12 | -1 | -1 | -1 | -1 | 1 | 1 | 1 | 2.00 |

The following are the different analyses as they are computed by SPSS version 15.

Type I A then B then AB
Tests of Between-Subjects Effects
Dependent Variable: DV

|  | Type I Sum <br> of Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Source | $216.017(\mathrm{a})$ | 7 | 30.860 | 5.046 | .001 |
| Intercept Model | 3410.526 | 1 | 3410.526 | 557.709 | .000 |
| Agrp | 9.579 | 1 | 9.579 | 1.566 | .220 |
| Bgrp | 186.225 | 3 | 62.075 | 10.151 | .000 |
| Agrp * Bgrp | 20.212 | 3 | 6.737 | 1.102 | .364 |
| Error | 183.457 | 30 | 6.115 |  |  |
| Total | 3810.000 | 38 |  |  |  |
| Corrected Total | 399.474 | 37 |  |  |  |

a R Squared $=.541$ (Adjusted R Squared $=.434$ )

Type I B then A then $A B$
Tests of Between-Subjects Effects
Dependent Variable: DV

| Source | Type I Sum <br> of Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Corrected Model | $216.017(\mathrm{a})$ | 7 | 30.860 | 5.046 | .001 |
| Intercept | 3410.526 | 1 | 3410.526 | 557.709 | .000 |
| Bgrp | 193.251 | 3 | 64.417 | 10.534 | .000 |
| Agrp | 2.553 | 1 | 2.553 | .418 | .523 |
| Bgrp Agrp | 20.212 | 3 | 6.737 | 1.102 | .364 |
| Error | 183.457 | 30 | 6.115 |  |  |
| Total | 3810.000 | 38 |  |  |  |


| Corrected Total | 399.474 | 37 |
| :--- | ---: | ---: |

a R Squared $=.541$ (Adjusted R Squared $=.434$ )

## Type II

## Tests of Between-Subjects Effects

Dependent Variable: DV

| Source | Type II Sum <br> of Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Corrected Model | $216.017(\mathrm{a})$ | 7 | 30.860 | 5.046 | .001 |
| Intercept | 3410.526 | 1 | 3410.526 | 557.709 | .000 |
| Agrp | 2.553 | 1 | 2.553 | .418 | .523 |
| Bgrp | 186.225 | 3 | 62.075 | 10.151 | .000 |
| Agrp *Bgrp | 20.212 | 3 | 6.737 | 1.102 | .364 |
| Error | 183.457 | 30 | 6.115 |  |  |
| Total | 3810.000 | 38 |  |  |  |
| Corrected Total | 399.474 | 37 |  |  |  |

a R Squared $=.541$ (Adjusted R Squared $=.434$ )

## Type III

## Tests of Between-Subjects Effects

Dependent Variable: DV

| Source | Type III Sum <br> of Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Corrected Model | $216.017(\mathrm{a})$ | 7 | 30.860 | 5.046 | .001 |
| Intercept | 3163.841 | 1 | 3163.841 | 517.370 | .000 |
| Agrp | 3.464 | 1 | 3.464 | .566 | .458 |
| Bgrp | 185.172 | 3 | 61.724 | 10.093 | .000 |
| Agrp * Bgrp | 20.212 | 3 | 6.737 | 1.102 | .364 |
| Error | 183.457 | 30 | 6.115 |  |  |
| Total | 3810.000 | 38 |  |  |  |
| Corrected Total | 399.474 | 37 |  |  |  |

a R Squared $=.541$ (Adjusted R Squared $=.434$ )

Now I will run a whole set of regressions. Bear with me.
Using contrast coding for variable A alone: (i.e. $\mathrm{Y}=\mathrm{b}_{1} \mathrm{~A}$ )
ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| 1 | Regression | 9.579 | 1 | 9.579 | .884 | $.353(\mathrm{a})$ |
|  | Residual | 389.894 | 36 | 10.830 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), A
b Dependent Variable: DV
Using contrast coding for variable $B$ alone: (i.e. $Y=b_{1} B_{1}+b_{2} B_{2}+b_{3} B_{3}$ )

## ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | :---: |
| 1 | Regression | 193.251 | 3 | 64.417 | 10.621 | $.000(\mathrm{a})$ |
|  | Residual | 206.222 | 34 | 6.065 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), B3, B2, B1
b Dependent Variable: DV

Using contrast coding for the interaction terms alone (i.e. $Y=b_{1} A B_{1}+b_{2} A B_{2}+b_{3} A B_{3}$ )
ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| 1 | Regression | 20.538 | 3 | 6.846 | .614 | $.610(\mathrm{a})$ |
|  | Residual | 378.936 | 34 | 11.145 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), AB3, AB2, AB1
b Dependent Variable: DV

Using contrast terms for $A$ and $B$ but not $A B$ (i.e. $Y=b_{1} A+b_{2} B_{1}+b_{3} B_{2}+b_{4} B_{3}$ )
ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | :---: |
| 1 | Regression | 195.805 | 4 | 48.951 | 7.931 | $.000(\mathrm{a})$ |
|  | Residual | 203.669 | 33 | 6.172 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), B3, A, B2, B1

Using contrasts for $A$ and $A B$ but not $B \quad\left(\left(i . e . Y=b_{1} A+b_{2} A B 1+b_{3} A B_{2}+b_{4} A B_{3}\right)\right.$
ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| 1 | Regression | 30.845 | 4 | 7.711 | .690 | $.604(\mathrm{a})$ |
|  | Residual | 368.629 | 33 | 11.171 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), AB3, A, AB2, AB1
b Dependent Variable: DV

Using contrasts for $B$ and $A B$ but not $A$ (i.e. $Y=b_{1} B_{1}+b_{2} A B_{2}+b_{3} A B_{3} b_{4} A B_{1}+$ $\mathrm{b}_{5} \mathrm{AB}_{2}+\mathrm{b}_{6} \mathrm{AB}_{3}$ )

ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | :--- |
| 1 | Regression | 212.552 | 6 | 35.425 | 5.875 | $.000(\mathrm{a})$ |
|  | Residual | 186.921 | 31 | 6.030 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), AB3, B2, AB2, B3, B1, AB1
b Dependent Variable: DV

Now contrast codes for $\mathrm{A}, \mathrm{B}$, and AB (This is the full model)

## ANOVA(b)

| Model |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | :--- |
| 1 | Regression | 216.017 | 7 | 30.860 | 5.046 | $.001(\mathrm{a})$ |
|  | Residual | 183.457 | 30 | 6.115 |  |  |
|  | Total | 399.474 | 37 |  |  |  |

a Predictors: (Constant), AB3, B2, A, AB2, B3, B1, AB1
b Dependent Variable: DV

## Type I sums of squares

For Type I each term is adjusted only for the terms that were entered before it. I will just use the orders $\mathrm{A}, \mathrm{B}, \mathrm{AB}$ and $\mathrm{B}, \mathrm{A}, \mathrm{AB}$, but you could, if you wanted, us $\mathrm{AB}, \mathrm{B}, \mathrm{A}$ or A , $\mathrm{AB}, \mathrm{B}$, etc. I can't think why.

Let's start with Type I SS where the order of entry is $\mathrm{A}, \mathrm{B}, \mathrm{AB}$
$\operatorname{SSreg}(\mathrm{A})=9.579$
This will be $\mathrm{SS}(\mathrm{A})$ in the Anova
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B})=195.805$
Gain is SSreg $=195.805-9.579=186.225$
This will be $\mathrm{SS}(\mathrm{B})$ in the Anova
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B}, \mathrm{AB})=216.017$
Gain from previous model is $216.017-195.805=20.212$
This is $\operatorname{SS}(\mathrm{AB})$ in the Anova
Now Type I SS where the order of entry is B, A, AB
$\operatorname{SSreg}(\mathrm{B})=193.251$
This is $\operatorname{SS}(\mathrm{B})$ in that Anova
SSreg(A,B) - 195.805
Gain is $\mathrm{SS}(\mathrm{Reg})$ is $195.805-193.251=2.554$
This is $\operatorname{SS}(\mathrm{B})$
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B}, \mathrm{AB})=216.017$
Gain is $216.017-195.805=20.212$
This is $\operatorname{SS}(\mathrm{AB})$

For Type II you adjust B for A, A for B, and then AB for A and B. (If we had a three way, we would adjust A for B and C, B for A and C, C for A and B, AB for A, B, C, BC for $\mathrm{A}, \mathrm{B}, \mathrm{C}, \mathrm{AC}$ for $\mathrm{A}, \mathrm{B}, \mathrm{C}$, and ABC for $\mathrm{A}, \mathrm{B}, \mathrm{C}, \mathrm{AB}, \mathrm{AC}, \mathrm{BC}$. I have never tried that.)
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B})=195.805$
$\operatorname{SSreg}(\mathrm{A})=9.579$
Difference is $195.805-9.579=186.226$
$\operatorname{SSreg}(B)=193.251$

Difference from A,B is $195.805-193.251=2.554$
These are $\mathrm{SS}(\mathrm{A})$ and $\mathrm{SS}(\mathrm{B})$ in Anova.
For AB we take the full model
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B}, \mathrm{AB})=216.017$
$\operatorname{SSreg}(A, B)=195.805$
Difference is $216.017-195.805=20.212$
This is $\operatorname{SS}(\mathrm{AB})$

Type III
Here everything is adjusted for everything else.
$\operatorname{SSreg}(\mathrm{A}, \mathrm{B}, \mathrm{AB})=216.017$
$\operatorname{SSreg}(A, B)=195.805$
Difference is $216.017-195.805=20.212$
This is $\mathrm{SS}(\mathrm{AB})$
$\operatorname{SSreg}(\mathrm{A}, \mathrm{AB})=30.845$
Diff from full model is $216.017-30.845=185.172$
$\operatorname{SSreg}(\mathrm{B}, \mathrm{AB})=212.552$
Diff from full model is $216.017-212.552=3.465$

## MS(Error)

In all of these analyses the error term would come from MSresidual of the full model, which is 6.115

## References

Howell, D. C. (2010). Statistical Methods for Psychology. Wadsworth:Cengage.

