

Research Design - - Topic 15
Reporting Model II in MRC with Completely
Randomized Designs and a Continuous Factor

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General Description

Step I. Computing statistics for the “main effects”

Step II. Computing statistics for the “two way interactions”

Step III. Computing statistics for the “three way interaction”

Statistics that should be reported

Running Model II in SPSS GLM Univariate?

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General Description

To date, we have discussed the use of MRC for various designs. Generally, we have considered both Model I and Model II, and have discussed the direct relevance of these to the analysis of variance model. In this lecture, we will discuss presenting the results from Model II, focusing on the statistics reported and their appropriate interpretation.

This presentation includes example statistics for a three factor completely randomized design, A (with 3 levels), B (with 2 levels) and C (a centred continuous variable), where Dummy coding is used for the categorical factors. The description, however, can be generalized to any number of factors with any number of levels.

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Sample
Data

| | B1 | | B2 | |
|----|----|----|----|----|
| | C | X | C | X |
| A1 | 5 | 31 | -4 | 36 |
| | -4 | 33 | -4 | 37 |
| | 6 | 37 | 6 | 39 |
| | 3 | 29 | 3 | 33 |
| | -1 | 37 | -6 | 34 |
| A2 | -3 | 31 | | |
| | -5 | 32 | 2 | 36 |
| | 2 | 36 | 7 | 27 |
| | 4 | 30 | -6 | 28 |
| | -3 | 33 | 2 | 30 |
| A3 | -3 | 37 | | |
| | 4 | 38 | 1 | 36 |
| | -5 | 29 | -1 | 37 |
| | 3 | 36 | 2 | 36 |
| | 5 | 32 | -3 | 46 |
| | | -5 | 45 | |
| | | -2 | 42 | |

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Step I. Computing statistics for the “main effects”

Compute the relevant squared multiple correlations

- Enter a1, a2, b1, c $R^2 = .31795$

Use the regression coefficients from this run to compute intercepts and within cells slope

- Enter b1, c $R^2 = .10521$

- Enter a1, a2, c $R^2 = .26991$

- Enter a1, a2, b1 $R^2 = .30289$

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Compute the squared semi-partial multiple correlations, and associated F-ratios

$$\hat{R}_A^2 = R_{A,B,C}^2 - R_{B,C}^2 = .21274$$

$$\hat{R}_B^2 = R_{A,B,C}^2 - R_{A,C}^2 = .04804$$

$$\hat{R}_C^2 = R_{A,B,C}^2 - R_{A,B}^2 = .01506$$

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General Form of the F-ratios

$$F = \frac{\hat{R}^2 / v_1}{R_{error}^2 / v_2}$$

Using Model 1 error terms

$$F_A = \frac{.21274/2}{(1 - .31795)/25} = 3.899 \quad ns$$

$$F_B = \frac{.04804/1}{.68205/25} = 1.761 \quad F_C = \frac{.01506/1}{.68205/18} = 0.552$$

Using Model 2 error terms

$$F_A = \frac{.21274/2}{1 - .67404/18} = 5.873 \quad p < .05$$

$$F_B = \frac{.04804/1}{.32596/18} = 2.653 \quad F_C = \frac{.01506/1}{.32596/18} = 0.832$$

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Solution of the regression equation

Coefficients^a

| Model | | Unstandardized Coefficients | | Standardized Coefficients | t | Sig. |
|-------|------------|-----------------------------|------------|---------------------------|----------|---------|
| | | B | Std. Error | Beta | | |
| 1 | (Constant) | 38.49660 | 1.43556 | | 26.81638 | 0.00000 |
| | a1 | -3.10582 | 1.80666 | -0.32993 | -1.71909 | 0.09796 |
| | a2 | -5.25968 | 1.90036 | -0.53132 | -2.76773 | 0.01047 |
| | b1 | -2.02647 | 1.52715 | -0.22336 | -1.32696 | 0.19651 |
| | c | -0.13984 | 0.18822 | -0.12382 | -0.74294 | 0.46444 |

a. Dependent Variable: x

The following equations yield the cell estimates on the next slide.

$$\text{int} = 38.4966 - 3.10582 * a1 - 5.25968 * a2 - 2.02647 * b1$$

$$\text{slo} = -.13984$$

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Intercepts for the A and B factors

| | B1 | B2 | Means |
|-------|--------|--------|--------|
| A1 | 33.364 | 35.391 | 34.378 |
| A2 | 31.210 | 33.237 | 32.224 |
| A3 | 36.470 | 38.497 | 37.483 |
| Means | 33.682 | 35.708 | |

Note: If you were to plot these intercepts, the lines would be parallel. That is, there is no AB interaction at this stage.

Main effect slope = - 0.140

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Interpretation

The significant F-ratio for A indicates that the means of the intercepts for A differ more than expected by chance.

As described in earlier lectures, the regression coefficient for a1 indicates that:

$$I_{A1B2} - I_{A3B2} = 35.391 - 38.497 = -3.106$$

But since there is no interaction, this is equivalent to:

$$I_{A1} - I_{A3} = 34.378 - 37.483 = -3.105$$

Similarly: $I_{A2} - I_{A3} = 32.224 - 37.483 = -5.259$

If the F-ratio for B is significant, a similar logic applies.

If the F-ratio for C was significant, the within cells slope differs from 0.

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Step II. Computing the two-way interaction statistics.

Compute the relevant squared multiple correlations

- Enter a1, a2, b1, c, a1b1, a2b1, a1c, a2c, b1c

$$R^2 = .51808$$

Use the regression coefficients from this run to compute intercepts and two-way slopes

- Enter a1, a2, b1, c, a1c, a2c, b1c $R^2 = .37051$
- Enter a1, a2, b1, c, a1b1, a2b1, b1c $R^2 = .50727$
- Enter a1, a2, b1, c, a1b1, a2b1, a1c, a2c $R^2 = .49766$

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Compute the squared semi-partial multiple correlations, and associated F-ratios

$$\hat{R}_{AB}^2 = R_{A,B,C,AB,AC,BC}^2 - R_{A,B,C,AC,BC}^2 = .14757$$

$$\hat{R}_{AC}^2 = R_{A,B,C,AB,AC,BC}^2 - R_{A,B,C,AB,BC}^2 = .01081$$

$$\hat{R}_{BC}^2 = R_{A,B,C,AB,AC,BC}^2 - R_{A,B,C,AB,AC}^2 = .02042$$

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F-ratios for Step II:

Using Model 1 error terms

$$F_{AB} = \frac{.14757/2}{(1-.51806)/20} = 3.062 \quad ns$$

$$F_{AC} = \frac{.01081/2}{.48194/20} = 0.449 \quad F_{BC} = \frac{.02042/1}{.48194/20} = 0.847$$

Using Model 2 error terms

$$F_{AB} = \frac{.14757/2}{(1-.67404)/18} = 4.074 \quad p < .05$$

$$F_{AC} = \frac{.01081/2}{.32596/18} = 0.298 \quad F_{BC} = \frac{.02042/1}{.32596/18} = 1.128$$

Coefficients^a

| Model | | Unstandardized Coefficients | | Standardized Coefficients | t | Sig. |
|-------|------------|-----------------------------|------------|---------------------------|----------|---------|
| | | B | Std. Error | Beta | | |
| 1 | (Constant) | 39.83013 | 1.69157 | | 23.54627 | 0.00000 |
| | a1 | -4.08616 | 2.41441 | -0.43407 | -1.69240 | 0.10609 |
| | a2 | -9.33586 | 2.62664 | -0.94309 | -3.55430 | 0.00199 |
| | b1 | -6.02426 | 2.77130 | -0.66399 | -2.17380 | 0.04190 |
| | c | -0.37740 | 0.46357 | -0.33416 | -0.81411 | 0.42517 |
| | a1b1 | 2.99084 | 3.66649 | 0.26372 | 0.81572 | 0.42427 |
| | a2b1 | 9.28004 | 3.85859 | 0.76238 | 2.40503 | 0.02597 |
| | a1c | 0.32138 | 0.48529 | 0.18697 | 0.66223 | 0.51538 |
| | a2c | 0.18198 | 0.51630 | 0.09148 | 0.35246 | 0.72818 |
| | b1c | 0.34547 | 0.37529 | 0.21172 | 0.92054 | 0.36826 |

a. Dependent Variable: x

int = 39.83013 - 4.08616*a1 - 9.33586*a2 - 6.02426*b1 + 2.99084*a1*b1 + 9.28004*a2*b1

slo = -.37740 + .32138*a1 + .18198*a2 + .34547*b1

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| | B1 | B2 |
|------------|--------|--------|
| Intercepts | | |
| A1 | 32.711 | 35.744 |
| A2 | 33.750 | 30.494 |
| A3 | 33.806 | 39.830 |

Note: Plotting these intercepts would show the influence of the AB interaction.

| | B1 | B2 |
|--------|-------|-------|
| Slopes | | |
| A1 | .289 | -.056 |
| A2 | .150 | -.195 |
| A3 | -.032 | -.377 |

Note: Plotting these slopes would produce parallel lines because the three-way interaction is excluded.

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Interpretation of Intercepts

The regression coefficients for the ab terms on Slide 13 are equal to the following contrast-contrast interactions based on the intercepts in Slide 14:

$$b_{a1b1} = 32.711 - 35.744 - 33.806 + 39.830 = 2.991$$

$$b_{a2b1} = 33.750 - 30.494 - 33.806 + 39.830 = 9.280$$

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Interpretation of Slopes

The regression coefficients for the ac and bc terms on Slide 13 are equal to contrasts of slopes in Step 14:

$$b_{a1c} = -.056 - (-.377) = .321$$

$$b_{a2c} = -.195 - (-.377) = .182$$

$$b_{b1c} = -.032 - (-.377) = .345$$

Note, however, that because the three way interaction is defined as 0 in this step, these values are the same as contrasts involving the corresponding marginal means of the slopes.

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Step III. Computing the three-way interaction statistics.

Compute the relevant squared multiple correlations

- Enter a1, a2, b1, c, a1b1, a2b1, a1c, a2c, b1c, a1b1c, a2b1c

$$R^2 = .67404$$

Use the regression coefficients from this run to compute intercepts and three-way slopes

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Compute the squared semi-partial multiple correlation, and associated F-ratio

$$R_{ABC}^2 = R_{A,B,C,AB,AC,BC,ABC}^2 - R_{A,B,C,AB,AC,BC}^2 = .15592$$

For a completely randomized design, compute the residual R^2

$$R_{error}^2 = 1 - R_{A,B,C,AB,AC,BC,ABC}^2 = .32596$$

For studies involving repeated measures, other error terms will be required, as discussed in earlier topics.

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General Form of the F-ratio

$$F = \frac{\hat{R}^2 / v_1}{R_{error}^2 / v_2}$$

$$F_{ABC} = \frac{.15592 / 2}{.32596 / 18} = 4.305 \quad p < .05$$

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Coefficients^a

| Model | | Unstandardized Coefficients | | Standardized Coefficients | t | Sig. |
|-------|------------|-----------------------------|------------|---------------------------|----------|---------|
| | | B | Std. Error | Beta | | |
| 1 | (Constant) | 38.20000 | 1.56840 | | 24.35601 | 0.00000 |
| | a1 | -2.25185 | 2.19081 | -0.23921 | -1.02786 | 0.31763 |
| | a2 | -7.97522 | 2.33594 | -0.80564 | -3.41414 | 0.00309 |
| | b1 | -5.53068 | 2.40851 | -0.60959 | -2.29631 | 0.03388 |
| | c | -1.60000 | 0.57917 | -1.41668 | -2.76257 | 0.01282 |
| | a1b1 | 2.53808 | 3.18253 | 0.22380 | 0.79751 | 0.43555 |
| | a2b1 | 8.73348 | 3.35062 | 0.71748 | 2.60653 | 0.01785 |
| | a1c | 1.74815 | 0.66255 | 1.01706 | 2.63853 | 0.01669 |
| | a2c | 1.62017 | 0.68142 | 0.81443 | 2.37765 | 0.02871 |
| | b1c | 2.21753 | 0.71668 | 1.35899 | 3.09418 | 0.00626 |
| | a1b1c | -2.32123 | 0.86104 | -0.90960 | -2.69584 | 0.01478 |
| | a2b1c | -2.41012 | 0.91395 | -0.76480 | -2.63704 | 0.01674 |

a. Dependent Variable: x

Note. 9 of the regression coefficients are significant but only the last two are used in Model II. See slide 23 for the values that should be reported for Model II.

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$$\text{Int} = 38.20 - 2.25185 * a1 - 7.97522 * a2 - 5.53068 * b1 + 2.53808 * a1 * b1 + 8.73348 * a2 * b1$$

Intercepts

| | B1 | B2 |
|----|--------|--------|
| A1 | 32.956 | 35.948 |
| A2 | 33.428 | 30.225 |
| A3 | 32.669 | 38.200 |

$$\text{slo} = -1.60 + 1.74815 * a1 + 1.62017 * a2 + 2.21753 * b1 - 2.32123 * a1 * b1 - 2.41012 * a2 * b1$$

Slopes

| | B1 | B2 |
|----|-------|--------|
| A1 | .044 | .148 |
| A2 | -.172 | .020 |
| A3 | .618 | -1.600 |

Note: Plotting these values shows the complete results.

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The meaning of the regression coefficients for abc in Slide 20 based on the statistics in Slide 21 are:

$$b_{a1b1c} = .044 - .148 - .618 + (-1.600) = -2.322$$

$$b_{a2b1c} = -.172 - .020 - .618 + (-1.600) = -2.410$$

Note. Each of these describe a contrast-contrast interaction of slopes. The first is (A1-A3 with B1-B2)C and the second is (A2-A3 with B1-B2)C. They do not involve any pairwise contrasts of slopes.

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Statistics that should be reported drawn from the 3 steps

| | b | S.E. | t | β | F-ratio | |
|---|-------|--------|-------|----------|---------|--------|
| 1 | a1 | -3.106 | 1.807 | -1.719 | -.330 | |
| | a2 | -5.260 | 1.900 | -2.768** | -.531 | 5.873* |
| | b1 | -2.026 | 1.527 | -1.327 | -.223 | 2.653 |
| | c | -.140 | 0.188 | -0.743 | -.124 | 0.832 |
| 2 | a1b1 | 2.991 | 3.666 | .816 | .264 | |
| | a2b1 | 9.280 | 3.859 | 2.405* | .762 | 4.074* |
| | a1c | .321 | 0.485 | .662 | .187 | |
| | a2c | .182 | .516 | .352 | .091 | 0.298 |
| 3 | b1c | .345 | .375 | .921 | .212 | 1.128 |
| | a1b1c | -2.321 | .861 | -2.696** | -.910 | |
| | a2b1c | -2.410 | .914 | -2.637** | -.765 | 4.305* |

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Interpretation

For main effects, only A was significant, $F(2,18) = 5.873, p < .05$

- We could thus perform post hoc tests comparing the three intercepts for A on Slide 8, 34.378, 32.224, and 37.483. The regression coefficients on Slide 7 for a1 and a2 test two of the three contrasts of interest. Only the regression coefficient for a2 was significant, $b = -5.260, t(18) = -2.768, p < .02$, indicating that the intercept for A2 was significantly less than that for A3. To compare A1 with A2, the analysis must be redone with A2 coded with all 0's; or some other procedure should be used.

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For the two way interactions, only that for AB was significant, $F(2,18) = 4.074, p < .05$.

- Because AB was significant, this suggests that contrasts involving any of the cell intercepts in Slide 14 could be made, as in tests of simple main effects of means. This application of multiple regression does not provide any such contrasts.
- The regression coefficients for $a1b1$ and $a2b1$ describe contrast-contrast interactions. Of these, only the regression coefficient for $a2b1$ was significant, $b=9.280, t(18) = 2.405, p < .05$ which refers to the contrast-contrast interaction $(A2-A3)*(B1-B2)$. If it was desired to test the $(A1-A2)*(B1-B2)$ contrast-contrast interaction, another multiple regression analysis would be required.

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The three way interaction was significant, $F(2,18) = 4.305, p < .05$.

- Because this F-ratio was significant, it suggests that contrasts could be made of any of the cell slopes in Slide 21. This analysis does not provide any such contrasts. One could perform specific contrasts using formulae described in earlier lectures.
- The regression coefficients for $a1b1c$ and $a2b1c$ describe contrast-contrast interactions for these slopes. Both were significant. The three way interaction of slopes $(A1-A3)*(B1-B2)*C, b = -2.32123, t(18)=-2.696, p < .02$ and $(A2-A3)*(B1-B2)*C, b = -2.41012, t(18)=-2.637$ were significant. As can be seen in Slide 21, this interaction occurs because the differences between the slopes for the two levels of A are different at the B1 level than at the B2 level.

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The meaning of the regression coefficients in Slide 20 given the intercepts and slopes in Slide 21.

| | | Meaning |
|------------|----------|---------------------|
| Intercepts | Constant | INT(A3B2) |
| | a1 | A1B2-A3B2 |
| | a2 | A2B2-A3B2 |
| | b1 | A3B1-A3B2 |
| | a1b1 | A1B1-A1B2-A3B1+A3B2 |
| | a2b1 | A2B1-A2B2-A3B1+A3B2 |
| Slopes | c | A3B2 |
| | a1c | A1B2-A3B2 |
| | a2c | A2B1-A3B2 |
| | b1c | A3B1-A3B2 |
| | a1b1c | A1B1-A1B2-A3B1+A3B2 |
| | a2b1c | A2B1-A2B2-A3B1+A3B2 |
| | a2b3c | A2B1-A2B2-A3B1+A3B2 |

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Running Model II in SPSS GLM Univariate?

In previous examples, we showed how the analysis using Model I could also be run in SPSS GLM Univariate, making appropriate changes to the Design statement where required. This was true whether one of the factors was categorical or continuous. If all factors are categorical, they could also be run in SPSS GLM Univariate using Model II, however, Model II cannot be run in SPSS GLM Univariate if any factor is continuous. Again, this is because SPSS GLM solves specific equations and does not compute a series of squared multiple correlations and resulting squared semi-partial multiple correlations.

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