

## Cross-level Interaction Example (Predicting Random Slopes)

### SPSS

```
get file='c:\jason\spsswin\mlrclass\hsbmerged.sav'.
```

```
* Following recommendations of Curran & Bauer (2006) this analysis group-centered SES.1
*easiest in this case to just use the existing meanses variable.
compute cses=ses - meanses.

*but aggregate command would normally be needed.
* AGGREGATE creates group means for ses variable if you use the group id on the break subcommand.
*AGGREGATE
  /BREAK schoolid
  /mnses=MEAN(ses).
*compute cses=ses - mnses.
```

```
** Test the SES and sector cross-level interaction.
MIXED mathach WITH cses sector
  /CRITERIA=MXITER(1000) SCORING(1)
  /METHOD = REML
  /PRINT = SOLUTION TESTCOV HISTORY
  /FIXED = cses sector cses*sector| SSTYPE(3)
  /RANDOM = INTERCEPT cses | SUBJECT(schoolid) COVTYPE(UN).
```

#### Information Criteria<sup>a</sup>

-2 Restricted Log Likelihood	46638.605
Akaike's Information Criterion (AIC)	46646.605
Hurvich and Tsai's Criterion (AICC)	46646.610
Bozdogan's Criterion (CAIC)	46678.122
Schwarz's Bayesian Criterion (BIC)	46674.122

The information criteria are displayed in smaller-is-better forms.

a. Dependent Variable: mathach.

### Fixed Effects.

Type III Tests of Fixed Effects<sup>a</sup>

Source	Numerator df	Denominator df	F	Sig.
Intercept	1	157.820	1516.671	.000
cses	1	132.895	327.151	.000
sector	1	153.252	40.568	.000
cses * sector	1	144.133	32.912	.000

a. Dependent Variable: mathach.

Estimates of Fixed Effects<sup>a</sup>

Parameter	Estimate	Std. Error	df	t	Sig.	95% Confidence Interval	
						Lower Bound	Upper Bound
Intercept	11.41064	.2929978	157.820	38.944	.000	10.8319397	11.9893453
cses	2.8028147	.1549603	132.895	18.087	.000	2.4963069	3.1093224
sector	2.7995407	.4395387	153.252	6.369	.000	1.9312037	3.6678778
cses * sector	-1.34109	.2337667	144.133	-5.737	.000	-1.8031459	-.8790385

a. Dependent Variable: mathach.

### Covariance Parameters

Estimates of Covariance Parameters<sup>a</sup>

Parameter	Estimate	Std. Error	Wald Z	Sig.	95% Confidence Interval		
					Lower Bound	Upper Bound	
Residual	36.70560	.6257696	58.657	.000	35.4993786	37.9528069	
Intercept + cses	UN (1,1)	6.7504155	.8668911	7.787	.000	5.2483024	8.6824473
[subject = schoolid]	UN (2,1)	1.0507234	.3425159	3.068	.002	.3794047	1.7220422
	UN (2,2)	.2656916	.2288151	1.161	.246	-.0491268	1.4369356

a. Dependent Variable: mathach.

<sup>1</sup> These authors also recommend centering the level-2 variable, which I did not do here, but that would be perfectly acceptable to do. The AGGREGATE command can be used to derive the full sample mean if the /BREAK subcommand is left off.

```
AGGREGATE
  /mnsector=mean(sector).
  compute gndsector=sector - mnsector.
```

## R

```
> #Model with group-centered SES
> # group-mean centering of original ses variable called cses
> mydata$cses <- mydata$ses - ave(mydata$ses, mydata$schoolid)
> # grand-mean centering of sector (not used in the example)
> #mydata$gndsector <- mydata$sector - mean(mydata$sector)

> library(lme4)
> model1 <- lmer(mathach ~ cses + sector + cses*sector + (cses|schoolid), data = mydata, REML = TRUE)
> summary(model1)
Linear mixed model fit by REML ['lmerMod']
Formula: mathach ~ cses + sector + cses * sector + (cses | schoolid)
Data: mydata

REML criterion at convergence: 46638.6

Scaled residuals:
    Min       1Q   Median       3Q      Max
-3.06490 -0.73237  0.01564  0.75373  2.94191

Random effects:
 Groups Name Variance Std.Dev. Corr
schoolid (Intercept) 6.7378 2.5957
cses      cses      0.2657 0.5155 0.78
Residual                    36.7056 6.0585
Number of obs: 7185, groups: schoolid, 160

Fixed effects:
              Estimate Std. Error t value
(Intercept)  11.3939    0.2928   38.92
cses          2.8028    0.1550   18.09
sector        2.8075    0.4392    6.39
cses:sector  -1.3411    0.2338   -5.74

Correlation of Fixed Effects:
      (Intr) cses  sector
cses      0.259
sector    -0.667 -0.173
cses:sector -0.172 -0.663  0.261
> #random effects confidence intervals using full ML
> model2 <- lmer(mathach ~ cses + sector + cses*sector + (cses|schoolid), data = mydata, REML = FALSE)
> rand(model1)
Analysis of Random effects Table:
              Chi.sq Chi.DF p.value
cses:schoolid  11.9      2    0.003

> confint(model2)
Computing profile confidence intervals ...
      2.5 %      97.5 %
.sig01  2.2789614  2.9292406
.sig02  0.3036104  1.0000000
.sig03  0.1777120  0.8711717
.sigma  5.9586839  6.1611543
(Intercept) 10.8202813 11.9674334
cses        2.4993136  3.1070907
sector      1.9471145  3.6681425
cses:sector -1.7987557 -0.8826638
There were 50 or more warnings (use warnings() to see the first 50)
```

The warnings concern zeta values, which refer to transformed values of the change in deviance, values that can be used to assess the behavior of the iterative solution. Essentially, this warning suggests that some of the simulated values were not unique, which can potentially signal a problem. As the slope variance and intercept-slope covariance are both significant according to the confidence intervals, the results seem consistent with the `rand()` function joint test of these two parameters.

## HLM

HLM provides options for group or grand centering variables as they are added to the equation. SES was group-mean centered in this example and HLM prints a note about that. HLM also provides convenient graphing, which I will illustrate in more detail later.

The maximum number of level-1 units = 7185  
 The maximum number of level-2 units = 160  
 The maximum number of iterations = 100  
 Method of estimation: restricted maximum likelihood

The outcome variable is MATHACH

### Summary of the model specified

#### Level-1 Model

$$MATHACH_{ij} = \beta_{0j} + \beta_{1j}*(SES_{ij}) + r_{ij}$$

#### Level-2 Model

$$\beta_{0j} = \gamma_{00} + \gamma_{01}*(SECTOR_j) + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}*(SECTOR_j) + u_{1j}$$

SES has been centered around the group mean.

#### Mixed Model

$$MATHACH_{ij} = \gamma_{00} + \gamma_{01}*SECTOR_j$$

$$+ \gamma_{10}*SES_{ij} + \gamma_{11}*SECTOR_j*SES_{ij}$$

$$+ u_{0j} + u_{1j}*SES_{ij} + r_{ij}$$

### Final Results - Iteration 43

Iterations stopped due to small change in likelihood function

$\sigma^2 = 36.68995$

$\tau$

INTRCPT1,  $\beta_0$       6.73966      1.03763

SES,  $\beta_1$             1.03763      0.30405

$\tau$  (as correlations)

INTRCPT1,  $\beta_0$       1.000      0.725

SES,  $\beta_1$             0.725      1.000

Random level-1 coefficient	Reliability estimate
INTRCPT1, $\beta_0$	0.884
SES, $\beta_1$	0.138

The value of the log-likelihood function at iteration 43 = -2.331840E+004

### Final estimation of fixed effects:

Fixed Effect	Coefficient	Standard error	t-ratio	Approx. d.f.	p-value
For INTRCPT1, $\beta_0$					
INTRCPT2, $\gamma_{00}$	11.393836	0.292784	38.915	158	<0.001
SECTOR, $\gamma_{01}$	2.807465	0.439216	6.392	158	<0.001
For SES slope, $\beta_1$					
INTRCPT2, $\gamma_{10}$	2.802449	0.156523	17.904	158	<0.001
SECTOR, $\gamma_{11}$	-1.340634	0.236028	-5.680	158	<0.001

### Final estimation of fixed effects (with robust standard errors)

Fixed Effect	Coefficient	Standard error	t-ratio	Approx. d.f.	p-value
For INTRCPT1, $\beta_0$					
INTRCPT2, $\gamma_{00}$	11.393836	0.292348	38.974	158	<0.001
SECTOR, $\gamma_{01}$	2.807465	0.435634	6.445	158	<0.001
For SES slope, $\beta_1$					
INTRCPT2, $\gamma_{10}$	2.802449	0.157937	17.744	158	<0.001
SECTOR, $\gamma_{11}$	-1.340634	0.230324	-5.821	158	<0.001

### Final estimation of variance components

Random Effect	Standard Deviation	Variance Component	d.f.	$\chi^2$	p-value
INTRCPT1, $u_0$	2.59609	6.73966	158	1383.78477	<0.001
SES slope, $u_1$	0.55141	0.30405	158	175.31196	0.164
level-1, $r$	6.05722	36.68995			

### Statistics for current covariance components model

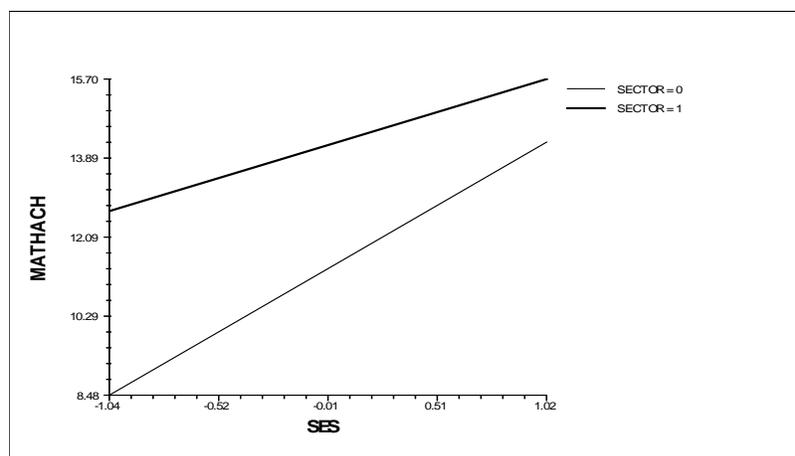
Deviance = 46636.802657

Number of estimated parameters = 4

## Write Up

A multilevel regression model was tested to investigate whether student socioeconomic status (SES) and school type (public vs. private) were predictive of a student's math achievement. Both intercepts and slopes were allowed to vary across schools. The cross-level interaction between school type and student SES was included in order to investigate whether the relationship between SES and math achievement depended on the type of school the student attended. In accordance with several recommendations about independent variable scaling with cross-level interaction tests (e.g., Bauer & Curran, 2005; Enders & Tofighi, 2007), the SES variable was group-mean centered in this analysis in order to improve the interpretation of the main effect. The fixed effects intercept was 11.39, which represents the average math achievement score in public schools, because school type was coded 0 for public schools. The test of the random effect indicated that the average math achievement within the public sector varied significantly across schools when SES was at the school average,  $\tau_0^2 = 6.74$ ,  $\chi^2(158) = 1383.78$ ,  $p < .001$ . SES was significantly related to math achievement, with an increase of 2.80 points in the math achievement test for each unit increase in SES,  $\gamma_{10} = 2.80$ ,  $SE = .16$ ,  $p < .001$ . The slopes for SES did not vary significantly across schools,  $\tau_1^2 = .30$ ,  $\chi^2(158) = 175.311$ , ns, however. The type of school was also significantly related to math achievement,  $\gamma_{01} = 2.81$ ,  $SE = .44$ ,  $p < .001$ . Students attending a private school had a math achievement score approximately 2.81 points higher than those attending a public school. These "main effects" should be interpreted in light of the significant cross-level, interaction, however. The effects of SES on math achievement depended on whether the student attended a private or public school,  $\gamma_{11} = -1.34$ ,  $SE = .23$ ,  $p < .001$ . Figure 1, which plots the interaction, shows that SES was more strongly related to math achievement for students in the public sector than the private sector. [I include the following text, because it is along the lines of what I would recommend for a write-up of a cross-level interaction in a journal article. This is not required for the homework assignment, however. Values are taken from the handout "Simple Slope Tests of Cross-level Interactions" available on the course website]. Simple slope tests (e.g., Preacher, Curran, & Bauer, 2006) indicated that SES was significantly related to math achievement for both school types. Students in public schools had math achievement scores that were 2.8 points higher for each unit increase in SES,  $\gamma_{SES|pub} = 2.802$ ,  $p < .001$ , whereas, students in private schools had math achievement scores only approximately one and one-half points higher for every unit increase in SES,  $\gamma_{SES|pri} = 1.462$ ,  $p < .01$ .

Figure 1



## References

- Bauer, D.J., & Curran, P.J. (2005). Probing interactions in fixed and multilevel regression: Inferential and graphical techniques. *Multivariate Behavioral Research*, 40, 373-400.
- Enders, C.K., & Tofighi, D. (2007). Centering predictor variables in cross-sectional multilevel models: A new look at an old issue. *Psychological Methods*, 12, 121-138.