

# Adding a Level-1 Predictor

PSYC 575

August 25, 2020 (updated: 9 September 2021)

# Week Learning Objectives

- Explain what the **ecological fallacy** is
- Use **cluster-mean/group-mean centering** to **decompose** the effect of a lv-1 predictor
- Define **contextual effects**
- Explain the concept of **random slopes**
- Analyze and interpret **cross-level interaction** effects

# Adding Level-1 Predictors

- E.g., student's SES
- Both predictor (ses) and outcome (mathach) are at level 1
- OLS still has Type I error inflation problem
  - Unless ICC = 0 for the predictor
- MLM can answer additional research questions
  - Within-Between effects and contextual effects
  - Random (varying) slopes
  - Cross-level interactions

# Research Questions

- Does math achievement vary across schools? How much is the variation?
- Do schools with higher mean SES have students with higher math achievement?
- Do students with higher SES have higher math achievement? Is the relation similar at the individual and cluster levels? Is this relation similar across schools?
- Is the relation between SES and math achievement moderated by some types of schools (e.g., Catholic vs. Public, high mean SES vs low mean SES)?

# The Same Predictor?

- Is it different to use MEANSES vs. SES as predictor?
  - MEANSES  $\rightarrow$  MATHACH is positive
  - $\gamma_{01} = 5.72$  ( $SE = 0.18$ )
- Should the coefficient be the same with SES?

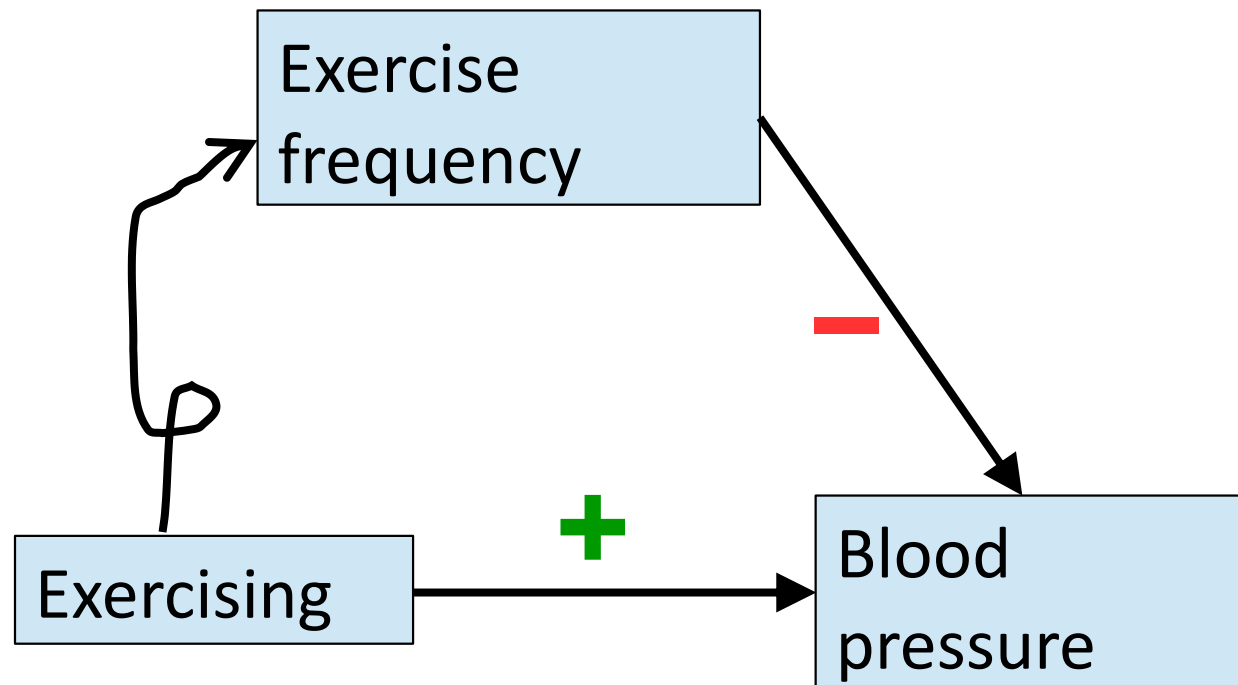
# Ecological Fallacy

# Ecological Fallacy

- Robinson's paradox (% immigrant and % illiterate)
- Errors in assuming that relationships at one level are the same moving to another level
- Failure to account for the clustering structure
  - ➔ Misleading results

# “Same” Predictor, Different Effects

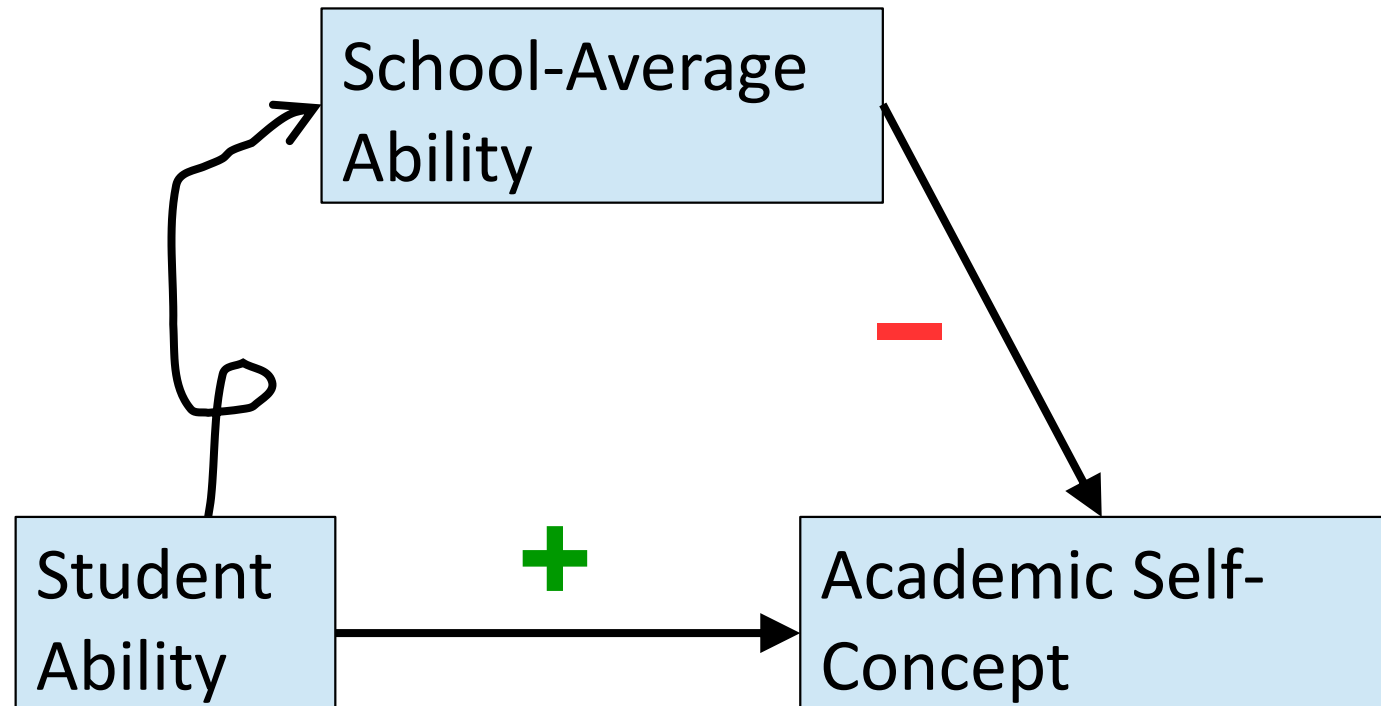
- Example: Exercise and blood pressure



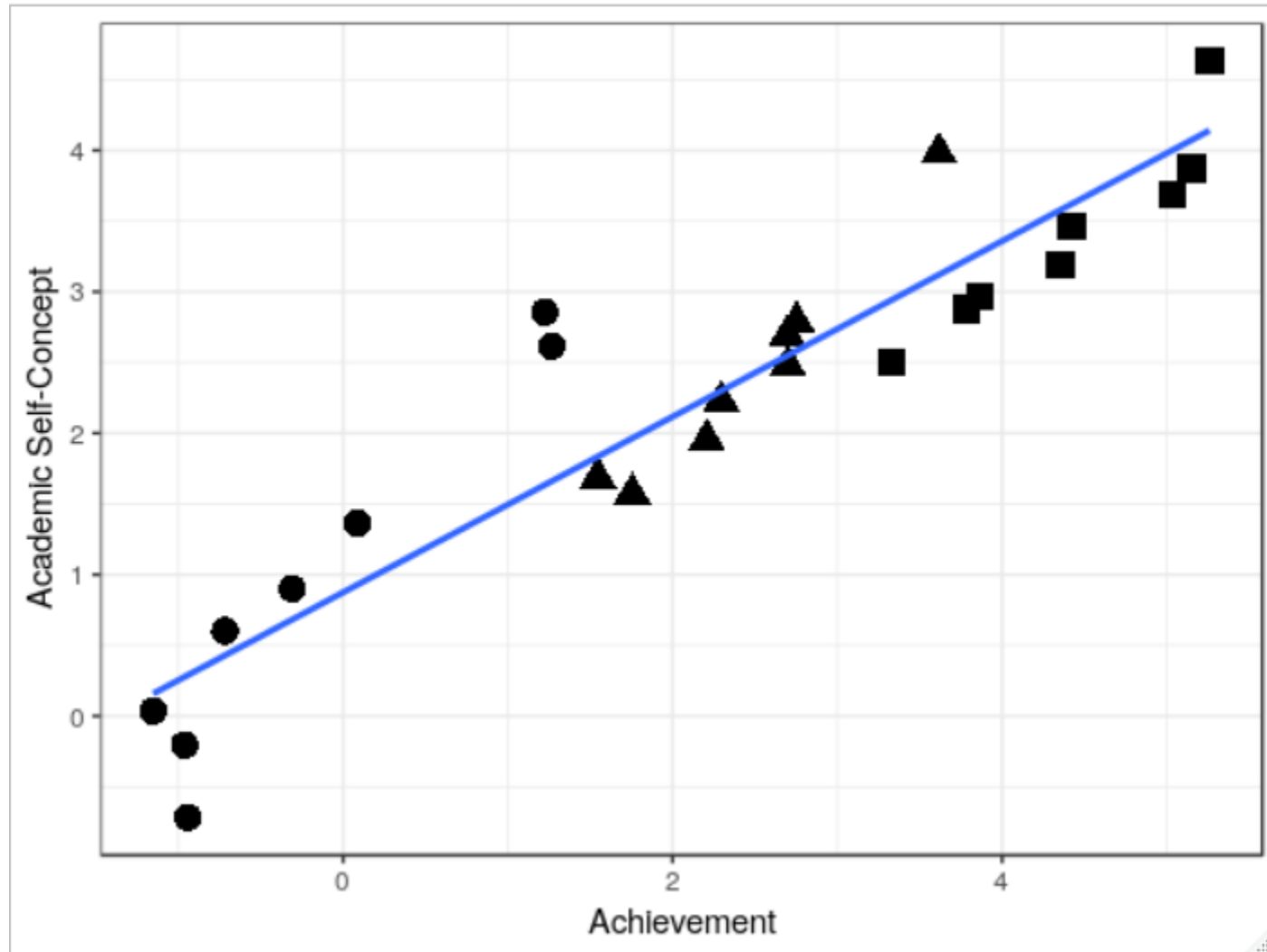


# “Same” Predictor, Different Effects

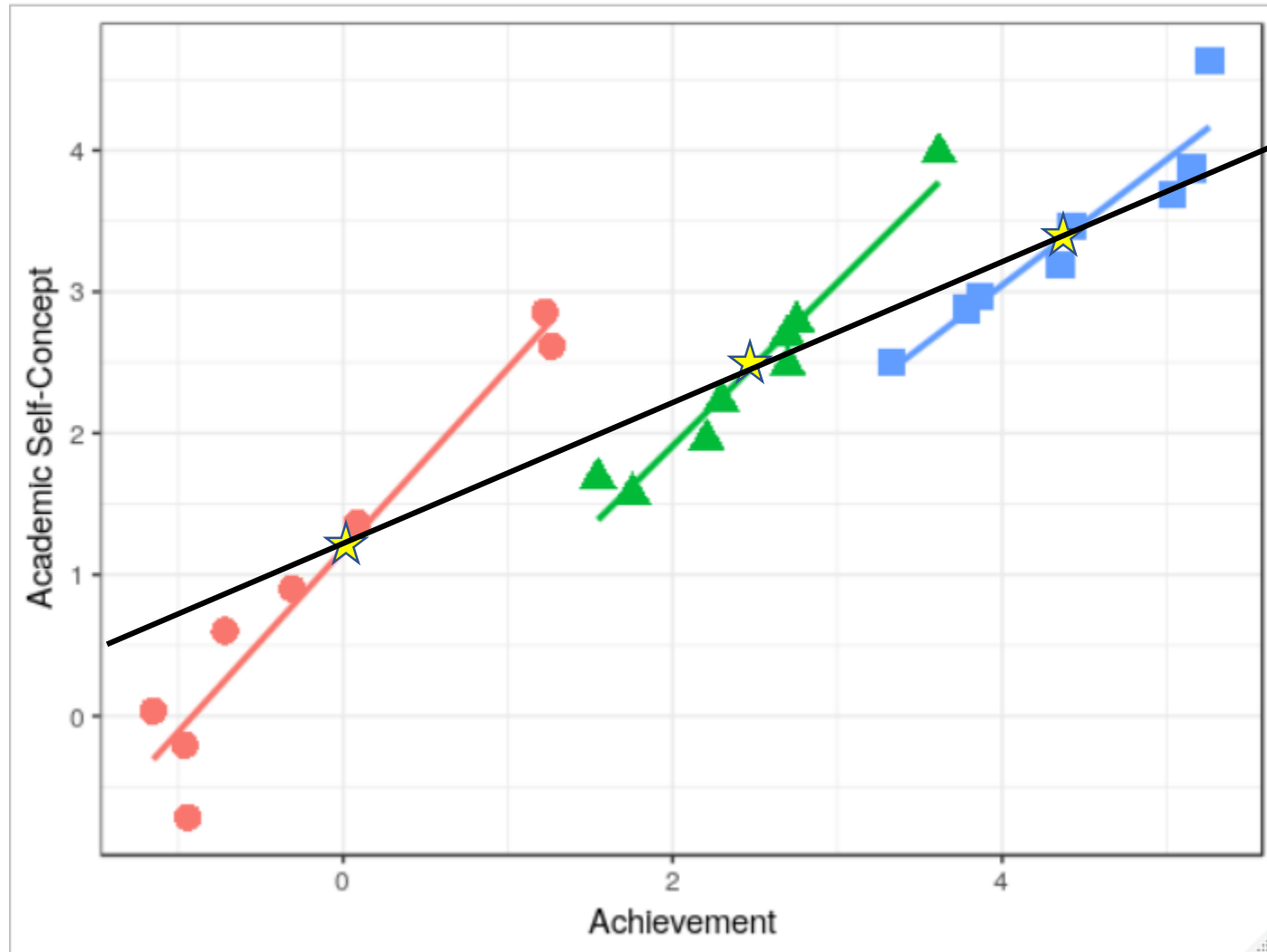
- Example: Big-Fish-Little-Pond Effect (Marsh & Parker, 1984)



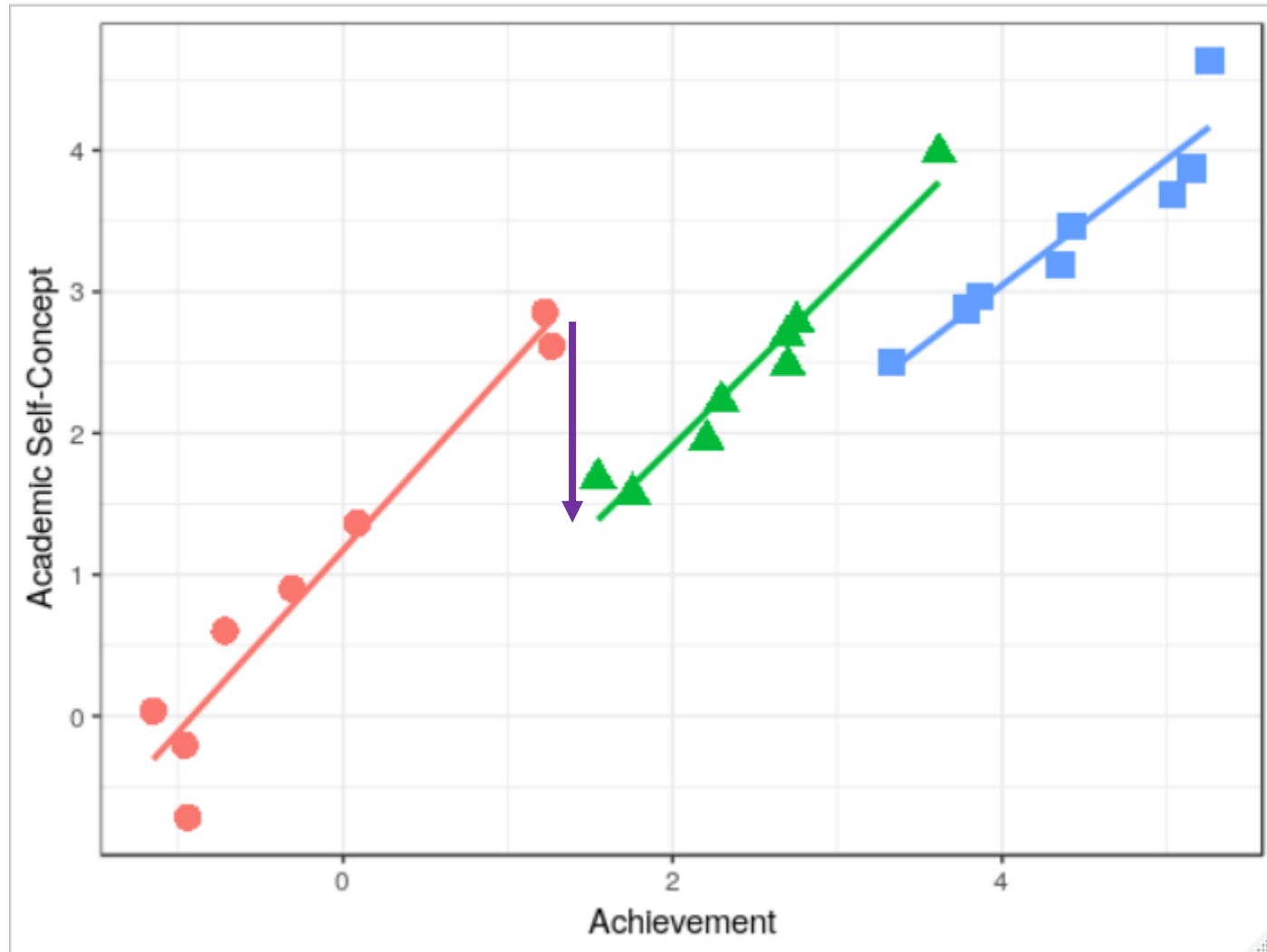
# Overall Effect



# Within & Between Effects



# Within & Contextual Effects



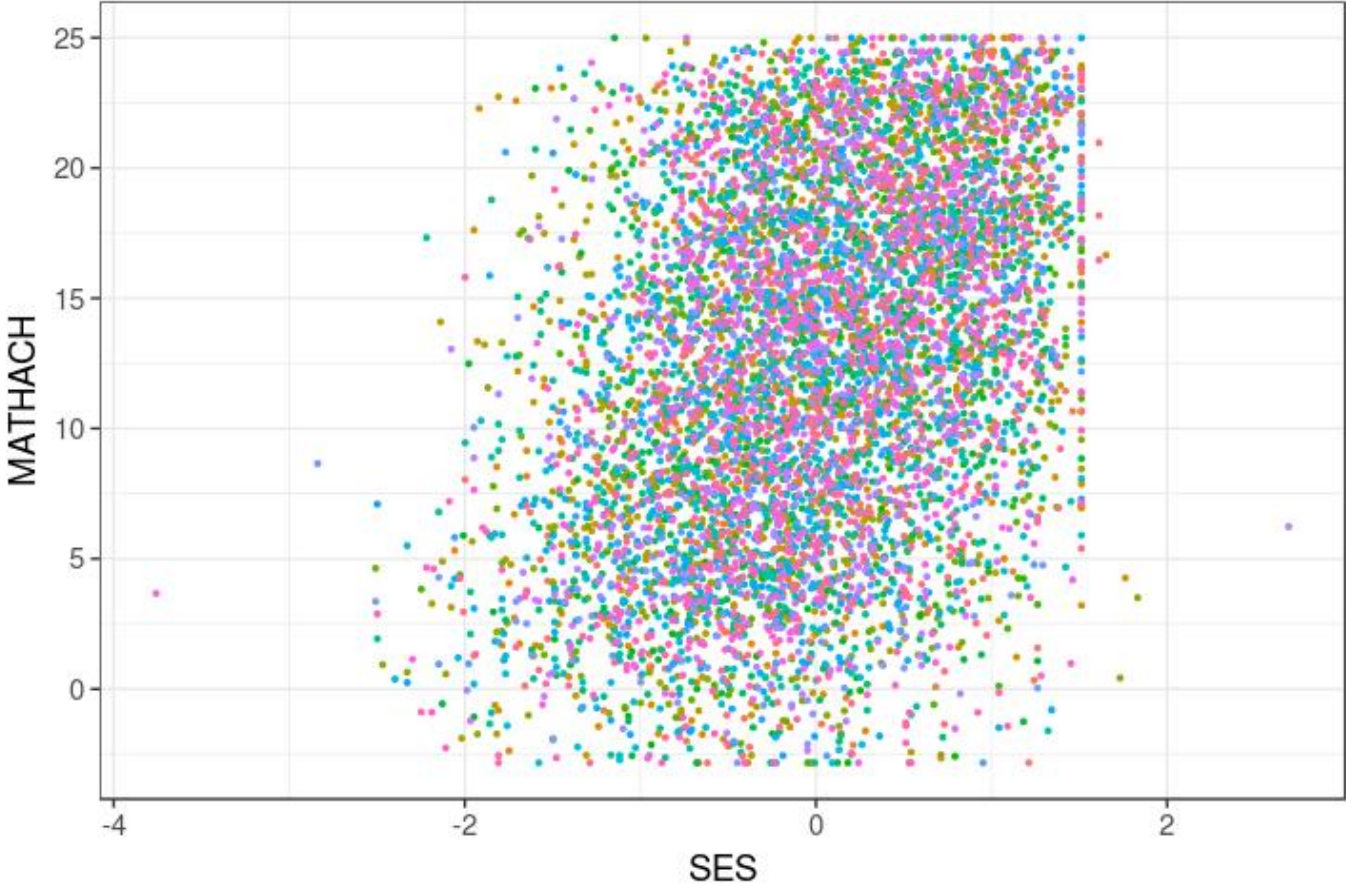
# Never simply include a level-1 predictor

Unless it has the same values for every cluster

# Two Approaches

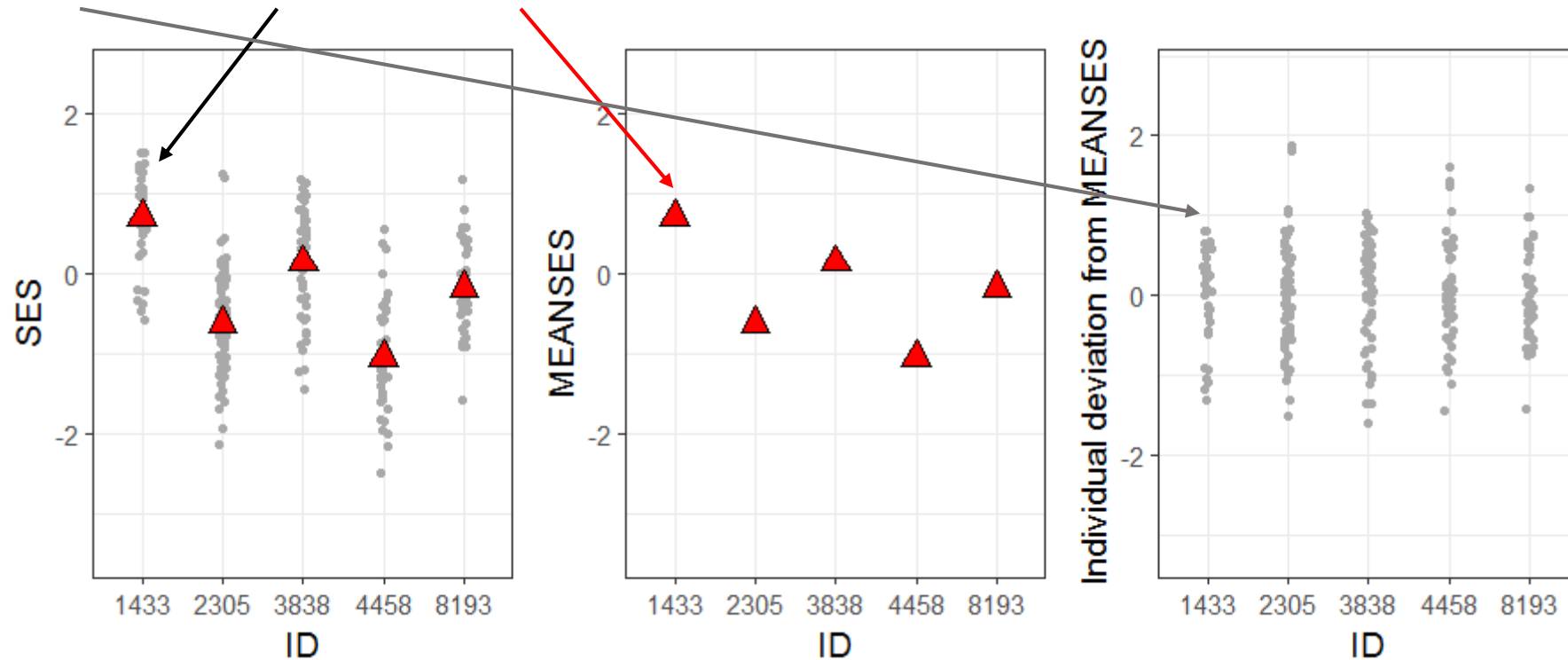
- Both involves computing the cluster means
  - E.g.,  $ses \rightarrow meases$
- 1. Cluster-mean centered (cmc) variable + cluster mean
  - Between-within method
  - Decompose into between-within effects
- 2. Raw/uncentered predictor + cluster mean
  - Study contextual effects (i.e., between minus within)

# mathach vs. ses



# Decomposing Into Lv-2 and Lv-1 Components

- Group-mean centering
  - $ses\_cmc = ses_{ij} - meanses_j$





# Between-Within Decomposition

- Lv 1:  

$$\text{mathach}_{ij} = \beta_{0j} + \beta_{1j} \text{ses\_cmc}_{ij} + e_{ij}$$
- Lv 2:  

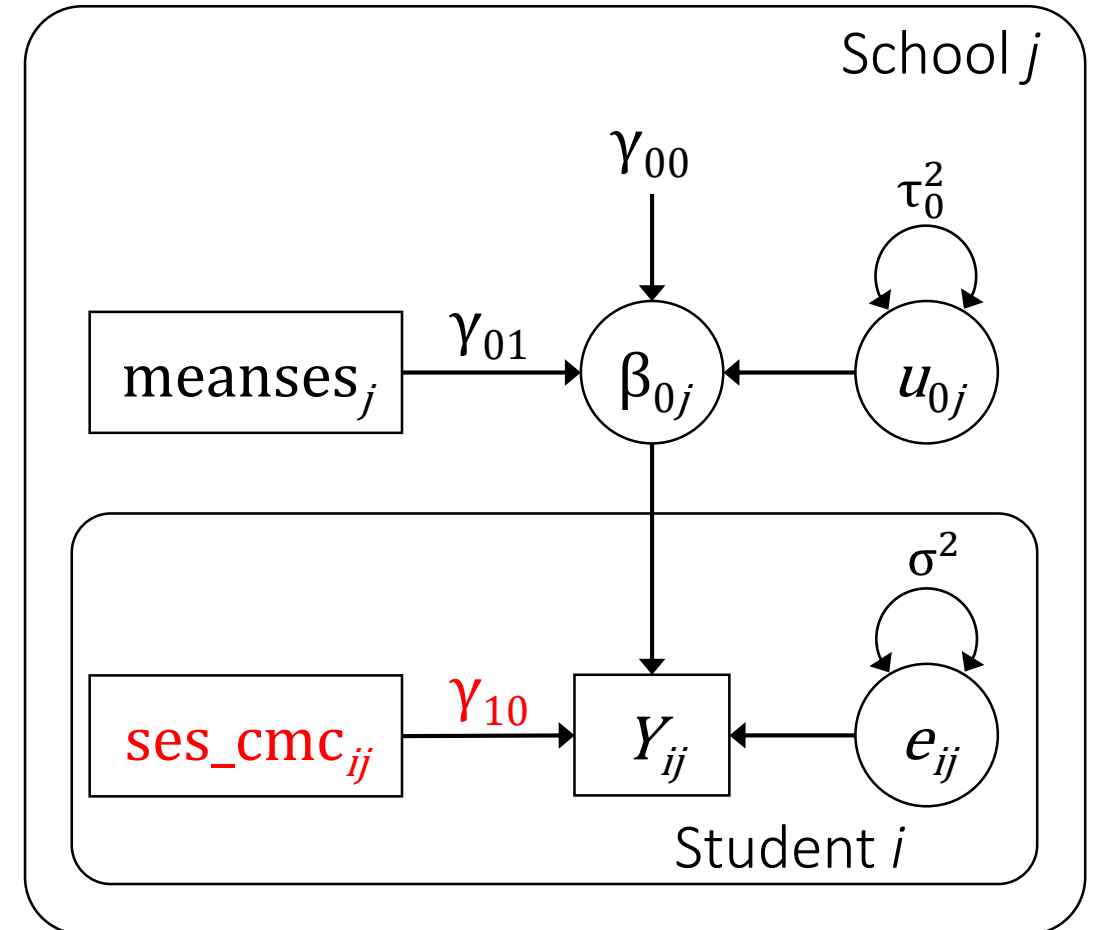
$$\beta_{0j} = \gamma_{00} + \gamma_{01} \text{meanses}_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10}$$
- Combined:  

$$\text{mathach}_{ij} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{10} \text{ses\_cmc}_{ij} + u_{0j} + e_{ij}$$

Student-level  
Effect

School-level  
Effect



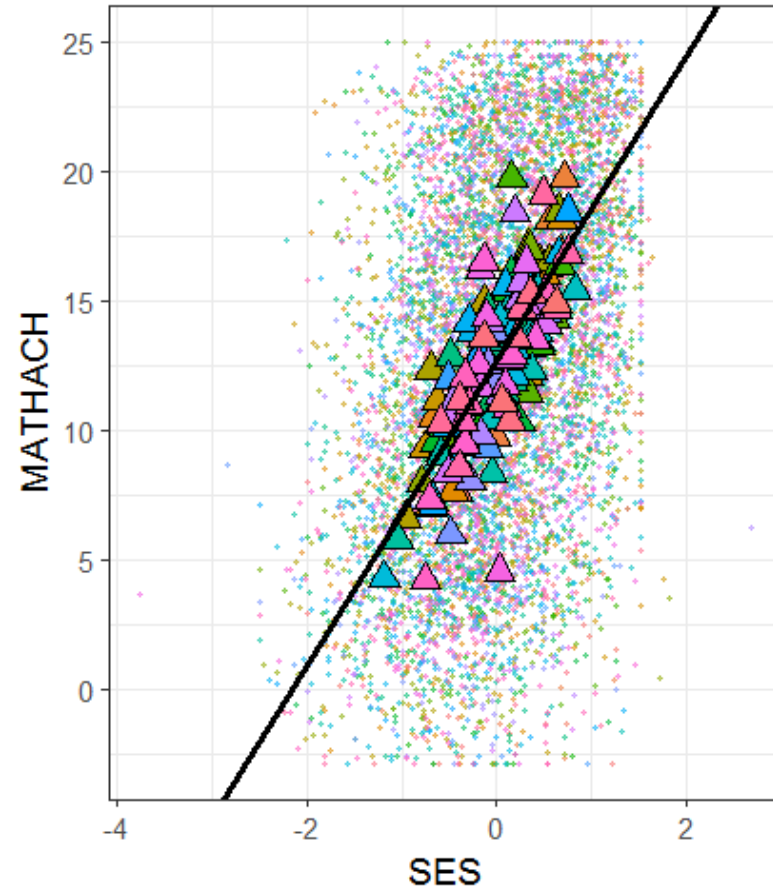
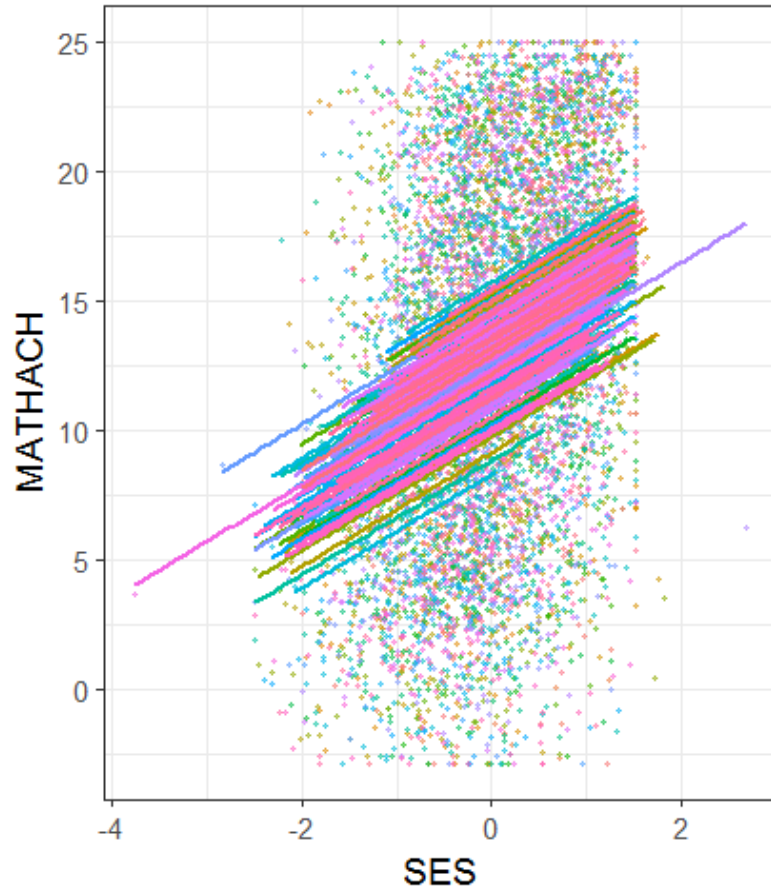
```
># Linear mixed model fit by REML ['lmerMod']
># Formula: mathach ~ meanses + ses_cmc + (1 | id)
># Data: hsball
```

```
># Fixed effects:
```

```
>#           Estimate Std. Error t value
># (Intercept) 12.6481    0.1494   84.68
># meanses      5.8662    0.3617   16.22
># ses_cmc      2.1912    0.1087   20.16
```

The student-level effect is 2.19  
The school-level effect is 5.87

# Visualizing the Difference



# Interpret the Coefficients

- Student A

- From a school of average SES

- SES level at the school mean

- $ses = \underline{\quad}$ ,  $meansas = \underline{\quad}$ ,  $ses\_cmc = \underline{\quad}$

- Predicted mathach =  $\underline{\quad} + \underline{\quad} (\underline{\quad}) + \underline{\quad} (\underline{\quad})$   
=  $\underline{\quad}$

# Interpret the Coefficients

- Student B
  - From a school of average SES
  - SES level 1 unit higher than the school mean
  - $\text{meanses} = \underline{\hspace{1cm}}$ ,  $\text{ses\_cmc} = \underline{\hspace{1cm}}$
- Predicted mathach =  $\underline{\hspace{1cm}} + \underline{\hspace{1cm}} (\underline{\hspace{1cm}}) + \underline{\hspace{1cm}} (\underline{\hspace{1cm}})$   
=  $\underline{\hspace{1cm}}$

# Interpret the Coefficients (Cont'd)

- Student C

- From a high SES school (one unit higher than average)
- SES level 1 unit below the school mean

→  $meanses = \underline{\hspace{1cm}}$ ,  $ses\_cmc = \underline{\hspace{1cm}}$

- Predicted mathach =  $\underline{\hspace{1cm}} + \underline{\hspace{1cm}} (\underline{\hspace{1cm}}) + \underline{\hspace{1cm}} (\underline{\hspace{1cm}})$   
=  $\underline{\hspace{1cm}}$

# Contextual Effects

# Contextual Effect<sup>1</sup>

- $\gamma_{01} - \gamma_{10} = 5.87 - 2.19 = 3.68$
- Effect of School SES (context) on individuals:
  - Expected difference in achievement between two students with same SES, but from schools with a 1 unit difference in meanses

[1]: When there is no random slopes, the contextual effect model is a reparameterization of the between-within model, meaning that they have the same fit



```
># Linear mixed model fit by REML ['lmerMod']
># Formula: mathach ~ meanses + ses + (1 | id)
># Data: hsball
```

```
># Fixed effects:
```

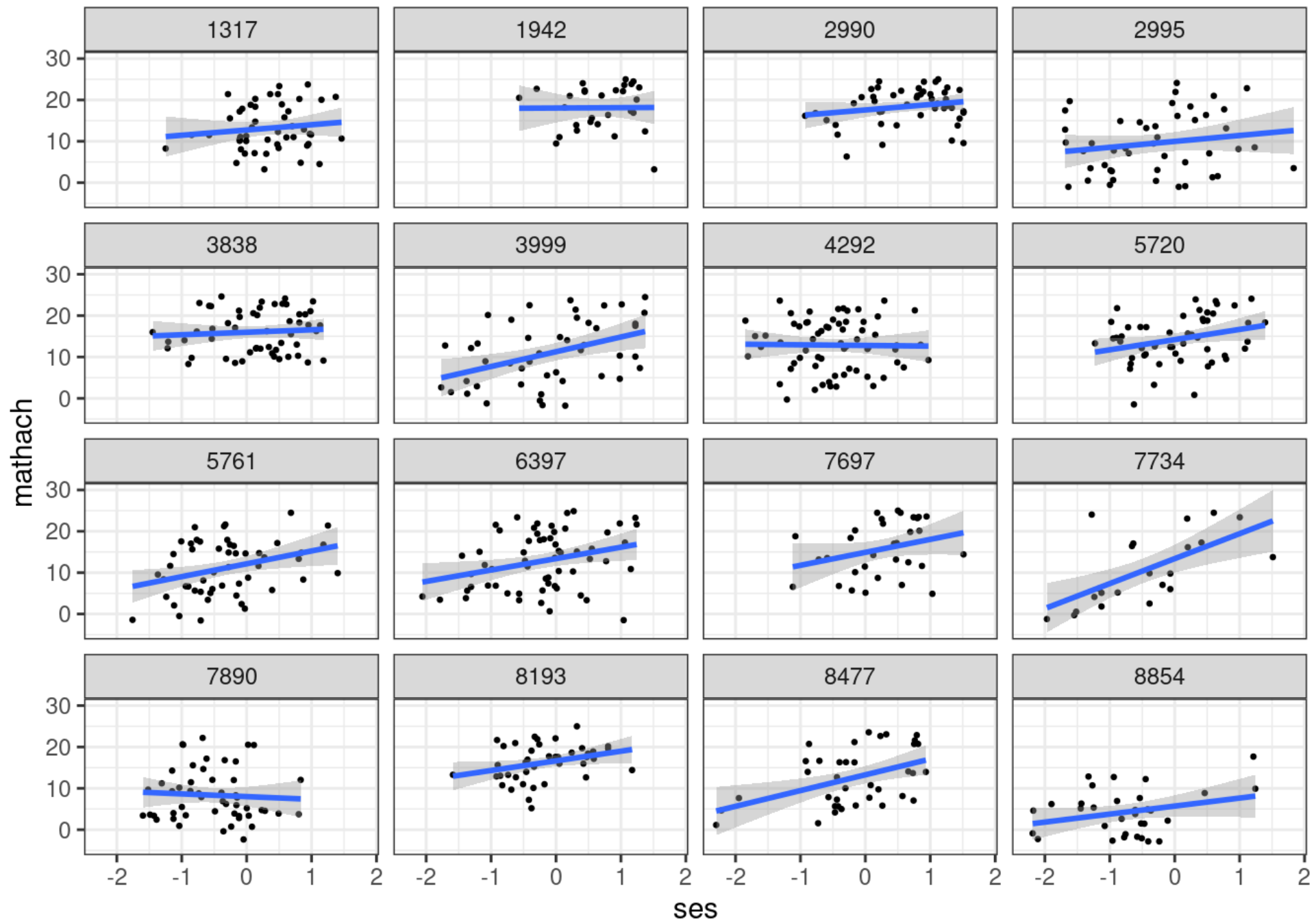
```
>#           Estimate Std. Error t value
># (Intercept) 12.6613    0.1494  84.763
># meanses     3.6750    0.3777   9.731
># ses         2.1912    0.1087  20.164
```

The student-level effect is 2.19;  
the contextual effect  
= 3.68 = 5.87 - 2.19

# Random Slopes/Random Coefficients

# Research Questions

- Does math achievement varies across schools? How much is the variation?
- Do schools with higher mean SES have students with higher math achievement?
- Do students with higher SES have higher math achievement? Is the relation similar at the individual and cluster levels? **Is this relation similar across schools?**
- Is the relation between SES and math achievement moderated by some types of schools (e.g., Catholic vs. Public, high mean SES vs low mean SES)?

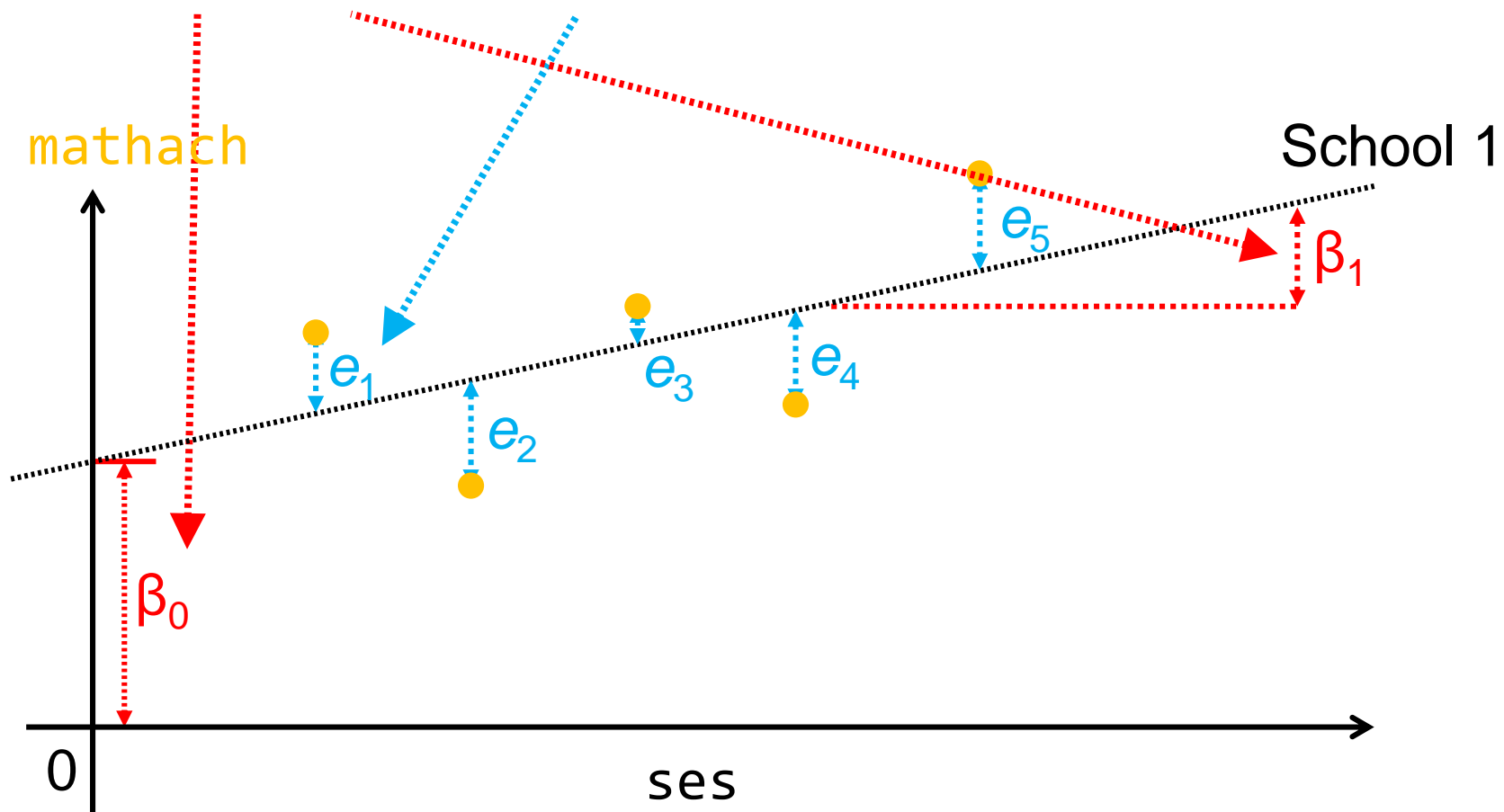


# Varying Regression Lines

- Decomposing effect model
  - Assumes constant slope across schools for  $ses \rightarrow mathach$
- Instead, one can investigate whether that relation changes across schools

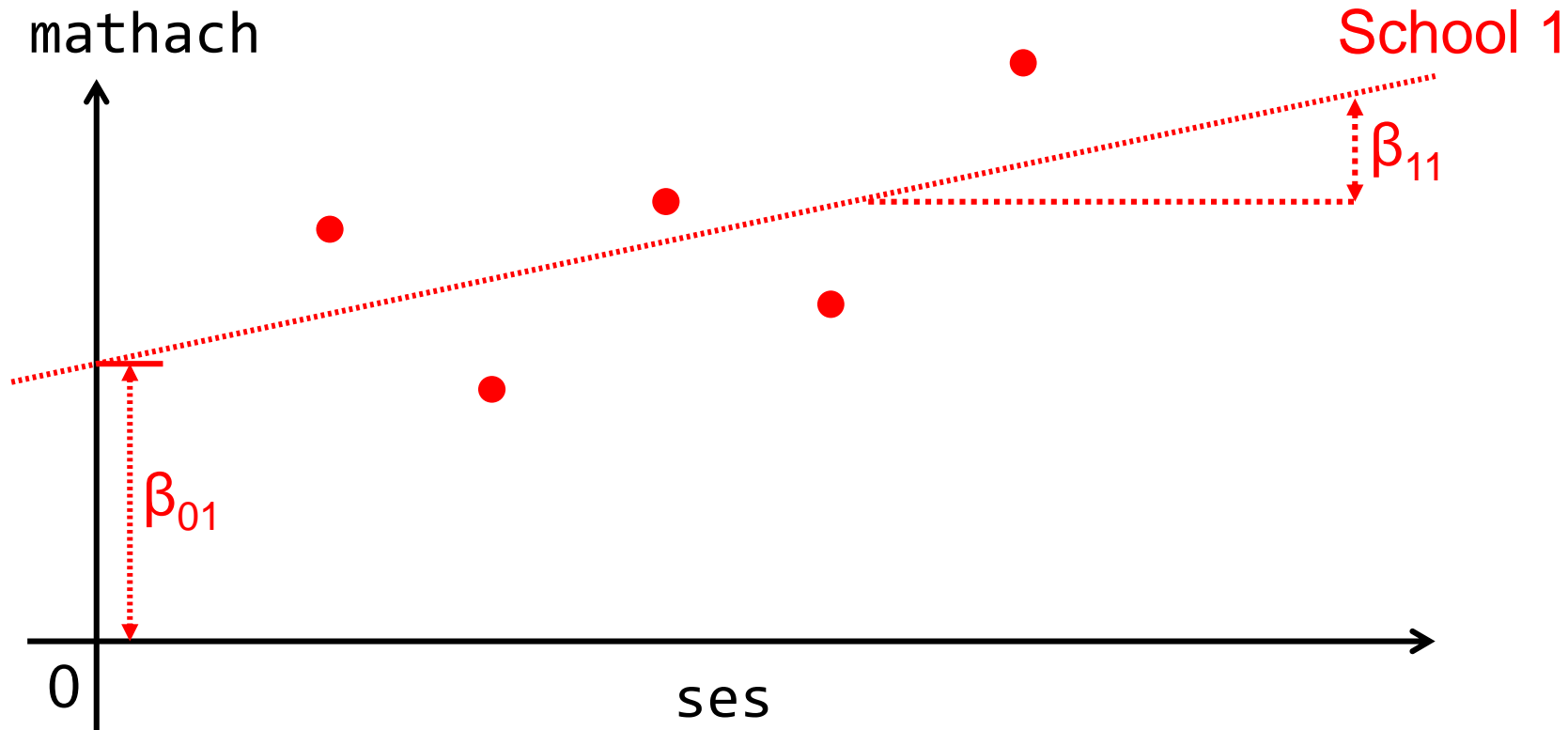
# Let's Focus on One School

- $\text{mathach}_i = \beta_0 + \beta_1 \text{ses}_i + e_i$



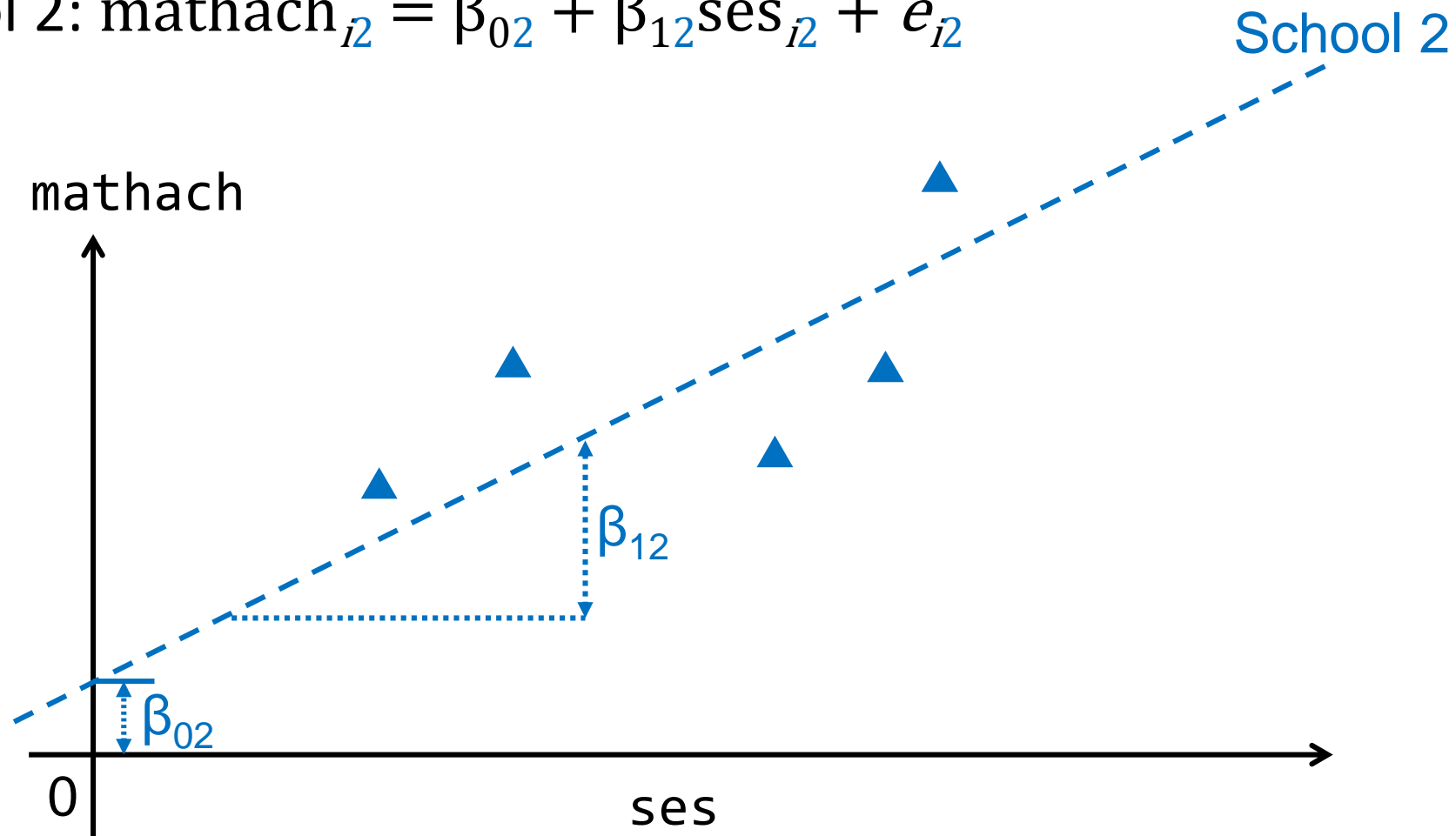
# Multi-Level Model (MLM)

- School 1:  $\text{mathach}_{i1} = \beta_{01} + \beta_{11}\text{ses}_{i1} + e_{i1}$



# Consider a Second School

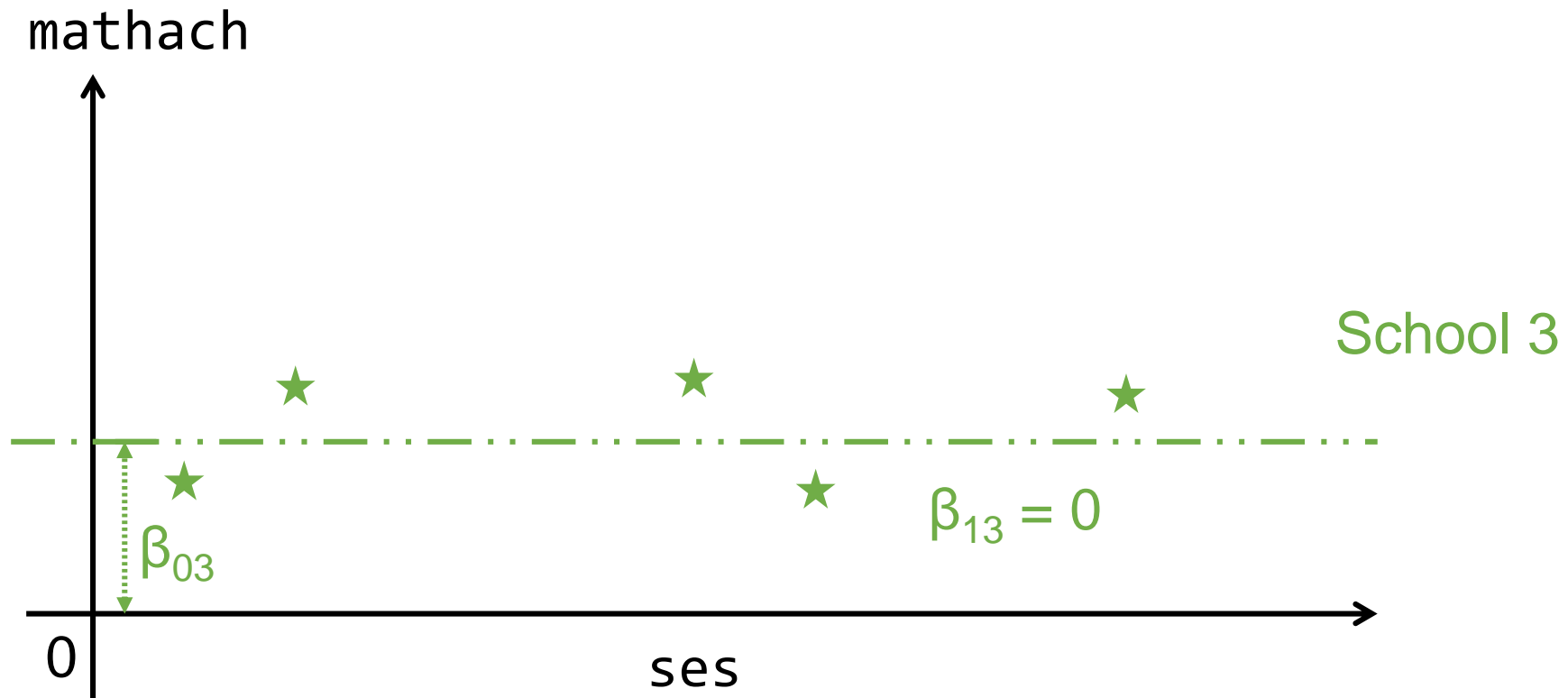
- School 2:  $\text{mathach}_{i2} = \beta_{02} + \beta_{12}\text{ses}_{i2} + e_{i2}$



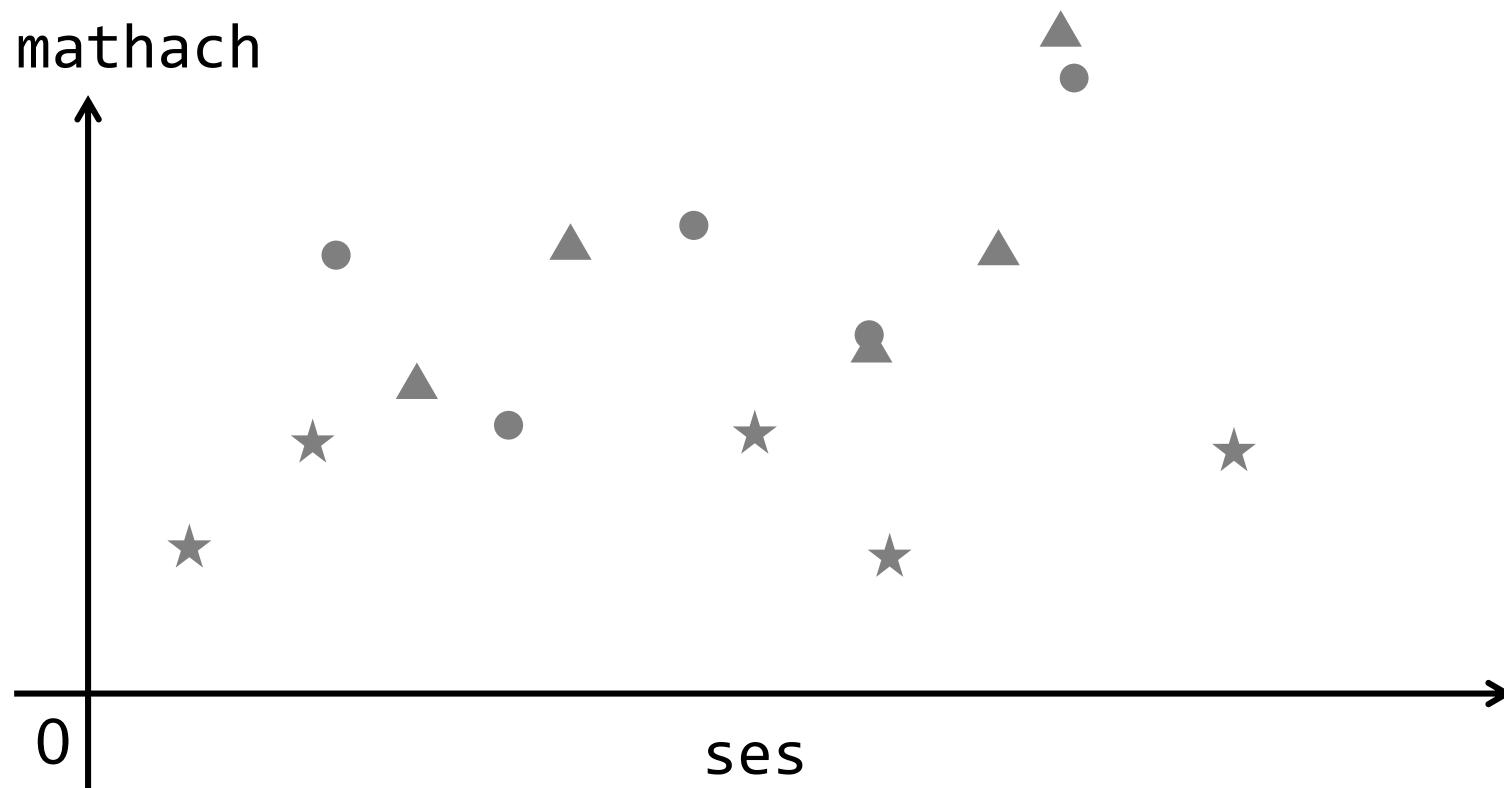


# Consider a Third School

- School 3:  $\text{mathach}_{i3} = \beta_{03} + \beta_{13}\text{ses}_{i3} + e_{i3}$

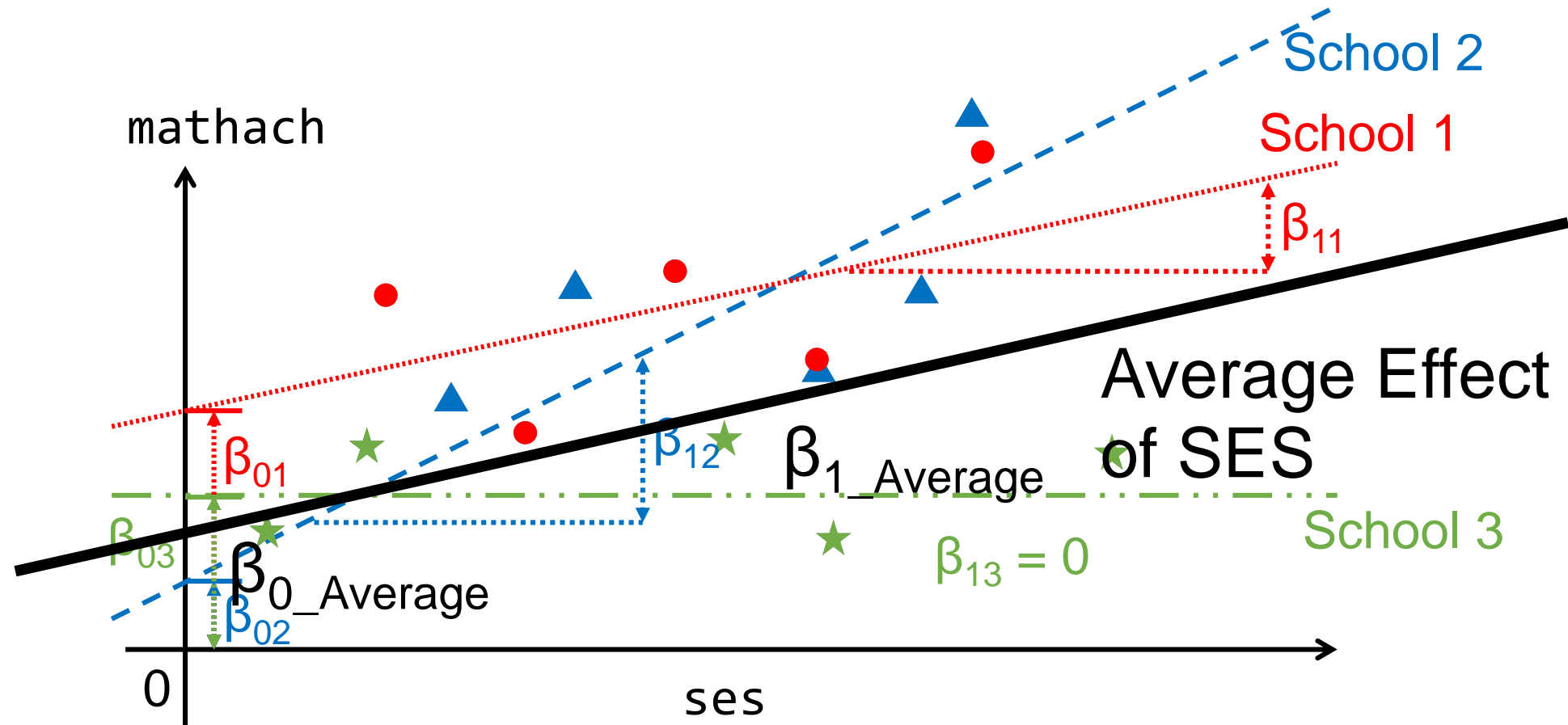


# Combining All Schools



# Combining All Schools

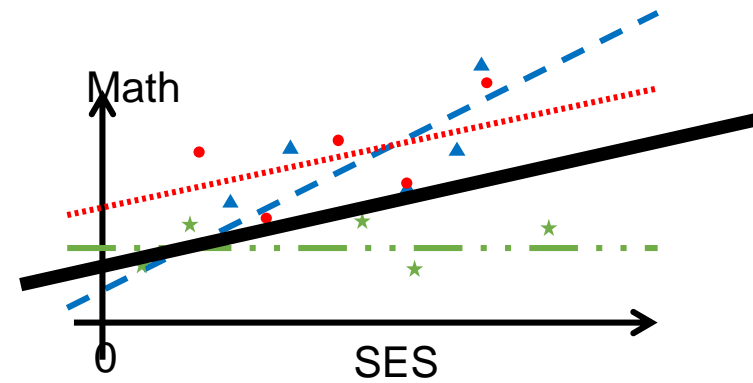
- $\text{mathach}_{ij} = \beta_{0j} + \beta_{1j}\text{ses}_{ij} + e_{ij} \quad (j = 1, 2, \dots, 160)$



# Combining All Schools

160  
Schools

School	$\beta_{0j}$	$\beta_{1j}$
1224	11.06	2.50
1288	13.07	2.48
1296	9.20	2.35
1308	14.38	2.31
...		
9397	10.40	1.87
9508	13.69	2.52
9550	11.29	2.67
9586	13.37	2.27
Mean	13.01	2.39
Variance	4.83	0.41



Random  
Intercepts

Random  
Slopes

$$\beta_{1\_Average} = \gamma_{10}$$

$$\beta_{0\_Average} = \gamma_{00}$$

# Random-Coefficient Model

- Lv 1:

- $\text{mathach}_{ij} = \beta_{0j} + \beta_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

- Lv 2:

- $\beta_{0j} = \gamma_{00} + \gamma_{01} \text{meanses}_j + u_{0j}$

- $\beta_{1j} = \gamma_{10} + u_{1j}$

- Combined:

- $\text{mathach}_{ij} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{10} \text{ses\_cmc}_{ij} + u_{0j} + u_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

Deviation of school  $j$ 's slope from the average

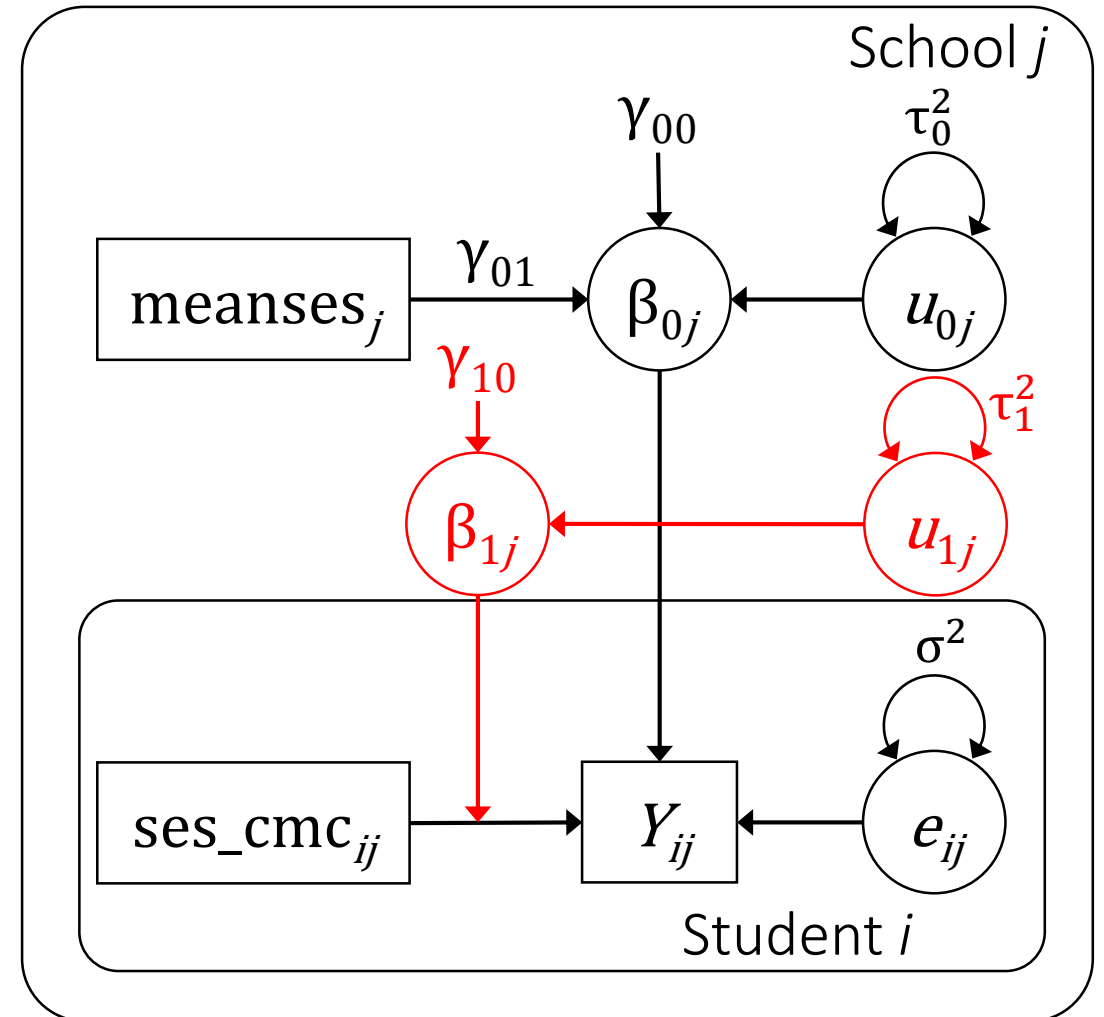
Average slope of SES

# Centering

- Raudenbush & Bryk (2002) noted that slope variance were better estimated with cluster mean centering
  - However, Snijders & Bosker (5.3.1) suggested it should be based on theory
- Remember to add the cluster means
- See also consult Enders & Tofighi (2007)<sup>1</sup>

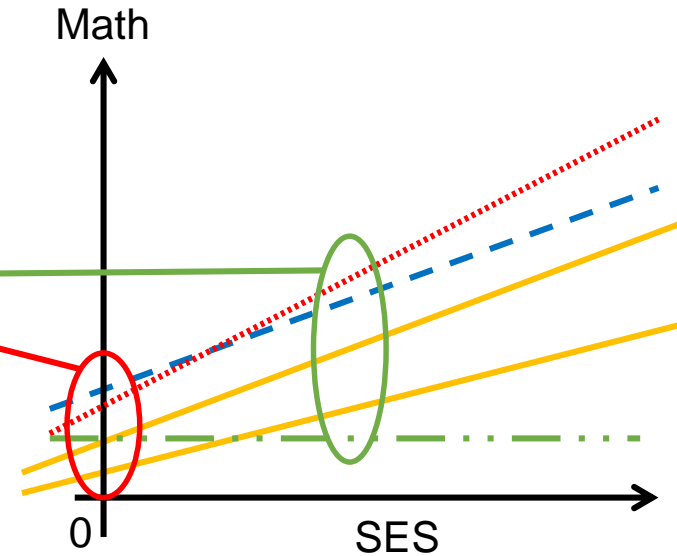
[1]: <https://doi.org/10.1037/1082-989X.12.2.121>

# Path Diagram



# Variance Components

- $\text{Var}(u_{0j}) = \tau_0^2$
- $\text{Var}(u_{1j}) = \tau_1^2$



$$\text{Var} \begin{pmatrix} u_{0j} \\ u_{1j} \end{pmatrix} = \mathbf{G} = \begin{bmatrix} \tau_0^2 & \tau_{01} \\ \tau_{01} & \tau_1^2 \end{bmatrix}$$

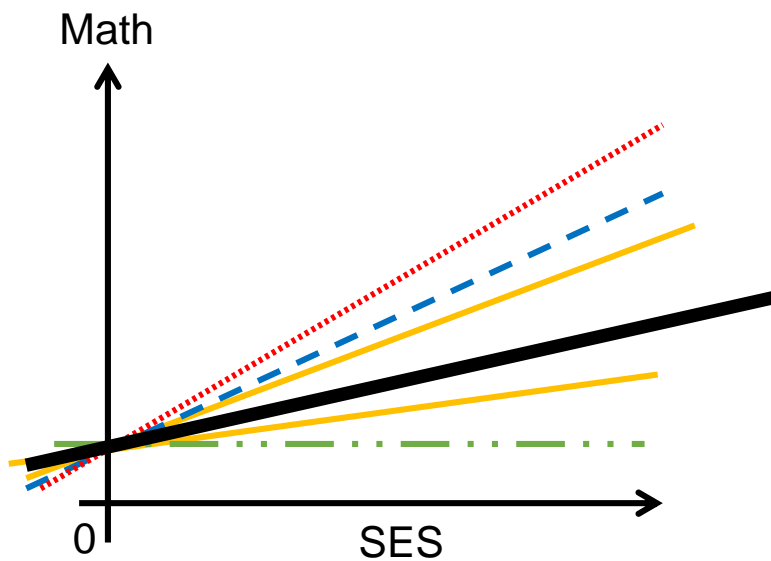
Variance of the school intercepts

Covariance of the intercepts and slopes, which are *seldom interpreted*

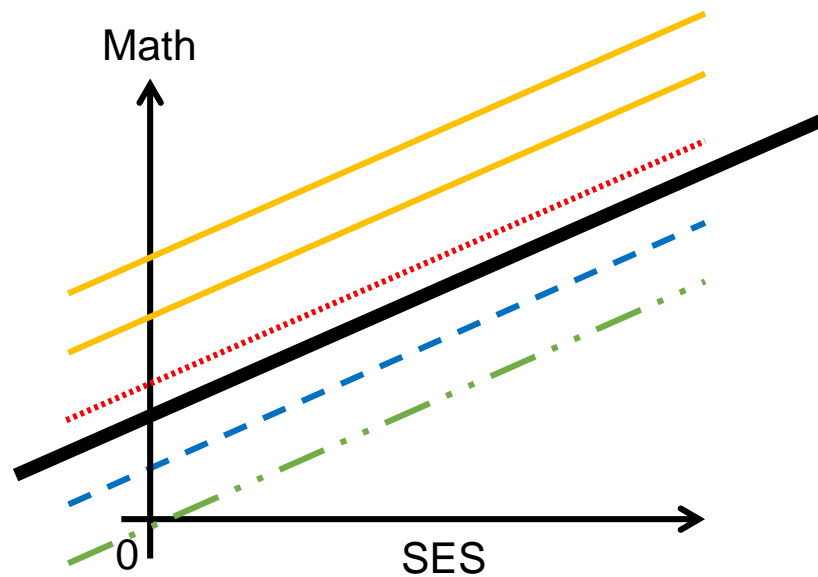
Variance of the school slopes



No random intercepts  
 $\text{Var}(u_{0j}) = \tau_0^2 = 0$



No random slopes  
 $\text{Var}(u_{1j}) = \tau_1^2 = 0$



# Full Equations

$$\text{mathach}_{ij} = \gamma_{00} + \gamma_{01}\text{meanses}_j + \gamma_{10}\text{ses\_cmc}_{ij} \\ + u_{0j} + u_{1j}\text{ses\_cmc}_{ij} + e_{ij}$$

$$\begin{pmatrix} u_{0j} \\ u_{1j} \end{pmatrix} \sim N \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \tau_0^2 & \tau_{01} \\ \tau_{01} & \tau_1^2 \end{bmatrix} \right)$$
$$e_{ij} \sim N(0, \sigma)$$

# Look at the *SEs* of Fixed Effects

```
> lmer(mathach ~ meanses + ses_cmc + (ses_cmc | id), data = hsball)
```

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	12.6454	0.1492	84.74
meanses	5.8963	0.3600	16.38
ses_cmc	2.1913	0.1280	17.12

*SE* = 0.109 when random slopes not included  
→ underestimated

# Random Effect Estimates

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
id	(Intercept)	2.6931	1.6411	
	ses_cmc	0.6858	0.8282	-0.19
Residual		36.7132	6.0591	

Number of obs: 7185, groups: id, 160

- $\tau_0^2 = 2.69 =$  variance of intercepts
- $\tau_1^2 = 0.69 =$  slope variance

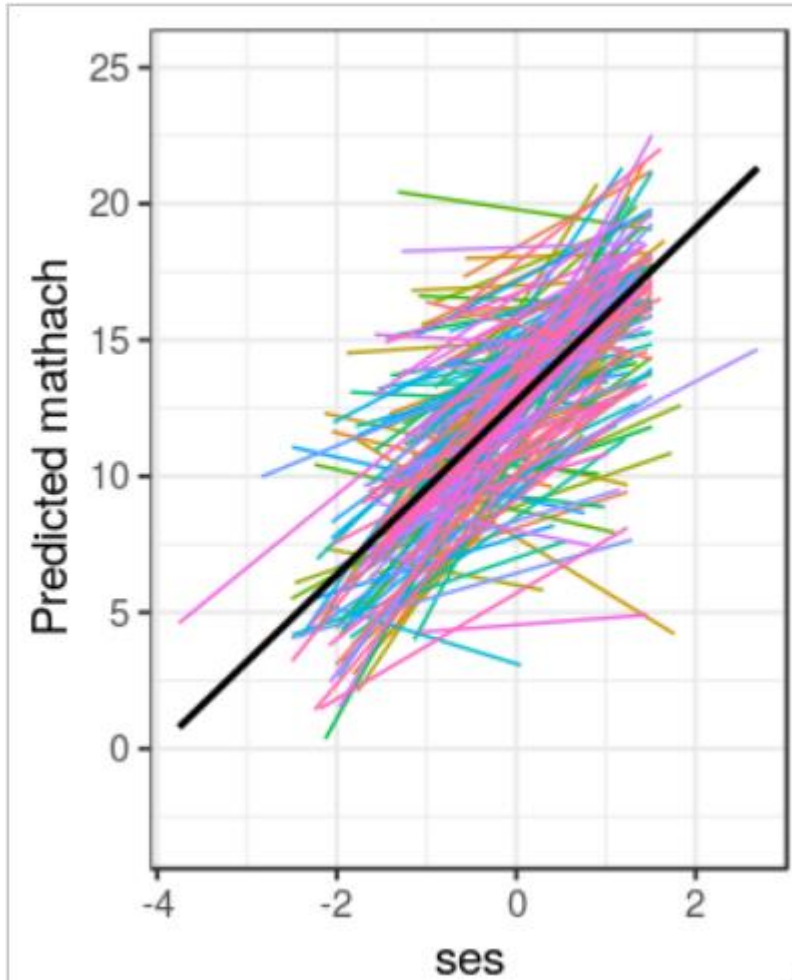
# Interpreting Random Slopes

- Average slope =  $\gamma_{10} = 2.19$
- *SD* of slopes =  $\tau_1 = 0.83$
- 68% Plausible range
  - $\gamma_{10} \pm \tau_1 = [\gamma_{10} - \tau_1, \gamma_{10} + \tau_1]$   
= [\_\_\_\_, \_\_\_\_]

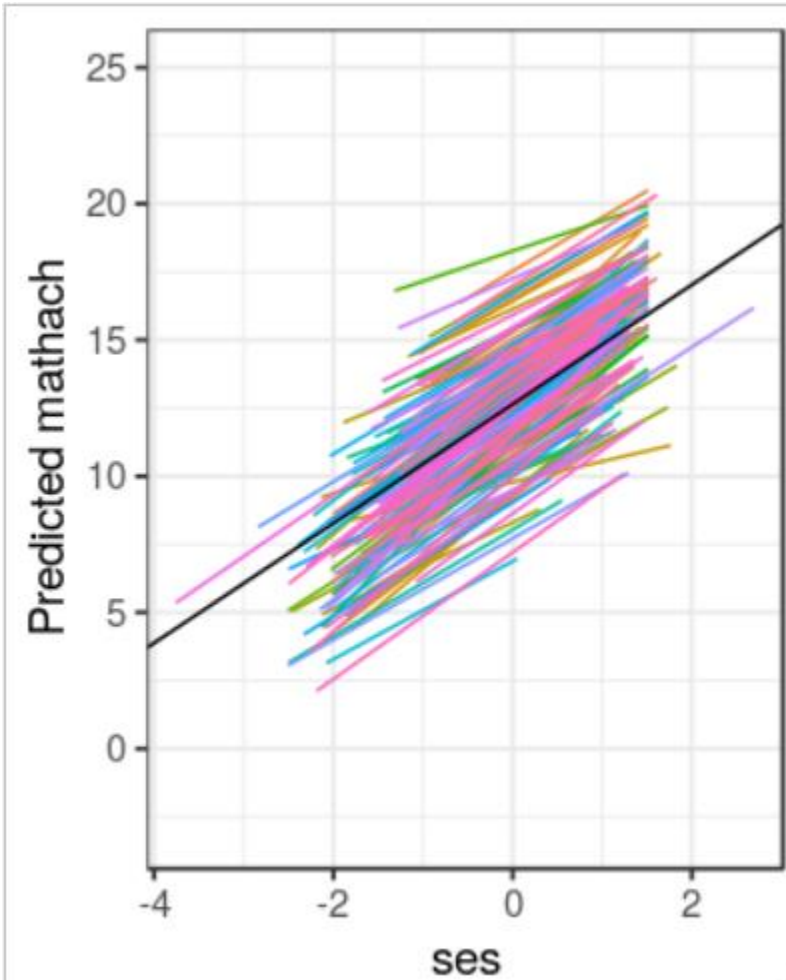
For majority of schools, SES and achievement are positively associated, with regression coefficients between \_\_\_\_ and \_\_\_\_

# Visualize the Varying Slopes

OLS



Shrinkage (EB)



# Cross-Level Interaction

# Research Questions

- Does math achievement vary across schools? How much is the variation?
- Do schools with higher mean SES have students with higher math achievement?
- Do students with higher SES have higher math achievement? Is the relation similar at the individual and cluster levels? Is this relation similar across schools?
- Is the relation between SES and math achievement moderated by some types of schools (e.g., Catholic vs. Public, high mean SES vs low mean SES)?



# Cross-Level Interaction

- Whether school-level variables moderate student-level relationships between variables
- Also called an intercepts and slopes-as-outcomes model
- Let's add another school-level variable: sector
  - 1 = Catholic ( $n = 70$ ), 0 = Public ( $n = 90$ )

# Model Equations

- Lv 1:

- $\text{mathach}_{ij} = \beta_{0j} + \beta_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

- Lv 2:

- $\beta_{0j} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{02} \text{sector}_j + u_{0j}$

- $\beta_{1j} = \gamma_{10} + \gamma_{11} \text{sector}_j + u_{1j}$

- Combined:

- $\text{mathach}_{ij} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{10} \text{ses\_cmc}_{ij} + \gamma_{02} \text{sector}_j + \gamma_{11} \text{sector}_j \times \text{ses\_cmc}_{ij} + u_{0j} + u_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

Main Effect of  
SECTOR

Cross-level product  
(interaction) term

# Model Equations (cont'd)

- Lv 1:

- $\text{mathach}_{ij} = \beta_{0j} + \beta_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

- Lv 2:

- $\beta_{0j} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{02} \text{sector}_j + u_{0j}$

- $\beta_{1j} = \gamma_{10} + \gamma_{11} \text{sector}_j + u_{1j}$

- Combined:

- $\text{mathach}_{ij} = \gamma_{00} + \gamma_{01} \text{meanses}_j + \gamma_{10} \text{ses\_cmc}_{ij} + \gamma_{02} \text{sector}_j + \gamma_{11} \text{sector}_j \times \text{ses\_cmc}_{ij} + u_{0j} + u_{1j} \text{ses\_cmc}_{ij} + e_{ij}$

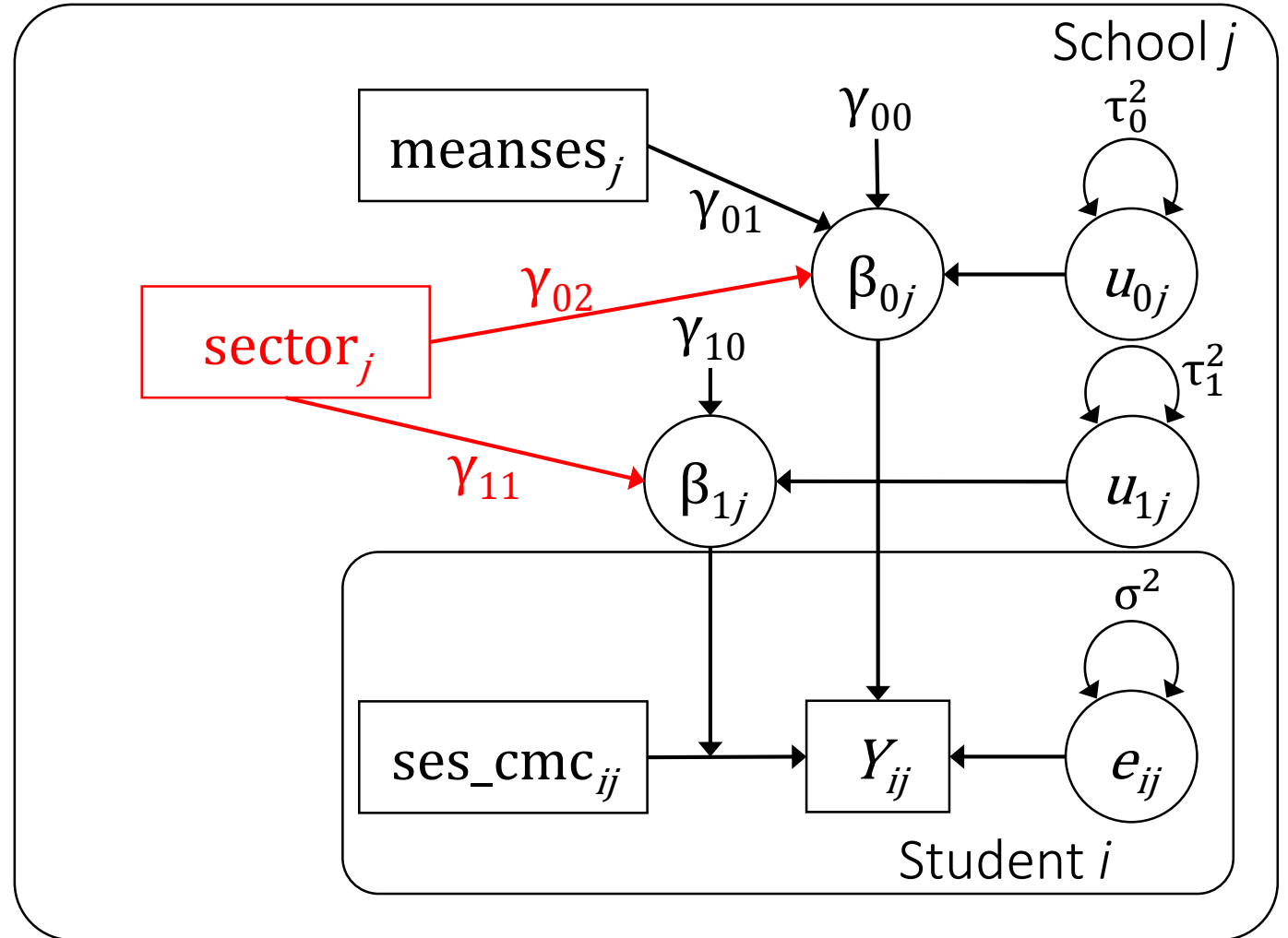
Deviation of intercept for School  $j$

Deviation of slope for School  $j$

# Path Diagram

$$\begin{pmatrix} u_{0j} \\ u_{1j} \end{pmatrix} \sim N \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \tau_0^2 & \tau_{01} \\ \tau_{01} & \tau_1^2 \end{bmatrix} \right)$$

$$e_{ij} \sim N(0, \sigma)$$



# Fixed Effect Estimates

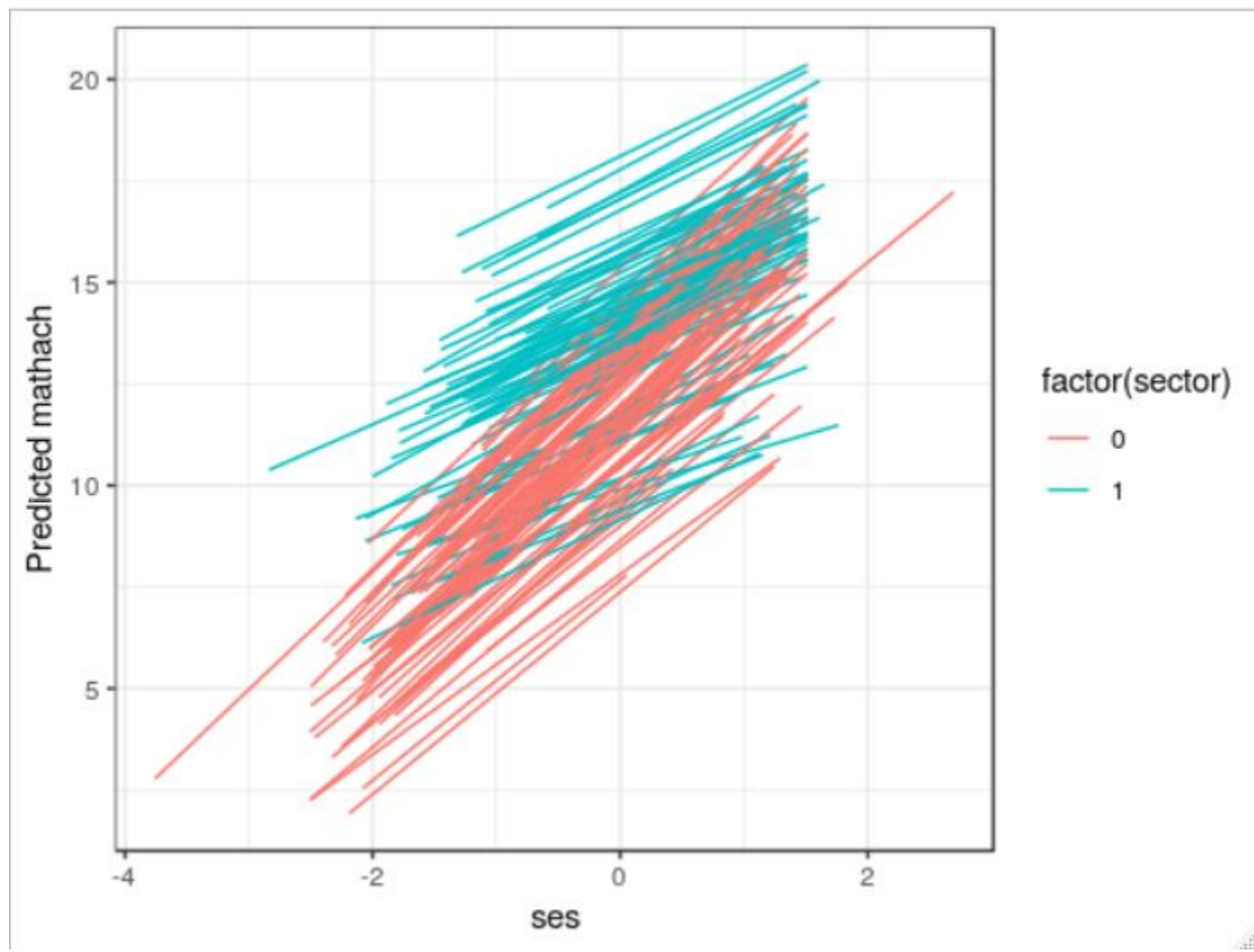
Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	12.0846	0.1987	60.81
meanes	5.2450	0.3682	14.24
sectorCatholic	1.2523	0.3062	4.09
ses_cmc	2.7877	0.1559	17.89
sectorCatholic:ses_cmc	-1.3478	0.2348	-5.74

Average slope for SES is estimated as **2.79** for Public schools (i.e., sector = 0)

Average slope for SES is estimated as **2.79** – **1.35** = **1.44** for Catholic schools (i.e., sector = 1)

# Plot the Interaction



# Things to Remember

- A level-1 predictor can have differential relationships with the outcome, depending on the level of analysis
  - **Ecological fallacy**: assume constant relationship across levels
- **Cluster/group-mean centering**: decompose a level-1 predictor into its cluster means and deviations from the cluster means
- MLM provides a way to efficiently model variability of regression lines (i.e., intercepts and slopes) across clusters
  - Through the use of **random slopes/coefficients**
- Cross-level interaction  
= Including a lv-2 predictor in the slope equation