

Two-Level Models for Cross-Classified Data

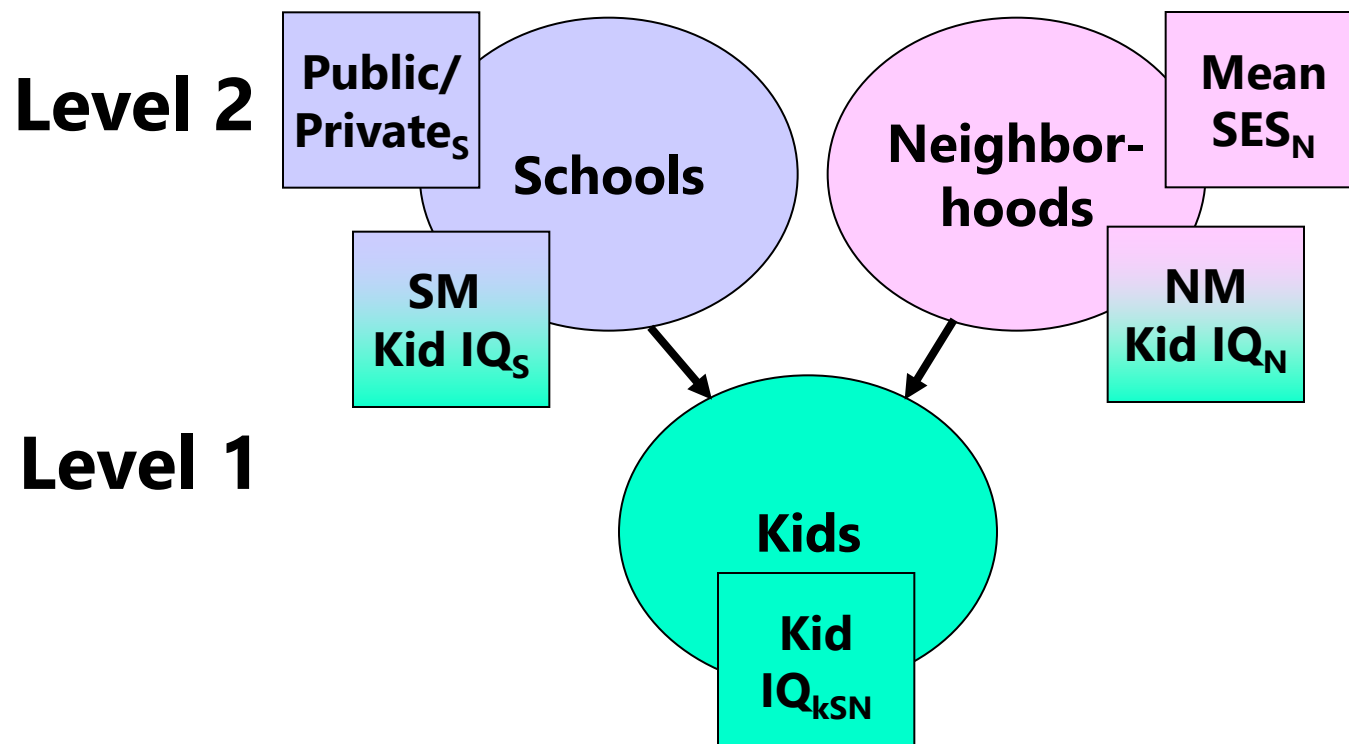
- Topics:
 - Cross-sectional cross-classification (time-invariant groups)
 - Longitudinal cross-classification (time-varying groups)
 - No smushing allowed in cross-classified models, either

More Complex Multilevel Designs

- Multilevel models are specified based on the relevant dimensions by which observations differ each other, and how the units are organized
- Two-level models have at least two piles of variance, in which level-1 units are nested within level-2 units:
 - Longitudinal Data: Occasions nested within Persons
 - Clustered Data: Students nested within Teachers
- Three-level models have at least three piles of variance, in which level-2 units are nested within level-3 units (stay tuned):
 - Longitudinal Data: Occasions nested within Persons within Families
 - Clustered Data: Students nested within Teachers within Schools
- In other designs, multiple sources of systematic variation may be present, but the sampling may be crossed instead...
 - Models with crossed random effects are known as “cross-classified” models in the context of clustered data
 - Here are a few examples on when this might happen...

Kids, Schools, and Neighborhoods

- Kids are nested within schools AND within neighborhoods
- Not all kids from same neighborhood live in same school, so schools and neighborhoods are crossed at level 2
- Can include predictors for each source of variation



Kids, Schools, and Neighborhoods

$$\begin{aligned} y_{ksn} = & \mathbf{Y}_{000} && \rightarrow \text{fixed intercept (all } x\text{'s} = 0) \\ & + \mathbf{Y}_{010}(\text{Private}_s) + \mathbf{Y}_{020}(\text{SMIQ}_s) && \rightarrow \text{school effects} \\ & + \mathbf{Y}_{001}(\text{SES}_n) + \mathbf{Y}_{002}(\text{NMIQ}_n) && \rightarrow \text{neighborhood effects} \\ & + \mathbf{Y}_{100}(\text{KidIQ}_{ksn}) && \rightarrow \text{kid effects} \\ & + \mathbf{U}_{0s0} && \rightarrow \text{random intercept of school} \\ & + \mathbf{U}_{00n} && \rightarrow \text{random intercept of neighborhood} \\ & + \mathbf{e}_{ksn} && \rightarrow \text{residual kid-to-kid variation} \end{aligned}$$

Because cross-classified models have two types of level-2 units, it can be much easier to write a single, composite equation

Time (t), Students (s), and Classes (c)

- Students are nested within Classes at each occasion...
- But if students move into different classes across time...
 - Time at level 1 is nested within Student AND within Classes
 - Student is crossed with Class at level 2
- How to model a time-varying random classroom effect?
 - This is the basis of so-called “value-added models”
- (At least) Two options via fixed or random effects:
 - Acute effect: Effect for class operates only when kids are in the class
 - e.g., Class effect \leftarrow teacher bias
 - Once a student is out of the class, class effect is no longer present
 - Transfer effect: Effect for class operates now and in the future...
 - e.g., Class effect \leftarrow differential learning
 - Effect stays with the student in the future (i.e., a “layered” value-added model)

Time (t), Students (s), and Classes (c)

- Custom-built intercepts for time-varying effects of classes
 - An intercept is usually a column of 1's, but ours will be 0's and 1's to serve as switches that turn on/off class effects

Student ID	Class ID	Grade	Year	Per-Year Class ID (-99 = missing)			Intercepts for Acute Effects			Intercepts for Transfer Effects		
				Year 0 Class	Year 1 Class	Year 2 Class	Year 0 Intercept	Year 1 Intercept	Year 2 Intercept	Year 0 Effect	Year 1 Effect	Year 2 Effect
101	1	3	0	1	-99	43	1	0	0	1	0	0
101	-99	4	1	1	-99	43	0	0	0	0	0	0
101	43	5	2	1	-99	43	0	0	1	1	0	1
102	3	3	0	3	21	42	1	0	0	1	0	0
102	21	4	1	3	21	42	0	1	0	1	1	0
102	42	5	2	3	21	42	0	0	1	1	1	1

Time (t), Students (s), and Classes (c)

- Hoffman (2015) Equation 11.3: fixed effects model for class as a categorical time-varying predictor:

- Allows for control of classes only....

$$\begin{aligned} \text{Effort}_{tsc} = & \gamma_{000} + \gamma_{100} (\text{Year}01_{tsc}) + \gamma_{200} (\text{Year}12_{tsc}) + U_{0s0} + e_{tsc} \\ & + \gamma_{001}^0 (\text{Class}1_c)(\text{Int}0_{tsc}) + \gamma_{002}^0 (\text{Class}2_c)(\text{Int}0_{tsc}) \cdots + \gamma_{00c}^0 (\text{Class}C_c)(\text{Int}0_{tsc}) \\ & + \gamma_{001}^1 (\text{Class}1_c)(\text{Int}1_{tsc}) + \gamma_{002}^1 (\text{Class}2_c)(\text{Int}1_{tsc}) \cdots + \gamma_{00c}^1 (\text{Class}C_c)(\text{Int}1_{tsc}) \\ & + \gamma_{001}^2 (\text{Class}1_c)(\text{Int}2_{tsc}) + \gamma_{002}^2 (\text{Class}2_c)(\text{Int}2_{tsc}) \cdots + \gamma_{00c}^2 (\text{Class}C_c)(\text{Int}2_{tsc}) \end{aligned}$$

- Hoffman (2015) Equation 11.4: class as a random effects crossed with students at level 2:

- Controls and models class-related variance so it can be predicted

$$\begin{aligned} \text{Effort}_{tsc} = & \gamma_{000} + \gamma_{100} (\text{Year}01_{tsc}) + \gamma_{200} (\text{Year}12_{tsc}) + U_{0s0} + e_{tsc} \\ & + U_{00c}^0 (\text{Int}0_{tsc}) + U_{00c}^1 (\text{Int}1_{tsc}) + U_{00c}^2 (\text{Int}2_{tsc}) \end{aligned}$$

More on Cross-Classified Models

- In crossed models, lower-level predictors can have random slopes of over higher levels AND random slopes of the other crossed factor at the same level
 - Example: Kids, Schools, and Neighborhoods (data permitting)
 - Kid effects could vary over schools AND/OR neighborhoods
 - School effects could vary over neighborhoods (both level 2)
 - Neighborhood effects could vary over schools (both level 2)
- Concerns about smushing still apply over both level-2's
 - Separate contextual effects of kid predictors for schools and neighborhoods (e.g., after controlling for how smart you are, it matters incrementally whether you go to a smart school AND if you live in a neighborhood with smart kids)