# Three-Level Random Effects Models

- Topics:
  - > Examples of three-level designs of time, persons, and groups
  - Partitioning variation across three levels in clustered longitudinal data (occasions within persons within groups)
  - > Unconditional (time only) model specification
  - Conditional (other predictors) model specification
  - Partitioning variation across three levels in intensive longitudinal data (occasions within days within persons)

### What determines the number of levels?

- Answer: the model for the outcome variance ONLY
- How many dimensions of sampling in the <u>outcome</u>?
  - > Longitudinal, one person per family?  $\rightarrow$  2-level model
  - > Longitudinal, 2+ people per family?  $\rightarrow$  3-level model
  - > Longitudinal, 2+ people per family, many cities?  $\rightarrow$  4-level model
  - Sampling dimensions may also be crossed instead of nested, or may be modeled with fixed effects if the # units is small
- Need at least one pile of variance per dimension (for 3 levels, that's 2 sets of random effects and a residual)
  - Include whatever predictors you want for each level, but keep in mind that the usefulness of your predictors will be constrained by how much Y variance exists in its relevant sampling dimension

#### Kinds of 3-Level Designs: Clustered Longitudinal

 First example: Predicting time-specific respondent outcomes for people nested in countries, collected over several years (<u>all same people</u> and <u>same countries</u> are measured over time)



- Country predictors can be included at level 3 only (no random effects)
- Person predictors should be included at levels 2 and 3 (+random over 3)
- What about effects of <u>time-varying predictors</u>?
  - > For <u>People</u>: effects should be included at all 3 levels (+random over 2 and 3)
  - > For <u>Countries</u>: effects are only possible at levels 1 and 3 (+random over 3)

## Other Examples of 3-Level Designs

- The sampling design for the outcome (not the predictors) dictates what your levels will be, so time may not always be level 1
- Example: Predicting answer compliance in respondents nested in interviewers, collected over several years (<u>all different people</u>)



- Based on the sampling of time, time may be modeled...
  - > As fixed effects in the model for the means  $\rightarrow$  2-level model instead
    - Best to use dummy codes for time if few occasions OR no time-level predictors of interest
  - > As a random effect in the model for the variance  $\rightarrow$  3-level model
    - Then differences in compliance rates over time can be predicted by time-level predictors

## Other Examples of 3-Level Designs

 Another example: Predicting time-specific respondent outcomes for people nested in countries, collected over several years (<u>all different people</u>, but the <u>same countries</u> measured over time)



- Before including any fixed effects of time, country and time are actually crossed, not nested as shown here
  - Are nested after controlling for which occasion is which via fixed effects (using dummy codes per mean or a time trend that describes the means)
  - > Time is still a level because not all countries change the same way

#### 3-Level Designs: Predictors vs. Outcomes

• Same example: What if, instead of respondent outcomes, we wanted to predict **time-varying country outcomes**?



Because the outcome was measured at level 2 (country per time):

- Respondents are no longer a level at all (no outcomes for them)
- So there is nothing for respondent predictors to do, except at higher levels
  - > **Time-specific averages** of respondent predictors  $\rightarrow$  time-level outcome variation
  - ▶ Across time, country averages of respondent predictors  $\rightarrow$  country-level outcome variation

**Empty Means, 3-Level Random Intercept Model:** Example for Clustered Longitudinal Data

Notation: t = level-1 time, i = level-2 person, j = level-3 group

Level 1: 
$$y_{tij} = \beta_{0ij} + e_{tij}$$
  
Residual = time-specific deviation  
from person's predicted outcome  
Level 2:  $\beta_{0ij} = \delta_{00j} + U_{0ij}$   
Person Random Intercept  
= person-specific deviation  
from group's predicted outcome  
Level 3:  $\delta_{00j} = \gamma_{000} + V_{00j}$   
Fixed Intercept  
= grand mean  
(because no  
(composite equation:  

= group-specific deviation

from fixed intercept

Btw: My bad for reusing "V"

predictors yet)

+e<sub>tij</sub>

## 2-Level Random Intercept Model

- Where does each kind of person dependency go? Into a new random effects variance component (or "pile" of variance):
- Let's start with an empty means, random intercept 2-level model for time within person:



## **3-Level Random Intercept Model**

• Now let's see what happens in an empty means, random intercept 3-level model of time within person within groups:



#### ICCs in a 3-Level Random Intercept Model Example: Time within Person within Group

• ICC for level 2 (and level 3) relative to level 1:

• ICC<sub>L2</sub> = 
$$\frac{\text{Between-Person}}{\text{Total}} = \frac{\text{L3+L2}}{\text{L3+L2+L1}} = \frac{\tau_{V_{00}}^2 + \tau_{U_0}^2}{\tau_{V_{00}}^2 + \tau_{U_0}^2 + \sigma_e^2}$$

→ This ICC expresses similarity of occasions from same person (and by definition, from the same group) → of the **total variation in Y**, how much of it is **between persons, or cross-sectional (not due to time)**?

• ICC for level 3 relative to level 2 (ignoring level 1):

• ICC<sub>L3</sub> = 
$$\frac{\text{Between-Group}}{\text{Between-Person}} = \frac{\text{L3}}{\text{L3+L2}} = \frac{\tau_{V_{00}}^2}{\tau_{V_{00}}^2 + \tau_{U_0}^2}$$

→ This ICC expresses similarity of persons from same group (ignoring within-person variation over time) → of **that total between**-**person variation in Y**, how much of that is actually **between groups**?

# 3-Level Model for Intensive Longitudinal Data (occasions, days, persons)



Useful ICC variants for this type of design:

#### $ICC_{L3B} = L3 / total$

- % Between Persons
- Note: this is what is given by STATA and Mplus as "level-3 ICC"

#### $ICC_{L2B} = L2 / L2 + L1$

- Proportion of timerelated variance for day
- Tests if occasions on same day are more related than occasions on different days (i.e., is day needed?)

## 2-Level Random Slope Model

• What about time? After adding fixed effects of time, we can add random effects of time over persons in a 2-level model:



## **3-Level Random Slope Model**

 In a 3-level model, we can have random effects of time over persons AND groups:



**3-Level Random Time Slope Model** 

Notation: t = level-1 time, i = level-2 person, j = level-3 group

**Level 1:** 
$$y_{tij} = \beta_{0ij} + \beta_{1ij}$$
 (Time<sub>tij</sub>) +  $e_{tij} \leftarrow \frac{\text{Residual}}{\text{deviation from person's predicted growth line ( $\sigma_e^2$ )}$ 

Level 2: 
$$\beta_{0ij} = \delta_{00j} + U_{0ij}$$
  
 $\beta_{1ij} = \delta_{10j} + U_{1ij}$ 
Person Random Intercept and Slope =  
person-specific deviations from group's  
predicted intercept, slope ( $\tau_{U_0}^2, \tau_{U_1}^2, \tau_{U_01}$ )

Level 3: 
$$\delta_{00j} = \gamma_{000} + V_{00j}$$
  
 $\delta_{10j} = \gamma_{100} + V_{10j}$ 

<u>Group Random Intercept and Slope</u> = group-specific deviations from fixed intercept, slope ( $\tau_{V_{00}}^2, \tau_{V_{10}}^2, \tau_{V_{00,10}}$ )

Fixed Intercept, Fixed Linear Time Slope Composite equation (9 parameters):  $y_{tij} = (\gamma_{000} + V_{00j} + U_{0ij}) + (\gamma_{100} + V_{10i} + U_{1ii}) (Time_{tii}) + e_{tii}$ 

# Random Time Slopes at both Levels 2 AND 3? An example with family as group:



### ICCs for Random Intercepts and Slopes

 Once random slopes are included at both level-3 and level-2, ICCs can be computed for the random intercepts and slopes specifically (which would be the level-3 type of ICC)

$$ICC_{Int} = \frac{Between - Group}{Between - Person} = \frac{L3 Int}{L3 Int + L2 Int} = \frac{\tau_{V_{00}}^2}{\tau_{V_{00}}^2 + \tau_{U_0}^2}$$
$$ICC_{Slope} = \frac{Between - Group}{Between - Person} = \frac{L3 Slope}{L3 Slope + L2 Slope} = \frac{\tau_{V_{10}}^2}{\tau_{V_{10}}^2 + \tau_{U_1}^2}$$

- Can be computed for any level-1 slope that is random at both levels (e.g., linear and quadratic time, time-varying predictors)
- Be careful when the model is uneven across levels, though Random Level 2: int, linear, quad  $\rightarrow$  Linear is when time = 0

Random Level 3: int, linear

Linear is at any occasion

#### More on Random Slopes in 3-Level Models

- Any level-1 predictor can have a random effect over level 2, level 3, or over both levels, but I recommend working your way UP the higher levels for assessing random effects...
  - > e.g., Does the effect of time vary over level-2 persons?
  - > If so, does the effect of time vary over level-3 groups, too?  $\rightarrow$  Is there a commonality in how people from the same group change over time?
- ... because random effects at level 3 only are possible but unlikely (e.g., means everyone in the group changes the same)
- Level-2 predictors can also have random effects over level 3
  e.g., Does the effect of a L2 person characteristic vary over L3 groups?
- Level-1, level-2, and level-1 by level-2 cross-level interactions can all have random effects over level 3, too, in theory
  - But tread carefully! The more random effects you have, the more likely you are to have convergence problems ("G matrix not positive definite")

# **Conditional Model Specification**

- Remember separating between- and within-person effects? Now there are 3 potential effects for any level-1 predictor!
  - Example in a Clustered Longitudinal Design: Effect of stress on wellbeing, both measured over time within person within families:
  - Level 1 (Time): During Times of more stress, people have lower (time-specific) wellbeing than in times of less stress
  - Level 2 (Person): People in the family who have more stress have lower (person average) wellbeing than people in the family who have less stress
  - Level 3 (Family): Families who have more stress have lower (family average) wellbeing than families who have less stress
- <u>2 potential effects for any level-2 predictor, also</u>
  - > Example: Effect of baseline level of person coping skills in same design:
  - Level 2 (Person): People in the family who cope better have better (person average) wellbeing than people in the family who cope worse
  - Level 3 (Family): Families who cope better have better (family average) wellbeing than families who cope worse

## Option 1: Separate Total Effects Per Level Using Variable-Based-Centering

- Level 1 (Time): Time-varying stress relative to person mean
  - → WPstress<sub>tij</sub> = Stress<sub>tij</sub> PersonMeanStress<sub>ij</sub>
  - $\rightarrow$  Directly tests if within-person effect  $\neq$  0?
  - $\rightarrow$  **Total** within-person effect of more stress **than usual**  $\neq$  0?
- Level 2 (Person): Person mean stress relative to family
  - → WFstress<sub>ii</sub> = PersonMeanStress<sub>ii</sub> FamilyMeanStress<sub>i</sub>
  - $\rightarrow$  Directly tests if within-family effect  $\neq$  0?
  - → Total effect of more stress *than other members of one's family* ≠ 0?
- Level 3 (Family): Family mean stress relative to all families (from constant)
  - $\rightarrow$  BFstress<sub>i</sub> = FamilyMeanStress<sub>i</sub> C
  - $\rightarrow$  Directly tests if between-family effect  $\neq$  0?
  - → Total effect of more stress *than other families* ≠ 0?

## Option 1: Separate Total Effects Per Level Using Variable-Based-Centering

Notation: t = level-1 time, i = level-2 person, j = level-3 group PM = person mean, FM = family mean, C = centering constant

Level 1:  $y_{tij} = \beta_{0ij} + \beta_{1ij}$ (Time<sub>tij</sub>) +  $\beta_{2ij}$ (Stress<sub>tij</sub> – PMstress<sub>ij</sub>) +  $e_{tij}$ 

Level 2:  $\beta_{0ij} = \delta_{00i} + \delta_{01i}$  (PMstress<sub>ii</sub> – FMstress<sub>i</sub>) + U<sub>0ii</sub>  $\boldsymbol{\beta}_{1ii} = \boldsymbol{\delta}_{10i} + \boldsymbol{U}_{1ii}$  $\beta_{2ii} = \delta_{20i} + (\mathbf{U}_{2ii})$ Fixed intercept, **Between-family** Level 3:  $\delta_{00j} = \gamma_{000} + \gamma_{001} (FMstress_i - C) + V_{00i}$ stress main effect  $\delta_{01i} = \gamma_{010} + (V_{01i})$ Within-family stress main effect  $\delta_{10i} = \gamma_{100} + V_{10i}$ Time main effect  $\delta_{20i} = \gamma_{200} + (V_{20i})$ Within-person stress main effect

## Option 2: Contextual Effects Per Level Using Constant-Based-Centering

- Level 1 (Time): Time-varying stress (relative to sample constant)
  - $\rightarrow$  TVstress<sub>tij</sub> = Stress<sub>tij</sub> C
  - $\rightarrow$  Directly tests if within-person effect  $\neq$  0?
  - $\rightarrow$  **Total** within-person effect of more stress **than usual**  $\neq$  0?
- Level 2 (Person): Person mean stress (relative to sample constant)
  - $\rightarrow$  BPstress<sub>ii</sub> = PersonMeanStress<sub>ii</sub> C
  - $\rightarrow$  Directly tests if within-person and within-family effects  $\neq$  ?
  - → Contextual effect of more stress *than other members of one's family* ≠ 0?
- Level 3 (Family): Family mean stress relative to all families (from constant)
  - → BFstress<sub>i</sub> = FamilyMeanStress<sub>i</sub> C
  - $\rightarrow$  Directly tests if within-family and between-family effects  $\neq$  ?
  - → **Contextual** effect of more stress **than other families** ≠ 0?

## Option 2: Contextual Effects Per Level Using Constant-Based-Centering

Notation: t = level-1 time, i = level-2 person, j = level-3 groupPM = person mean, FM = family mean, C = centering constant

Level 1: 
$$y_{tij} = \beta_{0ij} + \beta_{1ij}$$
(Time<sub>tij</sub>) +  $\beta_{2ij}$ (Stress<sub>tij</sub> - C) +  $e_{tij}$ 

Level 2: 
$$\beta_{0ij} = \delta_{00j} + \delta_{01j}$$
 (PMstress<sub>ij</sub>-C) + U<sub>0ij</sub>  
 $\beta_{1ij} = \delta_{10j} + U_{1ij}$   
 $\beta_{2ij} = \delta_{20j} + (U_{2ij})$   
Level 3:  $\delta_{00j} = \gamma_{000} + \gamma_{001}$  (FMstress<sub>j</sub>-C) + V<sub>00j</sub>  
 $\delta_{01j} = \gamma_{010} + (V_{01j})$  Contextual within-family stress main effect  
 $\delta_{10j} = \gamma_{100} + V_{10j}$  Time main effect  
 $\delta_{20j} = \gamma_{200} + (V_{20j})$  Within-person stress main effect

# What does it mean to omit higher-level effects under each centering method?

- Variable-Based-Centering: Omitting a fixed effect assumes that the effect at that level <u>does not exist</u> (= 0)
  - Remove L3 effect? Assume L3 Between-Family effect = 0
    - L1 effect = Within-Person effect, L2 effect = Within-Family effect
  - > Then remove L2 effect? Assume L2 Within-Family effect = 0
    - L1 effect = Within-Person effect
- **Constant-Based-Centering**: Omitting a fixed effect means the effect at that level is <u>equivalent to</u> the effect at the level below
  - Remove L3 effect? Assume L3 Between-Family = L2 Within-Family effect
    - L1 effect = Within-Person effect, L2 effect = 'smushed' WF and BF effects
  - > Then remove L2 effect? Assume L2 Between-Person effect = L1 effect
    - L1 'smushed' = Within-Person, Within-Family, and Between-Family effects

## Interactions belong at each level, too...

• Example: Is the effect of stress on wellbeing moderated by time-invariant person coping? Using variable-based-centering:

#### <u>Stress Effects</u>

- > Level 1 (Time): WPstress<sub>tij</sub> = Stress<sub>tij</sub> PersonMeanStress<sub>ij</sub>
- Level 2 (Person): WFstress<sub>ij</sub> = PersonMeanStress<sub>ij</sub> FamilyMeanStress<sub>j</sub>
- Level 3 (Family): BFstress<sub>j</sub> = FamilyMeanStress<sub>j</sub> C

#### <u>Coping Effects</u>

- Level 2 (Person): WFcope<sub>ij</sub> = Cope<sub>ij</sub> FamilyMeanCope<sub>j</sub>
- Level 3 (Family): BFcope<sub>j</sub> = FamilyMeanCope<sub>j</sub> C

#### Interaction Effects

- With level-1 stress: WPstress<sub>tij</sub> \* WFcope<sub>ij</sub>, WPstress<sub>tij</sub> \* BFcope<sub>j</sub>
- With level-2 stress: WFstress<sub>ij</sub> \* WFcope<sub>ij</sub>, (WFstress<sub>ij</sub> \* BFcope<sub>j</sub>)
- With level-3 stress: BFstress<sub>j</sub> \* BFcope<sub>j</sub>, (BFstress<sub>j</sub> \* WFcope<sub>ij</sub>)

### Interactions belong at each level, too...

Notation: t = level-1 time, i = level-2 person, j = level-3 group PM = person mean, FM = family mean, C = centering constant

Level 1:  $y_{tij} = \beta_{0ij} + \beta_{1ij}$ (Time<sub>tij</sub>) +  $\beta_{2ij}$ (Stress<sub>tij</sub> – PMstress<sub>ij</sub>) +  $e_{tij}$ 

Level 2: 
$$\beta_{0ij} = \delta_{00j} + \delta_{01j}$$
 (PMstress<sub>ij</sub> – FMstress<sub>j</sub>)  
+  $\delta_{02j}$  (Cope<sub>ij</sub> – FMcope<sub>j</sub>)  
+  $\delta_{03j}$  (PMstress<sub>ij</sub> – FMstress<sub>j</sub>) (Cope<sub>ij</sub> – FMcope<sub>j</sub>) +  $U_{0ij}$   
 $\beta_{1ij} = \delta_{10j} + U_{1ij}$   
 $\beta_{2ij} = \delta_{20j} + \delta_{21j}$  (Cope<sub>ij</sub> – FMcope<sub>j</sub>) + ( $U_{2ij}$ )

Level 3: 
$$\delta_{00j} = \gamma_{000} + \gamma_{001} (FMstress_j - C) + \gamma_{002} (FMcope_j - C) + \gamma_{003} (FMstress_j - C) (FMcope_j - C) + V_{00j}$$
  
 $\delta_{01j} = \gamma_{010} + (V_{01j}) \quad \delta_{02j} = \gamma_{020} + (V_{02j}) \quad \delta_{03j} = \gamma_{030} + (V_{03j})$   
 $\delta_{10j} = \gamma_{100} + V_{10j}$   
 $\delta_{20j} = \gamma_{200} + \gamma_{202} (FMcope_j - C) + (V_{20j}) \quad \delta_{21j} = \gamma_{210} + (V_{21j})$ 

#### Summary: Three-Level Random Effects Models

- Estimating 3-level models requires no new concepts, but everything is an order of complexity higher:
  - > Partitioning variance over 3 levels instead of 2  $\rightarrow$  many possible ICCs
  - > Random slope variance will come from the variance directly below:
    - Level-2 random slope variance comes from level-1 residual
    - Level-3 random slope variance comes from level-2 random slope (or residual)
  - > Level-1 effects can be random over level 2, level 3, or both
    - ICCs can be computed for level-1 slopes that are random over both level-2 and level-3 (assuming the L2 and L3 variance models match)
    - Smushing of level-1 fixed effects should be tested over levels 2 AND 3
  - > Level-2 effects can be random over level 3
    - Smushing of level-2 fixed effects should be tested over level 3
  - Level-3 effects cannot be random; no worries about smushing
  - > Phew....

#### Bonus: Pseudo-R<sup>2</sup> in Three-Level Models

- Although it may not work this neatly in real data, here is the logic for how each type of fixed effect should explain variance
- Main effects and purely same-level interactions are straightforward—they target their own level:
  - > L1 main effects and L1 interactions  $\rightarrow$  L1 residual variance
  - > L2 main effects and L2 interactions  $\rightarrow$  L2 random intercept variance
  - > L3 main effects and L3 interactions  $\rightarrow$  L3 random intercept variance
- For cross-level interactions, which variance gets explained depends on if random slopes are included at each level...
  - ▶ L3 \* L1 → L3 random variance in L1 slope if included, or L2 random variance in L1 slope if included, or L1 residual otherwise
  - ► L3 \* L2 → L3 random variance in L2 slope if included, or L2 random intercept otherwise
  - > L2 \* L1  $\rightarrow$  L2 random variance in L1 slope if included, or L1 residual otherwise