

Hierarchical Linear Modeling

Kenneth S. Law (羅勝強)

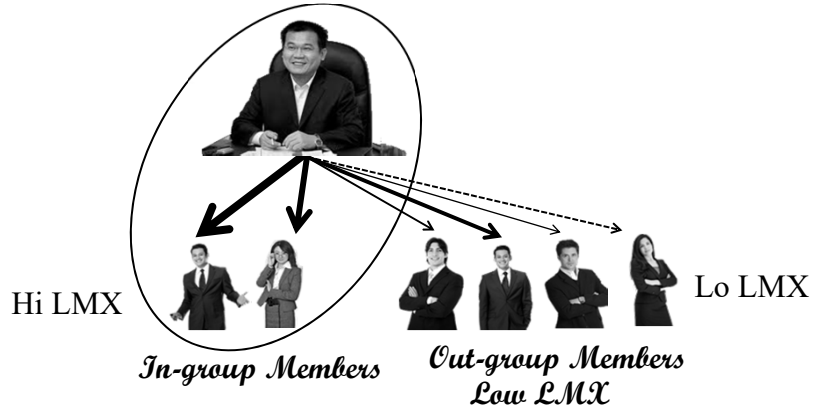
**Department of Management
The Chinese University of Hong Kong**

Hierarchical Linear Modeling

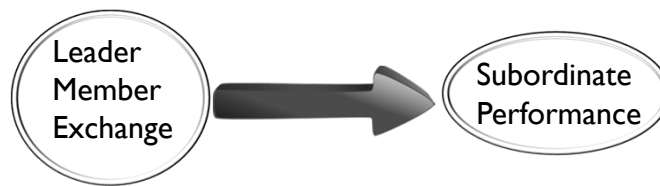
▶ The VDL model as an example	3-16
▶ Why not single level analysis	17-18
▶ Fixed effects and random effect	19
▶ Significance testing	20
▶ First Wrap up	21-24
▶ A realistic model: Helping	25-37
▶ Centering in HLM	38-54
▶ Examples of different HLM models	55-65
▶ Another example: math	66-78
▶ Mplus programming	79-91
▶ Some issues on HLM estimates	92-99
▶ % variance accounted for	100-108
▶ Multiple Xs	109
▶ Summary	110

The Vertical Dyadic Linkage Model

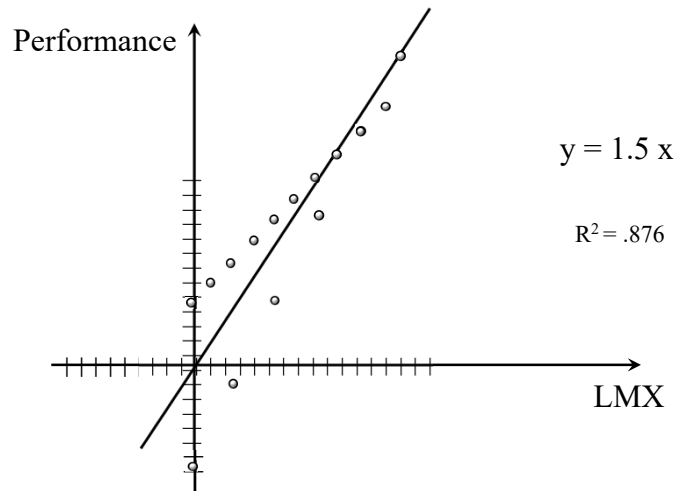
The VDL model says that supervisors deliberately develop different exchange relationship with different subordinates.



HLM



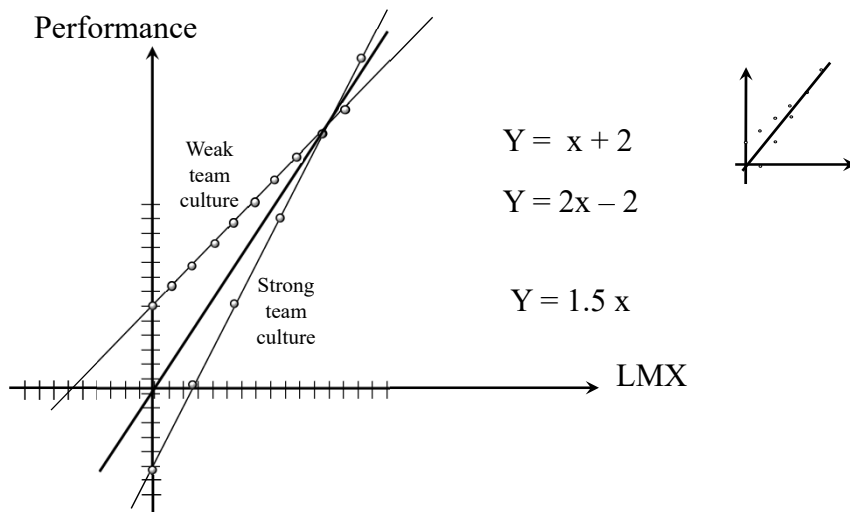
LMX → Performance



▷ 5

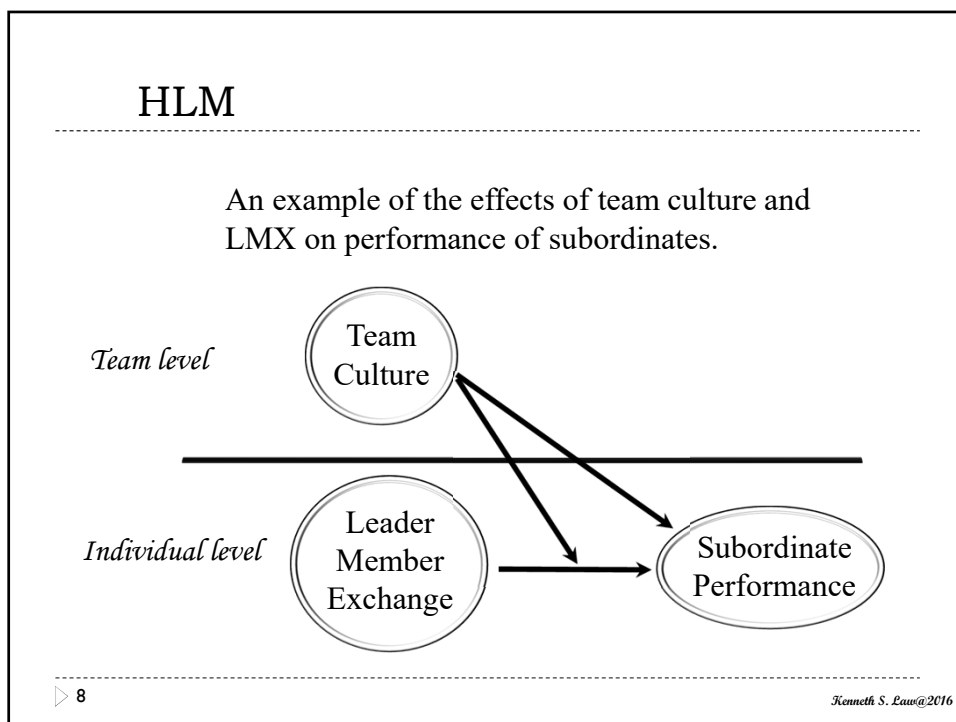
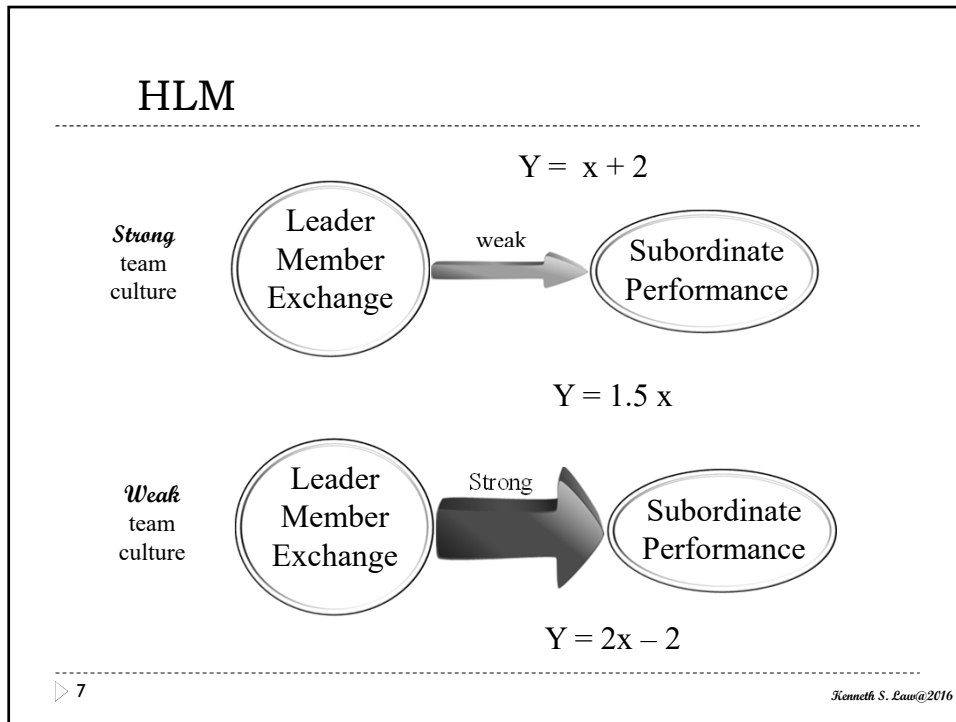
Kenneth S. Law@2016

Two levels of effects

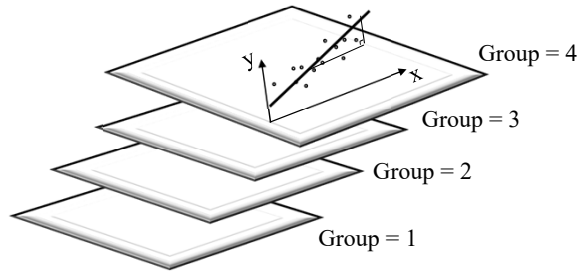


▷ 6

Kenneth S. Law@2016



Figurative Illustration of HLM

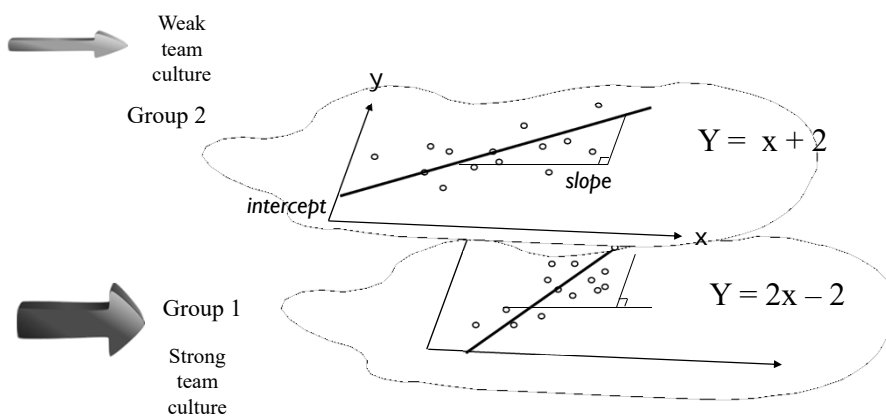


$x = \text{LMX}$
 $y = \text{performance}$

▷ 9

Kenneth S. Law@2016

A simple case of two groups



▷ 10

Kenneth S. Law@2016

The purpose of HLM

Group	Culture	Intercept	Slope	Equation
Group 2	Weak culture	-2	2	$Y = 2x - 2$
Group 1	Strong culture	+2	1	$Y = x + 2$

- We need to estimate the intercept and slope of each group separately.
 $y = \text{intercept} + \text{slope } x + \text{error}$
- What happened after estimating the intercept and slope of each group?
- We try to find a group-level variable to predict the variable intercept and slope in each group.

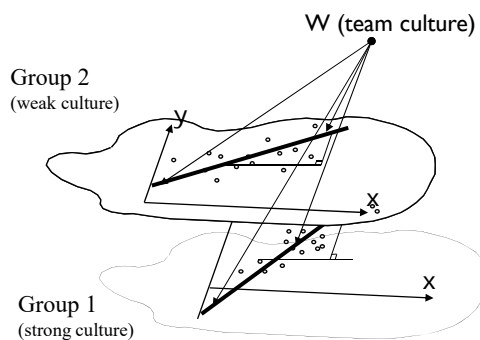
$$\text{Intercept} = \gamma_{00} + \gamma_{01}W + u_1$$

$$\text{slope} = \gamma_{10} + \gamma_{11}W + u_2$$

▷ 11

Kenneth S. Law@2016

Cross level analysis



1. Within each level, there is a regression analysis;

- Level 1
 $y = \beta_{01} + \beta_{11}x + \varepsilon_1$

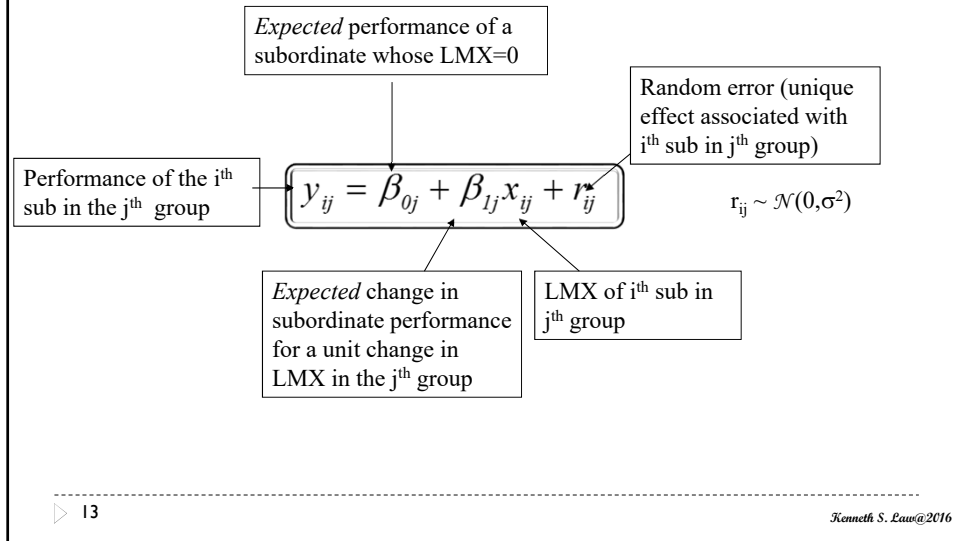
- Level 2
 $y = \beta_{02} + \beta_{12}x + \varepsilon_2$

2. The differential effect of x on y, (i.e., the intercept β_0 and slope β_1) at each level can be predicted by a group level variable (W).

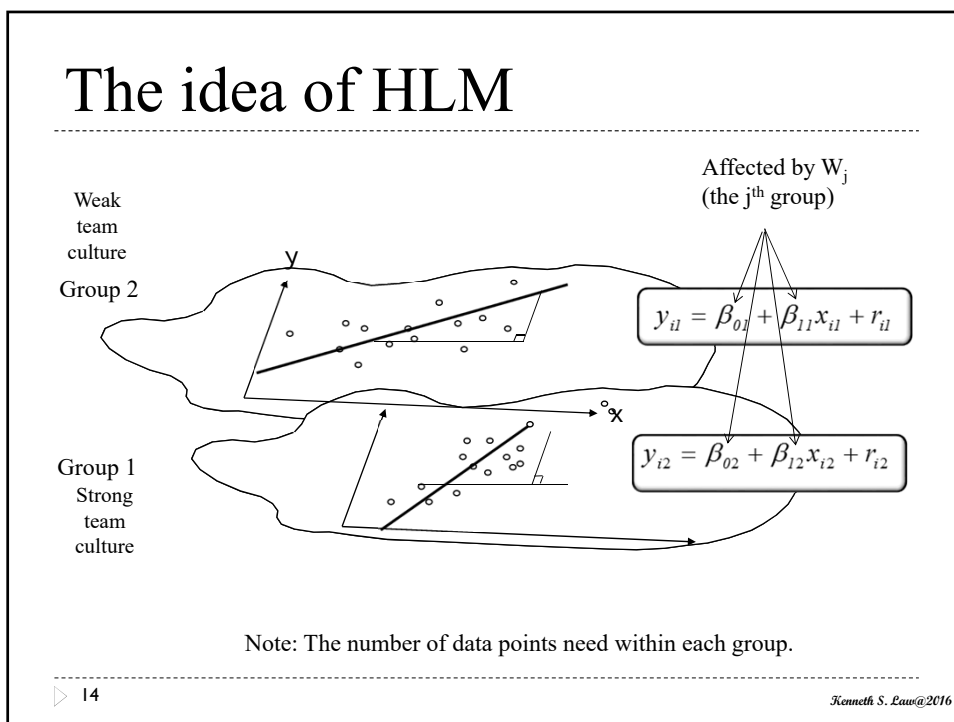
▷ 12

Kenneth S. Law@2016

The Individual level model



The idea of HLM



The database

Perf	LMX	
Y ₁₁	X ₁₁	team 1
Y ₂₁	X ₂₁	
Y ₃₁	X ₃₁	
Y ₄₁	X ₄₁	
Y ₅₁	X ₅₁	
Y ₁₂	X ₁₂	team 2
Y ₂₂	X ₂₂	
Y ₃₂	X ₃₂	
Y ₄₂	X ₄₂	
Y ₅₂	X ₅₂	

Weak $W_1 = 0$

Strong $W_2 = 1$

Varies within a group

Varies across different groups

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

Two types of team culture: strong culture vs. weak culture

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

Two types of team culture: strong culture vs. weak culture

15 Kenneth S. Law@2016

The effects of team culture and LMX on performance of subordinates

Team level

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

Individual level

Team Culture

Leader Member Exchange

Subordinate Performance

$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

weak $y_{i1} = \beta_{01} + \beta_{11}x_{i1} + r_{i1}$

strong $y_{i2} = \beta_{02} + \beta_{12}x_{i2} + r_{i2}$

W_j

16 Kenneth S. Law@2016

Why need two levels?

Y	LMX	Culture
Y ₁₁	X ₁₁	0
Y ₂₁	X ₂₁	0
Y ₃₁	X ₃₁	0
Y ₄₁	X ₄₁	0
Y ₅₁	X ₅₁	0
Y ₁₂	X ₁₂	1
Y ₂₂	X ₂₂	1
Y ₃₂	X ₃₂	1
Y ₄₂	X ₄₂	1

$$y_{i1} = \beta_{01} + \beta_{11}x_{i1} + r_{i1}$$

$$y_{i2} = \beta_{02} + \beta_{12}x_{i2} + r_{i2}$$

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

Why can't we use multiple regression and simply specify the following model?

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + \beta_{2j}w_j + r_{ij}$$

▷ 17

Kenneth S. Law@2016

Correlated Error Terms

Level 1 $y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + \beta_{2j}w_j + r_{ij}$$

Level 2 $\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

$$y_{ij} = [\gamma_{00} + \gamma_{01}w_j + u_{0j}] + [\gamma_{10} + \gamma_{11}w_j + u_{1j}]x_{ij} + r_{ij}$$

$$y_{ij} = \gamma_{00} + \gamma_{10}x_{ij} + \gamma_{01}w_j + \gamma_{11}x_{ij}w_j + [r_{ij} + u_{0j} + x_{ij}u_{1j}]$$

$$y_{ij} = b_{0j} + b_{1j}x_{ij} + b_{2j}w_j + b_{3j}x_{ij}w_j + r_{ij}$$

▷ 18

Fixed effect vs. Random effect

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

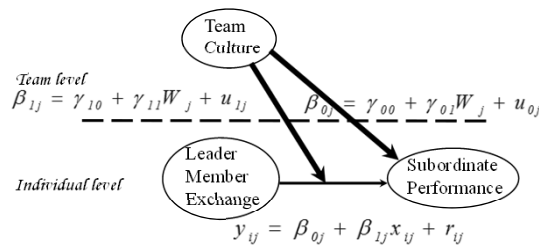
$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

Random effect:

- does not mean that it is not systematic or error
- effects that would change at level 2
- effects that would be different for each group

Fixed effect:

- effects that would not change within this model for all groups



Significance Tests

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij}$$

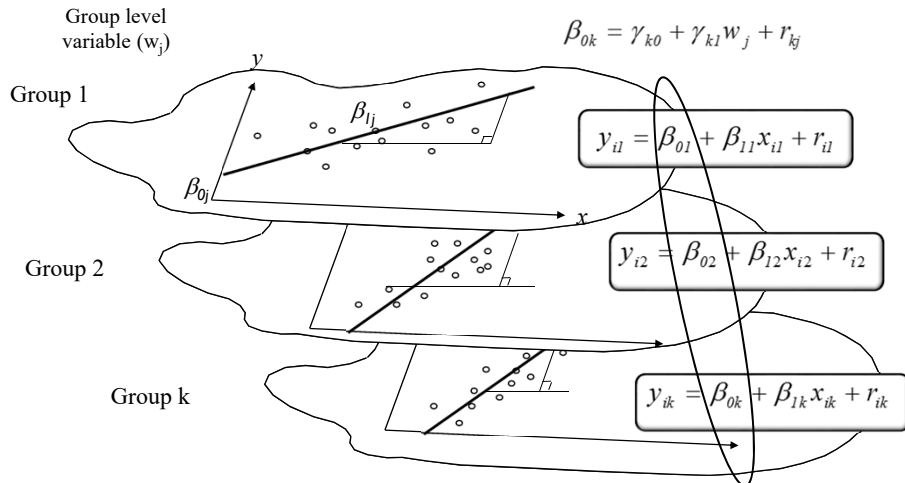
$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

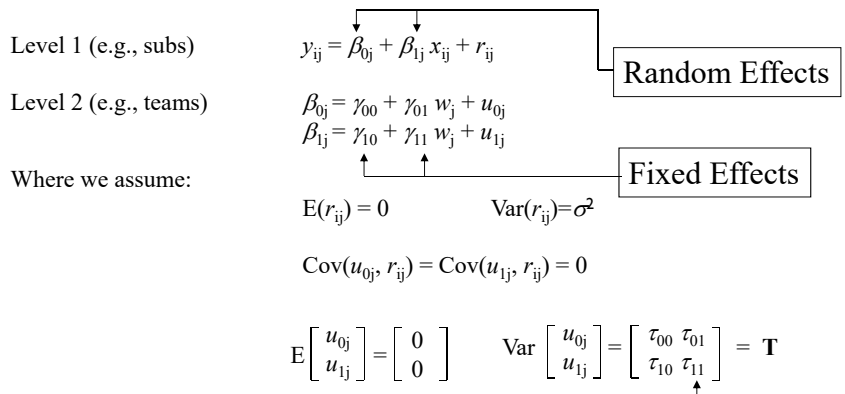
χ^2 -test of whether the residual variances are zero, i.e., How good is our estimation of the effect of team on “the effects of LMX on subordinates’ performance”?

t-test of whether these γ s are zero, i.e., Does team culture affects “the effects of LMX on subordinates’ performance”? (cross level interaction effect)

Wrap up: The idea of HLM

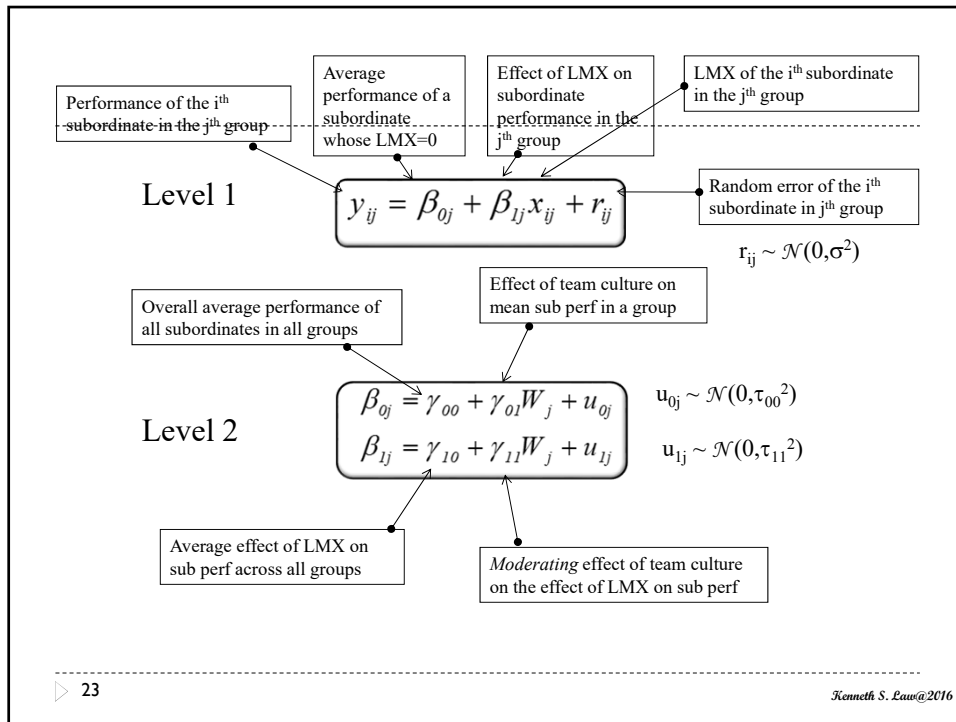


A simple Two-Level Model



In multi-level analyses, a quantity is random means that it fluctuates over units in the population. Therefore, estimates that are constant across levels is consider as fixed effects.

Source: Snijders, T.A.B. "Fixed and Random Effects" In B.S. Everitt and D.C. Howell (eds), Encyclopedia of Statistics in Behavioral Science. Vol. 2, 664-665. Chichester (etc.):Wiley, 2005.



Limitations of HLM

Level 1 (e.g., subs) $y_{ij} = \beta_{0j} + \beta_{1j} x_{ij} + r_{ij}$

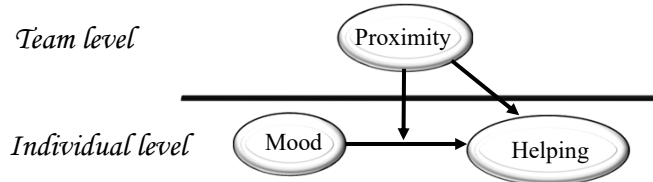
Level 2 (e.g., teams) $\beta_{0j} = \gamma_{00} + \gamma_{01} w_j + u_{0j}$
 $\beta_{1j} = \gamma_{10} + \gamma_{11} w_j + u_{1j}$

Higher level variables explaining intercepts and slope of lower level regressions

Level 1 (e.g., subs) $y_k = \beta_0 + \beta_1 x_{ik} + e_k$

Team performance Ability of team members

A Realistic Model



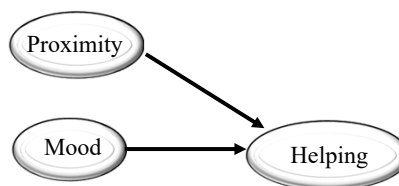
$$help_{ij} = \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01} prox_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11} prox_j + u_{1j}$$

Source: Hofmann, D.A. (1997). An Overview of the logic and rationale of hierarchical linear models, Journal of Management, 23(6), 723-744.

A single level approach



$$help_{ij} = \beta_0 + \beta_1 mood_{ij} + \beta_2 prox_j + r_{ij}$$

Source: Hofmann, D.A. (1997). An Overview of the logic and rationale of hierarchical linear models, Journal of Management, 23(6), 723-744.

Mplus Regression program

TITLE: Sample multi-level program on helping

DATA: FILE = helping.txt;

VARIABLE: NAMES = group help mood prox;

USEVARIABLES = help mood prox;

MODEL: help on mood prox;

OUTPUT: TECH1;

$$\text{helping} = \beta_0 + \beta_1 \text{ mood} + \beta_2 \text{ proximity}$$

▷ 27

Kenneth S. Law@2016

Mplus output

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
HELP ON				
MOOD	3.976	0.087	45.915	0.000
PROX	1.261	0.111	11.366	0.000
Intercepts				
HELP	1.502	0.833	1.803	0.071
Residual Variances				
HELP	40.962	1.832	22.359	0.000

$$\text{helping} = 1.502 + 3.976^{**} \text{ mood} + 1.261^{**} \text{ proximity}$$

▷ 28

Kenneth S. Law@2016

The Null model (random intercept only)

$$\begin{aligned} \text{help}_{ij} &= \beta_{0j} + r_{ij} \\ \beta_{0j} &= \gamma_{00} + u_{0j} \end{aligned}$$

Baseline model or null model means that there is no predictor in the model at all.

What is β_{0j} ? β_{0j} is the average helping behavior of each group.

What is r_{ij} ? r_{ij} is the random error associated with each employee in each group.

What is γ_{00} ? γ_{00} is the average helping of all employees in all groups (the grand mean).

What is u_{0j} ? u_{0j} is the deviation of "average helping of group j employees" from the grand mean.

$$\text{help}_{ij} = \gamma_{00} + u_{0j} + r_{ij}$$

The helping behavior of ith employee in the jth group is the sum of :

- the average helping of all employees in all groups
- the deviation of the jth group mean from the grand mean
- the deviation of the ith individual from the jth group mean

▷ 29

Kenneth S. Law@2016

Mplus Baseline Model

```
TITLE:      Sample multi-level program on helping
DATA:      FILE = helping.txt;
VARIABLE:  NAMES = group help mood prox;
           USEVARIABLES = help;
           WITHIN = ;
           BETWEEN = ;
           CLUSTER = group;
ANALYSIS:  TYPE = TWOLEVEL RANDOM;
MODEL:
  %WITHIN%
  help;

  %BETWEEN%
  help;
```

$$\begin{aligned} \text{help}_{ij} &= \beta_{0j} + r_{ij} \\ \beta_{0j} &= \gamma_{00} + u_{0j} \end{aligned}$$

▷ 30 OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus output

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Within Level Variances	σ^2			
HELP	31.757	1.763	18.011	0.000
Between Level Means	γ_{00}			
HELP	31.394	1.410	22.269	0.000
Variiances	τ_{00}			
HELP	97.786	15.287	6.397	0.000

$help_{ij} = \beta_{0j} + r_{ij}$
 $\beta_{0j} = \gamma_{00} + u_{0j}$

Intraclass Variable Correlation
HELP 0.755

$ICC = \frac{\tau_{00}}{\tau_{00} + \sigma^2} = \frac{97.786}{97.786 + 31.757} = .755$

What if ICC $\rightarrow 0$?

▷ 31 *Kenneth S. Law@2016*

Fixed coefficient model

Team level

Individual level

Mood

→

Helping

$help_{ij} = \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij}$
 $\beta_{0j} = \gamma_{00} + u_{0j}$
 $\beta_{1j} = \gamma_{10}$ ← *The effect of mood on helping is the same across all groups*

▷ 32 *Kenneth S. Law@2016*

Mplus Fixed coefficient model

```

TITLE:      Sample multi-level program on helping
DATA:      FILE = helping.txt;
VARIABLE:  NAMES = group help mood prox;
           USEVARIABLES = help mood;
           WITHIN = mood;
           CLUSTER = group;
ANALYSIS:  TYPE = TWOLEVEL(RANDOM);
MODEL:
  %WITHIN%
  help ON mood;
  help;
  %BETWEEN%
  help;
OUTPUT:    TECH1 TECH3;
    
```

$$\begin{aligned}
 help_{ij} &= \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij} \\
 \beta_{0j} &= \gamma_{00} + u_{0j} \\
 \beta_{1j} &= \gamma_{10}
 \end{aligned}$$

← *helping affected by mood*
 ← *allow mean difference across group*
 ← *Estimate the mean of each group at Level 2*

▷ 33

Kenneth S. Law@2016

Mplus HLM output

	Estimate	S.E.	Est./S.E.	
Within Level				
HELP ON MOOD	2.999	0.068	44.230	0.000
Residual Variances HELP	5.963	0.315	18.938	0.000
Between Level				
Means HELP	31.394	1.410	22.269	0.000
Variances HELP	99.076	15.275	6.486	0.000

$$\begin{aligned}
 help_{ij} &= \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij} \\
 \beta_{0j} &= \gamma_{00} + u_{0j} \\
 \beta_{1j} &= \gamma_{10}
 \end{aligned}$$

The within variable is helping, σ^2 dropped from 31.757 to 5.963

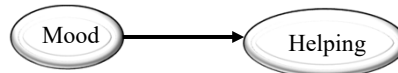
▷ 34

Kenneth S. Law@2016

Random coefficient model

Team level

Individual level



$$help_{ij} = \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + u_{1j} \leftarrow \text{The effect of mood on helping is allowed to vary across groups}$$

▷ 35

Kenneth S. Law@2016

Mplus Random coefficient model

TITLE: Sample multi-level program on helping

DATA: FILE = helping.txt;

VARIABLE: NAMES = group help mood prox;
 USEVARIABLES = help mood;
 WITHIN = mood;
 CLUSTER = group;

$$help_{ij} = \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

ANALYSIS: TYPE = TWOLEVEL(RANDOM);

MODEL:

- `%WITHIN%`
`help;`
`s | help on mood;`
 - we label the first-level effect of mood on help (i.e. β_{1j}) as s;
- `%BETWEEN%`
`help;`
`s;`
`help WITH s;`
 - mean help of each group (i.e. β_{0j}) as represented by “help” at the `%BETWEEN%` statement is estimated automatically.
 - The intercept (β_{0j}) and slope (β_{1j}) in level-2 are correlated.

▷ 36 OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus HLM output

$$help_{ij} = \beta_{0j} + \beta_{1j} mood_{ij} + r_{ij}$$

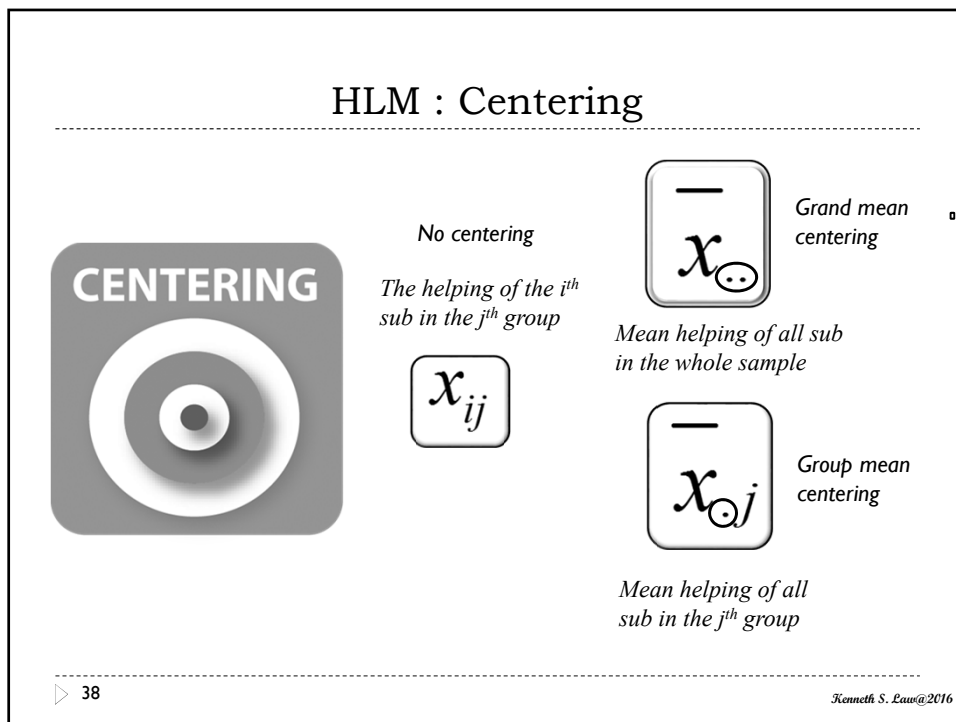
$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

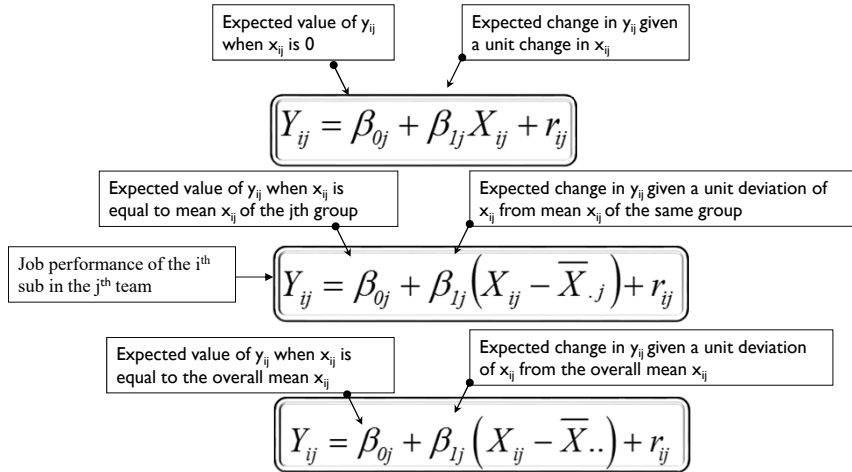
		Estimate	S.E.	Est./S.E.	
Within Level					
Residual Variances					
HELP	σ^2	5.611	0.304	18.452	0.000
The within variable is helping, σ^2 dropped from 5.963 to 5.611					
Between Level					
HELP WITH S		-0.105	0.443	-0.237	0.813
Means					
HELP	γ_{00}	13.837	0.959	14.436	0.000
S	γ_{10}	3.013	0.069	43.935	0.000
Original s fixed effect estimate was 2.999					
Variances					
HELP	τ_{00}	41.681	7.899	5.277	0.000
S	τ_{11}	0.125	0.037	3.360	0.001
Variance of s (τ_{11}) is highly significant					

▷ 37

Kenneth S. Law@2016



Three common centering



39

Group Mean Centering

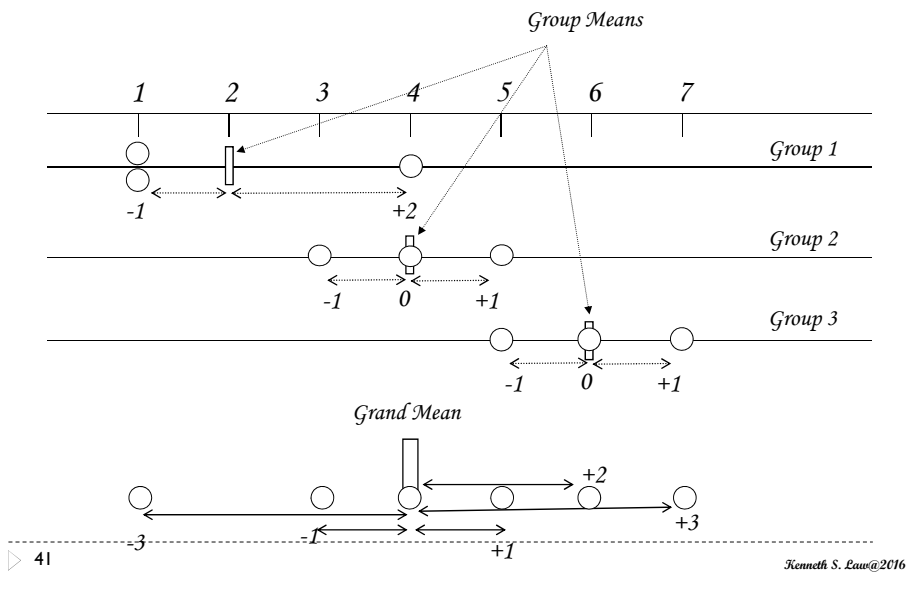
$$Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$$

Group	Person	x	x'	y
1	1	2	-1	3
1	2	3	0	3
1	3	3	0	4
1	4	4	1	4
2	1	3	-1	2
2	2	4	0	3
2	3	5	1	4
3	1	2	-0.5	4
3	2	3	0.5	3

40

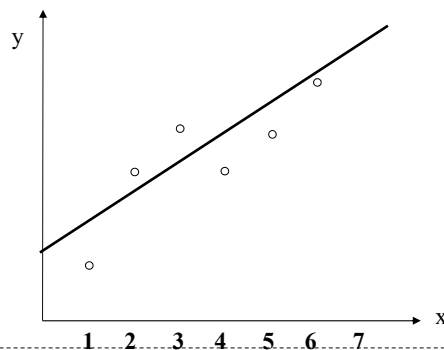
Kenneth S. Law@2016

Grand-mean vs. Group-mean centering



No Centering

$$\begin{array}{lll}
 Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij} & y = .80 + .77x & y = .878x \\
 Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij} & y = 3.5 + .77x & y = .878x \\
 Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij} & y = 3.5 + 1.5x & y = .816x
 \end{array}$$



y	Raw x	Grand	Group
1	1	-2.5	-1
3	2	-1.5	0
4	3	-0.5	1
3	4	0.5	-1
4	5	1.5	0
6	6	2.5	1
b0	0.80	3.50	3.50
b1	0.77	0.77	1.50

42

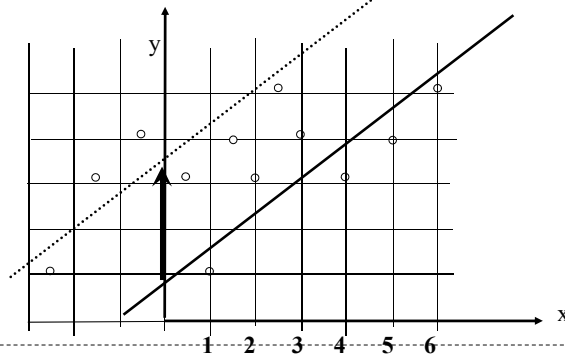
Kenneth S. Law@2016

Grand Mean Centering

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij} \quad y = .80 + .77x$$

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{..}) + r_{ij} \quad y = 3.5 + .77x$$

$$Y_{ij} = (\beta_{0j} - \beta_{1j}\bar{X}_{..}) + \beta_{1j}X_{ij} + r_{ij} \quad y = 3.5 + 1.5x$$



▷ 43

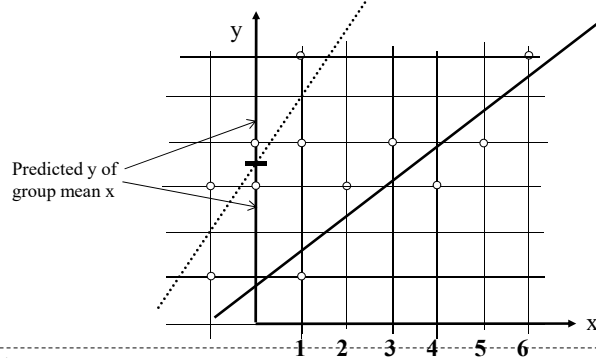
Kenneth S. Law@2016

Group Mean Centering

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij} \quad y = .80 + .77x$$

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij} \quad y = 3.5 + .77x$$

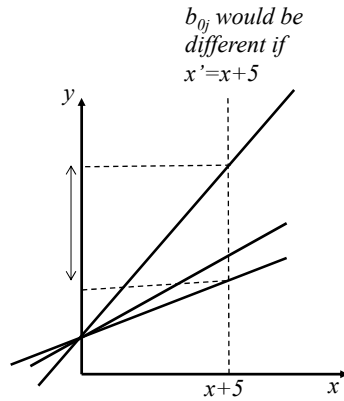
$$Y_{ij} = (\beta_{0j} - \beta_{1j}\bar{X}_{.j}) + \beta_{1j}X_{ij} + r_{ij} \quad y = 3.5 + 1.5x$$



▷ 44

Kenneth S. Law@2016

Effect of shifting on intercept & variance



- Not only the mean values of b_{0j} would be different, the variance of b_{0j} , i.e., τ_{00} would also be different with different centering options
- The same is true for b_{1j} and τ_{11} & τ_{01} .

For the original x values, the three groups would have exactly the same intercept.

Group Mean Centering

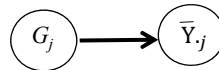
- when group mean centering is adopted, the level-1 intercept variance is equal to the between group variance in the outcome measure.
- As a result, the level-2 regression coefficients, under group mean centering, simply represent the group level relationship between the level-2 predictor and the outcome variable of interest (i.e., the relationship between the level-2 predictor, X_j , and Y_j).

$$\begin{aligned}
 Y_{ij} &= \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{\cdot j}) + r_{ij} \\
 \beta_{0j} &= \gamma_{00} + \gamma_{01} G_j + u_{0j} \\
 \beta_{1j} &= \gamma_{10} + \gamma_{11} G_j + u_{1j}
 \end{aligned}$$

$$E(Y_{ij}) = E(\beta_{0j}) + \beta_{1j} E(X_{ij} - \bar{X}_{\cdot j}) + E(r_{ij})$$

$$\bar{Y}_{\cdot j} = \beta_{0j}$$

$$\beta_{0j} = \bar{Y}_{\cdot j}$$



Grand Mean Centering

- when grand mean centering is adopted, the variance in the intercept term represents the between group variance in the outcome measure adjusted for the level-1 predictor(s).
- The level-2 regression coefficients represent the group level relationship between the level-2 predictor and the outcome variable less the influence of the level-1 predictor(s).

$$\begin{aligned}
 Y_{ij} &= \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{..}) + r_{ij} \\
 \beta_{0j} &= \gamma_{00} + \gamma_{01}G_j + u_{0j} \\
 \beta_{1j} &= \gamma_{10} + \gamma_{11}G_j + u_{1j}
 \end{aligned}$$

$$\begin{aligned}
 E(Y_{ij}) &= E(\beta_{0j}) + \beta_{1j}E(X_{ij} - \bar{X}_{..}) + E(r_{ij}) \\
 \bar{Y}_{.j} &= \beta_{0j} + \beta_{1j}\bar{X}_{.j} - \beta_{1j}\bar{X}_{..} \\
 \beta_{0j} &= \bar{Y}_{.j} - \beta_{1j}(\bar{X}_{.j} - \bar{X}_{..})
 \end{aligned}$$

$G_j \rightarrow \bar{Y}_{.j} - \beta_{1j}\bar{X}_{.j} + \beta_{1j}\bar{X}_{..}$

▷ 47

Kenneth S. Law@2016

Mplus Random coefficient model

TITLE: Sample multi-level program on helping

DATA: FILE = helping.txt;

DEFINE: center help (grandmean);

VARIABLE: NAMES = group help mood prox;

USEVARIABLES = help mood;

WITHIN = mood;

CLUSTER = group;

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:

%WITHIN%

help;

s | help on mood;

%BETWEEN%

help;

s;

help WITH s;

$$\begin{aligned}
 help_{ij} &= \beta_{0j} + \beta_{1j}(mood_{ij} - \overline{mood}_{..}) + r_{ij} \\
 \beta_{0j} &= \gamma_{00} + u_{0j} \\
 \beta_{1j} &= \gamma_{10} + u_{1j}
 \end{aligned}$$

▷ 48

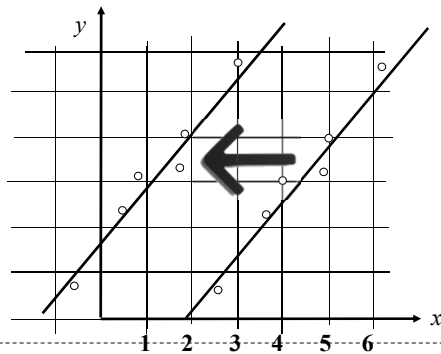
OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Grand Mean Centering

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij}$$

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{..}) + r_{ij}$$



▷ 49

Kenneth S. Law@2016

Mplus HLM output

$$help_{ij} = \beta_{0j} + \beta_{1j} (mood_{ij} - \overline{mood}_{..}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

	No Centering		Grand Mean centering	
	Estimate	P-Value	Estimate	P-Value
Within Level				
Residual Variances				
HELP	5.611	0.000	5.611	0.000
Between Level				
HELP WITH S	-0.105	0.813	0.622	0.156
Means				
HELP	13.837	0.000	31.428	0.000
S	3.013	0.000	3.013	0.000
Variances				
HELP	41.681	0.000	44.702	0.000
S	0.125	0.001	0.125	0.001

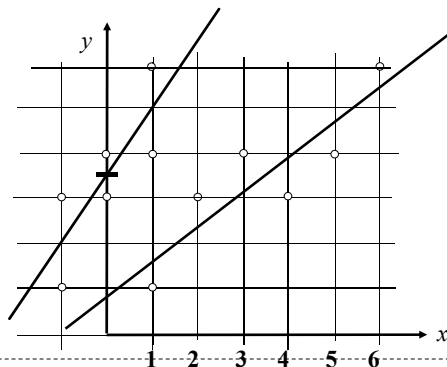
▷ 50

Kenneth S. Law@2016

Group Mean Centering

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij} \quad y = .80 + .77x$$

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{\cdot j}) + r_{ij} \quad y = 3.5 + 1.5x$$



▷ 51

Kenneth S. Law@2016

Mplus Random coefficient model

TITLE: Sample multi-level program on helping

DATA: FILE = helping.txt;

DEFINE: center help (groupmean);

VARIABLE: NAMES = group help mood prox;

USEVARIABLES = help mood;

WITHIN = mood;

CLUSTER = group;

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:

%WITHIN%

help;

s | help on mood;

%BETWEEN%

help;

s;

help WITH s;

$$help_{ij} = \beta_{0j} + \beta_{1j} (mood_{ij} - \overline{mood}_{\cdot j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

▷ 52

OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus HLM output

$$help_{ij} = \beta_{0j} + \beta_{1j} (mood_{ij} - \overline{mood_{.j}}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

	Estimate	S.E.	Est./S.E.	
Within Level				
Residual Variances				
HELP	5.607	0.304	18.445	0.000
Between Level				
HELP WITH S	0.239	0.639	0.374	0.708
Means				
HELP	γ_{00} 31.394	1.410	22.269	0.000
S	γ_{10} 2.999	0.069	43.632	0.000
Variances				
HELP	τ_{00} 99.095	15.275	6.488	0.000
S	τ_{11} 0.127	0.038	3.372	0.001

▷ 53 Kenneth S. Law@2016

Mplus HLM output

$$help_{ij} = \beta_{0j} + \beta_{1j} (mood_{ij} - \overline{mood_{.j}}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

	No Centering		Grand Mean centering		Group Mean centering	
	Estimate	P-Value	Estimate	P-Value	Estimate	P-Value
Within Level						
Residual Variances						
HELP	5.611	0.000	5.611	0.000	5.607	0.000
Between Level						
HELP WITH S	-0.105	0.813	0.622	0.156	0.239	0.708
Means						
HELP	13.837	0.000	31.428	0.000	31.394	0.000
S	3.013	0.000	3.013	0.000	2.999	0.000
Variances						
HELP	41.681	0.000	44.702	0.000	99.095	0.000
S	0.125	0.001	0.125	0.001	0.127	0.001

▷ 54 Kenneth S. Law@2016

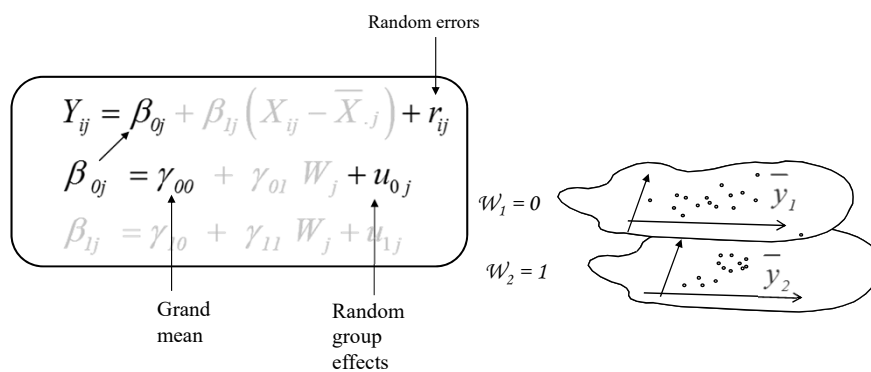
A recap

Examples of different HLM models

▷ 55

Kenneth S. Law@2016

Model 1



- Y_{ij} is affected by:
- (1) A grand mean;
 - (2) The random (group) effects
 - (3) Random (individual) errors

▷ 56

Kenneth S. Law@2016

One-way ANOVA random effects

Level 1 $Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$ random errors

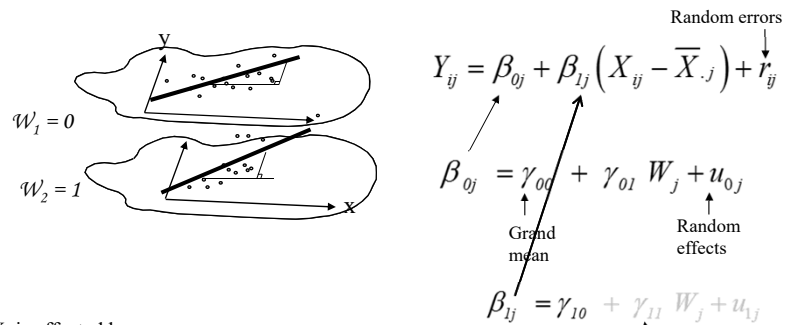
Level 2 $\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$ random effects

$\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$

$$Y_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

Model 2



- Y_{ij} is affected by:
- (1) A grand mean;
 - (2) Random (group) effects
 - (3) A third variable (X) effects
 - (4) Random (individual) errors

Note: the X-effect (covariance effect) is constant. It is neither affected by another variable W , nor having a random component.

One-way ANCOVA random effects

Level 1 $y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$

Level 2 $\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$
 $\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$

$$y_{ij} = \beta_{0j} + \beta_{1j}x_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

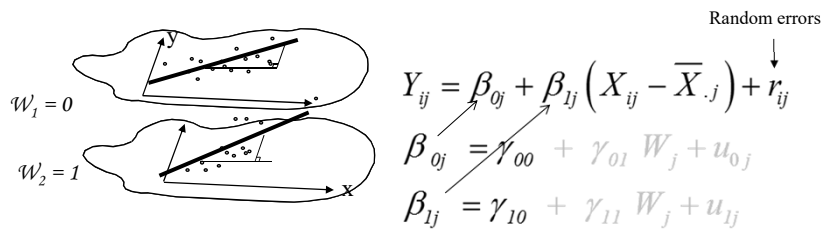
$$\beta_{1j} = \gamma_{10}$$

random effects

No random effects

59

Model 3



60

Kenneth S. Law@2016

Simple regression models

Level 1 $Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$

Level 2 $\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$

$\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$

$Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$

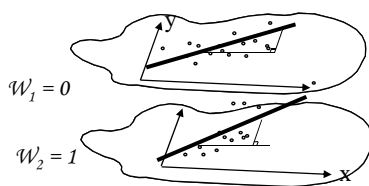
$\beta_{0j} = \gamma_{00}$

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

Level-2 effect is not modeled, it reduces to Level-1 effect only.

61

Model 4



$$Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$$

↑ Random errors
↓

$$\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$$

↑ Grand mean
↑ Random effects

- Y_{ij} is affected by:
- (1) A grand mean;
 - (2) Random (group) effects
 - (3) A third variable (X) effects
 - (4) Random (individual) errors

62

Kenneth S. Law@2016

Mean-as-outcome regression

Level 1 $Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij}$

Level 2 $\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$

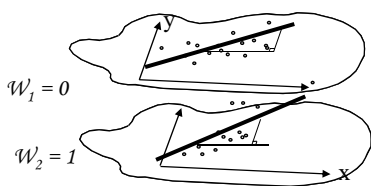
$$Y_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$$

The *means* from each of many groups *as an outcome* to be predicted by group characteristics.

63

Model 5



- Y_{ij} is affected by:
- (1) A grand mean;
 - (2) Random (group) effects
 - (3) A third variable (X) effects
 - (4) Random (individual) errors

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij}$$

↓ Random errors

$$\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$$

← Random effects

$$\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$$

↑ Grand mean

64

Kenneth S. Law@2016

Intercepts-and-Slopes-as-outcome regression

$$\text{Level 1} \quad Y_{ij} = \beta_{0j} + \beta_{1j} (X_{ij} - \bar{X}_{.j}) + r_{ij}$$

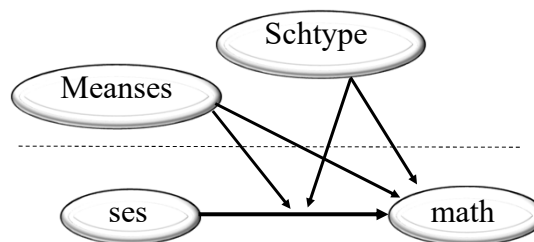
$$\text{Level 2} \quad \beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$$

- Both the *intercepts* and *slopes as outcomes* to be *regressed* (predicted by W_j).
- Have you studied all W_j of interests, or do you want to generalize to other W_j s.

65

Another example



66

Kenneth S. Law@2016

Model 1: Null model

```

TITLE:      Sample multi-level program on helping (hlm_sem1)

DATA:      FILE = hlm_sem.dat;

VARIABLE:  NAMES = schid minority female ses math size schtype meanses;
           USEVARIABLES = math;
           WITHIN = ;      ! level 1 variables here (none)
           BETWEEN = ;    ! level 2 variables here (none)
           CLUSTER = schid;

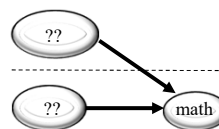
ANALYSIS:  TYPE = TWOLEVEL RANDOM;

MODEL:
  %WITHIN%
  math;      ! no fixed effects
  %BETWEEN%
  math;      ! no predictors of intercept

OUTPUT:    TECH1 TECH3;
    
```

$$math_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$



▷ 67

Kenneth S. Law@2016

Mplus output

	Estimate	S.E.	Est./S.E.	P-Value
Within Level				
error variance level-1				
Variiances				
MATH	39.148	0.835	46.876	0.000
Between Level				
Means				
MATH	12.637	0.244	51.823	0.000
Variiances				
MATH	8.562	1.057	8.100	0.000
error variance level-2				

$$math_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

The within variable is MATH, $\sigma^2 = 39.148$. This is the variance of r_{ij} , which is constant across level-2 units.

▷ 68

Kenneth S. Law@2016

Model 2: A level-2 predictor

```

TITLE:      Sample multi-level program on helping (hlm_sem2)

DATA:      FILE = hlm_sem.dat;

VARIABLE:  NAMES = schid minority female ses math size schtype meanses;
           USEVARIABLES = math meanses;
           WITHIN = ;
           BETWEEN = meanses;
           CLUSTER = schid;

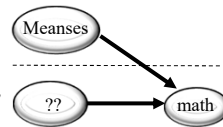
ANALYSIS:  TYPE = TWOLEVEL RANDOM;

MODEL:
  %WITHIN%
  math;
  %BETWEEN%
  math ON meanses; ! Level 2 predictor of mean ses

OUTPUT:    TECH1 TECH3;
    
```

$$math_{ij} = \beta_{0j} + r_{ij}$$

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



▷ 69

Kenneth S. Law@2016

Mplus output

	Estimate	S.E.	Est./S.E.	P-Value
Within Level				
Variances				
MATH	39.157	0.836	46.838	0.000
Between Level				
<i>Meanses can be used to predict mean Math between groups</i>				
MATH ON MEANSSES	5.863	0.321	18.276	0.000
Intercepts				
MATH	12.650	0.148	85.575	0.000
Residual Variances				
MATH	2.598	0.467	5.558	0.000

$$math_{ij} = \beta_{0j} + r_{ij}$$

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

▷ 70

Kenneth S. Law@2016

Model 3: A level-1 predictor

```

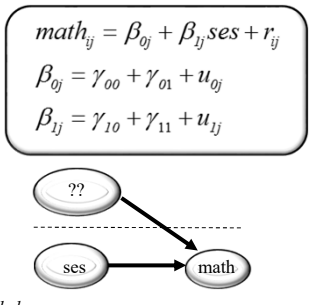
TITLE:      Sample multi-level program on helping (hlm_sem3)

DATA:      FILE = hlm_sem.dat;

VARIABLE:  NAMES = schid minority female ses math size schtype meanses;
           USEVARIABLES = math ses;
           WITHIN = ses;  ! Level 1 predictor of math
           BETWEEN = ;
           CLUSTER = schid;

ANALYSIS:  TYPE = TWOLEVEL RANDOM;

MODEL:
  %WITHIN%
  math; ! Mean (intercept) of math for each group
  s | math ON ses; ! slope for each group
  %BETWEEN%
  math; ! nothing predicts intercept
  s; ! nothing predicts slope
  math with s; ! Covariance between intercept and slope
  
```

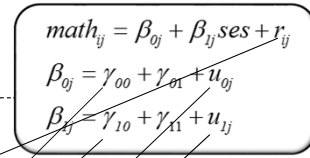


▷ 71 OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus output

	Estimate	S.E.	Est./S.E.	P-Value
Within Level	σ^2 dropped from 39.148 to 36.820			
Residual Variances				
MATH	36.831	0.725	50.779	0.000
Between Level				
MATH WITH				
S	-0.167	0.326	-0.513	0.608
Means				
MATH	12.666	0.203	62.281	0.000
S	2.394	0.128	18.646	0.000
Variances				
MATH	4.781	0.756	6.324	0.000
S	0.402	0.240	1.670	0.095



▷ 72

Kenneth S. Law@2016

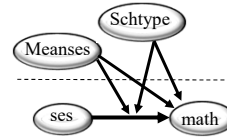
Model 4: Both level-1 & level-2 predictor

TITLE: Sample multi-level program on helping (hlm_sem4)

DATA: FILE = hlm_sem.dat;
 DEFINE: center ses (groupmean);
 VARIABLE: NAMES = schid minority female ses math size schtype meanses;
 USEVARIABLES = math ses schtype meanses;
 WITHIN = ses;
 BETWEEN = schtype meanses;
 CLUSTER = schid;

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:
 %WITHIN%
 math;
 s | math ON ses;
 %BETWEEN%
 math ON schtype meanses;
 s ON schtype meanses;
 math WITH s



$$math_{ij} = \beta_{0j} + \beta_{1j}(ses_{ij} - \overline{ses}_j) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schttype_j + \gamma_{02}meanses_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}schttype_j + \gamma_{12}meanses_j + u_{1j}$$

▷ 73 OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus output

	Estimate	S.E.	Est./S.E.	P-Value
Within Level	σ^2 dropped from			
Residual Variances	39.148 to 36.720			
MATH	σ^2 36.720	0.721		
Between Level				
S ON				
SCHTYPE	γ_{11} -1.640	0.238	-6.905	0.000
MEANSES	γ_{12} 1.033	0.333	3.100	0.002
MATH ON				
SCHTYPE	γ_{01} 1.227	0.308	3.982	0.000
MEANSES	γ_{02} 5.332	0.336	15.871	0.000
MATH WITH				
S	0.200	0.192	1.041	0.298
Intercepts				
MATH	γ_{00} 12.096	0.174	69.669	0.000
S	γ_{10} 2.938	0.147	19.986	0.000
Residual Variances				
MATH	τ_{00} 2.316	0.414	5.591	0.000
S	τ_{11} 0.071	0.201	0.352	0.725

$$math_{ij} = \beta_{0j} + \beta_{1j}(ses_{ij} - \overline{ses}_j) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schttype_j + \gamma_{02}meanses_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}schttype_j + \gamma_{12}meanses_j + u_{1j}$$

▷ 74

Kenneth S. Law@2016

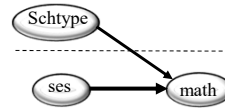
Model 5: Fixed effect for slope

TITLE: Sample multi-level program on helping (hlm_sem4)

DATA: FILE = hlm_sem.dat;

DEFINE: center ses (groupmean);

VARIABLE: NAMES = schid minority female ses math size schtype meanses;
 USEVARIABLES = math ses schtype;
 WITHIN = ses;
 BETWEEN = schtype;
 CLUSTER = schid;



$$math_{ij} = \beta_{0j} + \beta_{1j}(ses_{ij} - \overline{ses}_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schtype_j + u_{0j}$$

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

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:

```
%WITHIN%
  math ON ses;  Note: NO s| math on ses for fixed effect
%BETWEEN%
  math ON schtype;
  ! Nothing to predict s since it is a fixed effect
  ! NO math WITH s since s will not change across level
```

OUTPUT: TECH1 TECH3;

▷ 75

Kenneth S. Law@2016

Mplus output

$$math_{ij} = \beta_{0j} + \beta_{1j}(ses_{ij} - \overline{ses}_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schtype_j + u_{0j}$$

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

	Estimate	S.E.	Est./S.E.	P-Value
Within Level				
MATH ON				
SES	γ_{10} 2.191	0.129	16.938	0.000
Residual Variances				
MATH	σ^2 37.008	0.715	51.771	0.000
Between Level				
MATH ON				
SCHTYPE	γ_{01} 2.805	0.436	6.434	0.000
Intercepts				
MATH	γ_{00} 11.393	0.292	38.959	0.000
Residual Variances				
MATH	τ_{00} 6.643	0.869	7.645	0.000

▷ 76

Kenneth S. Law@2016

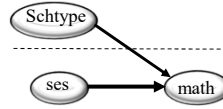
Model 6: Random and fixed effect for slope

TITLE: Sample multi-level program on helping (hlm_sem4)

DATA: FILE = hlm_sem.dat;

DEFINE: center ses (groupmean);

VARIABLE: NAMES = schid minority female ses math size schtype meanses;
 USEVARIABLES = math ses female schtype;
 WITHIN = ses female;
 BETWEEN = schtype;
 CLUSTER = schid;



$$math_{ij} = \beta_{0j} + \beta_{1j}ses + \beta_{2j}female + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schtype_j + u_{0j}$$

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

$$\beta_{2j} = \gamma_{20}$$

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:

%WITHIN%

s | math ON ses; *Note: NO s | math on ses for fixed effect*

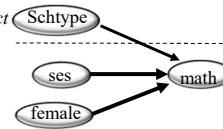
math ON female;

%BETWEEN%

math ON schtype;

s; *! Nothing to predict s*

math WITH s;



77 OUTPUT: TECH1 TECH3;

Kenneth S. Law@2016

Mplus output

	Estimate	S.E.		
Within Level				
MATH ON				
FEMALE	-1.199	0.182	-6.599	0.000
Residual Variances				
MATH	σ^2 36.602	0.714	51.262	0.000
Between Level				
MATH ON				
SCHTYPE	γ_{01} 2.548	0.405	6.286	0.000
MATH WITH				
S	0.704	0.347	2.030	0.042
Means				
S	γ_{10} 2.352	0.125	18.775	0.000
Intercepts				
MATH	γ_{00} 12.092	0.302	39.997	0.000
Variances				
S	τ_{11} 0.360	0.228	1.575	0.115
Residual Variances				
MATH	τ_{00} 3.626	0.629	5.766	0.000

$$math_{ij} = \beta_{0j} + \beta_{1j}ses + \beta_{2j}female + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}schtype_j + u_{0j}$$

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

$$\beta_{2j} = \gamma_{20}$$

78

Kenneth S. Law@2016

Mplus programming

▷ 79

Kenneth S. Law@2016

Mplus Programming: random intercept

TITLE: two-level regression with a random intercept and an observed covariate

DATA: FILE = ex9.1a.dat;

VARIABLE:

NAMES = y x w xm clus;

WITHIN = x; ← Level 1

BETWEEN = w xm; ← Level 2

CLUSTER = clus; ←

DEFINE:

CENTER x (GRANDMEAN);

ANALYSIS:

TYPE = TWOLEVEL; ←

MODEL:

%WITHIN%

y ON x;

%BETWEEN%

y ON w xm; ←

- (1) No need to declare whether a variable is between or within.
- (2) If you do not state whether a variable is WITHIN or BETWEEN, Mplus assumes that it has both between and within variances.

Grouping variable 1, 2, 3, 4, ...

Multilevel model with random intercept

$$Y_{ij} = \beta_{0j} + \beta_1 X_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + \gamma_{02} \bar{x}_{.j} + u_{0j}$$

▷ 80

Kenneth S. Law@2016

Mplus Output

SUMMARY OF DATA

Number of clusters 110

Average cluster size 9.091

Estimated Intraclass Correlations for the Y
Variables

	Intraclass Variable Correlation
--	------------------------------------

Y	0.570
---	-------

▷ 81

Kenneth S. Law@2016

Mplus Output

SUMMARY OF DATA

Number of clusters 110

Average cluster size 9.091

Estimated Intraclass Correlations for the Y
Variables

	Intraclass Variable Correlation
--	------------------------------------

Y	0.570
---	-------

THE MODEL ESTIMATION TERMINATED
NORMALLY

▷ 82

Kenneth S. Law@2016

Mplus Output

MODEL FIT INFORMATION

Number of Free Parameters 6

Loglikelihood

H0 Value -1525.938
 H0 Scaling Correction Factor 0.9402
 for MLR
 H1 Value -1525.938
 H1 Scaling Correction Factor 0.9402
 for MLR

RMSEA (Root Mean Square Error Of Approximation)

Estimate 0.000

CFI/TLI

CFI 1.000
 TLI 1.000

Information Criteria

Akaike (AIC) 3063.876
 Bayesian (BIC) 3093.322
 Sample-Size Adjusted BIC 3074.266
 (n* = (n + 2) / 24)

Chi-Square Test of Model Fit for the Baseline Model

Value 491.881
 Degrees of Freedom 3
 P-Value 0.0000

Chi-Square Test of Model Fit

Value 0.000*
 Degrees of Freedom 0
 P-Value 0.0000
 Scaling Correction Factor 1.0000
 for MLR

SRMR (Standardized Root Mean Square Residual)

Value for Within 0.000
 Value for Between 0.000

83 00

Kenneth S. Law@2016

Mplus Output

MODEL RESULTS

		Estimate	S.E.	Two-Tailed Est./S.E.	P-Value
Within Level					
Y	ON	β_1			
X		0.724	0.033	22.118	0.000
Residual Variances σ^2					
Y		1.022	0.041	25.117	0.000

$$Y_{ij} = \beta_{0j} + \beta_1 X_{ij} + r_{ij}$$

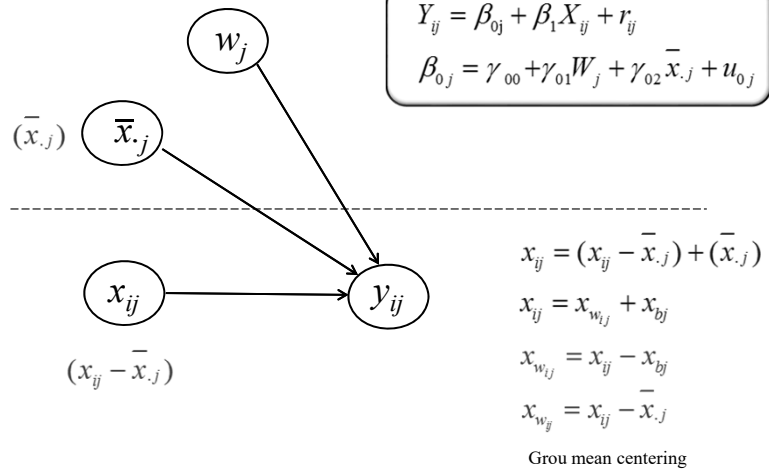
$$\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + \gamma_{02} \bar{x}_{.j} + u_{0j}$$

Between Level					
Y	ON	γ_{01}			
W		0.570	0.108	5.305	0.000
XM		0.976	0.160	6.107	0.000
Intercepts γ_{02}					
Y		1.991	0.080	24.804	0.000
Residual Variances γ_{00}					
Y		0.571	0.088	6.486	0.000
τ_{00}					

84

Kenneth S. Law@2016

Partitioning within and between variance



▷ 85

Kenneth S. Law@2016

Hypothetical data

	innovation		performance
Group	x	Mean x	y
1	3	4	5
1	4	4	5
1	5	4	5
2	1	2	3
2	2	2	3
2	3	2	3
3	2	3	4
3	3	3	4
3	4	3	4

$r_{xy} = 1.0 \quad r_{yx} = .71$

▷ 86

Kenneth S. Law@2016

Partitioning within and between variance

```

TITLE: Partitioning variances
DATA: FILE = ex9.1b.dat;
VARIABLE: NAMES = y x w clus;
BETWEEN = w;
CLUSTER = clus;
DEFINE: CENTER = x (GRANDMEAN);
ANALYSIS: TYPE = TWOLEVEL;
MODEL:
%WITHIN%
  y ON x (g01);
%BETWEEN%
  y ON w
  x (g10);
MODEL CONSTRAINT:
NEW(diff);
diff = g10 - g01;
    
```

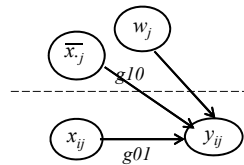
Note: the data does not have xm (we estimate xm from the data)

We do not declare x as WITHIN

Same random intercept model

Effect of within group x on y

Effect of mean x on y



▷ 87

Kenneth S. Law@2016

Mplus Programming: random intercept

```

TITLE: two-level regression with a
random intercept and an observed covariate
DATA: FILE = ex9.1a.dat;
VARIABLE:
NAMES = y x w xm clus;
WITHIN = x;
BETWEEN = w xm;
CLUSTER = clus
DEFINE:
  CENTER x (GRANDMEAN);
ANALYSIS:
  TYPE = TWOLEVEL;
MODEL:
%WITHIN%
  y ON x;
%BETWEEN%
  y ON w xm;
    
```

- (1) No need to declare whether a y variable is between or within.
- (2) If you do not state whether a variable is WITHIN or BETWEEN, Mplus assumes that it has both between and within variances.

▷ 88

Kenneth S. Law@2016

Full HLM model

Random slope and intercept

Slope and intercept as outcome regression

▷ 89

Kenneth S. Law@2016

Mplus : random slope & intercept

TITLE: two-level regression with a random intercept and an observed covariate

DATA: FILE = ex9.2a.dat;

VARIABLE:

NAMES = y x w xm clus;

WITHIN = x;

BETWEEN = w xm;

CLUSTER = clus;

DEFINE: CENTER x (GRANDMEAN);

ANALYSIS: TYPE = TWOLEVEL RANDOM;

MODEL:

%WITHIN%

s | y ON x;

%BETWEEN%

y s ON w xm;

y WITH s

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + \gamma_{02}\bar{x}_{.j} + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + \gamma_{12}\bar{x}_{.j} + u_{1j}$$

S is the slope of x→y for each group

w and xm affect both the intercept y and the slope s

The random slope is correlated with the random intercept

▷ 90

Kenneth S. Law@2016

Mplus Output

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
--	----------	------	-----------	--------------------

Within Level

Residual Variances	σ^2				
Y	1.032	0.047	22.126	0.000	

No slope or intercept estimate at the within level because they are random

Between Level

S	ON				
W	γ_{11}	0.396	0.097	4.063	0.000
XM	γ_{12}	0.542	0.136	4.001	0.000
Y	ON				
W	γ_{01}	0.874	0.118	7.387	0.000
XM	γ_{02}	1.345	0.164	8.186	0.000

$$Y_{ij} = \beta_{0j} + \beta_{1j}X_{ij} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + \gamma_{02}\bar{x}_{\cdot j} + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + \gamma_{12}\bar{x}_{\cdot j} + u_{1j}$$

Y	WITH				
S		0.306	0.067	4.572	0.000

Correlation of random slope and random intercept

Intercepts

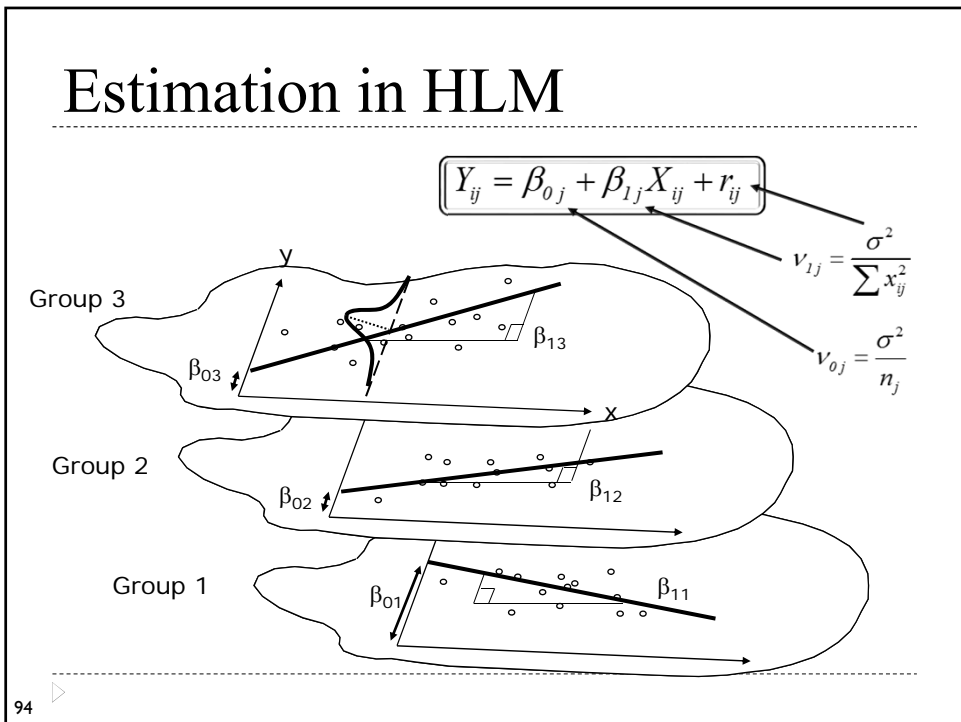
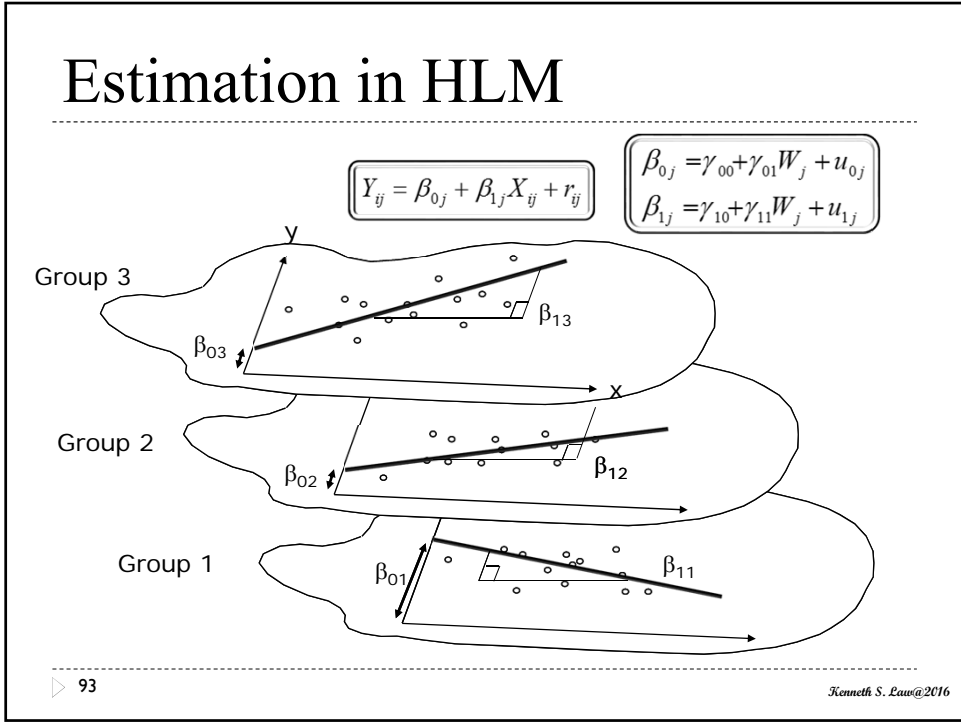
Y	γ_{00}	2.113	0.088	24.021	0.000
S	γ_{10}	1.039	0.075	13.813	0.000

Residual Variances

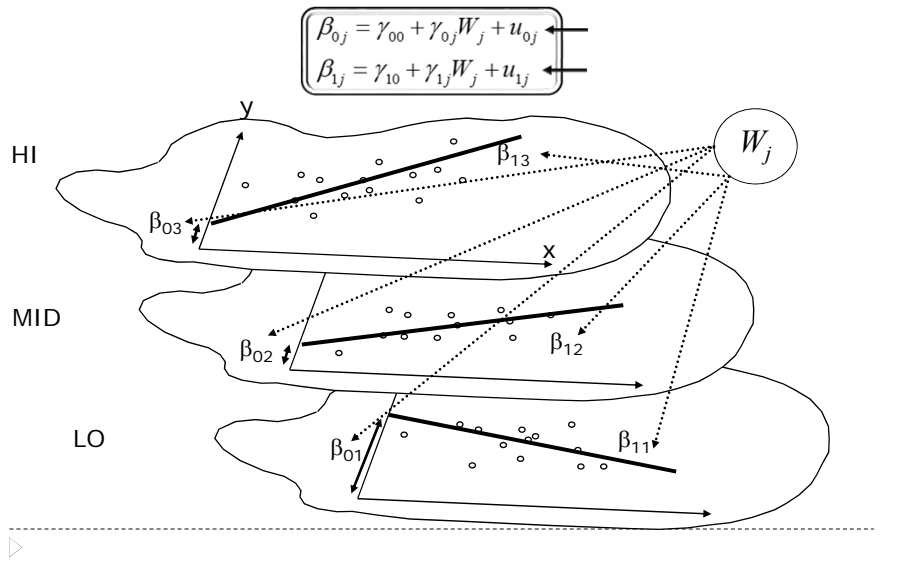
Y	τ_{00}	0.606	0.101	5.982	0.000
S	τ_{11}	0.334	0.056	5.932	0.000

Estimation

Some issues on HLM estimates



Estimation in HLM

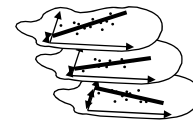


95

Two sets of estimates (I)

- ▶ For a particular team (e.g., Team 2), one can simply run an OLS regression and estimate the Level 1 effect (β_{02} and β_{12}) using the Level 1 equation:

$$Y_{i2} = \beta_{02} + \beta_{12}X_{i2} + r_{i2}$$



- ▶ After knowing β_{0k} and β_{1k} , one can then estimate γ_{00} , γ_{01} , γ_{10} , and γ_{11} using W_j in the following Level 2 equations:

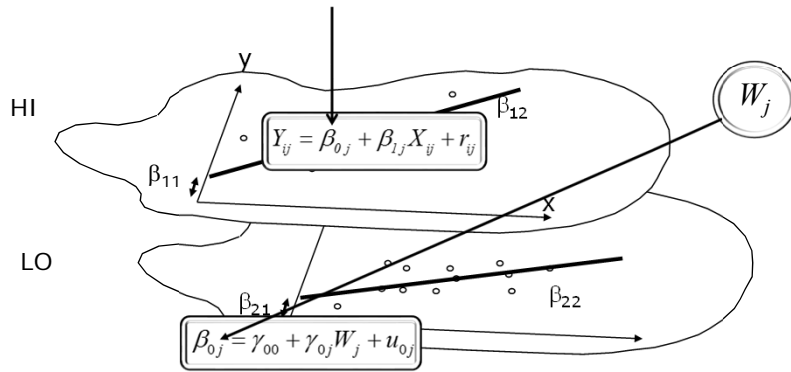
$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j} \quad (\text{when } j = 2)$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j} \quad (\text{when } j = 2)$$

- ▶ One then has two sets of β estimates, those estimated by Level 1 data and those estimated by Level 2 data.

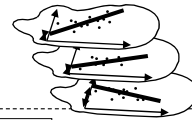
96

Estimation in HLM



97

Two sets of estimates (II)



- ▶ β s as estimated using Level 1 data $Y_{ijk} = \beta_{0k} + \beta_{1k} X_{ijk} + r_{ijk}$
- ▶ β s as estimated using Level 2 data $\beta_{0j} = \gamma_{00} + \gamma_{01} W_j + u_{0j}$
 $\beta_{1j} = \gamma_{10} + \gamma_{11} W_j + u_{1j}$

- ▶ One then has two set of β s estimates.
- ▶ HLM provides an overall weighted average estimate of the using these two sets of β estimates. The weights are obtained by partitioning the observed variances of β s into a true variance vs. error variance components. The resulting composite β estimates produces a smaller mean square error term than either the Level 1- and Level 2-estimates.

98

Estimation

- ▶ The first estimator of the β_j is simply the OLS regression estimator based on data from team j:

$$OLS(\hat{\beta}_j) = (X_j'X_j)^{-1}X_j'Y_j$$

- ▶ The second estimator is the predicted value of β_j given team culture captured by W_j :

$$Second(\hat{\beta}_j) = W_j \hat{\gamma}$$

- ▶ The optimal combination of these two estimators is:

$$optimal(\beta_j) = \Lambda_j \hat{\beta}_j + (I - \Lambda_j) W_j \hat{\gamma}$$

For each level of Level 2:

$$\lambda_{0j} = \tau_{00} / (\tau_{00} + \nu_{0j})$$

$$\lambda_{1j} = \tau_{11} / (\tau_{11} + \nu_{1j})$$

Where

$$\Lambda_j = T\Delta_j^{-1} = T(T + V_j)^{-1}$$

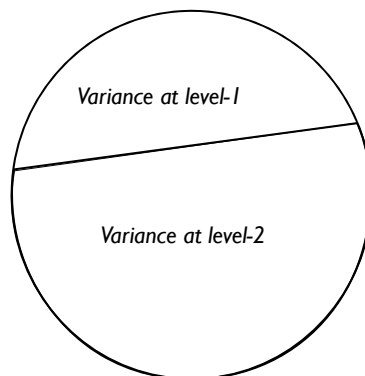
$$\nu_{0j} = \frac{\sigma^2}{n_j} \quad \nu_{1j} = \frac{\sigma^2}{\sum x_{ij}^2}$$

T is the parameter dispersion (variance of the β_j or the τ matrix)

V_j is the error dispersion (variance of random error within each group)

Λ_j is the ratio of true variance to total variance or the **reliability** of the parameters

Percentage of variance accounted for



$$y_{ij} = \beta_{0j} + \beta_{1j}(x_{ij} - \bar{x}_j) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}w_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}w_j + u_{1j}$$

- How much of the variances of y_{ij} is due to level-1 and how much to level-2?
- How much variance of y_{ij} is accounted for at level-1, how much as level-2?

Percentage of variance explained when there is a group-level predictor

$$\% \text{ variance explained in } \beta_{qj} = \frac{\hat{\tau}_{qq}(\text{random regression}) - \hat{\tau}_{qq}(\text{fitted model})}{\hat{\tau}_{qq}(\text{random regression})}$$

Random Regression Model

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

$\hat{\tau}_{qq}(\text{random regression})$

Fitted Model

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

$\hat{\tau}_{qq}(\text{fitted model})$

101 ▷

Percentage of variance explained in HLM

Model 1

$$y_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

Model 3

$$y_{ij} = \beta_{0j} + \beta_{1j}(SES_{ij} - MEANSES_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + u_{0j}$$

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

Model 2

$$y_{ij} = \beta_{0j} + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}(MEANSES)_{.j} + u_{0j}$$

Model 4

$$y_{ij} = \beta_{0j} + \beta_{1j}(SES_{ij} - MEANSES_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}(MEANSES)_{.j} + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}(MEANSES)_{.j} + u_{1j}$$

	Var	Model 1	Model 2	Model 3	Model 4
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

▷ 102

HLM manual

Kenneth S. Law@2016

% variance accounted for

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

$$Ratio\ 1 = \frac{\tau_{00}(M_1)}{\tau_{00}(M_1) + \sigma^2(M_1)} = \frac{8.62}{8.62 + 39.15} = .1804$$

18% of variance of Math attributable to level 2 (school level);
82% of variance is at the individual level (level 1).

% variance accounted for: Level 1

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

$$Ratio\ 2 = \frac{\sigma^2(M_1) - \sigma^2(M_3)}{\sigma^2(M_1)} = \frac{39.15 - 36.71}{39.15} = .0623$$

6% of individual Math variance is explained by SES.

% variance accounted for: level 2

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

$$Ratio\ 3 = \frac{\tau_{00}(M_1) - \tau_{00}(M_2)}{\tau_{00}(M_1)} = \frac{8.62 - 2.64}{8.62} = .6937$$

The level 1 variable SES helped tremendously ($\downarrow 69.37\%$) to explain the math performance across schools.

▷ 105

Kenneth S. Law@2016

% variance accounted for: intercept/slope

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

$$Ratio\ 4a = \frac{\tau_{00}(M_3) - \tau_{00}(M_4)}{\tau_{00}(M_3)} = \frac{8.68 - 2.65}{8.68} = .6947$$

$$Ratio\ 4b = \frac{\tau_{11}(M_3) - \tau_{11}(M_4)}{\tau_{11}(M_3)} = \frac{.68 - .66}{.68} = .0002$$

Adding MEANSES explained intercept, but not slope.

▷ 106

Kenneth S. Law@2016

% variance accounted for

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}	18%		.68	.66

$$Ratio\ 5 = \frac{\tau_{00}(M_4)}{\tau_{00}(M_4) + \sigma^2(M_4)} = \frac{2.65}{2.65 + 36.71} = .0673$$

When compare M_1 and M_4 , the % level 2 residual variance τ_{00} dropped from 18.04% of total residual variance to 6.73% (about 1/3), meaning that MEANSES is a more effective predictor than individual SES.

% variance accounted for

	Var	Model 1	Model 2	Model 3	Model 4
		baseline	Level2 predictor	Level1 predictor	Level1&2 predictor
r_{ij}	σ^2	39.15	39.16	36.71	36.71
u_{0j}	τ_{00}	8.62	2.64	8.68	2.65
u_{1j}	τ_{11}			.68	.66

σ^2 student variance in math;
 τ_{00} school variance in math)

Ratio 1
 (mainly σ^2 , small τ_{00})

Ratio 2
 (SES reduced σ^2 by only 6%)

Ratio 3
 (τ_{00} ↓ 69% by adding MEANSES)

Ratio 4
 (Add MEANSES reduced τ_{00} but not τ_{11})

Ratio 5
 (τ_{00} ↓ 1/3 by adding SES and MEANSES)

Multiple Xs

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{1ij} - \bar{X}_{1.j}) + \beta_{2j}(X_{2ij} - \bar{X}_{2.j}) + r_{ij}$$

As discussed β_{0j} and β_{1j} are modeled as left:

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

β_{2j} can be modeled in four different forms:

1. The effect of X_{2ij} is constrained to be invariant across Level-2 units;

$$\beta_{2j} = \gamma_{20}$$

2. The slope β_{2j} is a function of an average value γ_{20} , plus a random effect associated with each Level-2 unit;

$$\beta_{2j} = \gamma_{20} + u_{2j}$$

3. Part of the variation of the slope β_{2j} can be predicted by W_j , but a random components, u_{2j} , remains unexplained;

$$\beta_{2j} = \gamma_{20} + \gamma_{21}W_j + u_{2j}$$

4. Once the effect of W_j is taken into account, the residual variation in β_{2j} [$\text{Var}(u_{2j}) = \tau_{22}$] is negligible.

$$\beta_{2j} = \gamma_{20} + \gamma_{21}W_j$$

▷ 109

Kenneth S. Law@2016

Level 1 & 2 models

$$Y_{ij} = \beta_{0j} + \beta_{1j}(X_{ij} - \bar{X}_{.j}) + r_{ij}$$

$$\beta_{0j} = \gamma_{00} + \gamma_{01}W_j + u_{0j}$$

$$\beta_{1j} = \gamma_{10} + \gamma_{11}W_j + u_{1j}$$

Level	1	2
Units	subordinates	teams
Errors (random effects)	r_{ij}	u_{0j}, u_{1j}
Variance	$\text{Var}(r_{ij}) = \sigma^2$	$\text{Var}(u_{0j}), \text{Var}(u_{1j}), \text{Cov}(u_{0j}, u_{1j})$
Parameters	β	γ
Predictor	X_{ij}	W_j

▷ 110

Kenneth S. Law@2016

