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## **A Primer in Multilevel Modeling for Actuarial Applications**

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# Agenda

- Overall aim of the paper
- Motivation
- Overview of key points
- Conclusion
- Next steps



# Overall Aim of the Paper

- a simple *introduction*
- suitable for both *academics and practitioners*
- in order to encourage *more to use* this important technique in their future work.



# Motivation

- Hammad, Mona S. A., & Harby, Galal A. H. (2013). ***Using Multilevel Modeling for Group Health Insurance Ratemaking: A Case Study from the Egyptian Market.*** Paper presented at the Perspectives on Actuarial Risks in Talks of Young Researchers (PARTY2013), Ascona, Switzerland.



[www3.unil.ch/wpmu/party2013/](http://www3.unil.ch/wpmu/party2013/)

# Overview of key points

1. What is multilevel modeling?
2. Why to use multilevel modeling?
3. When to use multilevel modeling?
4. How to use multilevel modeling?



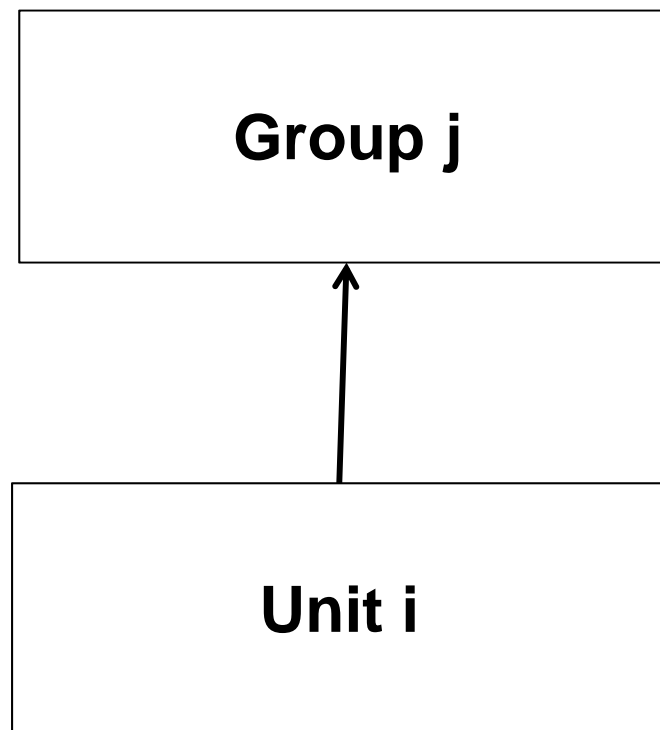
# 1- What is multilevel modeling?

- ***multilevel model*** can be defined as:
  - a model *“specified in stages, with each stage building upon another”* (Searle, Casella, & McCulloch, 2006, p. 315).
  - *“a regression (a linear or generalized linear model) in which the parameters—the regression coefficients—are given a probability model.”* (Gelman & Hill, 2007, p. 1)



## Simple Illustrative Example

- dataset of observations for units  $i$  nested groups  $j$ . (*examples?!*)



- Hierarchical/ clustered/ Nested data structures.
- → Reason for alternative name HLM / limitation
- More layers
- Panel/longitudinal data

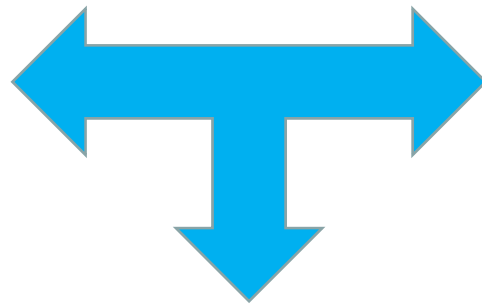


## Simple Illustrative Example (Cont.)

### Complete Pooling (one single model)

(-) ignore &  
correlation

(-) not suitable for  
all?



### No Pooling (separate models)

Not practical

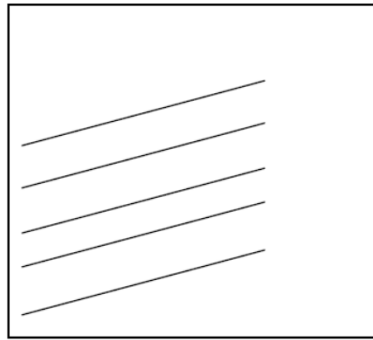
***Partial Pooling***  
***(Multilevel Model)***  
***a compromise***

single model but allows for heterogeneity  
between subjects and groups in the  
dataset. ([Gelman & Hill, 2007](#))

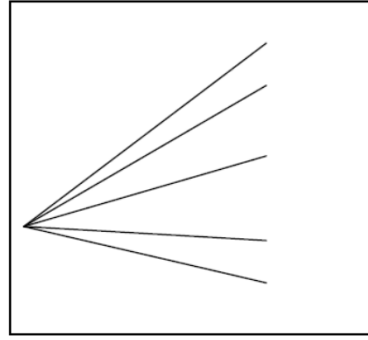




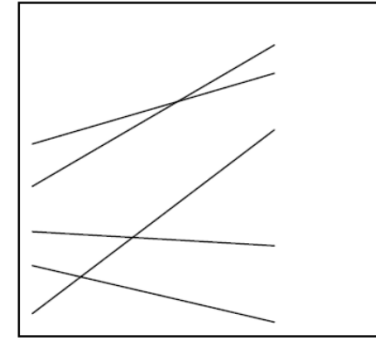
### A] Varying/Random intercept model



### B] Varying/Random slope model



### C] Varying/Random intercept and slope model



Note: each of these graphs assumes a hypothetical dataset with 6 groups (i.e. six level 2 units), with each fitted line representing a different group, in other words. Source: (Gelman & Hill, 2007, p. 238)

$$y_{ij} = \pi_{0j} + \pi_1 \alpha_{ij} + e_{ij}$$

$$\pi_{0j} = \beta_{00} + \beta_{01} x_j + r_{0j}$$

$$y_{ij} = \pi_0 + \pi_{1j} \alpha_{ij} + e_{ij}$$

$$\pi_{1j} = \beta_{10} + \beta_{11} x_j + r_{1j}$$

$$y_{ij} = \pi_{0j} + \pi_{1j} \alpha_{ij} + e_{ij}$$

$$\pi_{0j} = \beta_{00} + \beta_{01} x_j + r_{0j}$$

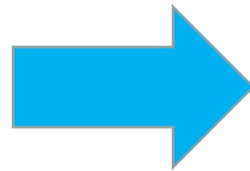
$$\pi_{1j} = \beta_{10} + \beta_{11} x_j + r_{1j}$$



Possibility of different explanatory variables at different levels

## 2- Why to use multilevel modeling?

- allowance for:
  - **heterogeneity** between different subjects in the dataset without the need to fit separate models.
  - possible **correlations over time** between observations related to the same subject and/or group.



It can handle  
complex data  
structures

### 3- When to use multilevel modeling?

- Complex data structures such as:
  - *Hierarchical data*
  - *Panel data*
  - *Cross-classified data*



# 4- How to use multilevel modeling?

- Specialized software vs. general purpose statistical packages (with mixed model procedure).

Hierarchical Linear Model (HLM)	Linear Mixed Model notation
Level 1: $y_{ij} = \pi_{0j} + e_{ij}$	$y_{ij} = \underbrace{\beta_{00}}_{Fixed} + \underbrace{r_{0j}}_{Random} + e_{ij}$
Level 2: $\pi_{0j} = \beta_{00} + r_{0j}$	



# Conclusion

- Powerful tool
- Open a wide spectrum of potential model designs
  - need for careful planning (i.e. a clear modeling strategy).



# Next Steps

- Read the paper?
- Send feedback?

*Thank You*



# References

- Gelman, A., & Hill, J. (2007). *Data Analysis Using Regression and Multilevel/Hierarchical Models*. New York: Cambridge University Press.
- Searle, S. R., Casella, G., & McCulloch, C. E. (2006). *Variance Components*. New Jersey: John Wiley & Sons, Inc.

