

S-043/Stat-151  
Analysis for Clustered and Longitudinal  
Data  
(Multilevel & Longitudinal Models)

An overview of what this course is about, and  
possibly even why you should take it.

Instructor: Prof. Luke W. Miratrix

[lmiratrix@g.harvard.edu](mailto:lmiratrix@g.harvard.edu)

Larsen 603

# S-043: multilevel and longitudinal models

- Who am I?
  - Luke Miratrix
  - Assistant professor at HGSE. Was assistant professor at Harvard Stat before this.
- Who are you?
  - HGSE PhD students
  - HGSE master's students
  - PhD/master's students from other schools (Harvard and others)
  - Undergrads?

# What is this course for?

---

This course is to provide you with the tools you need to do research using data with a multilevel or longitudinal structure (i.e., clustered or nested data)

# Multilevel data? What's that?

*In reality:* Data where you have groups of things

*Some examples:*

- Researchers sample schools from a district, then sample classes from those schools, then students from those classes (three levels)
- Researchers sample students, then measure each student at multiple different time points (two levels)

# Multilevel data? Why stress?

---

Ignoring nested structure is perilous (okay, not really, but it doesn't work well)

What is our sample size? 400?  
2? Something in between?



Universe  
of schools

A blue oval on the left contains the text 'Universe of schools'. A thin blue line extends from the right side of this oval to a larger blue rectangle on the right. The rectangle contains the text 'Sample: 1 private school, 1 public school, 200 kids from each school'.

Sample:  
1 private school, 1  
public school, 200  
kids from each  
school

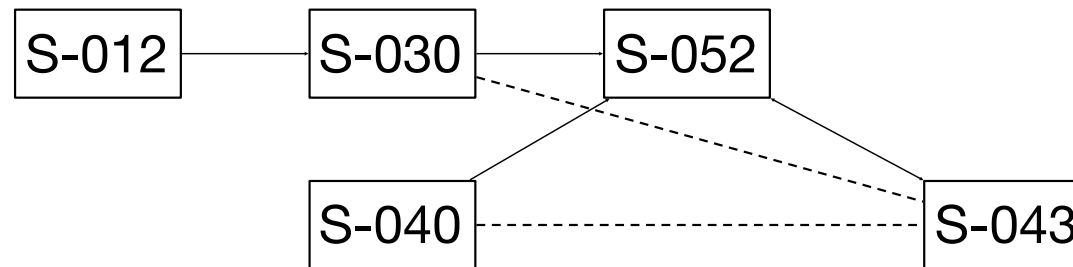
# Multilevel data? Hooray!

---

Multilevel data let us ask interesting questions:

- A. Do students from higher SES backgrounds tend to have higher levels of math achievement than students from lower SES backgrounds?
- B. Controlling for student SES, do *schools* with higher *mean* student SES tend to have higher *mean* levels of math achievement?
- C. How much did treatment impacts vary in my multi-site randomized controlled trial?
- D. Are racial achievement gaps bigger or smaller in higher SES schools than lower SES schools?

# Where does this course fit in the methods sequence?



This course...

- shows how to apply S-030/S-040 type analyses (specifically regression) to multilevel data
- Overlaps very slightly with S-052
- Uses much more mathematical formalism (greek letters) than S-052

To take this course, you should *definitely* know...

---

How to interpret estimates of linear models.

How to think critically about whether they are appropriate.

How to think at least a bit about testing and estimation.

How to use some statistical software (R ideal)

I.e., you should know:

**S-040 or Stat-139**

If these courses were hard for you, you should also have another more advanced quantitative course under your belt so you can have a stronger foundation. <sup>8</sup>



# Furthering your research

---

A major component of this course is the final project

This final project is ideally a vehicle to advance your research.

Ideally, you will come to this course with your own dataset to analyze.

(If not, you should start looking for one.)

# You might also want...

---

- ★ to be able to use R for simple tasks, including data management and fitting models
- ★ some knowledge of generalized linear regression (e.g., logistic regression)
- ★ some mathematical comfort (e.g., summation notation and matrices)
- ★ an intuitive comfort with linear regression (especially interpretation)

These are not “must haves,” but having one or more of these will make the course easier.

# To take this course

---

You don't need to...

- ★ be an expert using R

- ★ be experienced with mathematical statistics  
(especially proofs)

- ★ have any experience analyzing multi-level data

# General Content of Course

---

1. Grapple with clustered data (and see some common tools to handle it).
2. Introduce a formal model (the multilevel model) to handle and describe such data.
3. Examine how to assess uncertainty (and do inference) using such models.
4. Examine how to learn about variation using such models.
5. Extend the above to longitudinal data.
6. Learn how to assess uncertainty using simulation and bootstrap.
7. Generalize and extend to nonlinear data (e.g., counts).

**Core of Course**

Sure, but what will we *actually* be doing?

---

Two lectures per week (T/Th, 2-3:30).

Some readings

Weekly section

About three short problem sets (individual)

Two longer analyses (pairs)

Final project (ideally pairs and *very* student-driven)

# Primary Materials of the Course

---

- ★ Canvas - the central spot for announcements and resources. This is where you will submit data.
- ★ Raudenbush and Bryk Book - a concrete and clear textbook that focuses on analysis of specific forms of data
- ★ R Studio - a real tool for statisticians.

# Why take this course?

---

So you can do these things:

- ★ Think critically about data-based claims and quantitative arguments.
- ★ Explore and analyze real data with hierarchical or longitudinal aspects.
- ★ Present such data numerically and visually.
- ★ Know which statistical methods to use when, and how to use them.
- ★ Use R, a statistical software package.
- ★ Learn new statistical analysis techniques on your own.



# Finding Course Bureaucracy Details

---



Please see posted slides on the website.

**<https://canvas.harvard.edu/courses/67934>**

Also,  
**YOU MUST  
READ THE SYLLABUS**



# Further Details

# How homework is made

---

**Problem sets:** Individually completed work

- ★ Collaboration encouraged.
- ★ Individual write-ups.

**Projects:** Mandatory groups of 2

- ★ Single write-up submitted for the group.

**Final Project:** Groups of 1-3 students, with an ideal of 2.

- ★ Individual projects only if due to using dissertation/research work as one's final project (although we encourage collaboration nonetheless).

Further details, and further information on **collaboration vs. plagiarism**, in syllabus<sup>8</sup>

# Class Meetings, Attendance, Behavior

---

- ★ Lectures will be interactive. Questions are encouraged.
- ★ Lectures will be adapted to the needs of the students.
- ★ We will sometimes use polling systems to foster immediate feedback and discussion.
- ★ You are expected to come to every class meeting.
- ★ You are expected to behave professionally in lecture (no web surfing, phones are silent, no IM, etc) because it negatively impacts others.

# The work you will do

---

## ★ 2 Problem Sets

- Smaller questions targeting specific issues of understanding.

## ★ 2 Projects

- Analyze real-world data sets and write coherently about what you find.

## ★ Final Project

- Analyze your own data and discover something new.

## ★ A few mini-assignments, such as a warm-up learning R assignment.

# A Note on Papers and Examples

The original sources for the studies used in the course will be provided and linked when possible - if interested in the details of the study, please check out the original article!

# An incomplete take on Longitudinal Data

---

One approach to longitudinal data is to view a sequence of observations from people across time as nested inside people.

Other approaches not covered:

- ★ Survival Analysis
- ★ Hazard Models
- ★ Cox Regression Models

See Singer and Willet's  
“Applied Longitudinal Data Analysis”<sup>22</sup>

# Where is the Math?

---

Goal: teach good intuition as to how these models work, how they are fitted, and what to worry about.

Goal: teach literacy so students can read papers in the field, communicate their models and results, and engage with the research.

Math and mathematical notation help these goals, therefore:

- ★ Mathematical notation will be used extensively.
- ★ Informal mathematical argument will be used in class sometimes.
- ★ Optional readings will be given out for those who want to dive deeper.

diamondPricing.R\* x formatPlot.R x diamonds x

Source on Save

Source

```

1 library(ggplot2)
2 source("plots/formatPlot.R")
3
4 view(diamonds)
5 summary(diamonds)
6
7 summary(diamonds$price)
8 aveSize <- round(mean(diamonds$carat), 4)
9 clarity <- levels(diamonds$clarity)
10
11 p <- qplot(carat, price,
12           data=diamonds, color=clarity,
13           xlab="Carat", ylab="Price",
14           main="Diamond Pricing")
15

```

15:1 (Top Level) ↕

R Script ↕

Console ~/ ↕

```

      x              y              z
Min.   : 0.000   Min.   : 0.000   Min.   : 0.000
1st Qu.: 4.710   1st Qu.: 4.720   1st Qu.: 2.910
Median : 5.700   Median : 5.710   Median : 3.530
Mean   : 5.731   Mean   : 5.735   Mean   : 3.539
3rd Qu.: 6.540   3rd Qu.: 6.540   3rd Qu.: 4.040
Max.   :10.740   Max.   :58.900   Max.   :31.800
> summary(diamonds$price)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
   326    950   2401   3933   5324   18820
> aveSize <- round(mean(diamonds$carat), 4)
> clarity <- levels(diamonds$clarity)
> p <- qplot(carat, price,
+           data=diamonds, color=clarity,
+           xlab="Carat", ylab="Price",
+           main="Diamond Pricing")

```

Workspace

History

Load

Save

Import Dataset

Clear All

Data

diamonds 53940 obs. of 10 variables

Values

aveSize 0.7979

clarity character [8]

p ggplot [8]

Functions

format.plot(plot, size)

RStudio

Files

Plots

Packages

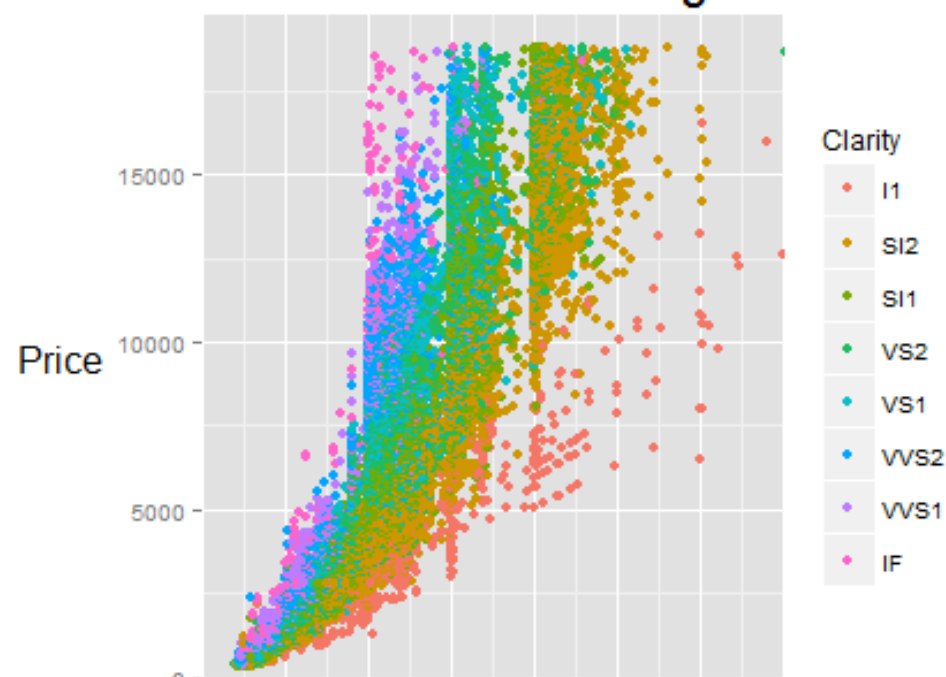
Help

Zoom

Export

Clear All

## Diamond Pricing





# Getting Started with R and RStudio on Your Computer

---

- ★ R is free and easy to install on your computer.

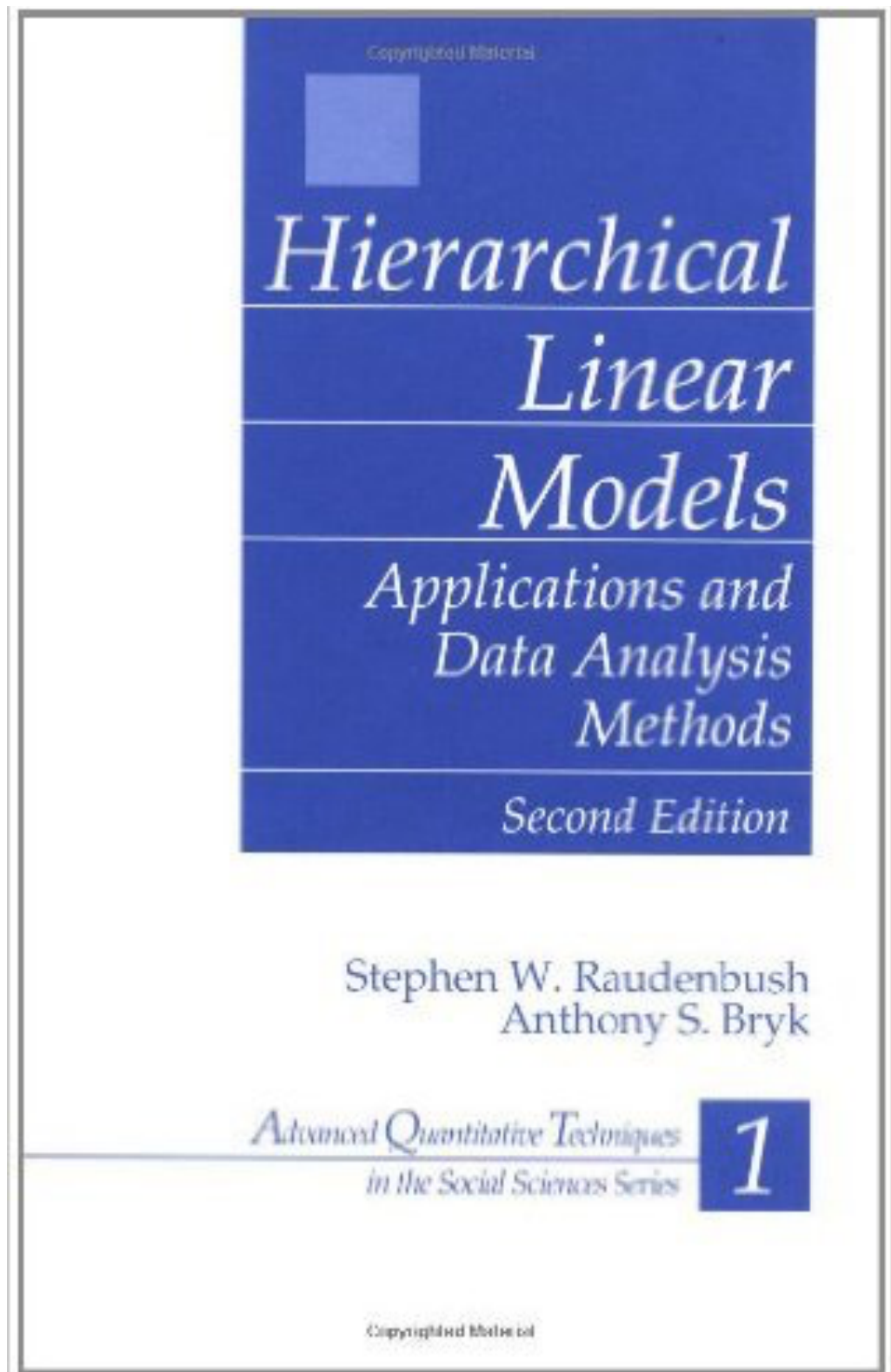
<http://lib.stat.cmu.edu/R/CRAN/>

- ★ Once you install R, download and install RStudio.

<http://www.rstudio.com>

*We will help you with this during office hours or section if you get stuck.*

# Primary Textbook

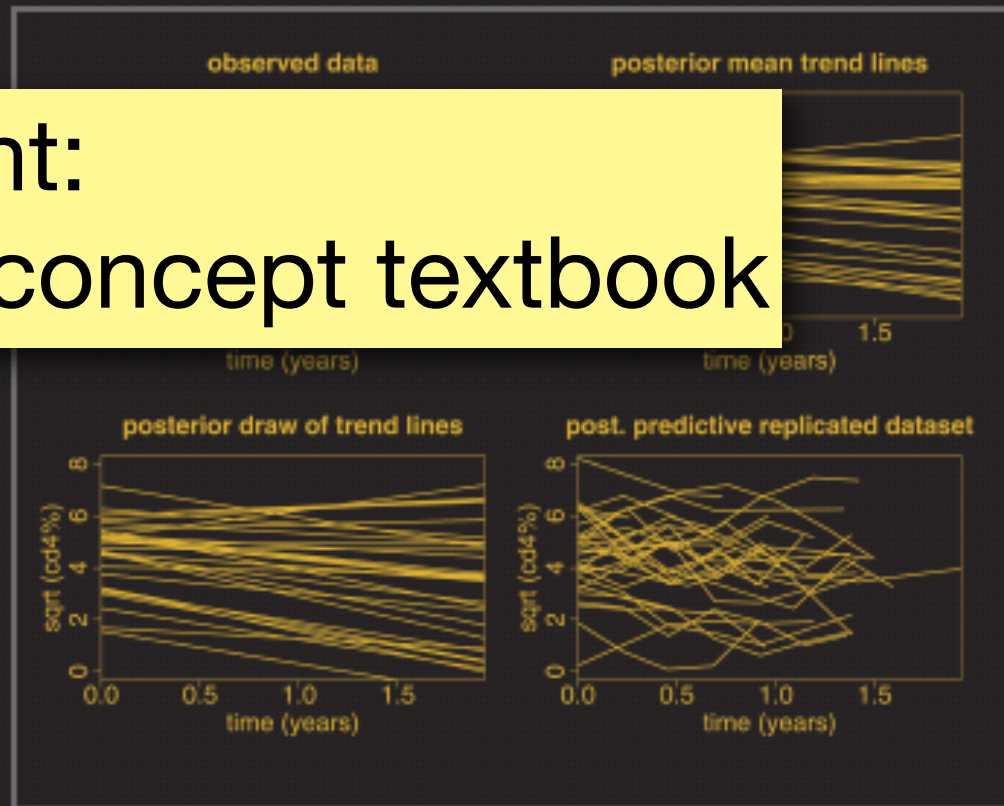


# Math, explanation, no code

---

- ★ Some mathematics showing where MLMs come from.
- ★ Excellent worked examples, connecting the fitted models to conceptual questions.
- ★ Case studies that bring the models to life.
- ★ Discussion of many classic concerns that arise with modeling.

# Supplement: High level concept textbook



## Data Analysis Using Regression and Multilevel/Hierarchical Models

ANDREW GELMAN  
JENNIFER HILL

# It is actually *two* textbooks

---

## Part 1: Single Level Regression

- ★ Part A - Basic fitting, logistic, generalized linear models
  - ★ Part B - Inference, simulation, causal inference
- 

## Part 2: Multilevel Regression

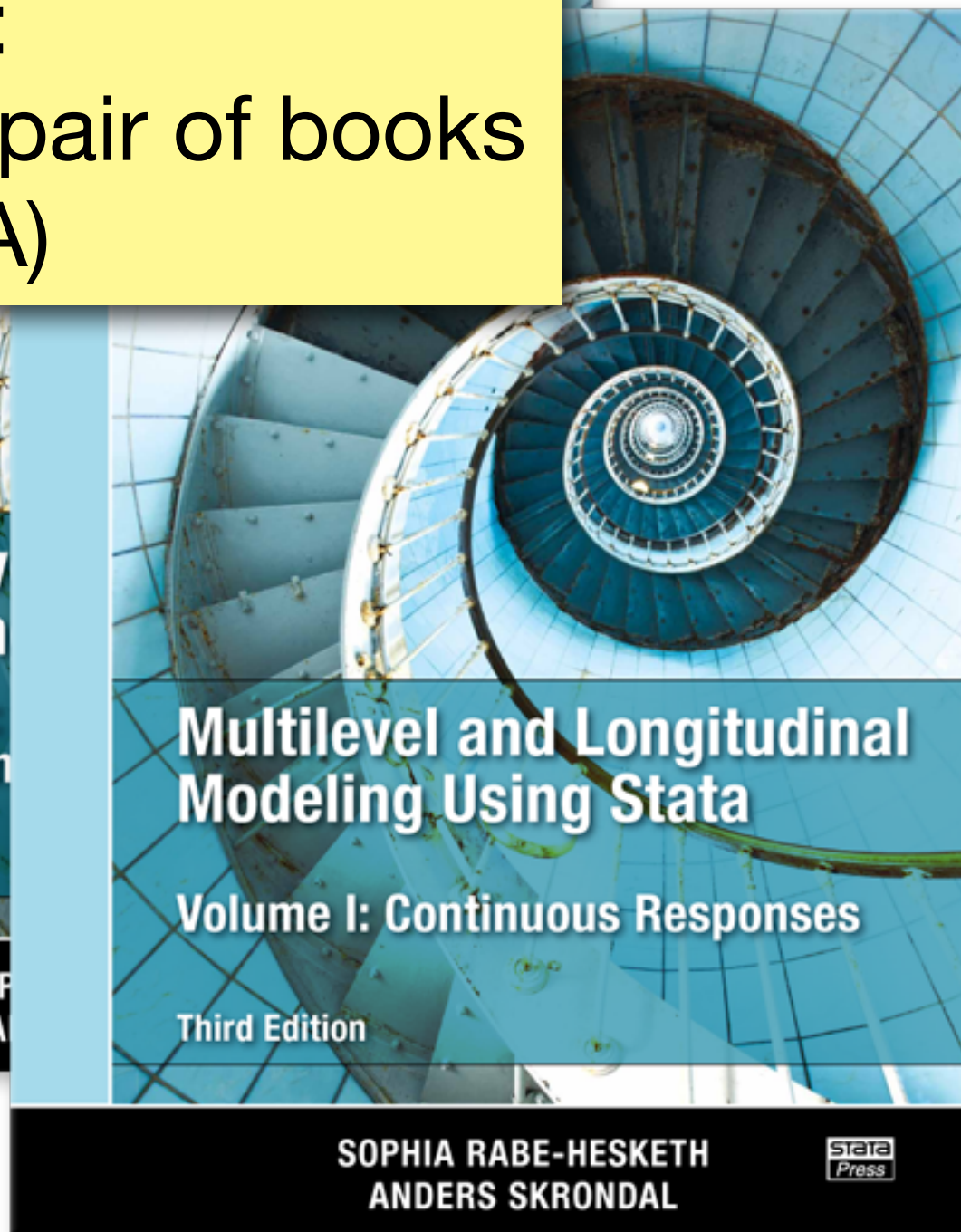
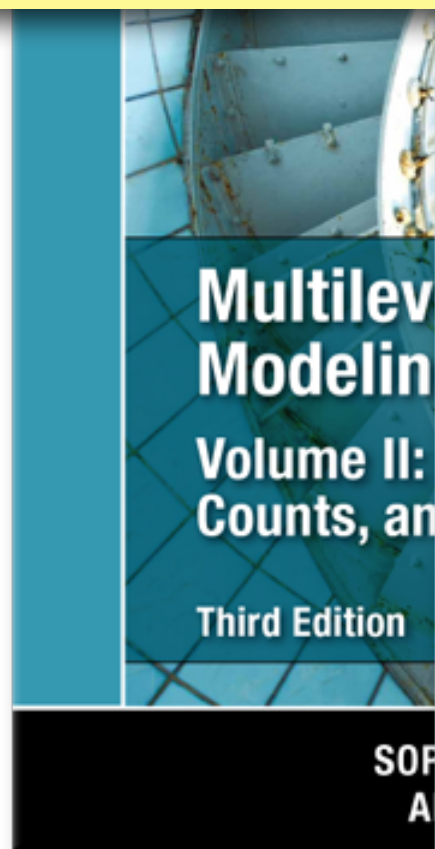
- ★ Part A - What these things are and how to think about them
- ★ Part B - Fitting them and getting Fancy

## Part 3: Getting fancy

- ★ Power and sample size
- ★ Further conceptual issues of multilevel models
- ★ More on inference

**The first part has some ideas of simulation for uncertainty that are important**

**Supplement:**  
**A very clear pair of books**  
**(but in STATA)**





# It is actually *two (three?)* textbooks

---

## Volume 1: Continuous outcomes

- ★ Part I: An intro on linear regression
- ★ Part II: Multilevel models
- ★ Part III: Longitudinal and panel data

**Part I should be review, basically.**

## Volume 2: Other outcomes

- ★ Each section starts with the “single level” (not multilevel) model and extends to multilevel models