

An Evaluative Comparison of Random Coefficient Growth Models for Individual Development

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Association for Psychological Science

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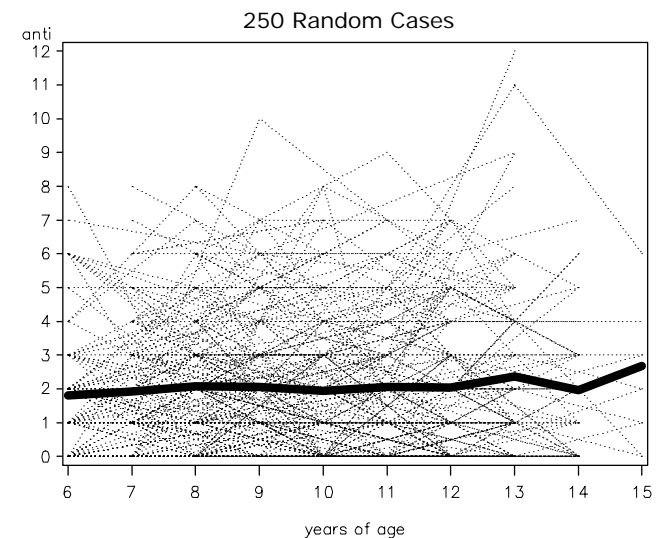
Overview of Talk

- Motivating example
- Definition of Random Coefficient Growth Model (RCGM)
- Three basic types of RCGMs
- Controversies in model selection and interpretation

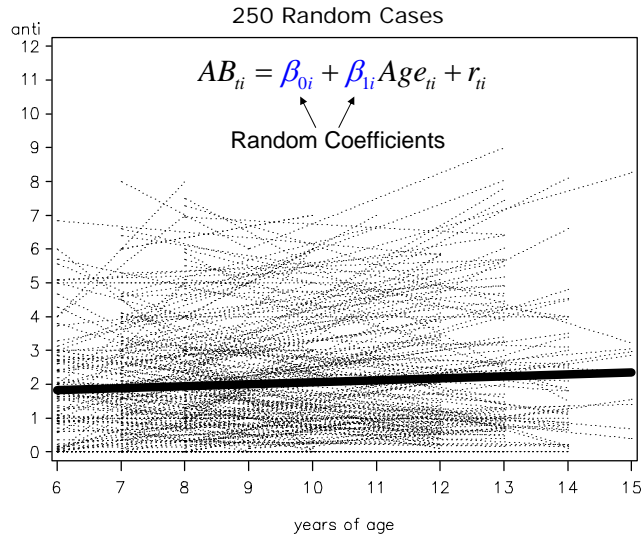
Motivating Example: Antisocial Behavior (AB)

- Concerns:
 - What do trajectories of change in AB look like from age 6 to age 15?
 - Are there sex-differences in these developmental trends?
 - Do supportive home environments protect against increases in AB?
 - Can AB be predicted by poor academic performance?
- Sample:
 - 894 children assessed biennially from 1986 to 1992 as part of the NLSY-Child Sample.
 - Between 6 and 8 years old in 1986.
- Measures:
 - Antisocial Behavior: Sum of 6 items from BPI
 - Early Home Environment: HOME-SF cognitive and emotional support scores from first assessment
 - Academic Performance: PIAT Math scores

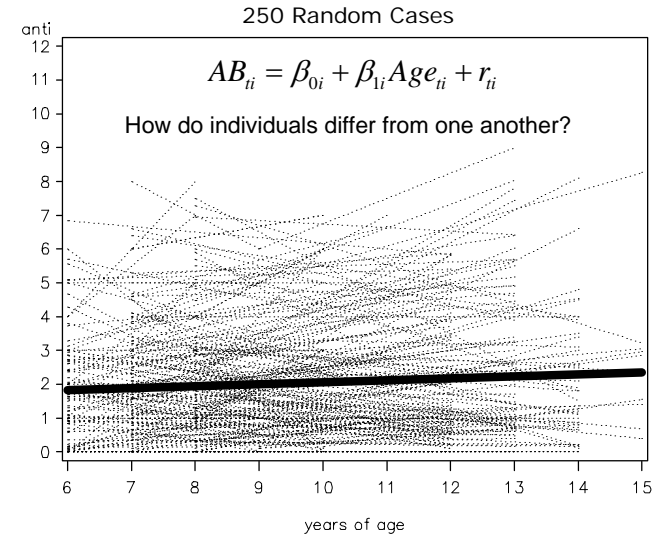
Raw Data



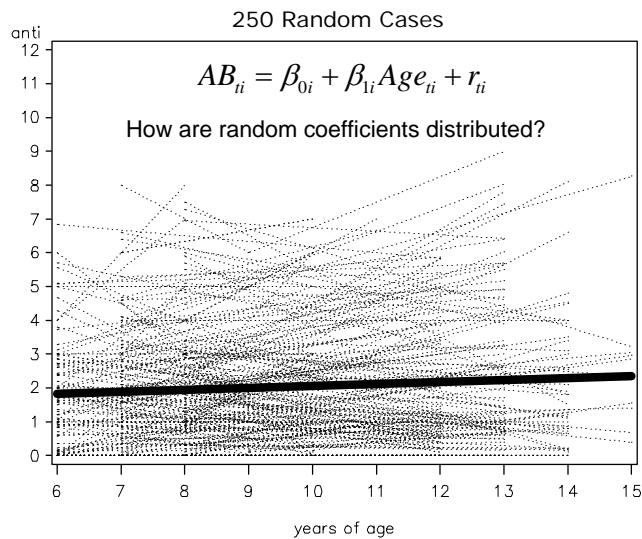
The Idea Behind RCGMs



The Idea Behind RCGMs



The Idea Behind RCGMs

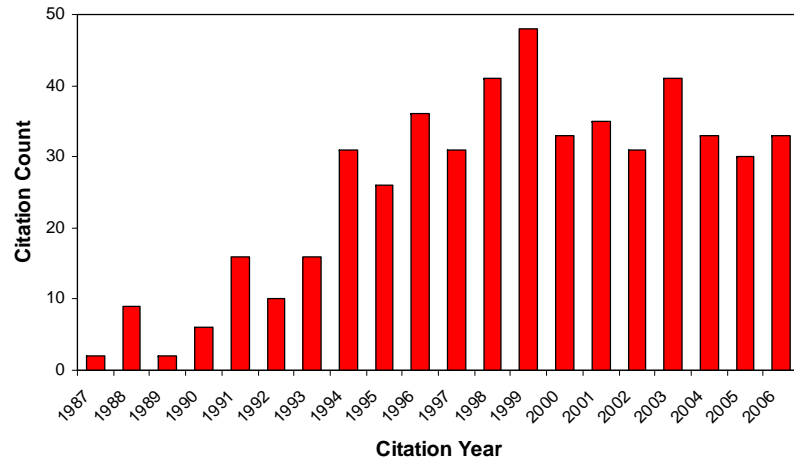


Three Types of RCGMs

1. RCN: Models that assume the random coefficients are normally distributed (conditional on predictors)
 - Latent curve/trajectory/growth models, HLM, multilevel growth models, mixed-effects models

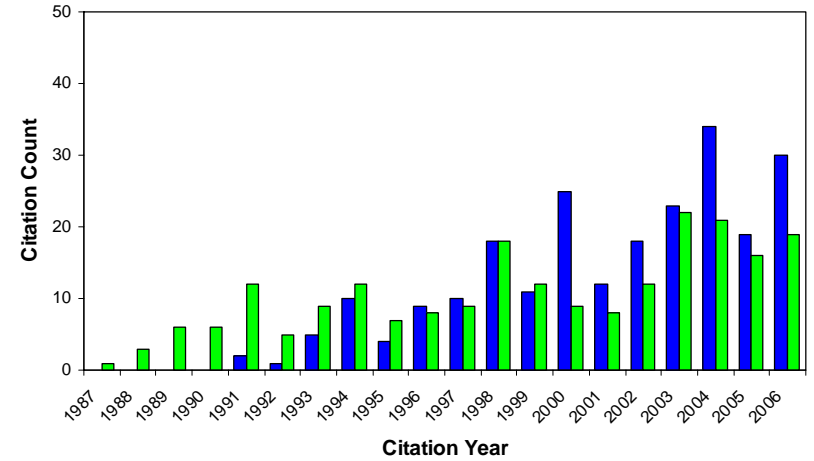
Trends in Use of RCN Models (HLMs)

Bryk & Raudenbush (1987)



Trends in Use of RCN Models (LCMs)

McArdle & Epstein (1987)
Meredith & Tisak (1990)



RCN

- Within-Person Model:

$$AB_{it} = \beta_{0i} + \beta_{1i}(Age_{it} - 6) + r_{it} \quad r_{it} \sim N(0, \sigma^2)$$

- Between-Person Model:

$$\begin{aligned} \beta_{0i} &= \gamma_{00} + u_{0i} \\ \beta_{1i} &= \gamma_{10} + u_{1i} \end{aligned} \quad \begin{pmatrix} u_{0i} \\ u_{1i} \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \tau_{00} & \\ & \tau_{11} \end{pmatrix} \right]$$

- Assumes continuous individual differences in change over time.

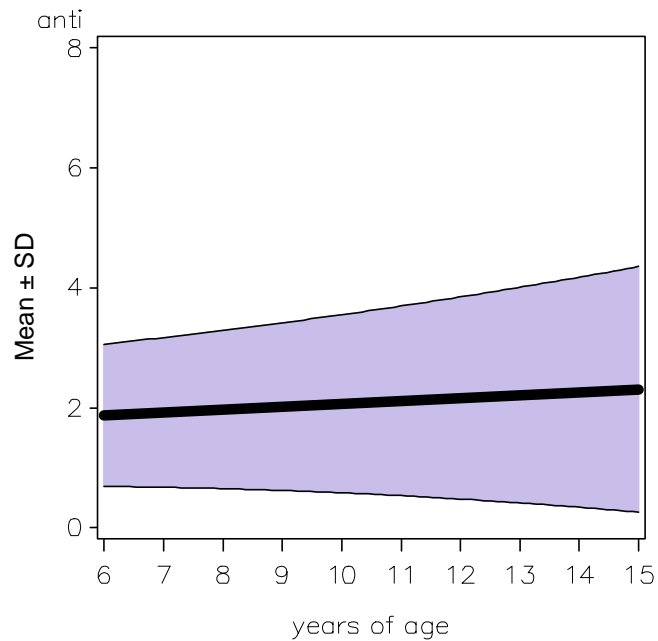
RCN Fit to Antisocial Data

| Parameter | Estimate (SE) |
|---|---------------|
| <i>Fixed effects</i> | |
| Intercept (γ_{00}) | 1.87 (0.07)** |
| Age (γ_{10}) | 0.05 (0.01)** |
| <i>Variance / Covariance Parameters</i> | |
| Intercept (τ_{00}) | 1.43 (0.25)** |
| Age (τ_{11}) | 0.02 (0.01)* |
| Covariance (τ_{10}) | 0.05 (0.04) |
| Residual (σ^2) | 2.09 (0.14)** |

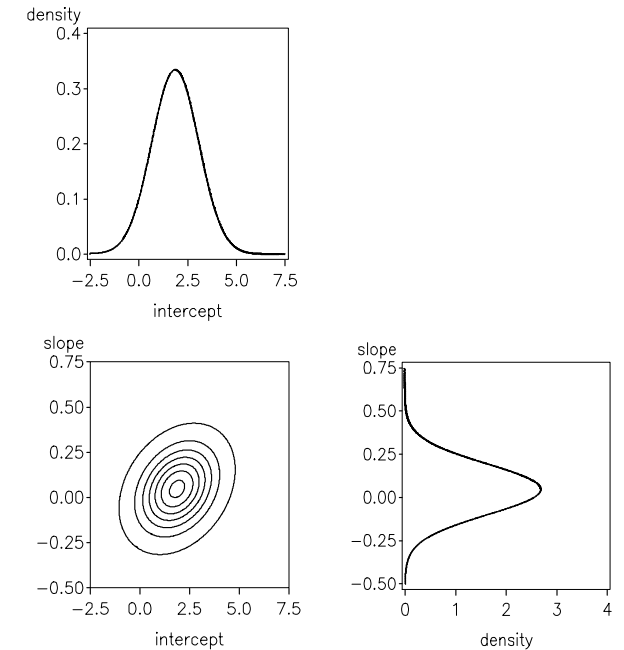
Note. Robust standard errors reported.

* p < .05 ; ** p < .01

Graphical Depiction of RCN



Graphical Depiction of RCN



Adding Predictors to RCN

| Parameter | Estimate (SE) |
|---|-----------------------|
| <i>Prediction of Intercepts (β_{0i})</i> | |
| Intercept | 1.18 (0.14)** |
| Sex | 0.85 (0.13)** |
| Home | -0.22 (0.04)** |
| <i>Prediction of Slopes (β_{1i})</i> | |
| Intercept | 0.12 (0.03)** |
| Sex | 0.01 (0.03) |
| Home | -0.01 (0.01) |
| <i>Time Varying Covariates</i> | |
| Math | -0.17 (0.04)** |
| <i>Variance / Covariance Parameters</i> | |
| Intercept | 1.05 (0.16)** |
| Age | 0.02 (0.01)* |
| Covariance | 0.06 (0.03)* |
| Residual | 2.11 (0.06)** |

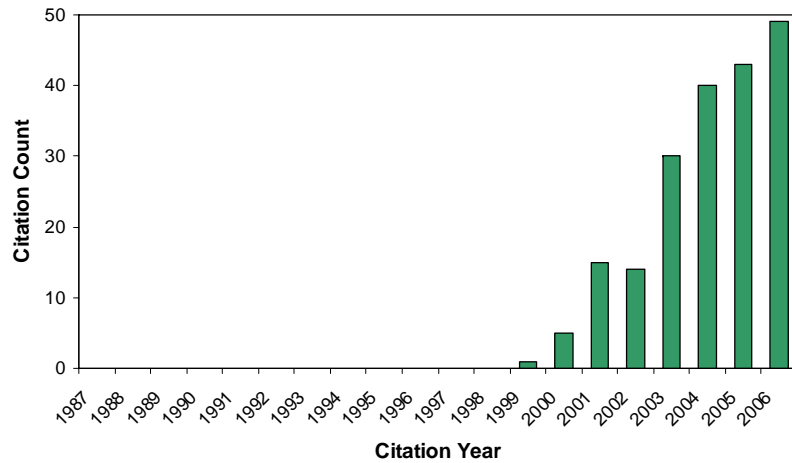
* p<.05; ** p<.01

Three Types of RCGMs

1. RCN: Models that assume the random coefficients are normally distributed (conditional on predictors)
 - Latent curve/trajectory/growth models, HLM, multilevel growth models, mixed-effects models
2. RCD: Models that assume the random coefficients are discretely distributed across K groups
 - Semi-parametric group-based trajectory method, non-parametric random coefficient model, latent class growth analysis, latent class regression

Trends in Use of RCD Models (SPGBM)

Nagin (1999)



RCD

- Within-Person Model:

$$AB_{it} = \beta_{0i} + \beta_{1i} (Age_{it} - 6) + r_{it} \quad r_{it} \sim N(0, \sigma^2)$$

- Between-Person Model:

$$\begin{aligned} \beta_{0i} &= \gamma_{00(k)} \text{ if } C_i = k & C_i &= 1, 2, \dots, K \\ \beta_{1i} &= \gamma_{10(k)} \text{ if } C_i = k & P(C_i = k) &= \pi_{(k)} \end{aligned}$$

- Assumes individual differences in change over time are discretely distributed – K types of trajectories.

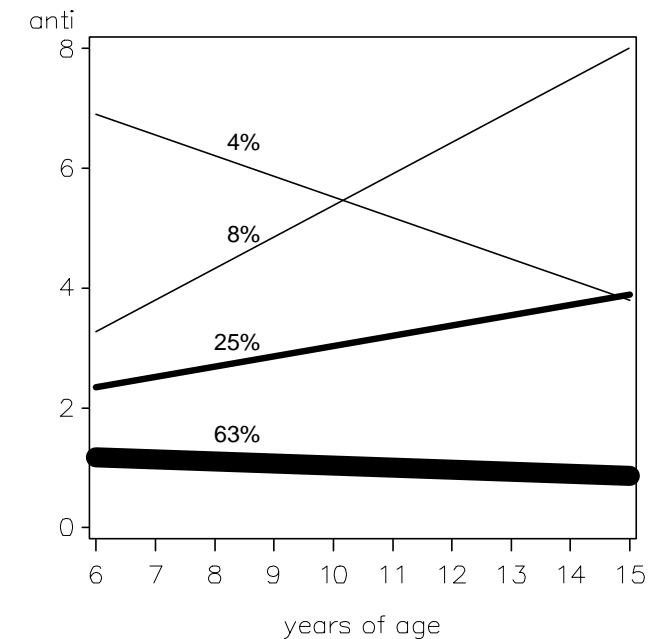
RCD Fit to Antisocial Data

| Parameter | Latent Class | | | |
|---|-------------------|------------------|------------------|---------------|
| | “High Increasing” | “Low Increasing” | “High Declining” | “Abstaining” |
| <i>Class Size</i> | | | | |
| Class probability ($\pi_{(k)}$) | .08 | .25 | .04 | .63 |
| Sample N | 70 | 222 | 39 | 562 |
| <i>Fixed effects</i> | | | | |
| Intercept ($\gamma_{00(k)}$) | 3.27 (0.91)** | 2.35 (0.22)** | 6.90 (1.95)** | 1.17 (0.07)** |
| Age ($\gamma_{10(k)}$) | 0.53 (0.20)** | 0.17 (0.04)** | -0.34 (0.37) | -0.03 (0.01)* |
| <i>Variance / Covariance Parameters</i> | | | | |
| Residual (σ^2) | 2.02 (0.10)** | 2.02 (0.10)** | 2.02 (0.10)** | 2.02 (0.10)** |

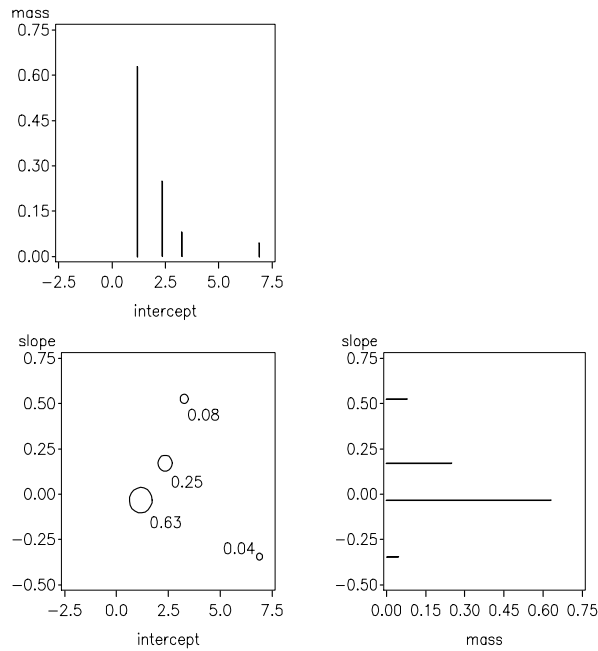
Note. Robust standard errors reported.

* $p < .05$; ** $p < .01$

Graphical Depiction of RCD



Graphical Depiction of RCD



Adding Predictors to RCD

| Parameter | Latent Class | | | |
|--|-------------------------|-------------------------|--------------------------|-----------------------|
| | "High Increasing" | "Low Increasing" | "High Declining" | "Abstaining" |
| <i>Class Size^a</i> | | | | |
| Class probability | .08 | .28 | .05 | .59 |
| Sample N | 71 | 236 | 39 | 498 |
| <i>Between-Class Effects^b</i> | | | | |
| Sex | 4.25 (2.15,8.43) | 3.70 (2.30,5.94) | 9.70 (3.15,29.88) | N/A |
| Home | 0.59 (0.47,0.72) | 0.65 (0.56,0.75) | 0.68 (0.48,0.96) | N/A |
| <i>Within-Class Effects</i> | | | | |
| Intercept | 3.08 (0.22)** | 2.23 (0.14)** | 6.77 (0.33)** | -0.91 (0.11)** |
| Age | 0.56 (0.05)** | 0.17 (0.03)** | -0.32 (0.07)** | 0.03 (0.03) |
| Math | -0.11 (0.04)** | -0.11 (0.04)** | -0.11 (0.04)** | -0.11 (0.04)** |
| <i>Variance / Covariance Parameters</i> | | | | |
| Residual | 1.96 (0.04)** | 1.96 (0.04)** | 1.96 (0.04)** | 1.96 (0.04)** |

* p < .05 ; ** p < .01

^a Based on estimated posterior probabilities

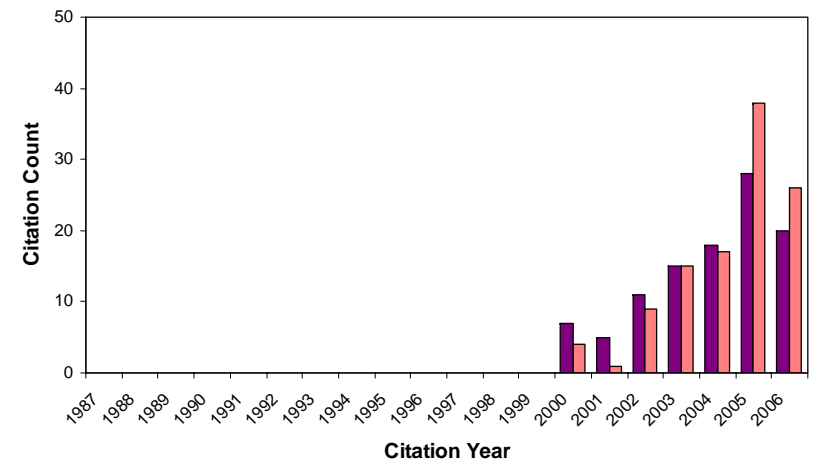
^b Odds Ratio (95% CI); Reference class is Abstaining

Three Types of RCGMs

1. RCN: Models that assume the random coefficients are (conditionally) normally distributed
 - Latent curve/trajectory/growth models, HLM, multilevel growth models, mixed-effects models
2. RCD: Models that assume the random coefficients are discretely distributed across K latent groups
 - Semi-parametric group-based trajectory method, non-parametric random coefficient model, latent class growth analysis, latent class regression
3. RCNM: Models that assume the random coefficients are from a (conditional) normal mixture distribution with K latent groups
 - General growth mixture models

Trends in Use of RCNM Models (GMMs)

Muthen & Shedden (1999)
Muthen & Muthen (2000)



RCNM

- Within-Person Model:

$$AB_{it} = \beta_{0i} + \beta_{1i} (Age_{it} - 6) + r_{it} \quad r_{it} \sim N(0, \sigma_{(k)}^2)$$

- Between-Person Model:

$$\begin{aligned} \beta_{0i} &= \gamma_{00(k)} + u_{0i(k)} \text{ if } C_i = k \\ \beta_{1i} &= \gamma_{10(k)} + u_{1i(k)} \text{ if } C_i = k \end{aligned} \quad \begin{pmatrix} u_{0i(k)} \\ u_{1i(k)} \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \tau_{00(k)} & & \\ & \tau_{10(k)} & \\ & & \tau_{11(k)} \end{pmatrix} \right]$$

$$C_i = 1, 2, \dots, K$$

$$P(C_i = k) = \pi_{(k)}$$

- Assumes K discrete trajectory groups, within which individual differences in change over time are continuously distributed.

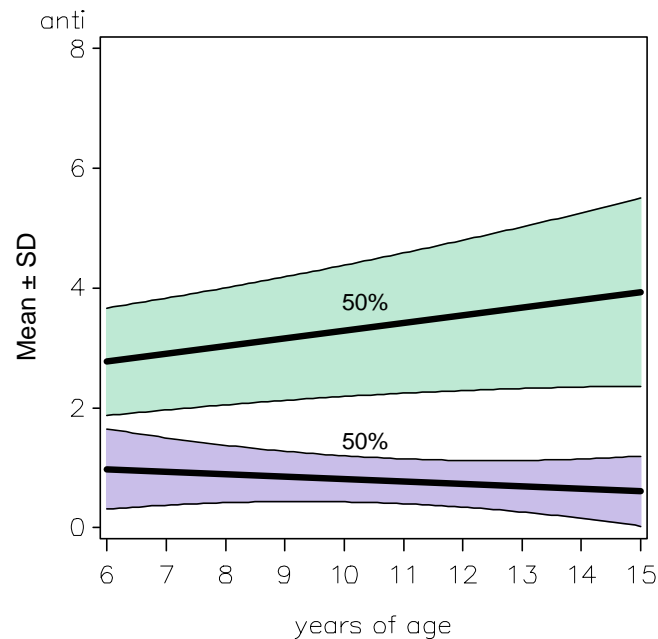
RCNM Fit to Antisocial Data

| Parameter | Latent Class | |
|---|---------------|----------------|
| | “Increasing” | “Decreasing” |
| <i>Class Size</i> | | |
| Class probability ($\pi_{(k)}$) | .50 | .50 |
| Sample N | 445 | 449 |
| <i>Fixed effects</i> | | |
| Intercept ($\gamma_{00(k)}$) | 2.78 (0.13)** | 0.98 (0.09)** |
| Age ($\gamma_{10(k)}$) | 0.13 (0.02)** | -0.04 (0.02)** |
| <i>Variance / Covariance Parameters</i> | | |
| Intercept ($\tau_{00(k)}$) | 0.83 (0.42)* | 0.43 (0.10)** |
| Age ($\tau_{11(k)}$) | 0.02 (0.02) | 0.01 (0.004)* |
| Covariance ($\tau_{10(k)}$) | 0.01 (0.07) | -0.06 (0.02)** |
| Residual ($\sigma_{(k)}^2$) | 3.55 (0.32)** | 0.62 (0.07)** |

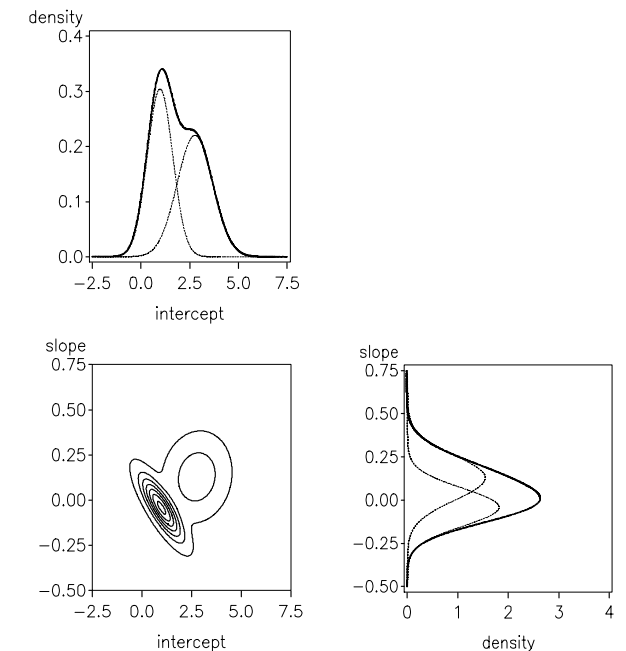
Note. Robust standard errors reported.

* $p < .05$; ** $p < .01$

Graphical Depiction of RCNM



Graphical Depiction of RCNM



Adding Predictors to RCNM

| Parameter | Latent Class | |
|--|-------------------------|----------------------|
| | "Increasing" | "Decreasing" |
| <i>Class Size^a</i> | | |
| Class probability | .51 | .49 |
| Sample <i>N</i> | 426 | 417 |
| <i>Between-Class Effects^b</i> | | |
| Sex | 2.82 (1.93,4.11) | N/A |
| Home | 0.66 (0.58,0.75) | N/A |
| <i>Within-Class Prediction of Intercepts (β_{0i})</i> | | |
| Intercept | 1.89 (0.24)** | 0.76 (0.10) |
| Sex | 0.62 (0.24)** | 0.31 (0.12)** |
| Home | -0.06 (0.04) | -0.06 (0.04) |
| <i>Within-Class Prediction of Slopes (β_{1i})</i> | | |
| Intercept | 0.26 (0.05)** | 0.01 (0.02) |
| Sex | -0.01 (0.05) | -0.05 (0.03)* |
| Home | 0.01 (0.01) | 0.01 (0.01) |
| <i>Time Varying Covariates</i> | | |
| Math | -0.30 (0.07)** | -0.06 (0.03) |
| <i>Variance / Covariance Parameters</i> | | |
| Intercept | 0.45 (0.34) | 0.35 (0.11)** |
| Age | 0.01 (0.02) | 0.01 (0.004)** |
| Covariance | 0.05 (0.06) | -0.05 (0.02)** |
| Residual | 3.60 (0.17)** | 0.61 (0.05)** |

* $p < .05$; ** $p < .01$

^aBased on estimated posterior probabilities

^bOdds Ratio (95% CI); Reference class is Decreasing

Comparison of Three Models

- Overall trends are similar
 - Most children had low levels of AB at age 6.
 - Some children showed increases in AB from age 6 to 15, others showed stable or decreasing AB.
 - Those with highest initial levels of AB tended to show greatest increases over time.
 - Male children, and children from less supportive home environments displayed higher levels of AB.
 - Children scoring poorly on the PIAT-M displayed higher levels of AB.
- Despite general consistency, much controversy over appropriate model

Controversy in the Selection of RCGMs

- 5 recent papers on RCD and RCNM models have been followed by commentaries and rejoinders
 - Bauer & Curran (2003, *Psychological Methods*), commentaries by Cudeck & Henly, Muthen, & Rindskopf.
 - Eggleston, Laub & Sampson (2004, *J. Quant. Criminology*), commentary by Nagin.
 - Nagin & Tremblay (2005, *Annals AAPSS*), commentaries by Maughan and Raudenbush.
 - Nagin & Tremblay (2005, *Criminology*), commentary by Sampson & Laub.
 - Connell & Frye (2006, *Infant & Child Dev*), commentaries by Hoeksma & Kelderman, Muthen, and Stanger.
- Reflects disagreements, misunderstandings about relative merits of different RCGMs.

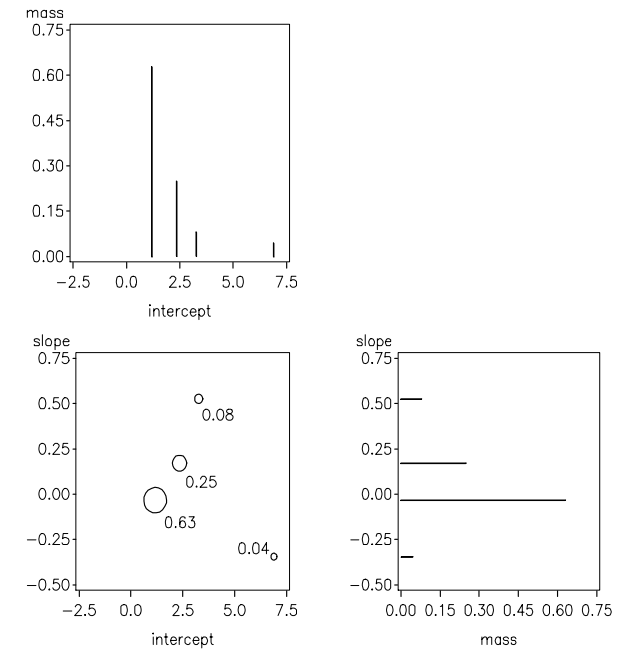
Key Issue: The Interpretation of Groups

- Often, the latent classes estimated from RCD or RCNM models are interpreted as true taxa (i.e., "real" groups).
- This can be problematic
 - Are there 4 groups (RCD), 2 groups (RCNM), or no groups (RCN)?
 - Should groups be strictly homogeneous (RCD) or do we permit within-group variability (RCNM)?
 - Spurious groups can compensate for errors in model specification, e.g., lack of normality of residuals (Bauer & Curran, 2003, 2004).
 - Number/nature of groups can change with minor alterations in model specification, covariates, measurement, or design (Eggleston et al, 2004; Jackson & Sher, 2005, 2006).
 - RCN does permit groups as a function of observed covariates (when strong etiological theory for taxa).

Key Issue: The Interpretation of Groups

- Nagin (2005) advocates use of groups as an approximating device even when groups are artificial.

Graphical Depiction of RCD



Key Issue: The Interpretation of Groups

- Nagin (2005) advocates use of groups as an approximating device even when groups are artificial.
- Drawbacks to Groups:
 - Artificial groups prone to reification.
 - Use of RCD and RCNM models may reduce power.
- Alternatives to Groups:
 - RCN models relatively robust to violation of normality assumption for random effects.
 - Other semiparametric RCGMs (Chen, Zhang & Davidian, 2002; Zhang & Davidian, 2001)
- Perhaps better to focus less on groups and more on overall trends.

Conclusions

- RCGMs offer many conceptual and statistical advantages for modeling individual change
- Many different RCGMs (HLM, LCM, GMM, LCGA, SPGBA, etc) can be organized into three categories
 - RCN assumes continuously distributed individual differences.
 - RCD assumes discretely distributed individual differences.
 - RCNM assumes continuously distributed individual differences within a small number of discrete groups.
- Key issue in choosing between models is utility of groups
 - Groups correspond nicely to taxonomic theory
 - Groups often improve model flexibility and fit.
 - Interpretation of latent groups can be risky.