

Osmotic-release methylphenidate randomized controlled trial for adolescents with attention-deficit/hyperactivity disorders and substance use disorders:

A missing data sensitivity analysis

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Network* 

PROGRAM OF EXCELLENCE
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RESEARCH

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CONFLICT OF INTEREST AND FUNDING

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BACKGROUND AND AIM

- Treatment efficacy can vary as a function of how missing data are handled in substance use disorder randomized clinical trials (RCTs)^{1,2}
- Bias can result from inappropriate handling⁷⁻¹¹
- Substance use disorder RCTs often use sub-optimal methods^{1,2}

BACKGROUND AND AIM

- Missing at Random (MAR):
 - Missing values depend on observed values that are included in the analysis
- Missing Not at Random (MNAR):
 - Missing values depend on unobserved values

BACKGROUND AND AIM

- *We compared 3 modeling strategies for handling of missing values (i.e., missing at random [MAR] model, 2 different missing not at random [MNAR] models; Diggle-Kenward and Wu-Carroll)*

METHOD: A PARALLEL GROWTH EXAMPLE

- Data from a National Drug Abuse Treatment Clinical Trials Network study (N=303)⁶
- Evaluated the safety/efficacy of a 16-week RCT of OROS vs. placebo in adolescents (aged 13-18) with ADHD, also receiving cognitive behavioral therapy for their substance use disorder
- Two primary outcomes were clinician-reported ADHD symptoms and self-reported past 28 days of substance use (SU)

METHOD: A PARALLEL GROWTH EXAMPLE

- Original RCT thoroughly evaluated missing data
 - MAR is most likely, but this dataset provides a unique example for demonstration purposes
- Out sensitivity analysis is intended to build on original analyses and demonstrate models for consideration when examining treatment for co-morbid populations

METHOD: A PARALLEL GROWTH EXAMPLE

- Used latent growth modeling to compare methods of handling missing data:
 - 1) Latent growth model using maximum likelihood for missing data (i.e. MAR model)^{3,7}
 - 2) Diggle and Kenward (1994) MNAR selection model, dropout is a function of previous and concurrent self-reported SU⁸
 - 3) Wu and Carroll (1988) MNAR selection model where dropout is a function of the random effects (i.e., intercept and linear slope growth factors)⁹

Figure 1. Parallel, piecewise latent growth curve model of substance use (SU) and attention-deficit/hyperactivity disorder (ADHD).

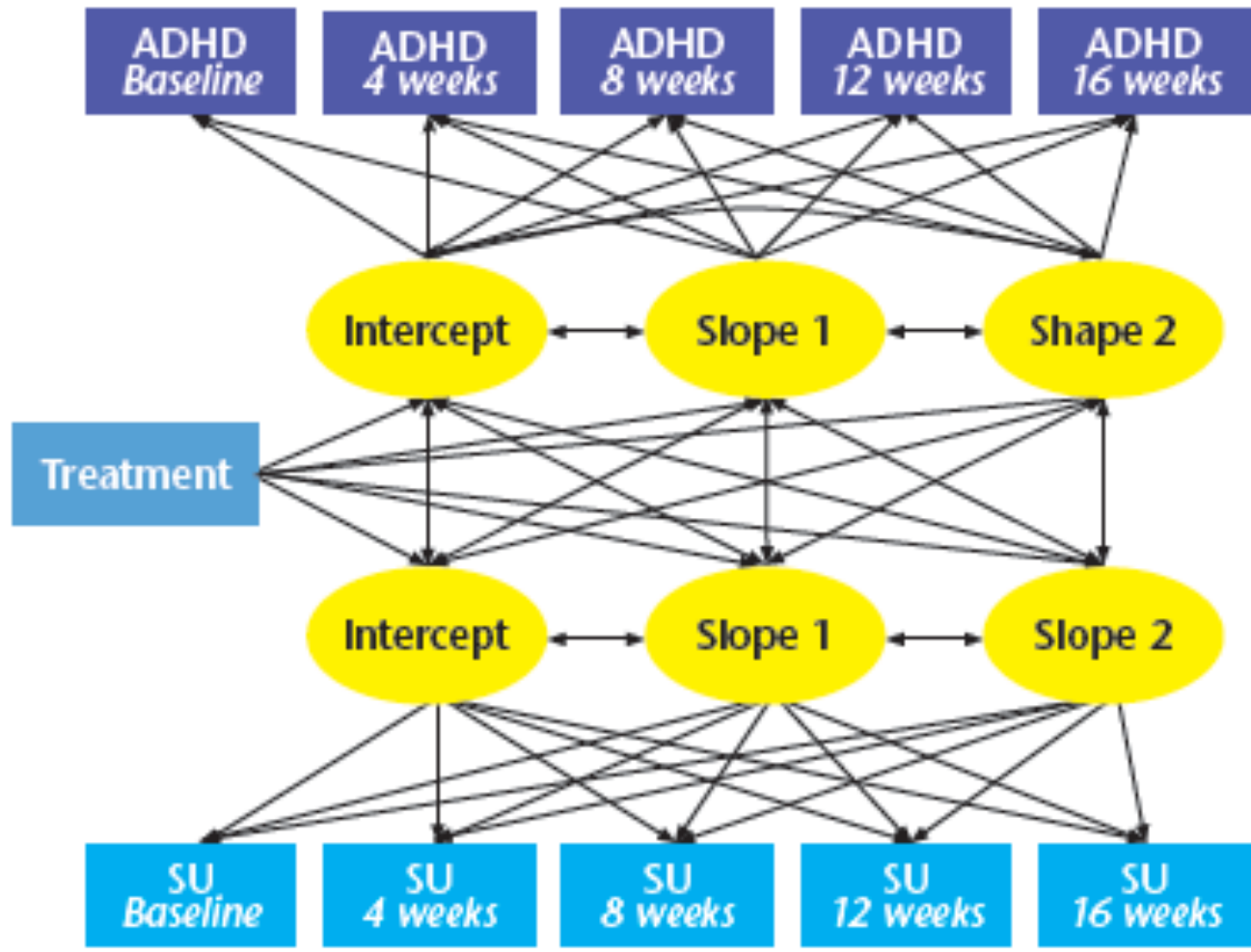


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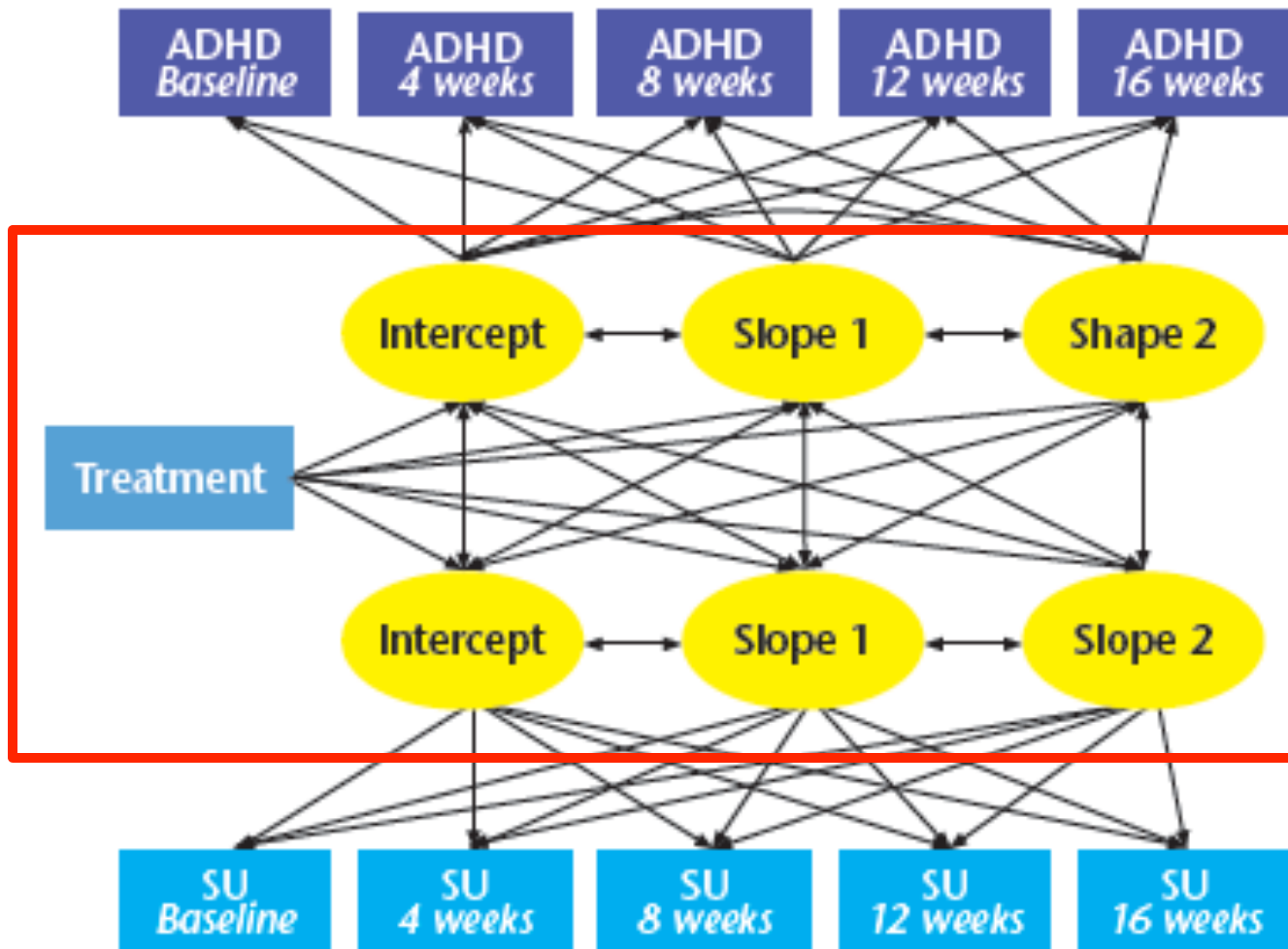


Table 1

Piecewise, Parallel Growth Model of Substance Use (SU) and Attention-Deficit/Hyperactivity Disorder Across 16-Weeks: Missing at Random Model

<i>Outcomes</i>	<i>Unstandardized Estimate</i>	<i>Standard Error</i>	<i>p-value</i>
ADHD Intercept	- 1.216	- 1.196	0.232
SU Intercept	-0.856	1.087	0.431
ADHD Slope 1	0.705	1.417	0.619
ADHD Slope 2	0.284	0.405	0.483
SU Slope 1	0.123	1.006	0.903
SU Shape 2	- 0.125	0.345	0.717

Unstandardized estimates represent the beta coefficient of the indicated random effect regressed onto treatment (osmotic-release methylphenidate + CBT = 1; 0 = placebo + CBT). *p < .05.

RESULTS: PARALLEL GROWTH MODEL

- MAR model: No significant treatment effect on ADHD or SU, small effect sizes for both ADHD and SU ($d=0.16, 0.10$, respectively)

Figure 2. Parallel, piecewise Diggle-Kenward latent growth curve model of substance use (SU) and attention-deficit/hyperactivity disorder (ADHD).

Note: Dashed lines represent parameters associated with the dropout indicators.

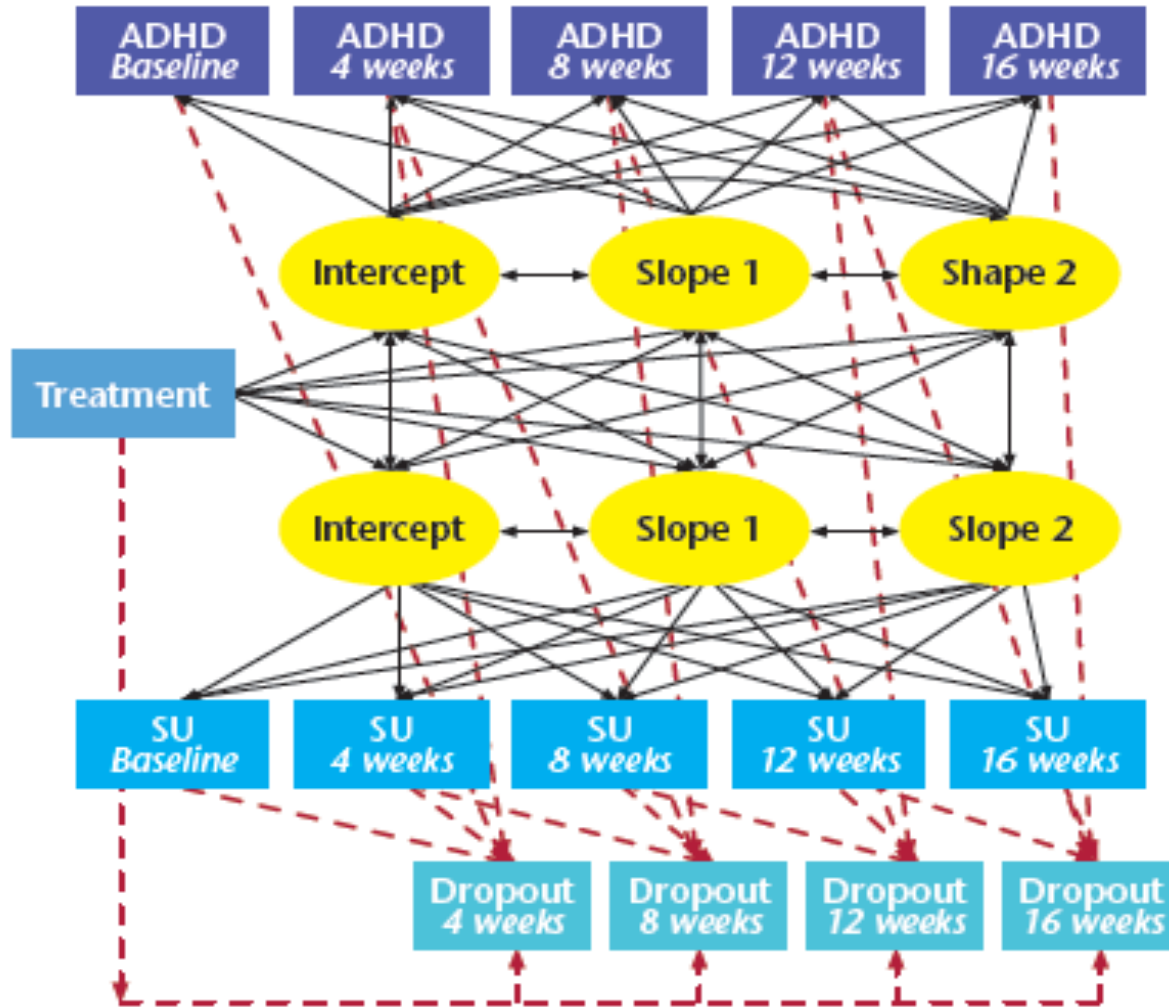


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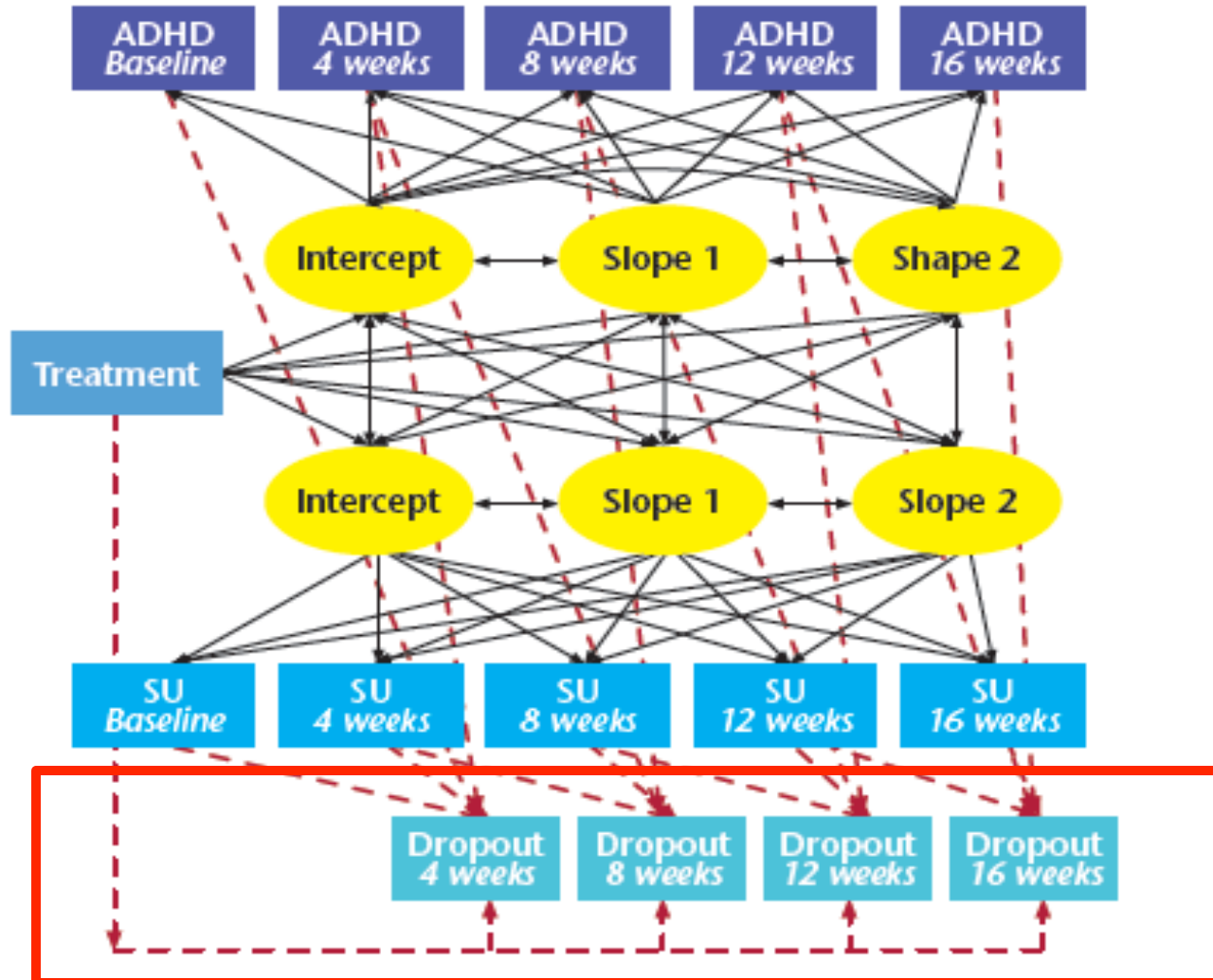


Table 2

Piecewise, Parallel Growth Model of Substance Use (SU) and Attention-Deficit/Hyperactivity Disorder Across 16-Weeks: Missing Not at Random (Diggle-Kenward) Model

<i>Outcomes</i>	<i>Unstandardized Estimate</i>	<i>Standard Error</i>	<i>p-value</i>
ADHD Intercept	- 1.214	1.017	0.233
SU Intercept	- 0.870	1.088	0.424
ADHD Slope 1	0.540	1.449	0.709
ADHD Shape 2	0.277	0.409	0.499
SU Slope 1	0.298	1.017	0.769
SU Slope 2	- 0.144	0.342	0.673
Dropout	- 0.375	0.248	0.130

Unstandardized estimates represent the beta coefficient of the indicated random effect regressed onto treatment (osmotic-release methylphenidate + CBT = 1; 0 = placebo + CBT). *p < .05.

RESULTS: PARALLEL GROWTH MODEL

- MNAR DK model: No significant treatment effects with similar effect sizes of ADHD ($d=0.03$) and SU ($d=0.11$)

Figure 3. Parallel, piecewise Wu-Carroll latent growth curve model of substance use (SU) and attention-deficit/hyperactivity disorder (ADHD).

Note: Dashed lines represent parameters associated with the dropout indicators.

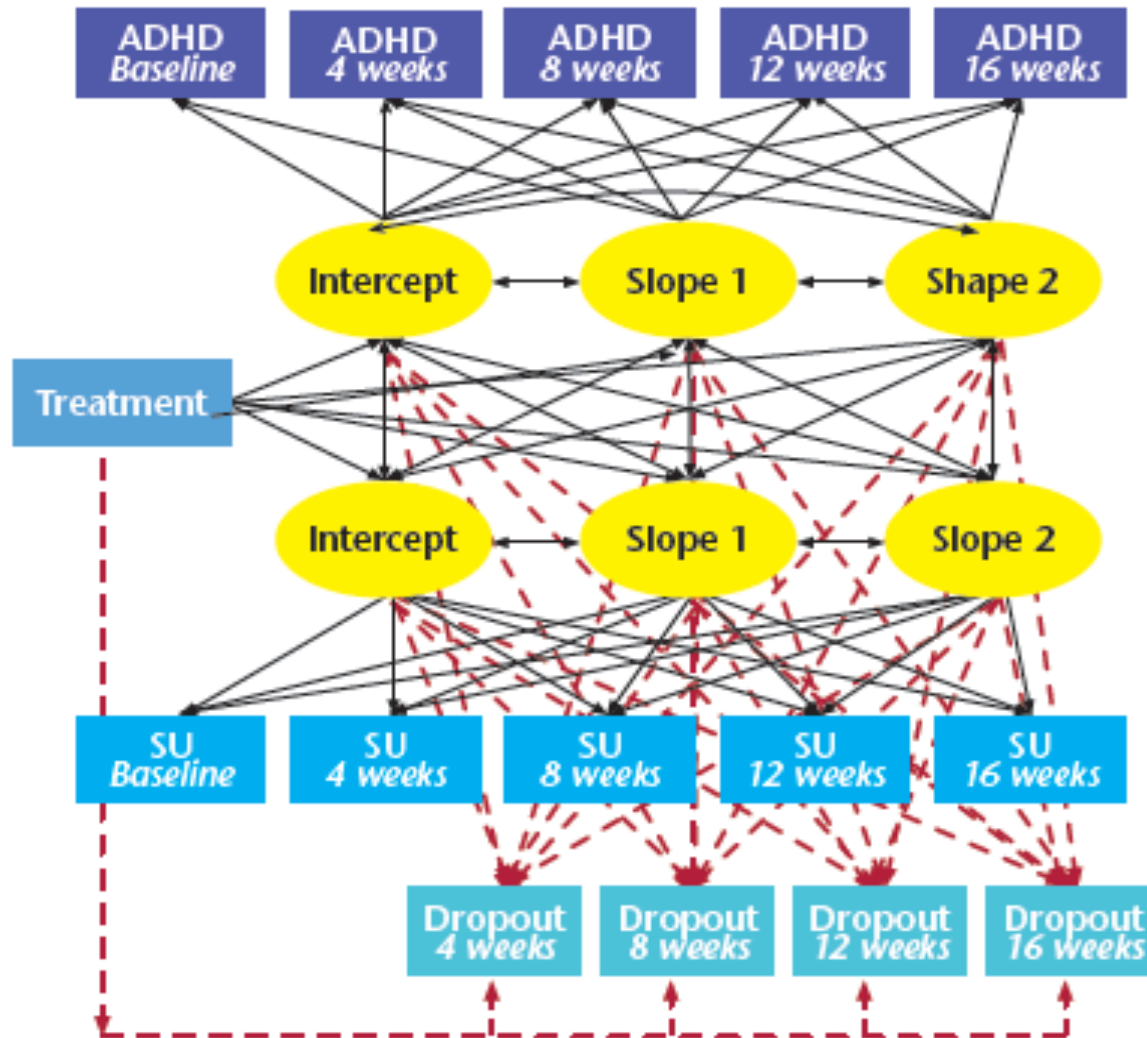


Figure 3. Parallel, piecewise Wu-Carroll latent growth curve model of substance use (SU) and attention-deficit/hyperactivity disorder (ADHD).

Note: Dashed lines represent parameters associated with the dropout indicators.

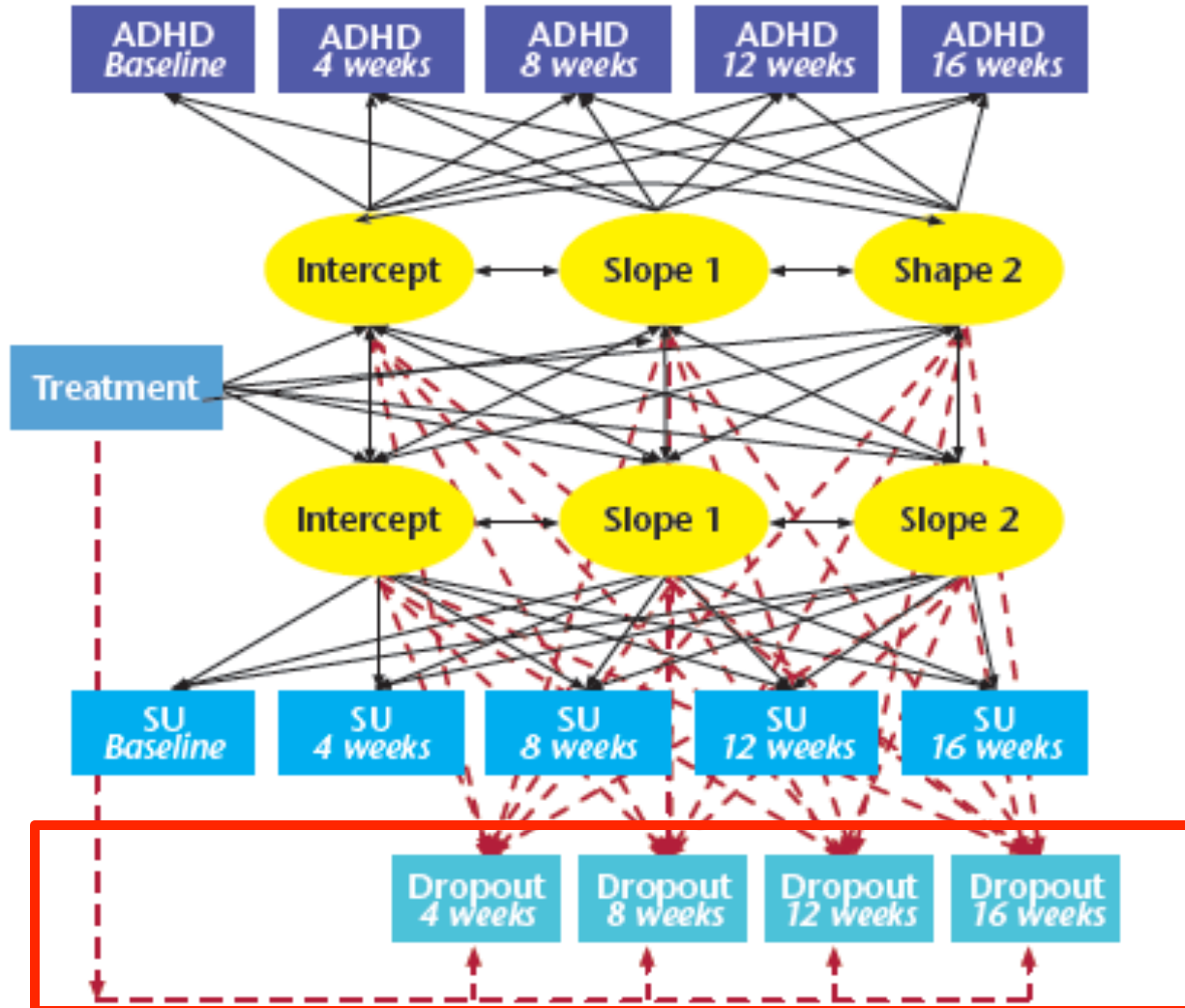


Table 3

Piecewise, Parallel Growth Model of Substance Use (SU) and Attention-Deficit/Hyperactivity Disorder Across 16-Weeks: Missing Not at Random (Wu-Carroll) Model

<i>Outcomes</i>	<i>Unstandardized Estimate</i>	<i>Standard Error</i>	<i>p-value</i>
ADHD Intercept	-1.355	1.198	0.258
SU Intercept	-3.745	2.532	0.139
ADHD Slope 1	0.760	1.688	0.653
ADHD Shape 2	0.333	1.172	0.776
SU Slope 1	6.431	0.834	<0.001*
SU Slope 2	- 1.947	1.429	0.173
Dropout	4.572	4.541	0.314

Unstandardized estimates represent the beta coefficient of the indicated random effect regressed onto treatment (osmotic-release methylphenidate + CBT = 1; 0 = placebo + CBT). *p < .05.

RESULTS: PARALLEL GROWTH MODEL

- The MNAR WC model evidenced a significant effect of OROS relative to placebo on SU, and the effect sizes for both ADHD ($d=2.11$) and SU ($d=1.09$) were larger than reported in the other models

CONCLUSIONS

- Consider whether or not the additional assumptions associated with each MNAR and MAR model are reasonable ^{3-5,10}
- Observing how treatment effects vary across missing data methods should move research teams to explore missing data patterns and mechanism(s)¹¹

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Thank You!

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