# Methodological and Analytical Aspects of Longitudinal Research

#### Koletić, Goran

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Methodological and Analytical Aspects of Longitudinal Research

COORDINATE project November 17, 2023





Goran Koletić, PhD Department of Sociology Faculty of Humanities and Social Sciences University of Zagreb

# LECTURE CONTENT

- 1. Aspects of longitudinal design (a brief overview)
  - Longitudinal hypotheses
  - Sample size
  - Number of waves
  - Measures
- 2. Challenges and recommendations
  - Logistic
  - Methodological
    - Attrition

3. Choosing an analysis framework (a brief overview + example)

Data collected using multiple measurement occasions across time nested within same entities (e.g., individuals – within-individual changes over time)

### Longitudinal study



### At least three measurement occasions (waves)

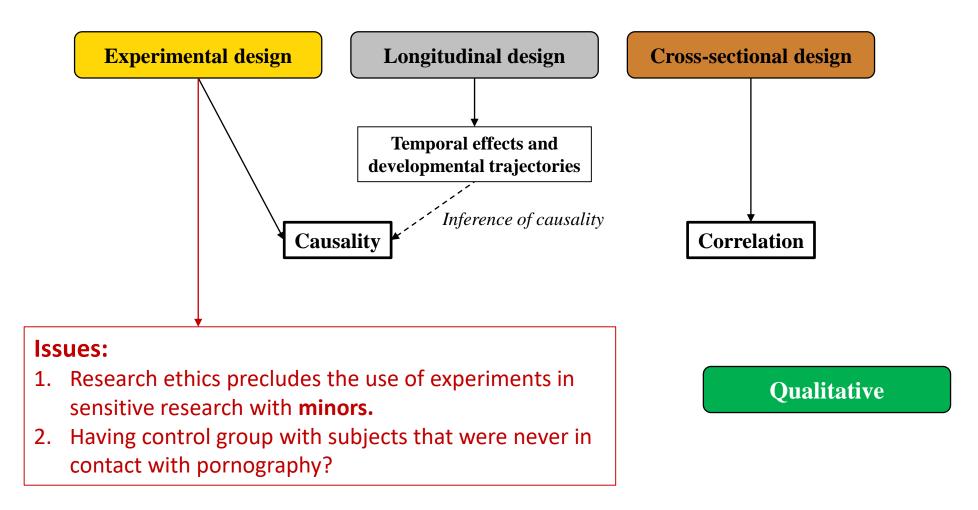
With only two waves of data: a) difficult to disentangle true change from measurement error;
b) impossible to model nonlinear forms of change.

Different from (econometric) time-series design.

https://www.scribbr.com/methodology/longitudinal-study/

## In some cases...THE BEST POSSIBLE DESIGN?

How to assess pornography use in adolescent population?



# NEED FOR LONGITUDINAL DESIGN

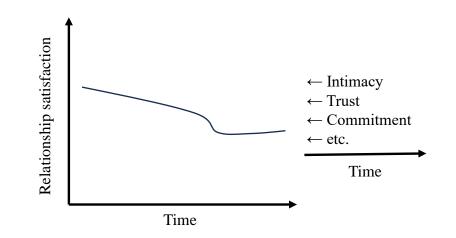
- 1. Assess temporal effects and developmental trajectories.
- 2. Test existing (cross-sectional) theory from a longitudinal perspective.
- 3. While cross-sectional studies render causal inference unwarranted, well-conducted longitudinal study will provide <u>an inference of causality</u>.
  - For example, should we pursuit experimental design for a target topic or not?
- 4. Do conclusions differ between cross-sectional and longitudinal designs?
  - Compare longitudinal effect sizes to the cross-sectional <u>effect sizes</u>.
  - If a longitudinal study makes the same predictions and leads to the same conclusions as a cross-sectional study, is there a unique theoretical contribution?

# **CONSTRUCTING LONGITUDINAL HYPOTHESES**

- Not uncommon that theories (or research-related conclusions) overlook **when** an effect is likely to occur or for what **duration.** 
  - Longitudinal versions of cross-sectional hypotheses
    - "A is associated with  $B" \rightarrow$  "A is associated with B *over time*"

### • Focus on unique change in a construct (vs. its static representation):

- 1. When does the change occur?
- 2. For how long it lasts and how it changes?
- 3. Why it changes?
- 4. What is <u>associated</u> with the change?
- 5. What is the nature of the association?
  - Decreasing/increasing trend
    - Less or more substantial change



# SAMPLE SIZE

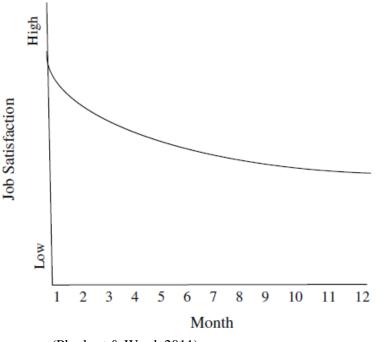
### As large as possible! (attrition)

### Keep in mind that:

- 1. Total sample size (observations) = subjects x measurement occasions
- 2. Within-subject designs have smaller error terms (compared to cross-sectional designs)
- 3. Power analysis is complex (subjects, measurement occasions, linear or nonlinear change, variability in change over time)
  - Adding more subjects should reflect on between-person effects
  - Adding more measurement occasions should reflect on within-person effects

# **Exactly equal spacing (less important) vs. number of time points (more important)**

• Frequent enough to detect hypothesized kind of change and that the occasions cover a reasonable duration of time



### **Possible assessments:**

- **T1, T12:** detecting linear decline
- T1, T6, T12: detecting nonlinearity
- T1, T2, T3: overestimating negative slope
- T10, T11, T12: underestimating negative slope

#### **Guidelines for number measurement occasions (and time lags)**

- 1. Review related literature.
- 2. When there is no "natural" measurement dynamic, conduct interviews or behavioral observations with relevant subjects to determine a measurement schedule.

<sup>(</sup>Ployhart & Ward, 2011)

# CHALLENGES

### LOGISTIC

- Time consuming (number of waves, time lags)
- Recruitment
  - Often requires larger baseline samples
- "Gatekeepers" (classroom-based)
  - School principals, etc.
- Motivating participants (online)
  - Incentives

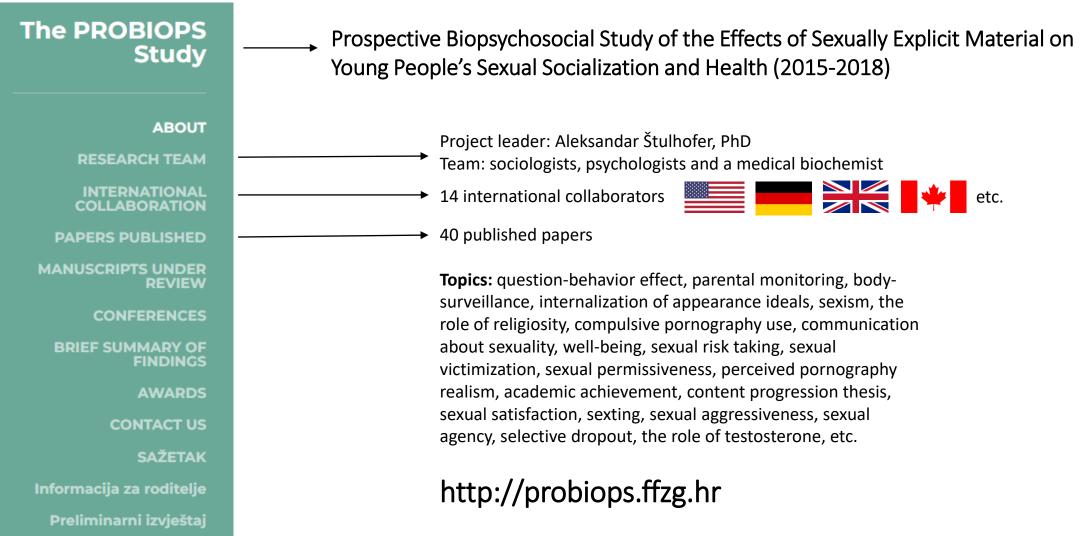
### METHODOLOGICAL

- Attrition
  - Reasons for lost to follow-up participants
  - Potential bias
  - Online vs. classroom-based (on-site) vs. commercial panel
- Familiarity with research topic and measures
- (Re)contacting participants and linking surveys
- Assuring anonymity (online)
- Assuring privacy (classroom-based)

### FINANCIAL

- Requires a research team
- Expensive

# **RECENT EXPERIANCE**



Funded by Croatian Science Foundation

# **PROBIOPS:** Participants and procedures

### ZAGREB

- Spring 2015.
- 59/90 high-schools
- 6 waves
- 6 month between waves
- Leaflet recruitment
- Online questionnaires
- N (T1<sub>baseline</sub>) = 2,235
- Lottery based incentives

**RIJEKA** (population wise, 3<sup>rd</sup> Croatian city)

- Winter 2015.
- 14/23 high-schools
- 6 waves
- 5-6 month between waves
- Classroom based
- Paper-pen questionnaires
- N (T1<sub>baseline</sub>) = 1,287
- No incentives

ZAGREB								
Wave	Year	Ν						
1	2015	2235						
2	2013	636 🔸						
3	2016	711						
4	2016	683						
5	2017	686						
6	2017	511						

Mean age (T1) = 16.2 41% M / 59% F

	RIJEKA								
Wave	Year	Ν							
1	2015	1287							
2	2016	1281							
3	2016	1232							
4	2017	1176							
5	2017	931							
6	2018	892							

Mean age (T1) = 15.9 44% M / 56% F

All 6 waves = 307

All 6 waves = 430

# **RECOMMENDATIONS (long before data collection)**



**Developing a catchy public name and an attractive visual identity** + feedback (e.g., focus groups)



Developing a "recruitment" leaflet and video tutorial



# **RECOMMENDATIONS** (long before data collection)

#### Setting up a registration website and social media sites

**Deciding on incentives model** (e.g., lottery based) + feedback

#### System for assuring confidentiality

(separate database for contact information and questionnaire data + linking database)

#### MEDIJI + JA

O ISTRAŽIVANJU SUDJELUJ U NASTAVKU ISTRAŽIVANJA

#### 1. Ulaz za registrirane korisnike

Prijavi se putem Facebook-a Klikni na

Prijavi se pomoću svoje E-mail adrese



Korisničko ime



Zaboravljeno korisničko ime?

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ZAGREB					
Wave	Collected				
1	27%				
2	57%				
3	63%				
4	61%				
5	67%				
6	62%				

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# A brief detour...INCENTIVES

### **Types of incentives**

- An incentive which shows respect for participants' time and effort
- Money, gift cards, food vouchers, school supplies, telephone cards, etc.

# Determining adequate incentive

- Incentive amounts vary depending on many factors, including:
  - Study budget
  - Standard of living in the study country
  - Population of interest
  - Institutional or governmental policies (monetary incentives not allowed, pre-established cap amount for incentives)

### Models

- 1. Each participant
- Each participant + extra for participating in each subsequent wave
- 3. One-price lottery
- 4. Horizontal lottery (a number of awards, same incentive amount)
- 5. Pyramidal lottery (a number of awards, increasing incentive amount)
- 6. Combining previous models

### Acquire feedback!

# **RECOMMENDATIONS** (before data collection)

#### Training a fieldwork force

### **Developing necessary planning/tracking sheets** (coordinating, contacting, and measures!)

A B C D E F G H I J K L M N O P Q R S T U V W X Y



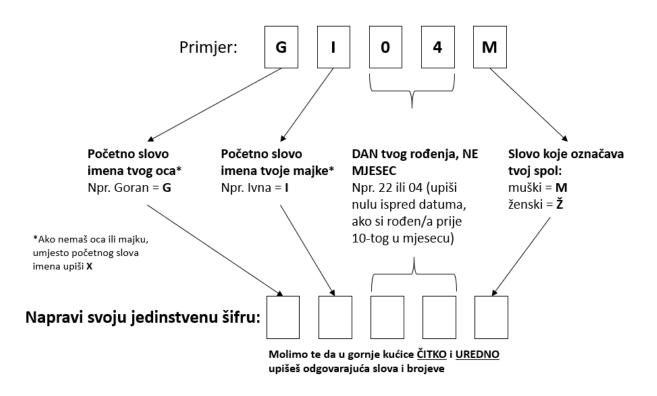
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### **Classroom-based data collection**

Use privacy panels



#### Develop coding system for linking participants across multiple study waves

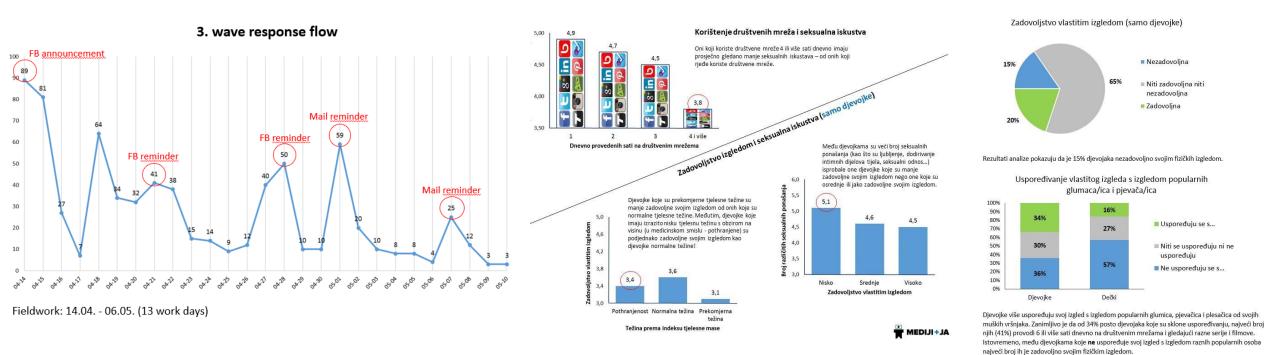


#### **Response tracking**

**Issues:** rarely checking email, changing email address, using "secondary" email for the initial registration

#### Maintaining communication before/during/after data collection periods

- Repeated in-person visits
- Social media posts with interesting results
- Social media and e-mail announcements
- Social media and e-mail <u>reminders</u>



#### Maintaining communication before/during/after data collection periods

• Repeated in-person visits

1. w	vave	2. w	vave	3. Wave			
Schools	Respondents	Visited	Respondents	Visited	Respondents		
59	2210	35	542 (M* = 32%)	31	601 (M* = 39%)		
		Not visited	Respondents	Not visited	Respondents		
		24	68 (M* = 15%)	28	124 (M* = 21%)		
		Schools	Respondents	Schools	Respondents		
		59	610	59	725		
		<u> </u>	"in-person effect"		"in-person effect"		
			+17%		+18%		

\*M = average response rate in (not)visited schools based on the number of baseline respondents

#### **Qualitative feedback** (intention to dropout, satisfaction with incentives, etc.)

BARRIERS	IVANIĆ GRAD	SESVETE	ZAGREB
Long time Intervals and forgetfulness	~		
Rare communication with participants, lack of reminders	~	~	
Image research too laid back. Unrecognized seriousness and importance of research	~		
Prizes are not particularly motivating			~
Use of e-mail compared to Facebook, WhatsApp or similar		~	~
Immaturity (boys)			~

RECOMMENDATIONS AND SUGGESTIONS:	IVANIĆ GRAD	SESVETE	ZAGREB
Joining the Facebook group	~		
Frequent e-mails (though rarely checked)	~	~	
More content on the Facebook group	~		
Be sure to visit schools	~		✓
Completing the questionnaire during class	~	~	~
emphasize prices		~	
Create an impression of obligation to an adult at school (teacher, psychologist)			✓
Some will be more interested if you have "dirty" questions		✓	

### **Document sampling flow**

ZAGREB	Data collection dates	Total sample size	Sample size after dataset cleaning	Sample size in MASTER data		ZAGREB - Number of	participants in al	l previous waves (l	inked)			ZAGREB	- Partic	ipared i	n: (number of wa	ives)
ZG W1	10 April - 11 May 2015	2655	2241	2235	w1 = 223	5 w12 = 636	w123 = 486	w1234 = 417	w12345 = 372			only 1 wave	1178	53%	at least 1 wave	2235
ZG W2	10 Nov - 27 Nov 2015	680	644	636	w2 = 636	w13 = 711	w124 = 462	w1235 = 400	w123456 = 307	7		only 2 waves	273	12%	any 2 waves	1057
ZG W3	14 April - 6 May 2016	766	727	711	w3 = 711	w14 = 683	w125 = 448	w1245 = 397				only 3 waves	172	8%	any 3 waves	784
ZG W4	7 Oct - 28 Oct 2016	739	692	683	w4 = 683	w15 = 686	w134 = 534	w1345 = 455				only 4 waves	145	6%	any 4 waves	612
ZG W5	15 Mar - 27 Mar 2017	761	693	686	w5 = 686	w23 = 486	w135 = 522	w2345 = 372				only 5 waves	160	7%	any 5 waves	467
ZG W6	17 Sep - 2 Oct 2017	542	517	511	w6 = 511	w24 = 462	w145 = 520					only 6 waves	307	14%	all 6 waves	307
						w25 = 448	w234 = 417					TOTAL	2235	100%		
						w34 = 534	w235 = 400									
						w35 = 522	w245 = 397									
						w45 = 520	w345 = 455									
RIJEKA	Data collection dates	Total sample size	Sample size after dataset cleaning	Sample size in MASTER data		RIJEKA - Num	nber of participan	ts in all previous w	vaves (linked)		Unique	RIJEKA	- Partici	ipared ir	: (number of way	ves)
RI W1	7 Dec 2015 - 1 Feb 2016	1307	1291	1287	w1 = 128	7 w12 = 1059	w123 = 883	w1234 = 758	w12345 = 534	w123456 = 430 w	1u = 69	only 1 wave	291	17%	at least 1 wave	1744
RI W2	18 April - 13 May 2016	1309	1283	1281	w2 = 128:	1 w13 = 1007	w124 = 862	w1235 = 588	w12346 = 516	w	2u = 68	only 2 waves	164	9%	any 2 waves	1453
RI W3	7 Oct - 27 Oct 2016	1252	1233	1232	w3 = 1232	2 w14 = 977	w125 = 665	w1236 = 582	w12356 = 473	w	3u = 49	only 3 waves	205	12%	any 3 waves	1289
RI W4	13 Mar - 12 Apr 2017	1202	1177	1176	w4 = 117	5 w15 = 757	w126 = 661	w1245 = 592	w12456 = 481	w	4u = 46	only 4 waves	285	16%	any 4 waves	1084
RI W5	2 Oct - 3 Nov 2017	944	931	931	w5 = 931	w16 = 746	w134 = 850	w1246 = 578	w13456 = 476	-	5u = 38	only 5 waves	369		any 5 waves	799
RI W6	5 Mar - 30 Mar 2018	924	892	892	w6 = 892	w23 = 1012	w135 = 665	w1256 = 534	w23456 = 469	w	6u = 21	all 6 waves	430	25%	all 6 waves	430
						w24 = 968	w136 = 651	w1345 = 596		Т	OTAL = 291	TOTAL	1744	100%		
						w25 = 744	w145 = 662	w1346 = 571								
						w26 = 735	w146 = 641	w1356 = 531								
						w34 = 968	w156 = 599	w1456 = 531								
	ZAGREB		RIJE			w35 = 756	w234 = 846	w2345 = 585								
w1	spring, 2015, 2. class		autumn, 2015, 2. cla	ass		w36 = 737	w235 = 654	w2346 = 562								
w2	autumn, 2015, 3. class		spring, 2016 2. clas			w45 = 751	w236 = 643	w2356 = 521								
w3	spring, 2016, 3. class		autumn, 2016, 3. cla	ass		w46 = 716	w245 = 653	w2456 = 525								
w4	autumn, 2016, 4. class		spring, 2017 3. clas			w56 = 696	w246 = 631	w3456 = 526								
w5	spring, 2017, 4. class		autumn, 2017, 4. cla				w256 = 589									
w6	autumn, 2017, out of sch	lool	spring, 2018, 4. clas	55			w345 = 660									

### Are we losing the most relevant cases first? (attrition)

- In longitudinal research, losing particular types of participants over the course of the study may become a serious analytical issue (e.g., identifying moderating effects, diminishing or inflating links between predictors and outcomes of interest).
  - *Štulhofer et al. (2021). Selective Dropout in Longitudinal Studies of Adolescent Pornography. Archives of sexual behavior, 50, 2215–2226.*

### • Using two independent panel samples, we examined:

- 1. Was attrition substantially different among adolescents who may be particularly <u>vulnerable to pornography use</u> compared to other participants?
  - Vulnerability indicators (measured at the baseline): adverse family situation, lower academic achievement, early biological maturation, lower self-esteem, sexual aggressiveness, earlier sexual debut.
- 2. Did panel type (online vs. classroom-based) moderate associations between attrition and the vulnerable group membership?

### Are we losing the most relevant cases first? (attrition)

• Based on attrition patterns in two panels, we distinguished: early attrition, later attrition, and participation gaps.

### RESULTS

- 1. Only <u>early attrition</u> was substantially higher among more vulnerable adolescents, compared with other participants.
- 2. Panel type moderated the associations between adolescent vulnerability and <u>participation gaps</u>, which was significant for the classroom-based but not the online panel.

Adolescents who are believed to be under increased risk of adverse outcomes associated with pornography use are less likely to complete longitudinal studies.

### Are we losing the most relevant cases first? (attrition)

**Pre-designed attrition reducing strategies (examples)** 

#### Modality of data collection

- Resources and required baseline sample
- Online data collections platforms vs. cell phone app

#### **Preparations for attrition**

- Short questionaries (and planned missing)
- Study's visual identity and presence
- Desirable incentives
- Focus groups (before and during data collection)

#### **Delaying selective dropout**

- Notifying participants about an upcoming study wave
- Communicating simple but interesting findings
- Adding or modifying incentives (e.g., adding bonus incentives tied to the number of waves completed)
- Seeding the panel with specially incentivized and committed peer leaders

### Are we losing the most relevant cases first? (attrition)

A simple analytic approach to assess attrition

For example, assessing attrition from T1 to T2

- N (T1, baseline) = 100
- N (T2) = 75

Binary logistic regression analysis

- Which participants have higher odds for dropping out?
- Use T1 data
- **DV** 
  - 0 = Participants in T2 (75)
  - 1 = Lost to follow-up (25)
- IV
  - Relevant predictors of attrition (age, gender, etc.)

### **CHOOSING AN ANALYSIS FRAMEWORK**



## The Hitchhiker's guide to longitudinal models: A primer on model selection for repeated-measures methods

Ethan M. McCormick <sup>a,b,c,\*</sup>, Michelle L. Byrne <sup>d,e</sup>, John C. Flournoy <sup>f</sup>, Kathryn L. Mills<sup>e</sup>, Jennifer H. Pfeifer <sup>e</sup>

<sup>a</sup> Methodology & Statistics Department, Institute of Psychology, Leiden University, Leiden, Netherlands

<sup>b</sup> Department of Psychology and Neuroscience, University of North Carolina, Chapel Hill, United States

<sup>c</sup> Cognitive Neuroscience Department, Donders Institute for Brain, Cognition and Behavior, Radboud University Medical Center, Nijmegen, Netherlands

<sup>d</sup> Turner Institute for Brain and Mental Health, School of Psychological Sciences, Monash University, Clayton, Australia

e Department of Psychology, University of Oregon, Eugene, United States

<sup>f</sup> Department of Psychology, Harvard University, Cambridge, United States



## **CHOOSING AN ANALYSIS FRAMEWORK**



#### Multilevel (mixed-effect / hierarchical) modeling

- Estimating higher levels of nesting (e.g., beyond individual)
- Limited with respect to measurement error in predictors or outcomes
- Simple inclusion of multiple time-variant covariates (e.g., relation satisfaction) and time-invariant covariates (e.g., gender)
- Relative model fit indices (AIC/BIC and likelihood ratio test) [model comparison]

#### Structural equation modeling (SEM)

- Repeated measures as multiple indicators on one or more <u>latent</u> factors
- Estimating and removing the effect measurement error in predictors or outcomes
- Absolute model fit indices (CFI, TLI, RMSEA)
- Mediated relationships between constructs

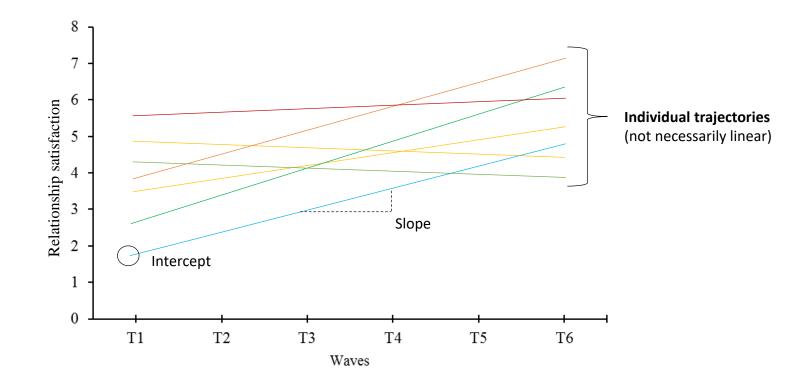
Similarities between the multilevel and SEM frameworks often outweigh the differences.

### **CHOOSING AN ANALYSIS FRAMEWORK: KEY CONSIDERATIONS**

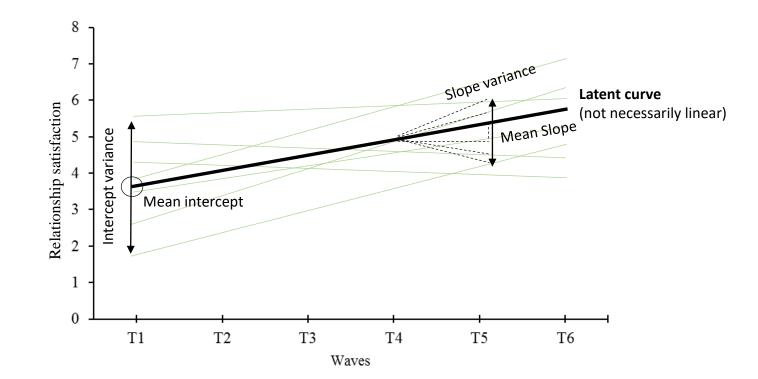
- Research question / hypothesis
- Variable type (categorical / quantitative) // (manifest / latent)
- Number of covariates
- Type of covariates (time-invariant / time-variant)
- (Un)balanced data (unequally spaced measurement occasions and/or missing data)
- Type of change (growth curve)
- Higher-order nesting
- <u>Software</u>

- Enables an assessment of <u>between-person differences</u> over time by estimating <u>within-person latent trajectories of change</u>
  - Observed repeated measures of a construct are represented by two latent factors (latent intercept and latent slope), and their means and variances
  - Latent intercept = initial level of a measured construct
  - Latent slope = measured construct's change over time

Assessing group means and between-person differences over time

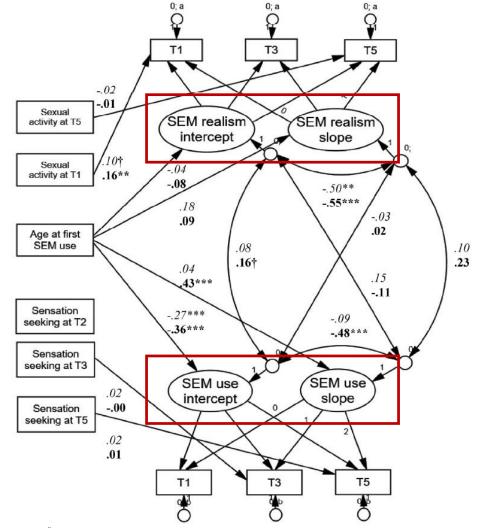


Estimating between-person differences in <u>within-person change</u> over time using **latent curve** and its **intercept** and **slope** 



### Advantages

- Assessing multiple constructs simultaneously (parallel LGCM)
- Ability to handle unequally spaced measurement occasions, nonlinear trajectories, and partially missing data
- Flexibility of including both timeinvariant and time-varying covariates



Wright, P. J., & Štulhofer, A. (2019). Adolescent pornography use and the dynamics of perceived pornography realism: Does seeing more make it more realistic? *Computers in Human Behavior*, *95*, 37–47. <u>https://doi.org/10.1016/j.chb.2019.01.024</u>

Interpretations of positive correlations between two latent constructs:

	Construc		
Construct B - SLOPE	Increasing trend Decreasing trend		Construct A - INTERCEPT
Increasing trend	The higher the increase in construct A, the more substantial the increase in construct B	The higher the increase in construct B, the less substaintal the decrease in construct A	The higher the baseline assessment of construct A, the <b>more</b> substantial the increase in construct B over time
Decreasing trend	The higher the increase in construct A, the less substaintal the decrease in construct B	The steeper the decrease in construct A, the more substantial the decrease in construct B (alternatively – both are decreasing less steeply)	construct A, the less substantial the

Interpretations of negative correlations between two latent constructs:

	Construc	t A - SLOPE			
Construct B - SLOPE		Construct A - INTERCEPT			
	Increasing trend Decreasing trend				
Increasing trend	The higher the increase in construct A, the less substantial the increase in construct B	The higher the increase in construct B, the more substaintal the decrease in construct A	The higher the baseline assessment of construct A, the less substantial the increase in construct B		
Decreasing trend	The higher the increase in construct A, the more substantial the decrease in construct B	The steeper the decrease in construct A, the less substantial the decrease in construct B	The higher the baseline assessment of construct A, the <b>more</b> substantial the decrease in construct B		

### Final remark...

Common statement ("mantra") in research papers: *More longitudinal research is needed*.

Time/effort/costs vs. sound empirical/theorical contribution

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