

Longitudinal Causal Inference in Psychology

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1. What is Causal Inference (Potential Outcomes) in longitudinal Settings?
2. Can we do that in psychological research?

What is Causal Inference (Potential Outcomes) in longitudinal Settings?



X_t

Potential Outcome:

$$Y_i^x \quad (1)$$



Y

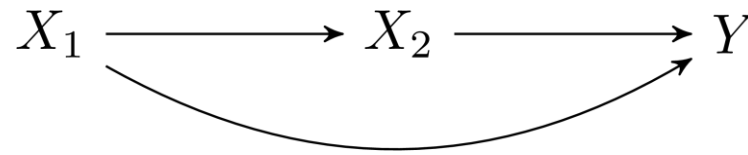
Individual Causal Effect:

$$Y_i^{x=1} - Y_i^{x=0} \quad (2)$$

Average Treatment Effect (ATE):

$$E[Y^{x=1}] - E[Y^{x=0}] \quad (3)$$

What is Causal Inference (Potential Outcomes) in longitudinal Settings?



X_t : Smartphone Usage



Y : Sleep quality

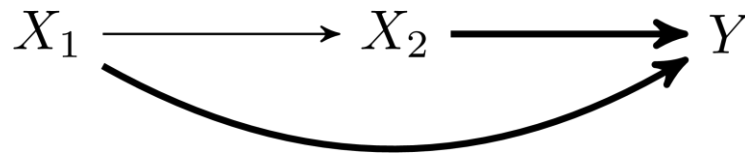
What is Causal Inference (Potential Outcomes) in longitudinal Settings?



X_t



Y



A Potential Outcome is defined under multiple interventions:

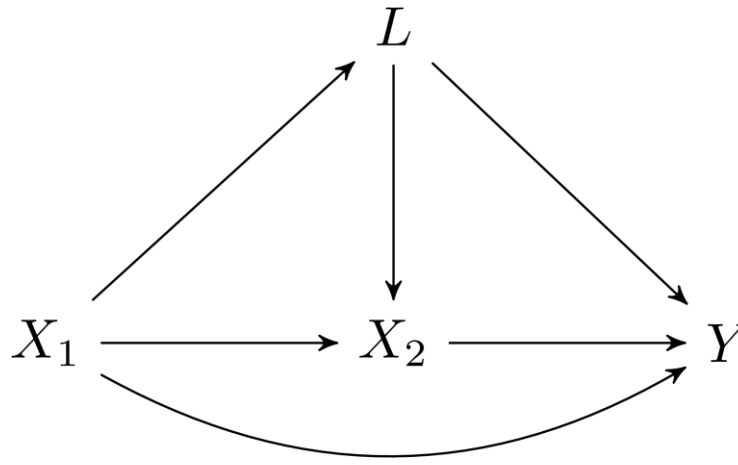
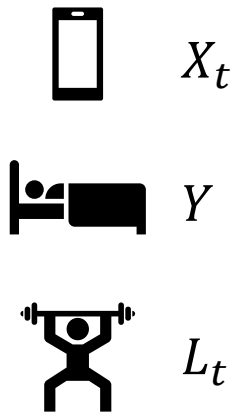
$$Y^{x_1, x_2}$$

The contrast of two Potential Outcomes is then the *joint effect*:

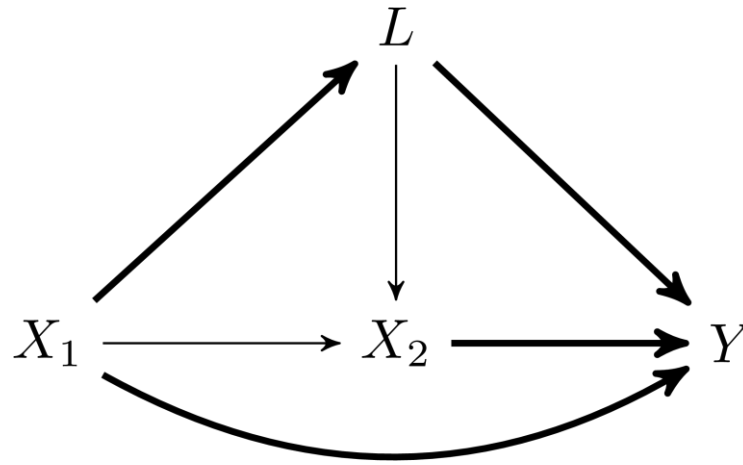
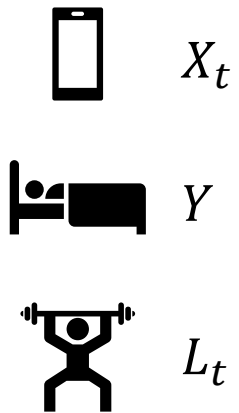
$$Y^{x_1=1, x_2=1} - Y^{x_1=0, x_2=0}$$

$$Y^{x_1=1, x_2=0} - Y^{x_1=0, x_2=0}$$

Longitudinal Causal Inference: It's complicated...



Longitudinal Causal Inference: It's complicated...



How to identify joint effects?

1. Experiment

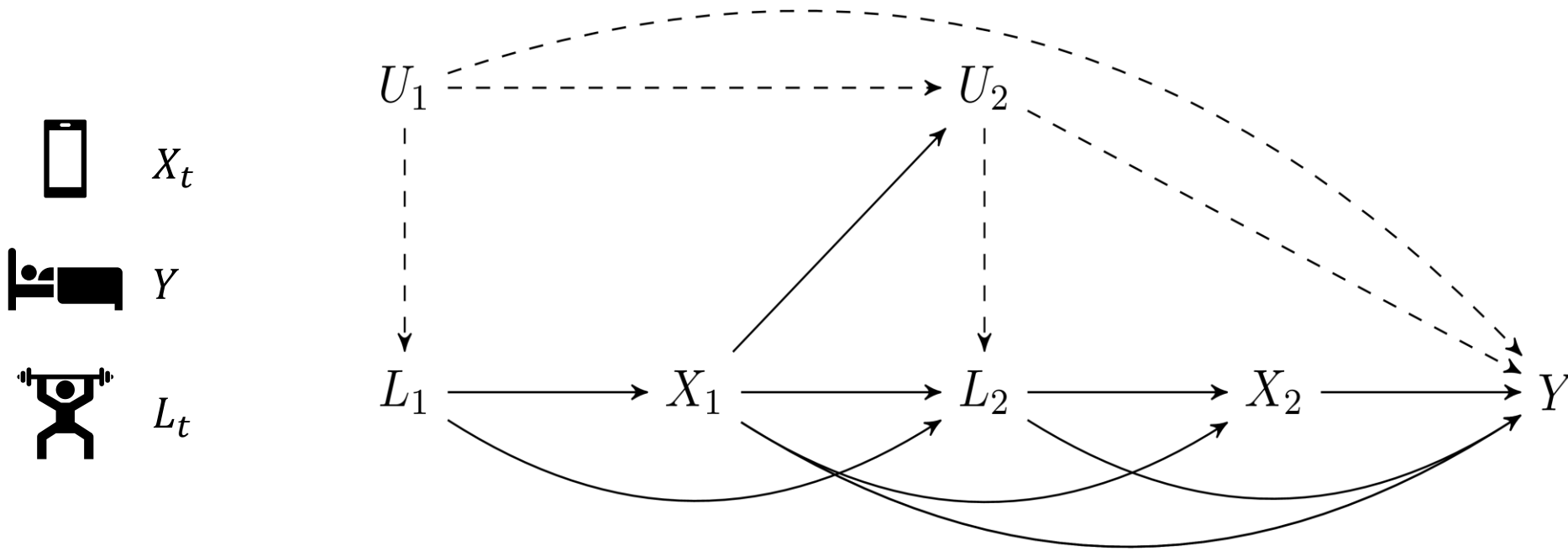
1. X_t is randomized at every time point t
2. Or: a full strategy is randomized ($x_1 = 1, x_2 = 1$)

2. Observational Study: Confounder adjustment

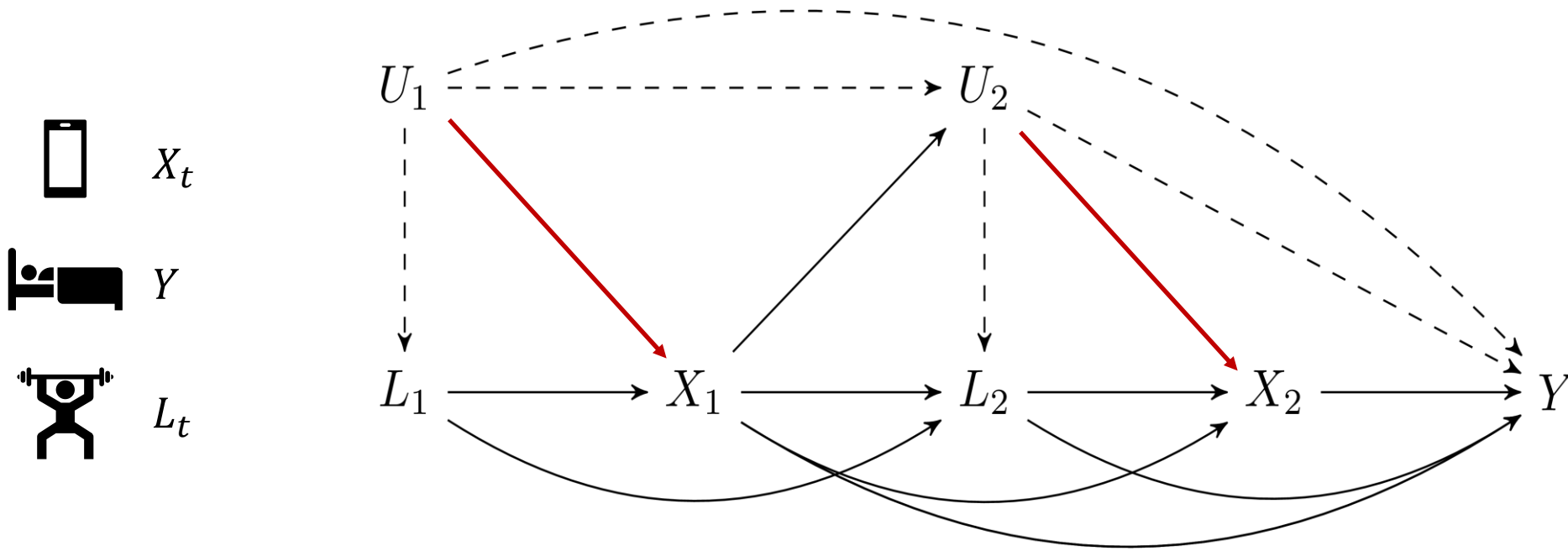
- If we know the relevant „reasons“ for X_t , we can adjust for them

1. What is Causal Inference (Potential Outcomes) in longitudinal Settings?
2. Can we do that in psychological research?

What do we need to identify joint effects?

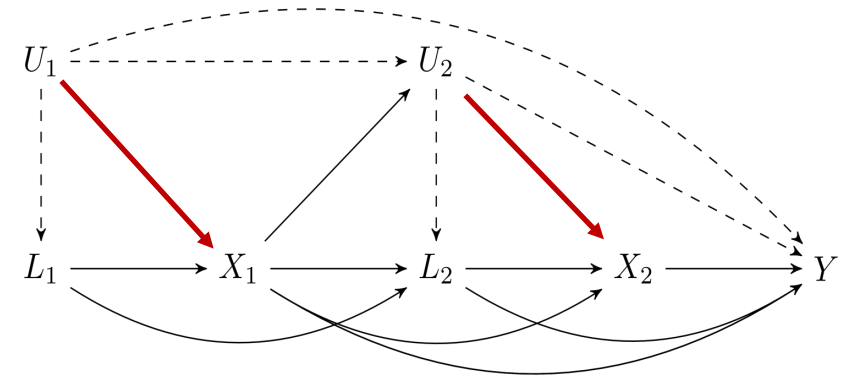


What do we need to identify joint effects?



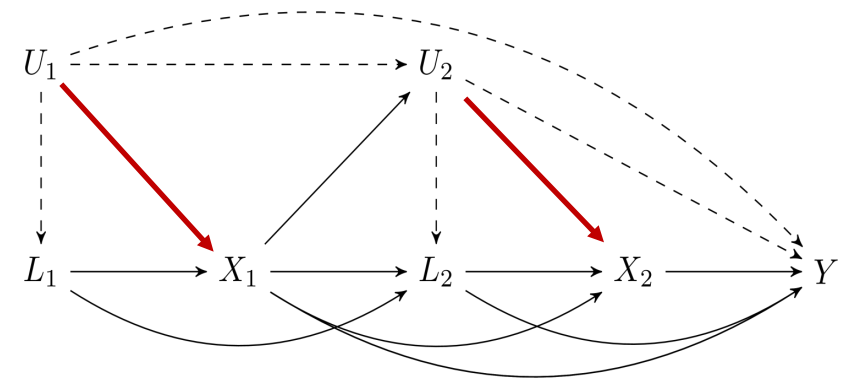
When are variables independent in psychology?

1. Randomization
- 2. Information Restriction**



When are variables independent in psychology?

1. If X is self-selected, U is not measureable
2. If X is assigned by a person **based on information**, it *cannot* be influenced by unobservables



Who assigns X based on observables L?

1. Doctors

X = Treatment

L = Biomarker

Y = Health

2. Psychotherapists

X = Treatment

L = Questionnaire

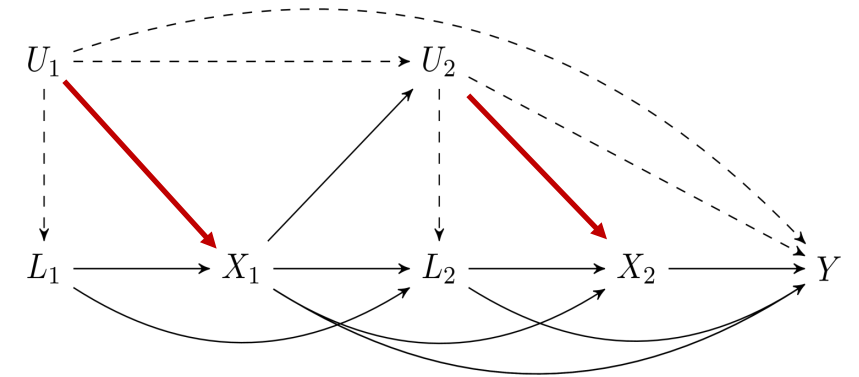
Y = Mental Health

3. Teachers

X = Extra Classes

L = Grades

Y = Passing the course



Summary

- Causal Inference is applicable in longitudinal settings
- Causal Inference is possible without full randomization
- Psychology needs clever research designs
- Leverage Information Restriction

Furthermore...

- SEM (path analysis) relies on unrealistic assumptions
- g-Methods from epidemiology rely on more reasonable assumptions

Junker, L., Schoedel, R., & Pargent, F. (2025). Towards a Clearer Understanding of Causal Estimands: The Importance of Joint Effects in Longitudinal Designs with Time-Varying Treatments. <https://doi.org/10.31234/osf.io/zmh5a>

Thank you!