



Practice: Evaluation Research

Session 03 - GCM and LCSM

psy112 - Evaluation Research

Faculty VI / UOL

Summer term 2025

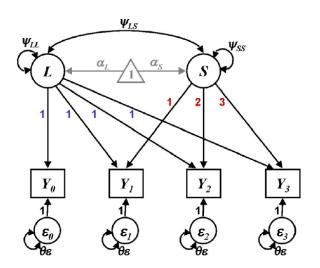
Content

GCMs

2 LCSMs

Summary

Model 1: Linear Growth Curve Model (GCM)



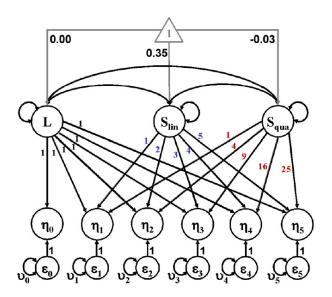
Model 1: Linear Growth Curve Model (GCM)

- **Description:** This is the most fundamental growth model. It captures a constant, straight-line pattern of change over time by estimating an initial level (Intercept, L) and a constant rate of change (Slope, S) for each individual.
- **Simple Equation:** The average score at time *t* is determined by the average starting point and the average slope.

$$E[Y_t] = \alpha_L + t \cdot \alpha_S$$

 When to Use: Use this model when your theory suggests that change should occur at a steady, constant rate over the entire measurement period.

Model 2: Quadratic (Second-Order) GCM



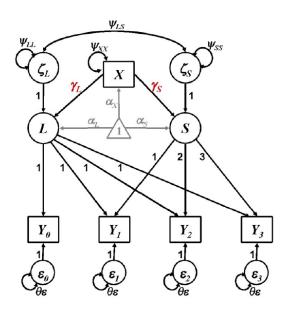
Model 2: Quadratic GCM

- **Description:** This model captures non-linear change by adding a quadratic slope factor (S_{qua}) . It allows for one curve in the trajectory, representing acceleration or deceleration in the rate of change.
- Simple Equation: The average score is now also a function of time squared.

$$E[Y_t] = \alpha_L + t \cdot \alpha_{S_{lin}} + t^2 \cdot \alpha_{S_{qua}}$$

 When to Use: Use this when you expect the rate of change to change over time. For example, learning might be rapid at first and then slow down.

Model 3: GCM with a Time-Invariant Covariate



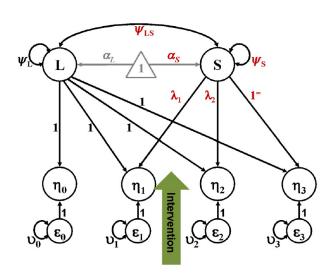
Model 3: GCM with a Time-Invariant Covariate

- **Description:** This model, also called a conditional GCM, explains individual differences in growth trajectories. It tests if a stable characteristic (X), like a personality trait or group membership, predicts the initial level (L) or rate of change (S).
- **Simple Equation:** The average Intercept and Slope now depend on the value of the covariate X.

$$E[L] = \alpha_L + \gamma_L X \quad | \quad E[S] = \alpha_S + \gamma_S X$$

• When to Use: Use this when your goal is to explain *why* individuals show different patterns of change.

Model 4: Nonlinear GCM (Latent Basis)



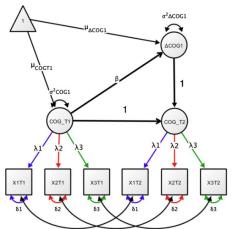
Model 4: Nonlinear GCM (Latent basis)

- **Description:** This model offers flexibility by letting the data determine the shape of the trajectory. Instead of fixing all slope loadings (e.g., 0, 1, 2, 3), some are freely estimated (λ_1, λ_2) , allowing the model to capture unspecified non-linear patterns.
- Simple Equation: The average score's change depends on the estimated time loadings (λ_t) .

$$E[Y_t] = \alpha_L + \lambda_t \cdot \alpha_S$$

• When to Use: Ideal for exploring the shape of change when theory is unclear, or for modeling abrupt shifts, such as those caused by an intervention.

Model 5: Latent Change Score (LCS) Model



Kievit et al. (2018)

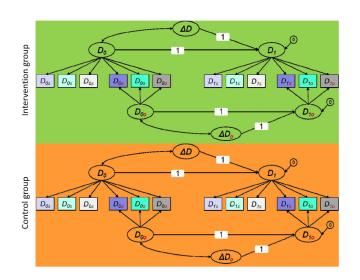
Model 5: Latent Change Score (LCS) Model

- **Description:** Unlike GCMs which model the overall trajectory, the LCS model focuses directly on the **change** between consecutive time points. It models the state at Time 2 as a function of the state at Time 1 plus a latent change score (Δ) .
- **Simple Equation:** The score at the next time point is the score from the previous time point plus the estimated change.

$$E[\mathsf{State}_{t+1}] = E[\mathsf{State}_t] + E[\Delta_{\mathsf{change}}]$$

 When to Use: When your research question is about the dynamics of change itself, such as what influences the transition from one state to the next.

Model 6: Multigroup LCS Model



Model 6: Multigroup LCS Model

- Description: This extends the LCS model to compare the dynamics of change between two or more distinct groups (e.g., an intervention group vs. a control group). The entire model is estimated separately for each group.
- Simple Equation: Equations are estimated for each group to test for differences in key parameters.

$$E[\Delta D_{\text{Intervention}}]$$
 vs. $E[\Delta D_{\text{Control}}]$

 When to Use: This is a powerful tool for randomized controlled trials (RCTs) to test if an intervention caused a different process of change compared to a control condition.

Summary of Model Differences

| Model Name | Primary Focus | Key Feature | Best Use Case |
|------------------|--------------------|------------------------|------------------------|
| Linear GCM | Overall Trajectory | Assumes a constant, | Testing hypotheses |
| | , , | linear rate of change. | about steady growth |
| | | | or decline. |
| Quadratic GCM | Overall Trajectory | Models one curve | When the rate of |
| | | in the trajectory | change is expected to |
| | | (acceleration/decel- | change over time. |
| | | eration). | |
| Conditional GCM | Explaining Differ- | Uses a time-invariant | Understanding |
| | ences | covariate to predict | *why* individuals or |
| | | intercept and/or | groups show different |
| | | slope. | growth patterns. |
| Latent Basis GCM | Exploring Shape | Freely estimates | When theory is |
| | | some slope loadings | unclear about the |
| | | to let the data define | shape of change, or |
| | | the shape. | for modeling inter- |
| | | · | ventions. |
| LCS Model | Dynamics of Change | Directly models the | Investigating the pro- |
| | | change between con- | cess of change itself |
| | | secutive time points. | and what influences |
| | | | transitions. |
| Multigroup LCS | Comparing Dynamics | Estimates separate | Testing if an inter- |
| | | LCS models for | vention or group |
| | | different groups. | membership al- |
| | | | ters the process of |
| | | | change. |

Next session

Topic: Causality

©Dr. Daniel Kristanto & Prof. Andrea Hildebrandt

 $daniel.kristanto@uni-oldenburg.de\ \&\ andrea.hildebrandt@uni-oldenburg.de$

