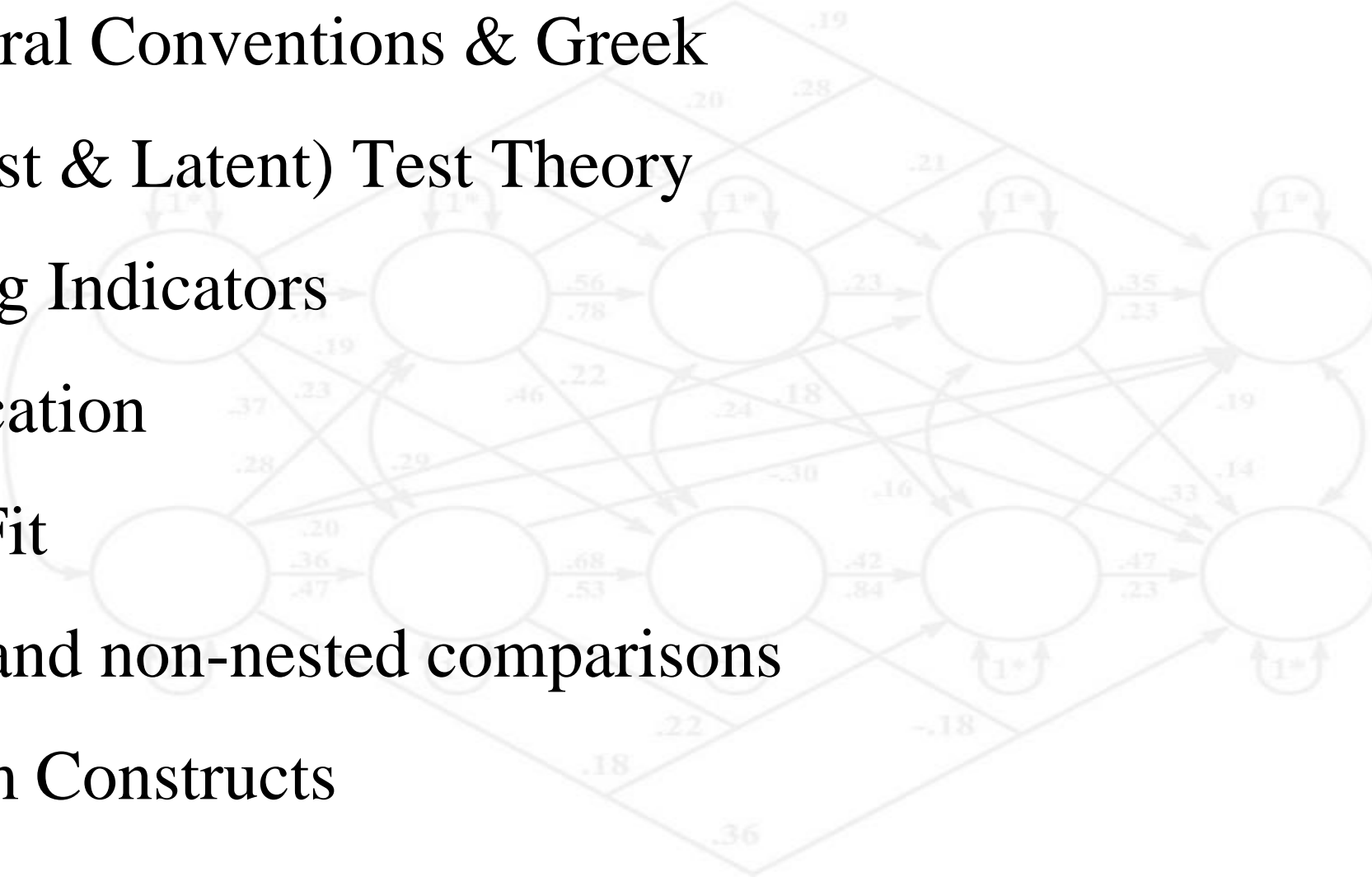


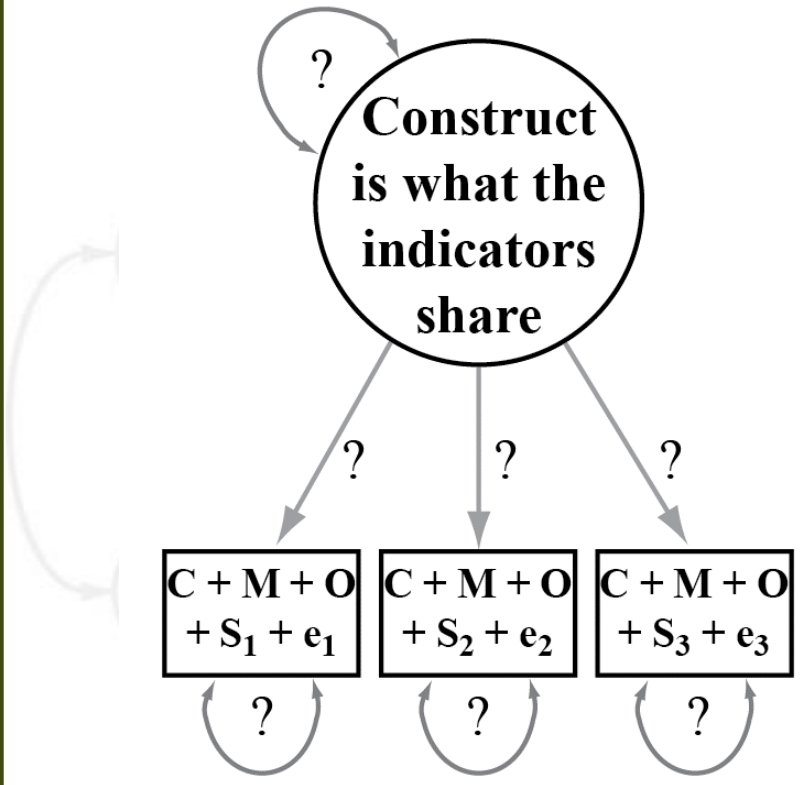
# An Overview of SEM Basics

- Configural Conventions & Greek
- (Manifest & Latent) Test Theory
- Selecting Indicators
- Identification
- Model Fit
- Nested and non-nested comparisons
- Phantom Constructs

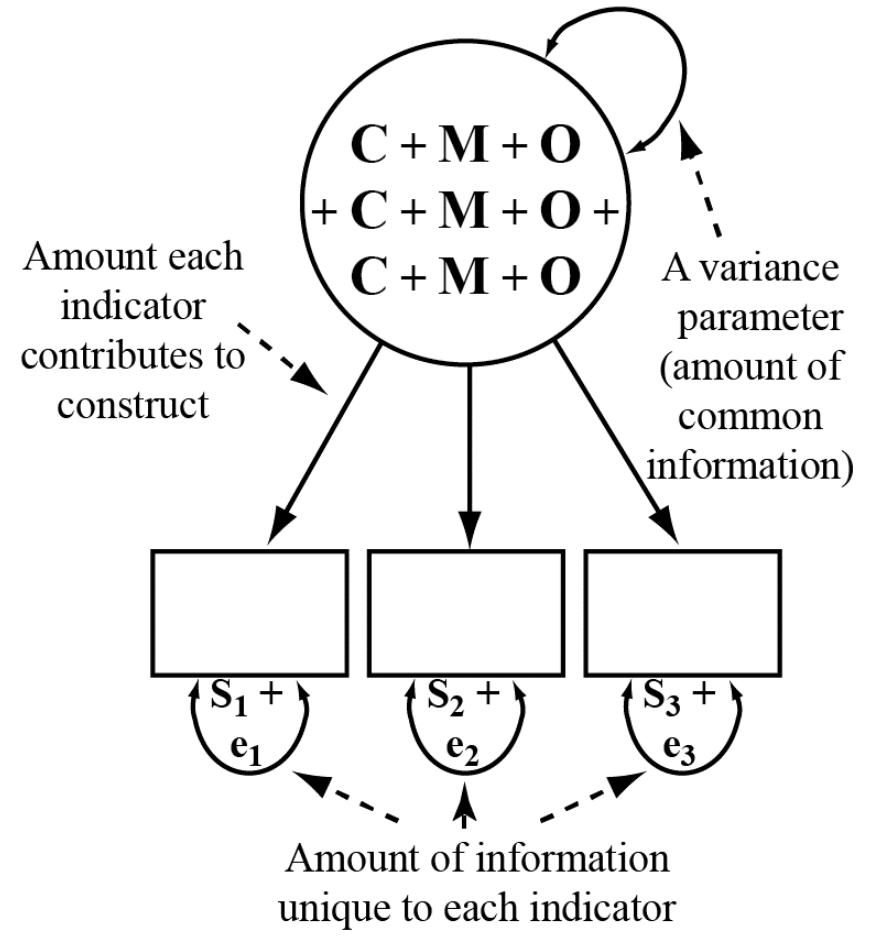


# A Construct is What the Indicators Share

A) Specification of a Construct (before estimation)

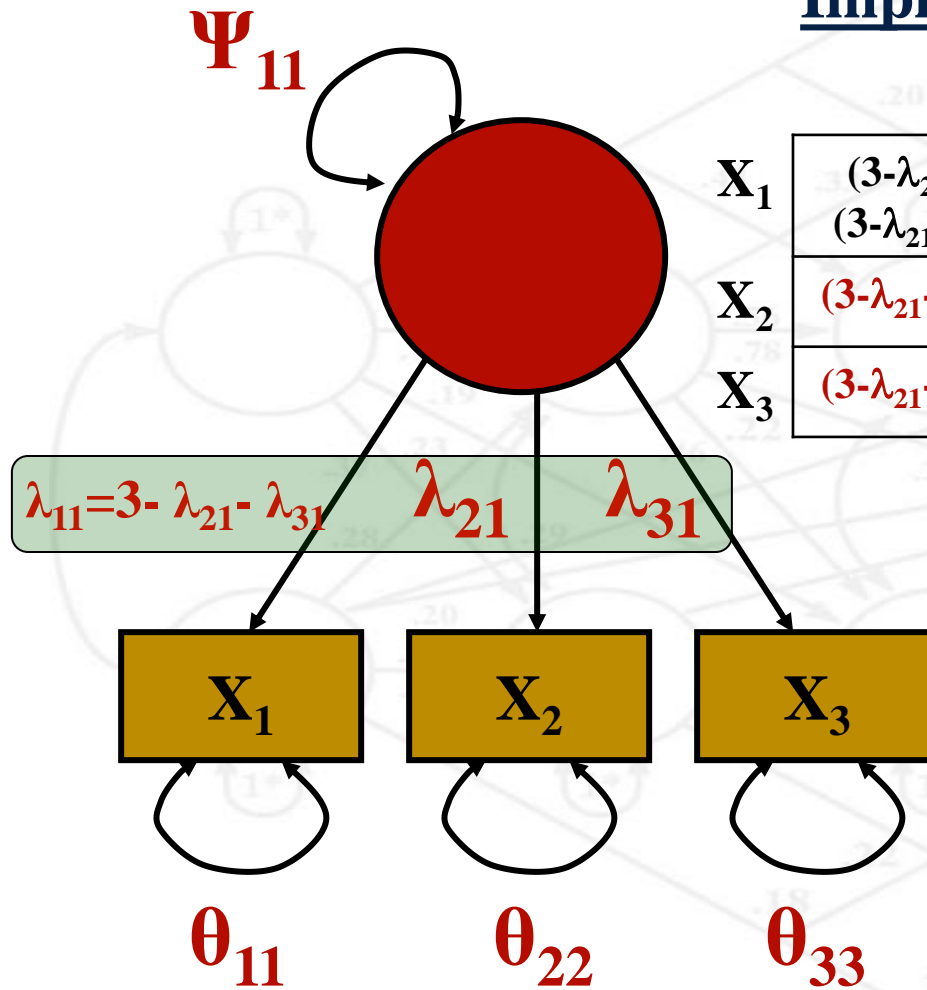


B) What's Shared among Indicators



# Effects Coding

## Implied variance/covariance matrix



	$X_1$	$X_2$	$X_3$
$X_1$	$(3 - \lambda_{21} - \lambda_{31}) \Psi_{11}$ $(3 - \lambda_{21} - \lambda_{31}) + \theta_{11}$		
$X_2$	$(3 - \lambda_{21} - \lambda_{31}) \Psi_{11} \lambda_{21}$	$\lambda_{21} \Psi_{11} \lambda_{21} + \theta_{22}$	
$X_3$	$(3 - \lambda_{21} - \lambda_{31}) \Psi_{11} \lambda_{31}$	$\lambda_{21} \Psi_{11} \lambda_{31}$	$\lambda_{31} \Psi_{11} \lambda_{31} + \theta_{33}$

Three methods of setting scale

3) Constrain loadings to average 1

# Evaluating Model Fit

**Evaluating model fit is a ‘gestalt’ process, not all indices need to meet ‘criteria’ to conclude you have an adequate model**

- **New areas of study with new measures may not yield ‘great’ fit at first until the measures are refined and the area of study is better understood.**
- **Check to see if the stem and leaf plot of the standardized residuals is approximately normal and if no modifications will improve model fit. In this situation, a ‘poor’ fitting model may be the best given newness of the study and measures.**

**Each fit index has ‘problems’:**

- **TLI/NNFI does not do well with small sample sizes**
- **RMSEA does not perform well in small models, especially in small models with very large sample sizes (the NCP is based on the chi-square)**
- **SRMR is good for 1-group 1-time-point models.**

# Building Parcels

- **Theory – Know thy S and the nature of your items**
- **Random assignment of items to parcels (e.g., fMRI)**
  - **Use Sterba’s calculator to find allocation variability when sampling error is high**
- **Balancing technique**
  - **Combine items with higher loadings with items having smaller loadings [Reverse serpentine pattern]**
- **Using a priori designs (e.g., CAMI)**
  - **Develop new tests or measures with parcels as the goal for use in research**

# Factorial Invariance

[a.k.a. Metric invariance, measurement equivalence]

Assumed when comparing groups, time points (ANOVA)

Establish equivalence of measurement (i.e., Are we measuring the same constructs in each group?)

Four levels of invariance:

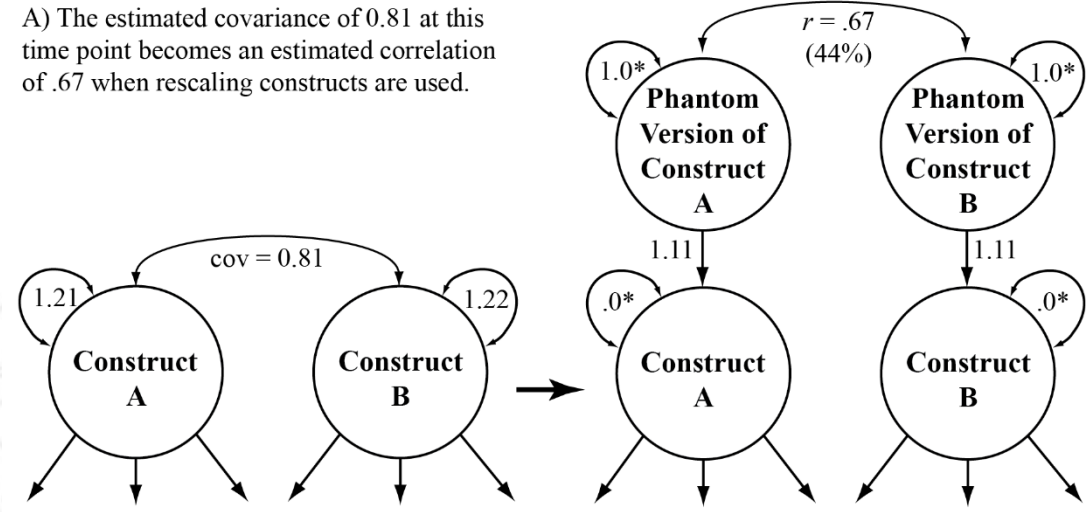
- 1) **Configural invariance:** Same pattern of fixed & free parameters
- 2) **Weak factorial invariance:** Corresponding factor loadings are equal across groups
- 3) **Strong factorial invariance:** Corresponding indicator means are equal across groups

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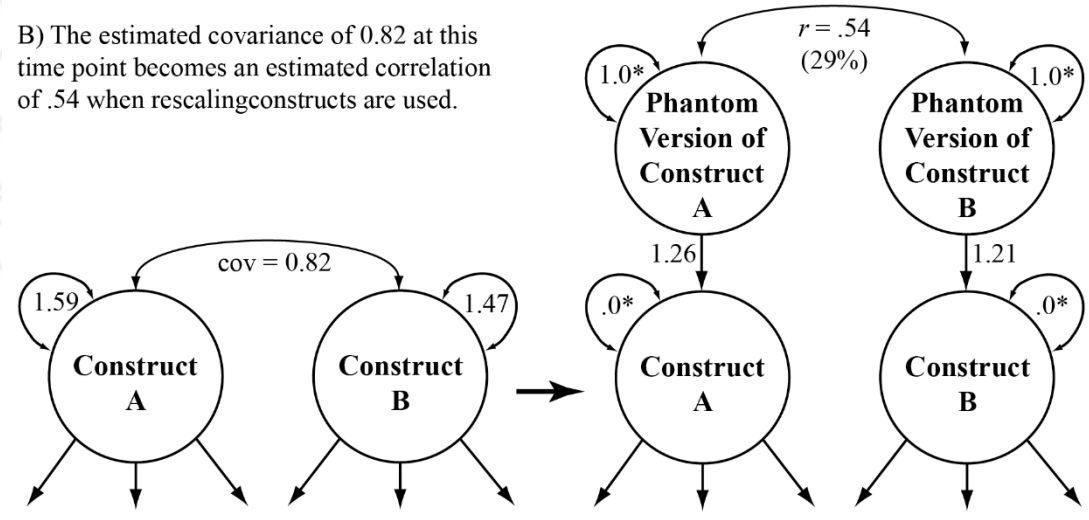
- 4) **Strict factorial invariance:** Corresponding indicator residuals are equal across groups  
(*this level is not recommended*)

# Phantom Constructs – Reason for Them

A) The estimated covariance of 0.81 at this time point becomes an estimated correlation of .67 when rescaling constructs are used.



B) The estimated covariance of 0.82 at this time point becomes an estimated correlation of .54 when rescaling constructs are used.



# Missing Data Key Considerations

- **Recoverability**
  - Is it possible to recover what the *sufficient statistics* would have been if there was no missing data?
    - (sufficient statistics = means, variances, and covariance)
  - Is it possible to recover what the *parameter estimates* of a model would have been if there was no missing data.
- **Bias**
  - Are the sufficient statistics/parameter estimates systematically different than what they would have been had there not been any missing data?
- **Power**
  - Do we have the same or similar rates of power (1-Type II error rate) as we would without missing data?