

Latent Variable Methods Workshop
Longitudinal Data Analysis Models

Tue Sep 2 - Thu Sep 4 2025

**Day 1 : Overview, Path Analysis, Factor
Analysis, Structural Equation Modeling, and
Effectively Using Mplus Software**

lvmworkshop.org

Brown University

Zoom

September 2, 2025

Day 1

- Orientation, overview and objectives
- Longitudinal data analysis (LDA)
- Structural equation modeling (SEM)
- Using Mplus & R/lavaan
- Lab exercise: CFA in Mplus & R/lavaan

Workshop Objectives

- ❶ Introduce LDA in LVM (Mplus, R/lavaan) framework
 - ▶ A range of questions addressed by LDA
 - ▶ How approached using Mplus, R/lavaan
- ❷ Experience applied LDA using Mplus, or R/lavaan
 - ▶ A range of models
 - ★ Latent growth curve
 - ★ latent change score models
 - ★ survival analysis (Mplus only)
 - ★ Mixture models (Mplus only)

Workshop Structure

- Morning Sessions

- ▶ Lecture
- ▶ Lab sessions
- ▶ Run through examples
- ▶ Ask questions at any time

- Afternoon Sessions

- ▶ More lecture
- ▶ More time to run through examples
- ▶ Or with your own data
- ▶ Individual assistance is the goal, emphasis on provided examples

Content Covered - Day 1

• Introduction to Longitudinal Data Analysis

- ▶ Overview
- ▶ Data Management for LDA
- ▶ Introduction to Structural Equation Modeling (with Mplus)
- ▶ Nuts and Bolts of Using Mplus
 - ★ How to communicate about models
 - ★ How to get data into Mplus
 - ★ The Statistical model in general (arrays)

Content Covered - Day 2

- Example data set
- Latent Growth Curve Models (LGCM)
 - ▶ Unconditional LGCM
 - ▶ Conditional LGCM
 - ▶ Non-Linear LGCM
- Random Effects Models for growth with Mplus
- Multilevel model approach for growth modeling with Mplus
- Growth Mixture Models
- Multiple indicator LGCM
- Modeling Retest, Practice and other methods artifacts

Content Covered - Day 3

- Advanced Topics

- ▶ LCGM

- ★ Multiple Group Models
 - ★ Randomized Controlled Trials
 - ★ Bivariate, Parallel, and Sequential Process Models

- ▶ Latent Change Score (Dual Change Score Models)

- ★ Dual Change Score Models (DSCM), Bivariate DSCM

- ▶ Bayesian Data Analysis and LGCM

- ▶ Survival Analysis with Mplus

- ★ Discrete time, Continuous time
 - ★ Simultaneous survival and growth modeling

- ▶ Modeling with complex sample weights

Where to go for more information

- see links at
- https://www.statmodel.com/courses_openenroll.shtml
- The Center for Statistical Training by Curran-Bauer Analytics
<https://centerstat.org/>
- <https://www.lvmworkshop.org/home/--deprecated-pages/usefullinks>

Other Resources

- What is longitudinal data analysis?
 - ▶ Singer JD & Willett JB. Applied longitudinal data analysis: Modeling change and event occurrence. 2003, New York: Oxford University Press.
- How do I do latent growth curve modeling?
 - ▶ Duncan TE, Duncan SC, & Strycker LA. An introduction to latent variable growth curve modeling: concepts, issues and applications. Second ed. 2006, Mahwah, New Jersey: LEA, Inc.
 - ▶ Newsom, J., Jones, R., & Hofer, S. (Eds.). (2011). Longitudinal Data Analysis: A Practical Guide for Researchers in Aging, Health and Social Sciences. New York: Routledge.
- Tell me more about the math behind latent curve methods
 - ▶ Bollen KA & Curran PJ. Latent curve models: a structural equation perspective. Wiley series in probability and statistics. 2006, Hoboken, N.J.: Wiley-Interscience.
- More about structural equation modeling
 - ▶ Bryne, B. (2011). Structural Equation Modeling with Mplus: Basic Concepts, Applications and Programming. New York: Routledge.

Other Key References

- McArdle J. Five steps in latent curve modeling with longitudinal life-span data. In: Levy R, Ghisletta P, Le Gorr J, Spini D, Widmer E, eds. Towards an Interdisciplinary Perspective on the Life Course (10). San Diego, CA: JAI Press; 2005
- McArdle J, Epstein D. Latent growth curves within developmental structural equation models. *Child Development*. 1987;58:110-133.
- McArdle JJ. Latent variable modeling of differences and changes with longitudinal data. *Annual review of psychology*. 2009;60:577-605.
- Curran, PJ, & Muthén, BO. Testing developmental theories in intervention research: Latent growth analysis and power estimation. *American Journal of Community Psychology*. 1999; 27:567-595.
- Jung T, & Wickrama, KAS. An introduction to latent class growth analysis and growth mixture modeling. *Social and Personality Psychology Compass*. 2008;2:302-317.

Introduction

Longitudinal Data Analysis

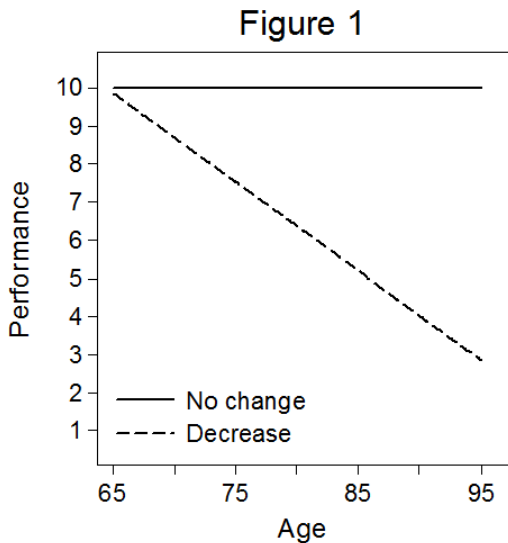
This Workshop

- Applied statistical data analysis of longitudinal data
 - ▶ Latent growth curve (LGC) or latent growth model (LGM)
 - ▶ Survival models
 - ▶ Latent change score models
 - ▶ Mixture models
 - ★ growth mixture
 - ★ change score mixture

The Motivating Questions

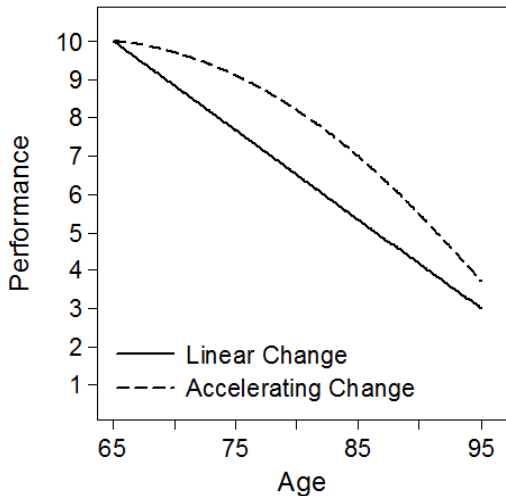
- How does Y change over time? Y could be
 - ▶ cognitive functioning, neuropsych performance
 - ▶ mood, depression
 - ▶ weight
 - ▶ number of falls
 - ▶ etc.
- What factors predict who changes slower or faster on Y

About Change



About Change

Figure 2



The M in LGM

Why modeling is important

Change Scores

Since life-span research is, in essence, concerned with change over time, it is tempting to assume that the basic psychometric datum with which one works is the change score. Thus, if one is making comparisons of intelligence test scores at two points in time (symbolized by X_1 and X_2), it is assumed that the interest is in $(X_2 - X_1)$ and that such scores should be computed as one of the first steps in performing analyses. Actually, both the history of the problem and the logic of investigation indicate that the last thing one wants to do is to think in terms of or compute such change scores unless the problem makes it absolutely necessary. (p87, emphasis added)

Nunnally, J. C. (1973). Research Strategies and Measurement Methods for Investigating Human Development. In J. R. Nesselroade & H. W. Reese (Eds.), Life-Span Developmental Psychology: Methodological Issues (pp. 87-88). New York: Academic Press.

The Problems with $Y_t - Y_{(t-1)}$

- Classical
 - ▶ Reliability of Change Scores
 - ▶ Regression to the Mean
 - ▶ "Law" of Initial Values
- Empirical
 - ▶ $\text{VAR}(\text{observed change}) > \text{VAR}(\text{true change})$
 - ▶ Floor/Ceiling Effects
 - ▶ Limited information / assumption of linear change

Change Scores

- Difference between current and prior state
- Easy to interpret
- Easy to communicate meaning
- Really only useful in completely randomized designs when comparison and control groups are (or can be assumed to be) equivalent at baseline
- ...and even still may not be most powerful approach

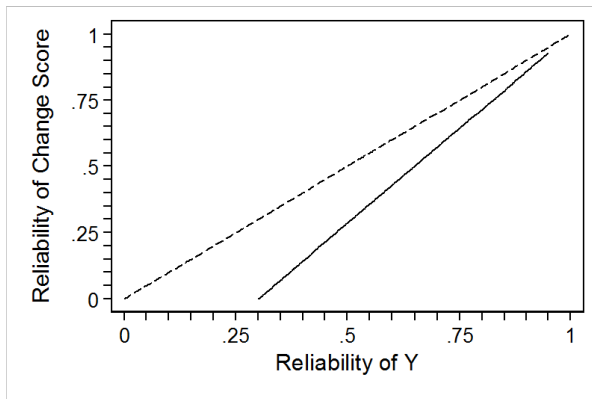
(Un)Reliability of Change Scores

- The first line of criticism of change scores
 - ▶ Cronbach, L., & Furby, L. (1970). How we should measure "change" - or should we? *Psych Bull* 74; 68-80.
 - ▶ Lord FM. The measurement of growth. *Educ Psychol Meas* 1956;16:421-437.
 - ▶ Lord FM. Further problems in the measurement of growth. *Educ Psychol Meas* 1958;18:437-454.
- Aging research example
 - ▶ van Belle, G., Uhlmann, R. F., Hughes, J. P., & Larson, E. B. (1990). Reliability of estimates of changes in mental status test performance in senile dementia of the Alzheimer type. *J Clin Epi*, 43(6), 589-95.
- See Also
 - ▶ Galton, F. (1886). Regression towards mediocrity in hereditary stature. *Journal of the Anthropological Institute*, 15, 246-263.

(Un)Reliability of Change Scores

- Premise is not universally accepted.
 - ▶ Rogosa D: Myths about longitudinal research. In Schaie K, Cambell R, Meredith W, Rawlings S (eds), Methodological issues in aging research. New York, Springer Publishing Company, 1988, 171-209
 - ▶ Nesselroade JR: Application of multivariate strategies to problems of measuring and structuring long-term change. In Goulet LR, Baltes PB (eds), Life-span developmental psychology: Research and theory. New York, Academic Press, 1970, 193-207
- Detailed review
 - ▶ Wittmann, W. (1997). The reliability of change scores: many misinterpretations of Lord and Cronbach by many others; revisiting some basics for longitudinal research. Methodology conference: Evaluation of Change in Longitudinal Data, Nürnberg.
<http://www.psychologie.uni-mannheim.de/psycho2/publi/papers/reliab.pdf>

(Un)Reliability of Change Scores



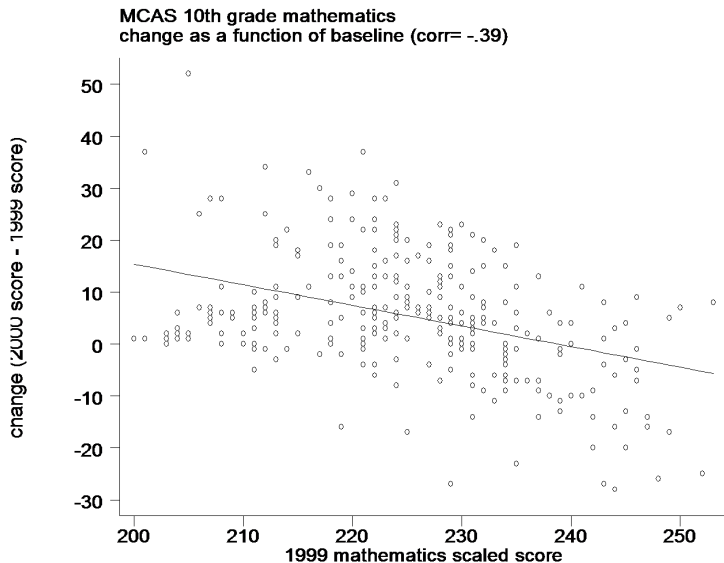
$$r_{\Delta} = \frac{r_{Y(t)} + r_{Y(t-1)} - 2 \times r_{Y(t), Y(t-1)}}{2 - 2 \times r_{Y(t), Y(t-1)}}$$

$r_{YY} = \text{reliability} = 0.80$; $r_{Y(t), Y(t-1)} = 0.3$

Streiner DL & Norman GR (1995). *Health Measurement scales*. New York. Oxford.

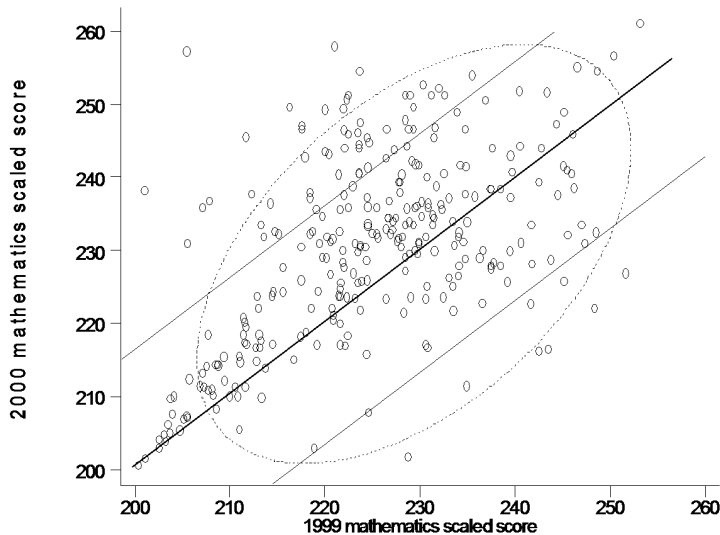
What Is Regression to the Mean?

- The only thing consistent about regression to the mean is *the absence of explicit, defensible definitions of the phenomenon* (p186).
- Rogosa, D. (1988). Myths about longitudinal research. In K. Schaie, R. Campbell, W. Meredith, & S. Rawlings (Eds.), *Methodological issues in aging research* (pp. 171-209). New York: Springer Publishing Company.



$$COR(y_2, y_1) = 0.54; COR(y_2 - y_1, y_1) = -0.39$$

MCAS results, regular students
correlation of mathematics scores 1999-2000 = 0.54



$$COR(y_2, y_1) = 0.54; COR(y_2 - y_1, y_1) = -0.39$$

Some top-scoring schools faulted

Many questioning MCAS assessment

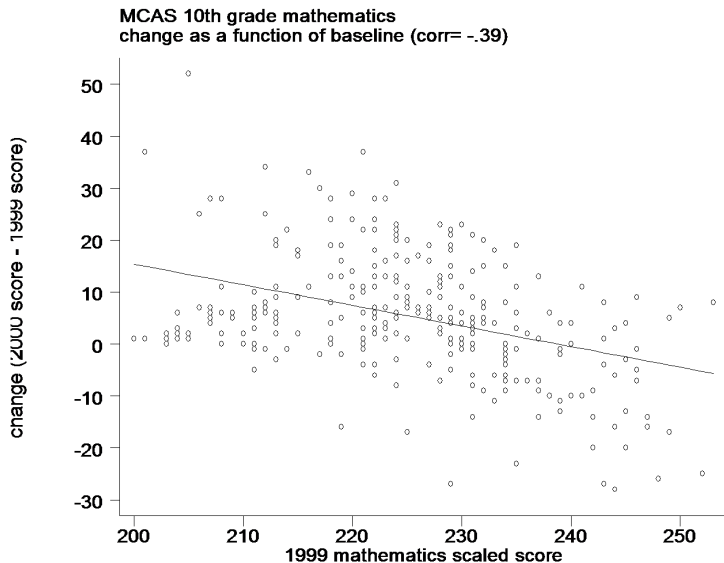
By Anand Vaishnav, Globe Staff, 1/10/2001

A state report rating public schools on three years of MCAS results shows some of Massachusetts's highest-performing schools "failed to meet" expectations, a description that angered many educators across the Commonwealth.

In the state's first attempt to give schools targets to reach on the Massachusetts Comprehensive Assessment System exams, 56 percent of 1,539 schools did not improve enough to meet those goals. The remaining 44 percent "exceeded, met, or approached" their growth targets, according to the report released yesterday.

The results largely bode well for urban districts that had the most ground to make up. But the state's top public schools - in wealthier districts such as Harvard, Newton, Wellesley - performed so well in 1998 that boosting scores any higher was nearly impossible. As a result, many of those schools received an "F," for "failed to meet" expectations.

[Boston Globe.](#)



$$COR(y_2, y_1) = 0.54; COR(y_2 - y_1, y_1) = -0.39$$

$$\text{VAR}(\text{observed change}) > \text{VAR}(\text{true change})$$

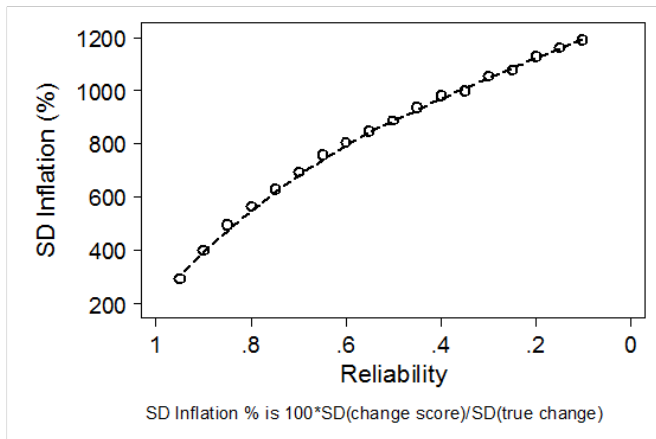
$$Y = T + e$$

$$Y_2 - Y_1 = (T_2 - T_1) + (e_2 - e_1)$$

$$\begin{aligned}\text{VAR}(Y_2 - Y_1) &= \text{VAR}(Y_2) + \text{VAR}(Y_1) - 2\text{COV}(Y_2, Y_1) \\ &= \text{VAR}(T_2) + \text{VAR}(T_1) - 2\text{COV}(T_2, T_1) + \\ &\quad \text{VAR}(e_2) + \text{VAR}(e_1) - 2\text{COV}(e_2, e_1)\end{aligned}$$

- Even if errors are uncorrelated, the variance of observed change score compounds measurement errors
- The lower the $\text{COR}(Y_t, Y_{t-1})$, the more variable is the change score
- Observed change score variance can be greater than baseline variance

$$\text{VAR}(\text{observed change}) > \text{VAR}(\text{true change})$$

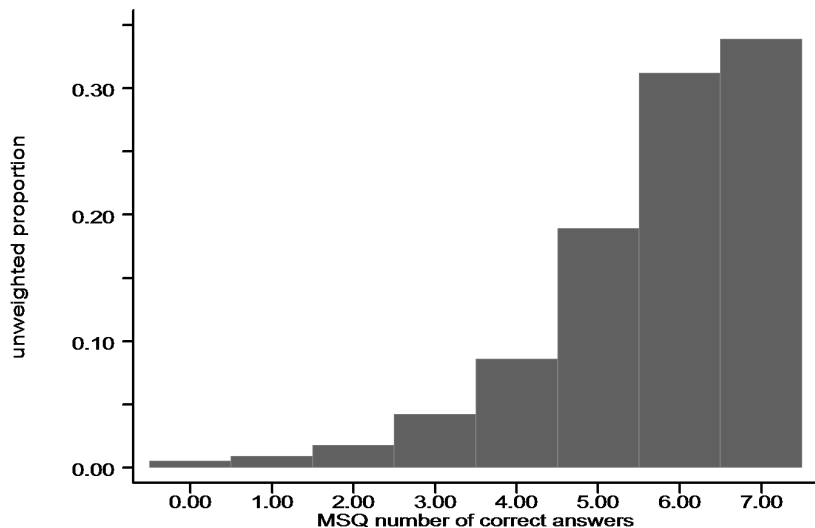


Change is random, additive, and correlated with baseline ($r = -0.40$)
 $\text{SD}(\text{true change}) = 0.1\text{SD}(\text{baseline})$

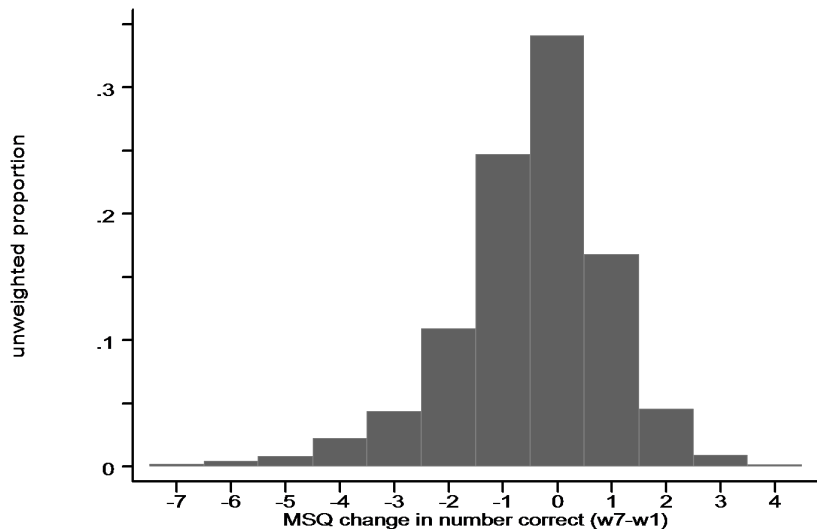
Floor/Ceiling Effect

- Many measures in health research suffer floor/ceiling effects: a large proportion of the sample score at lowest or highest possible level
- For Example
 - ▶ Cognitive function
 - ▶ Depressive symptoms
 - ▶ Physical functioning

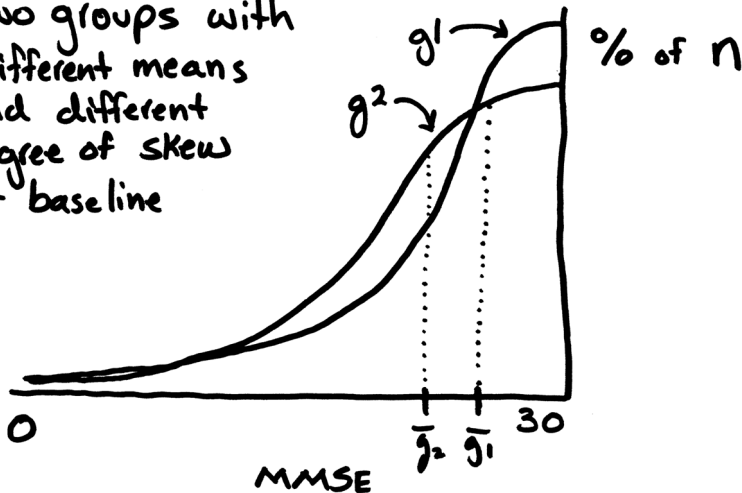
Distribution of Cognitive Functioning in EPESE



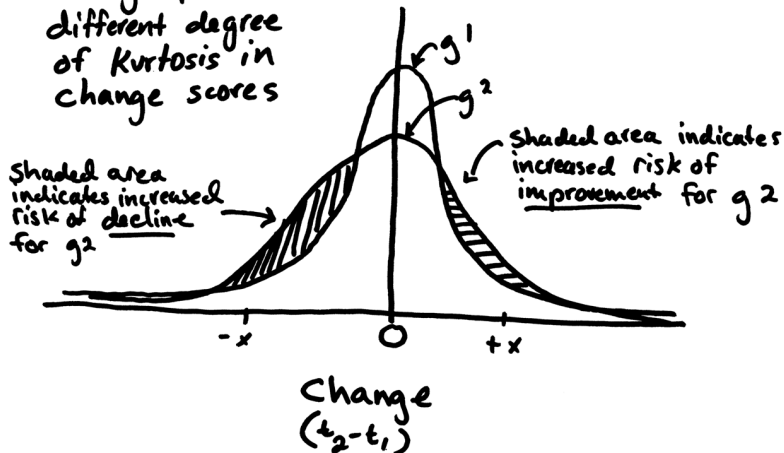
Distribution of Cognitive Functioning (w7-w1) in EPESE

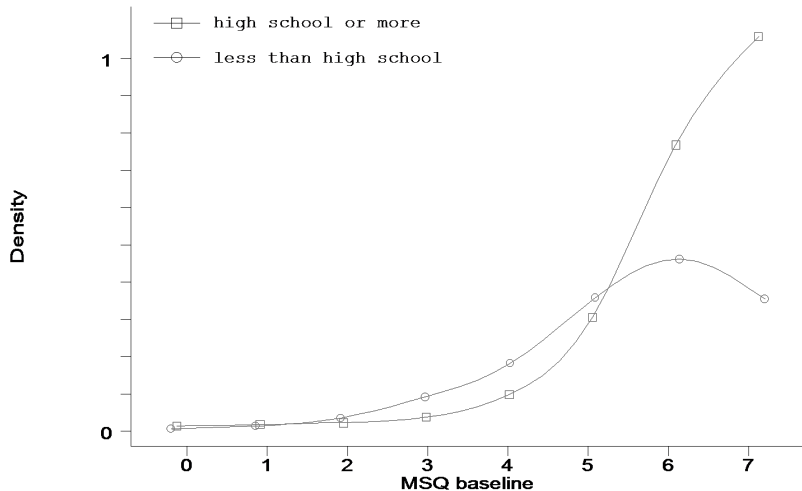


Two groups with
different means
and different
degree of skew
at baseline

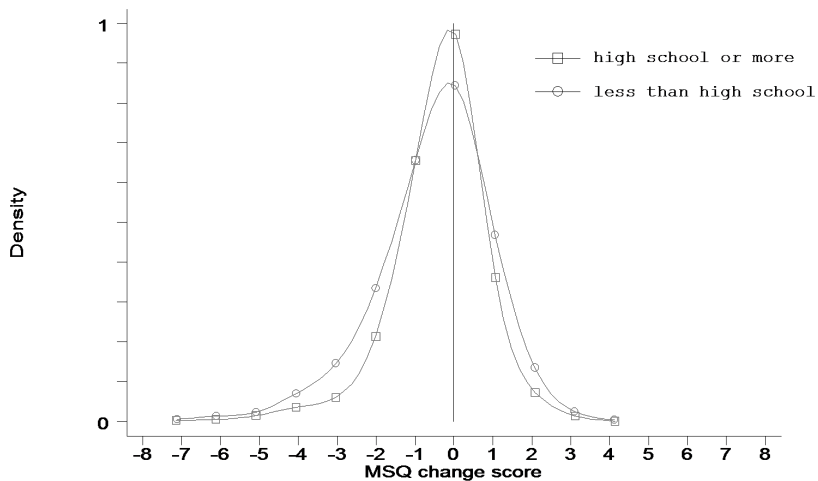


Two groups with
different degree
of Kurtosis in
change scores





Density plot. MSQ scores at year 1. EPESE (N=13,697) by education group

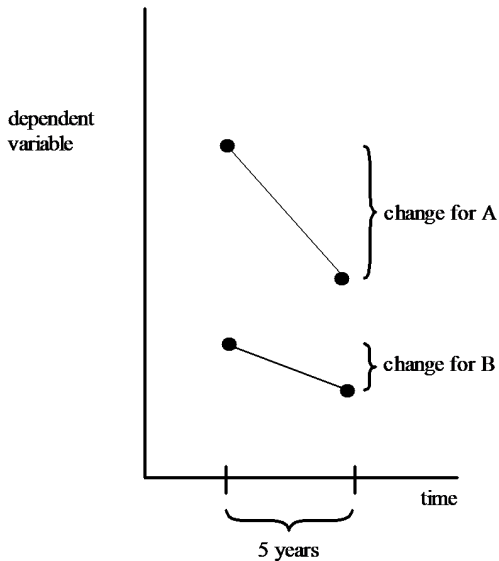


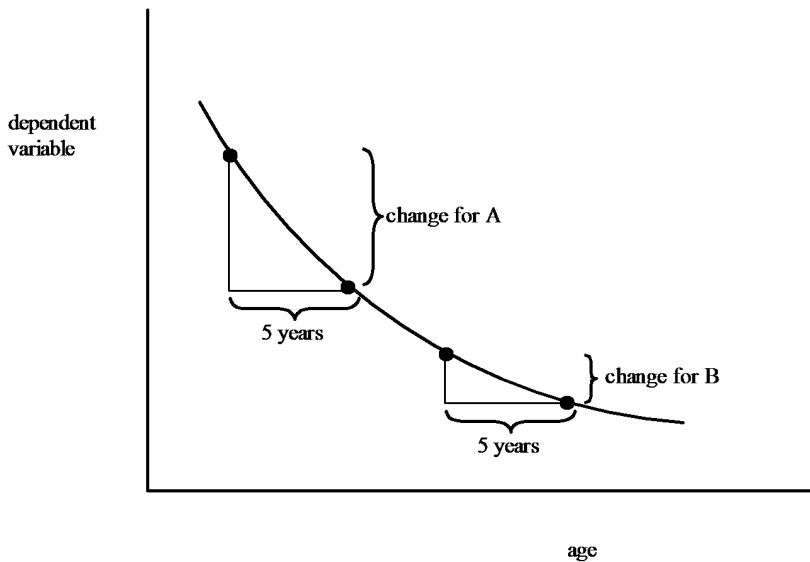
Density plot. MSQ change scores year 7 - year 1. EPESE (N=7,939) by education group

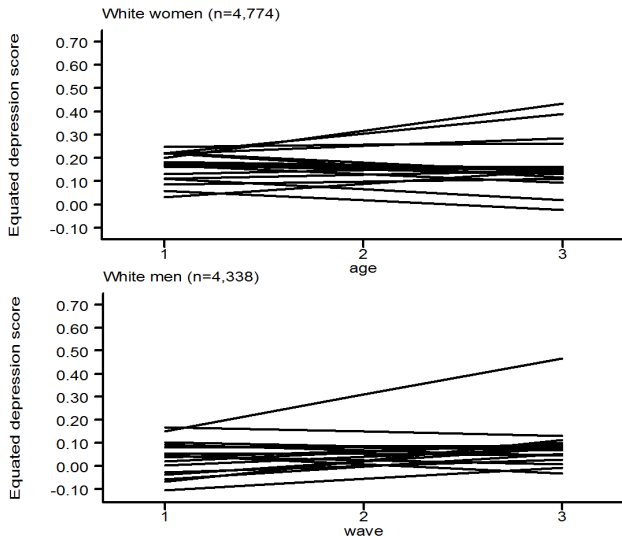
Table. Cognitive decline (2 or more points lower at follow-up) and cognitive improvement (2 or more points higher at 6th annual follow-up) among EPESE participants, by level of educational attainment.

Level of Education (years)	%	OR
Cognitive Decline		
less than 12	22	1.7 (1.5, 2.0)
12 or more	14	1.0
Cognitive Improvement		
less than 12	6	1.7 (1.4, 2.1)
12 or more	4	1.0

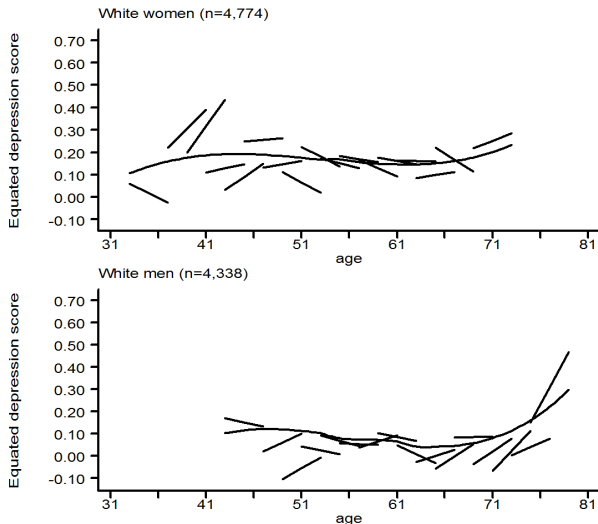
Assumption of Linear Change







Birth cohort-specific depression symptom trajectories, Health and Retirement Study waves 1,2,3



Birth cohort-specific depression symptom trajectories, Health and Retirement Study waves 1,2,3

An Approach

Model, Rather than Calculate, Change

Latent Growth Curve Models

- Appropriate when outcome
 - ▶ changes systematically over time, on a time interval relevant to the study
 - ▶ when there at least three, preferably four repeated observations
 - ▶ is discrete or continuous
- Alternatives
 - ▶ Survival models (for "absorbing events": events that can only be observed once)
 - ▶ Latent change score models
 - ▶ Autoregressive models

Latent Growth Curve Models

- Sometimes equivalent to
 - ▶ mixed effect models
 - ▶ random coefficient regression models
 - ▶ hierarchical or multilevel models
 - ▶ HLM, Mx, PROC Mixed, xtmixed
- Can be different
 - ▶ how time is handled (data vs. model parameter)
 - ▶ how variances are handled
 - ▶ how flexible the modeling framework is

Figure 3a

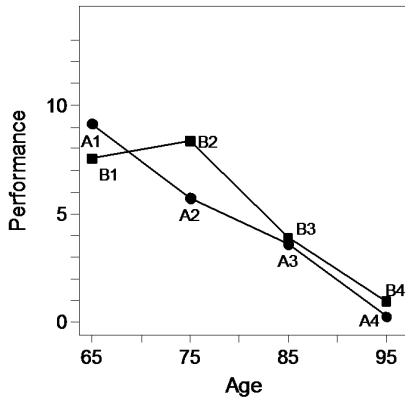
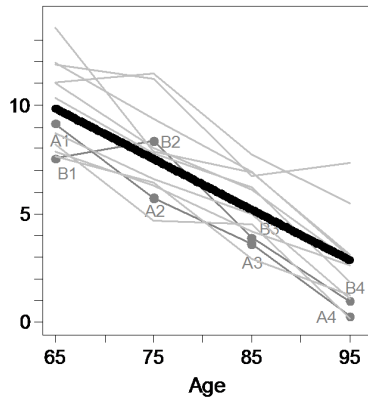


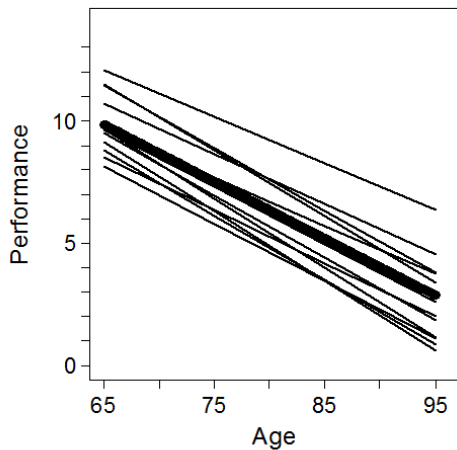
Figure 3b

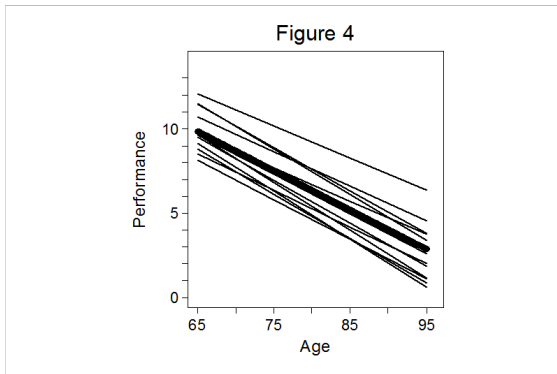


Observed scores and an overall sample mean trajectory

Panel 3a shows two persons (A and B) and their performance at four ages (A1=person A at age 65, A4=person A at age 95). Panel B shows how person A and B and their performance trajectory are distributed in a larger sample of persons. Panel 3b also shows a group average trajectory (heavy line).

Figure 4

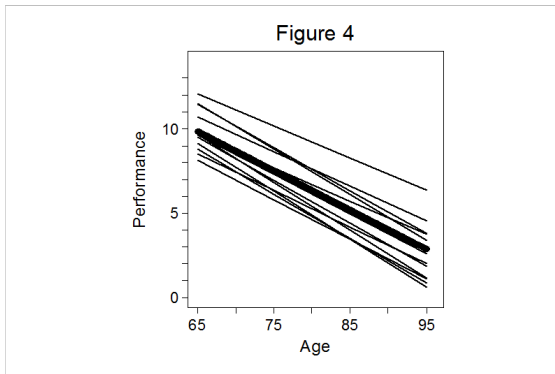




$$\text{Performance} = \text{Intercept} + \text{Slope} \times \text{Age} + \text{Residual}$$

$$\text{Intercept} = \text{Mean}_{65} + \text{Disturbance}$$

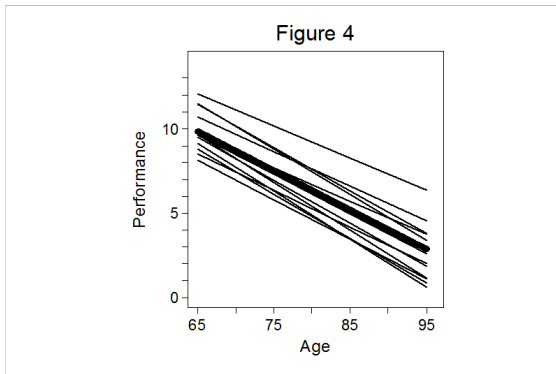
$$\text{Slope} = \text{Mean Change} + \text{Disturbance}$$



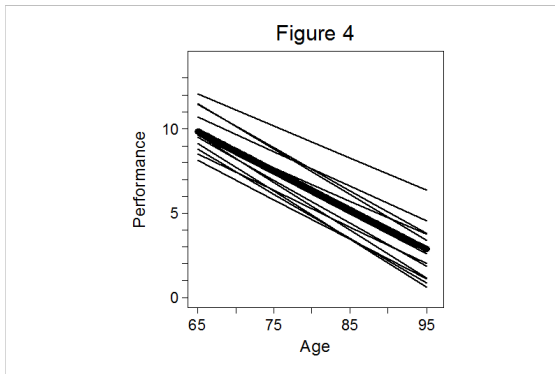
$$y_{it} = \eta_{i1} + \eta_{i2} \times \text{Age}_{it} + \epsilon_{it}$$

$$\eta_{i1} = \alpha_1 + \zeta_{i1}$$

$$\eta_{i2} = \alpha_2 + \zeta_{i2}$$



$$y_{it} = (\alpha_1 + \zeta_{i1}) + (\alpha_2 + \zeta_{i2}) \times \text{Age}_{it} + \epsilon_{it}$$



$$y_{it} = (\alpha_1 + \alpha_2 \times \text{Age}_{it}) + (\zeta_{i1} + \zeta_{i2} \times \text{Age}_{it} + \epsilon_{it})$$

$$y_{it} = (\text{fixed part}) + (\text{random part})$$

Time

- Time as data (Age_{it})
 - ▶ Random effects models, Mixed effect models, Multilevel/hierarchical models
 - ▶ Accommodates irregular assessment schedule
 - ▶ Time varying covariate effects vary over people
- Time as model parameter (Age_t)
 - ▶ LGCM
 - ▶ Assumes regular reassessments
 - ★ not necessarily evenly spaced
 - ▶ Allows flexible curve shape
 - ▶ Time varying covariate effects vary over time
- Mplus will do both

Summary

- Change scores have issues
- Modeling doesn't remove those issues but can accommodate (estimate) them
- There are many paths to enlightenment (hierarchical, multilevel, mixed, random, latent growth)
- Different approaches can be equivalent, mostly depending upon how time is handled

Don't get too hung up on the validity/invalidity of one model over another

- LGCM $\approx$ Mixed/Random Hierarchical
 - ▶ Curran PJ: Have Multilevel Models Been Structural Equation Models All Along? Multivariate Behavioral Research 38:529-569, 2003
- LGCM $\approx$ paired t-tests, RM-ANOVA
 - ▶ Voelkle MC: Latent growth curve modeling as an integrative approach to the analysis of change. Psychology Science 49:375, 2007
- But, mixed, random, hierarchical, change score, repeated measures ANOVA, all can be viewed as special cases of SEM models of change
- So SEM model is a flexible, extensible and general approach to change and worth mastering.

Questions

Structural Equation Modeling

Communicating and expressing modeling ideas
using latent variables

Learning Note

- Gist
 - ▶ What am I doing?
 - ▶ How do I explain it to my colleagues?
- Statistical concepts
 - ▶ What am I really doing?
 - ▶ How do I try to explain it to my statistician colleagues?
- Person-machine interface
 - ▶ How do I do this?

Covariance Structure Model

Regressions among dependent (or involving latent variables)	(Continuous) Latent Variables	
	Yes	No
Yes	Structural Equation Modeling	Path Analysis
No	Factor Analy- sis	Multivariate Regression

How to Learn about SEM

How

- Brown T. Confirmatory factor analysis for applied research. (2006). Guilford Press
- Jöreskog and Sörbom, LISREL user's reference guide. (from version 7)
- Raykov and Marcoulides, A first course in structural equation modeling. 2006
- Raykov T, Tomer A, Nesselroade JR. Reporting structural equation modeling results in Psychology and Aging: Some proposed guidelines. Psychol Aging. 1991;6:499-503.
- Resources to help you use Mplus. UCLA: Institute for Digital Research & Education (was: Academic Technology Services, Statistical Consulting Group).
<https://stats.idre.ucla.edu/mplus/>

Why

- De Stavola B et al. Statistical issues in life course epidemiology. Am J Epidemiol. 2006;163:84-96.
- Hoyle RH, Smith GT. Formulating clinical research hypotheses as structural equation models. J Consult Clin Psychol. 1994;62:429-440.
- Muthén B. Psychometric modeling in epidemiology with an emphasis on longitudinal analysis of alcohol use. Alcoholic Beverage Medical Research Foundation. 1994.

Latent Variable Models

Latent variables are variables that are not directly observed

- Presumed to exist
- Presence inferred from observed variables
- Account for covariation among observed variables

Latent and observed variables can be continuous or categorical

Ridiculously Brief and Highly Selected History of LV Models

- Spearman (1904) - Factor Analysis
- Wright (1934) - Path Analysis
- Lazarsfeld (1950) - Latent Class
- Lord (1952) - Item Response Theory
- Jöreskog (1973) - LISREL model
- Muthén (1987) - LISCOMP model
- Muthén (1998) - General LVM (Mplus)
- Asparouhov + Muthén (2008) - Exploratory SEM
- Rosseel (2010) - R/lavaan

Other Latent Variables

- Measurement errors
- Random effects
- Frailties, liabilities
- Variance components
- Missing data

see [www.statmodel.com/download/Topic 3.pdf](http://www.statmodel.com/download/Topic%203.pdf)

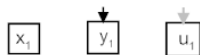
Path diagrams

Path analysis was invented by the geneticist Sewall Wright in 1918. Wright also made use of path diagrams. However, the diagrams used here follow a set of rules [loosely] for drawing path diagrams that has been described as the LISREL model (specifically, LISREL submodel 2). LISREL was developed in the 1960s and released widely in the 1970s by Karl G. Joreskog.

Challenges with path diagrams in LDA is that conventions for describing means and thresholds of observed (and latent) variables are not well developed. This owes to the historical fact that most early applications of covariance structure analysis (path analysis, factor analysis) concern the covariances among observed data, and variables were often standardized or centered (to mean 0).

The most widely used convention for describing threshold and mean structures is McArdle's RAM notation (Reticular Action Model).

Finally, it is worth mentioning that no specific knowledge of or ability to use and interpret path diagrams is required for using Mplus.



Observed variables in rectangles

Exogenous variables without residuals: covariate, cause

Endogenous variables (y, u) usually with residuals (errors): item, indicator, outcome

Sometimes y implies continuous, u implies discrete



Latent variables in ovals

Sometimes f and η imply continuous latent variables, c categorical

Continuous latent variables with disturbance terms

factors, random effects, classes, clusters, mixtures



Constant (=1)

Used to indicate means, thresholds, intercepts in regression

relationships. Several other conventions are used for this purpose.



Straight arrows indicate regression relationships including regression and factor loadings. Sometimes labeled with a regression coefficient ($b, \beta, \gamma, \kappa, \lambda, *$; * implies freely estimated). If not labeled and others in model are, might be constrained to be equal to 1. If labeled, subscripts in the order to-from (b_{yx}).



Curved arrows indicate covariance (correlation) relationships.

Often omitted from among exogenous variables, even when included in the model.

Notes:

There is no formal and universally accepted language or syntax rules for path diagrams.

Different authors have their own conventions.

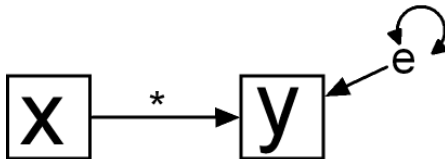
Path diagrams in publications are rarely complete images of the statistical model.

Path diagrams do not always translate directly into the syntax of an estimation program.

Path diagrams can be helpful tools. They can also be confusing and distracting.

Path diagrams as used in SEM are in general not the same as directed acyclic graphs (DAGs).

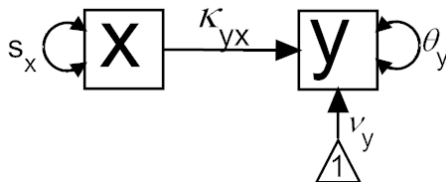
Simple Regression



$$y = \nu + kx + \epsilon$$

$$E(y) = \nu + kx$$

Simple Regression

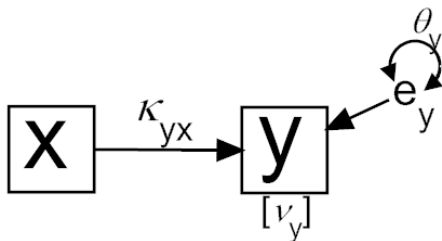


$$y = \nu + kx + \epsilon$$

$$E(y) = \nu + kx$$

RAM path diagram conventions described (James Steiger, Vanderbilt University)

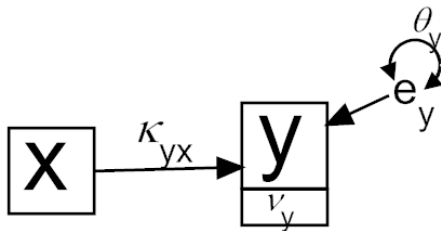
Simple Regression



$$y = \nu + kx + \epsilon$$

$$E(y) = \nu + kx$$

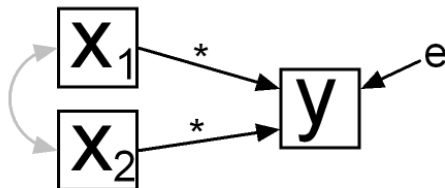
Simple Regression



$$y = \nu + kx + \epsilon$$

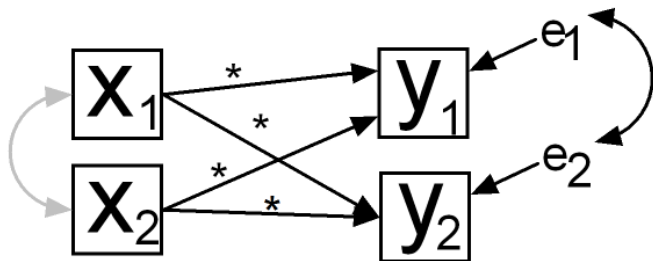
$$E(y) = \nu + kx$$

Multivariable Regression



$$y = \nu + k_1x_1 + k_2x_2 + \epsilon$$

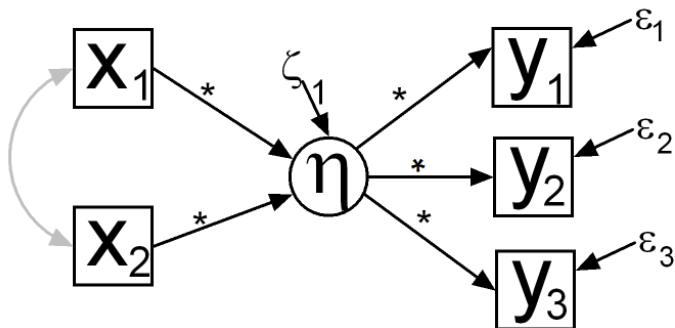
Multivariate Regression



$$y = Kx + \epsilon$$

(Confirmatory) Factor Analysis with Covariates

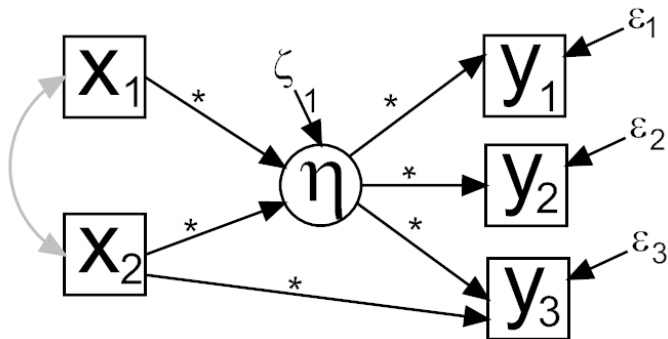
(MIMIC = Multiple Indicators Multiple Causes)



$$y = \nu + \Lambda\eta + Kx + \epsilon$$

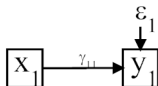
$$\eta = \alpha + \Gamma x + \zeta$$

MIMIC Model with Direct Effect

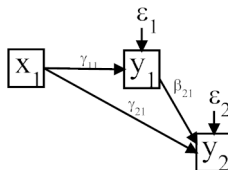


$$y = \nu + \Lambda\eta + Kx + \epsilon$$

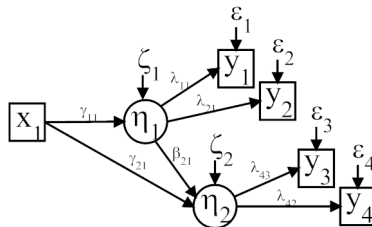
$$\eta = \alpha + \Gamma x + \zeta$$



regression

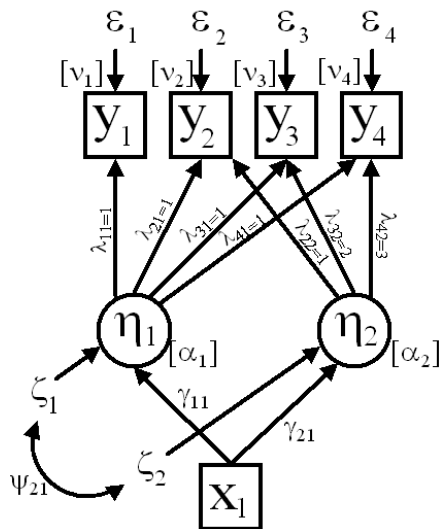


path analysis



Structural Equation Model

Latent Growth Model (Linear Change)

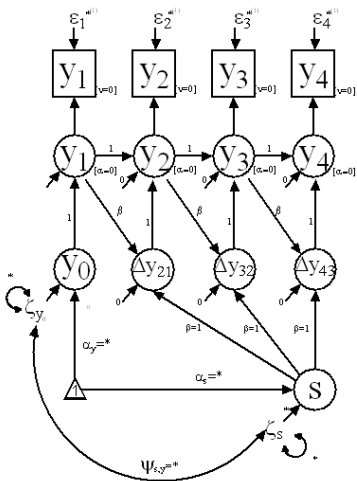


$$y_{it} = \nu + \eta_1 \lambda_{t1} + \eta_2 \lambda_{t2} + \epsilon_{it}$$

$$\eta_1 = \alpha_1 + \gamma_{11} x_{1i} + \zeta_1$$

$$\eta_2 = \alpha_2 + \gamma_{21} x_{1i} + \zeta_2$$

Dual Change Score Model



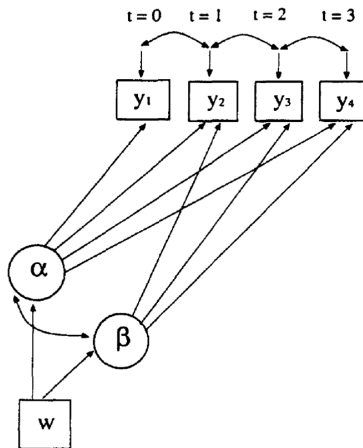
$$y_{it} = y_0 + \Delta y_{t(t-1)} + \epsilon$$

$$\Delta y_{t(t-1)} = \beta y_{(t-1)} + s$$

$$y_0 = \alpha_y + \zeta_{y_0}$$

$$s = \alpha_s + \zeta_s$$

FIGURE 3
Graphical representation of a growth model for four time points.



Muthén BO, Khoo ST. Longitudinal studies of achievement growth using latent variable modeling. *Learning and individual differences*. 1998;10(2):73-101.

The Basic Structure of Growth Models 455

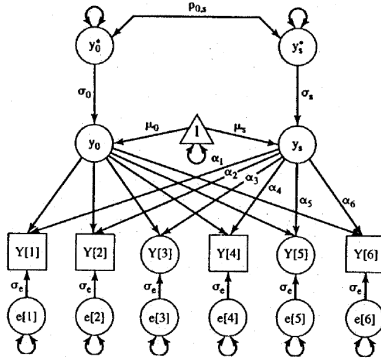


Figure 18.7 The basic latent growth structural model as a path diagram from McArdle & Epstein (1987) and McArdle & Hamagami (1992).

McArdle J, Nesselroade J. Growth curve analysis in contemporary psychological research. In: Schinka JA, Velicer W, eds. *Comprehensive Handbook of Psychology, Volume Two: Research Methods in Psychology*. New York: Wiley; 2002.

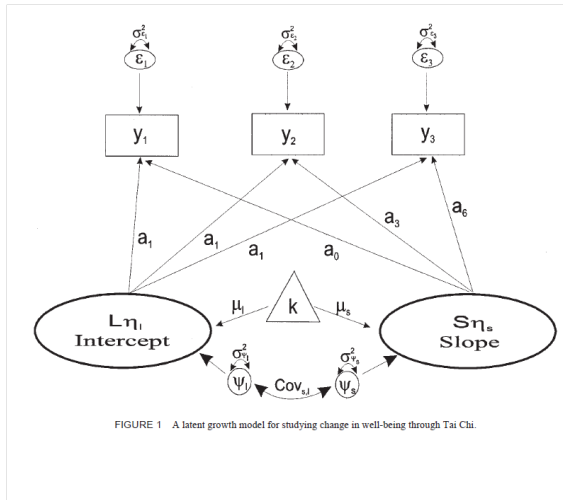
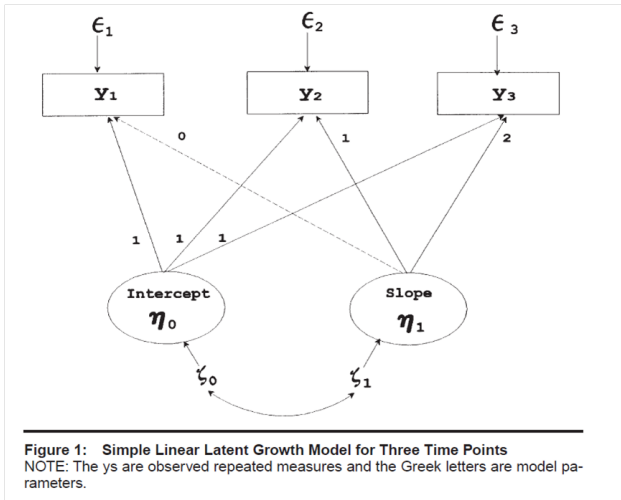


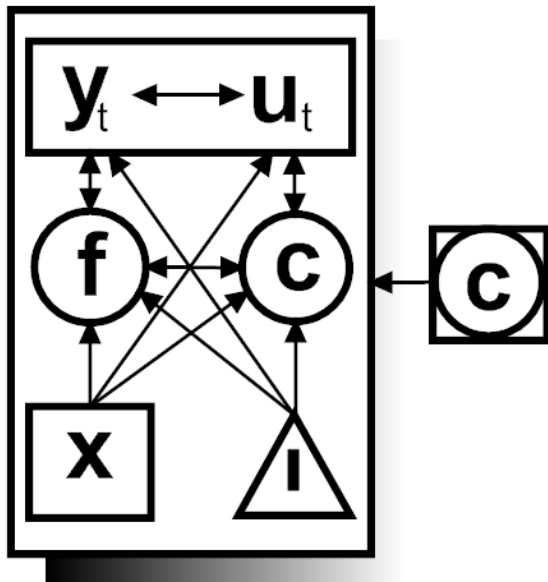
FIGURE 1 A latent growth model for studying change in well-being through Tai Chi.

Li et al. Enhancing the psychological well-being of elderly individuals through Tai Chi exercise: A latent growth curve analysis. Structural Equation Modeling. January 2001;8(1):53-83.

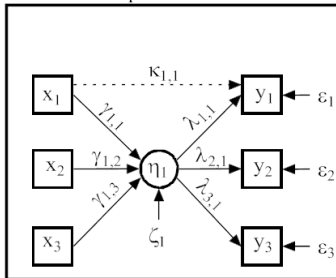


Wang et al. Evaluation of Hiv Risk Reduction Intervention Programs Via Latent Growth Model.
 Eval Rev 1999 23: 648-662

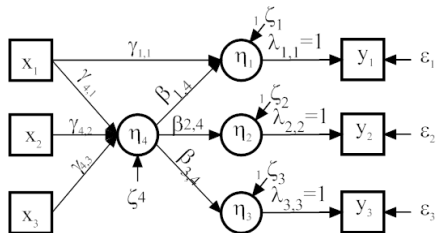
Mplus Framework



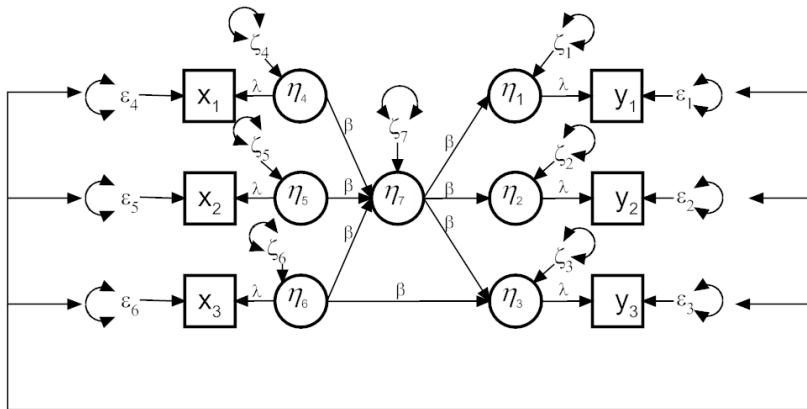
Mplus Model



LISCOMP Model



Mplus Model, All Y continuous



Mplus and LISCOMP Parameter Arrays

Mplus Variable and Parameter Arrays

K (kappa) a $p \times q$ matrix of slopes for the regressions of the p observed response variables (y 's) on the q x -variables

Measurement Model

$$y^* = v + \Lambda \eta + Kx + \varepsilon$$

Structural Regression model:

$$\eta = \alpha + B\eta + \Gamma x + \zeta$$

LISCOMP Variable and Parameter Arrays

- x (exogenous variables) a q -vector, i.e., $x_1, x_2, x_3 \dots x_q$.
- y (observed response variables) a p -vector i.e., $y_1, y_2 \dots y_p$
- τ (tau) a vector of length $(c-1)$, equal to the number of thresholds; where c is number of categories
- v (nu) a $p \times 1$ vector of measurement intercepts, where p is the number of response variables ($y_1, y_2 \dots y_p$)
- Λ (lambda) a $p \times m$ matrix of measurement slopes/loadings, where p is equal to the number of y 's, and m is equal to the number of latent constructs (η 's).
- I (identity matrix), a $m \times m$ identity matrix
- Θ (theta) a $p \times p$ covariance matrix for the residuals in the measurement relations
- α (alpha) a $m \times 1$ vector of the structural regression latent variable (η 's) means and intercepts. *May be fixed to zero in the for categorical variables.*
- B (beta) a $m \times m$ matrix of slopes for the regressions among the latent variables (η 's) with zeros as it's diagonal elements.
- Γ (gamma) a $m \times q$ matrix of slopes for the regressions of the m latent variables (η 's) on the q x -variables
- Ψ (psi) a $m \times m$ covariance matrix for the latent variables (η 's) and the residuals in the latent variable relations
- Δ (delta) diagonal of a $p \times p$ matrix of scaling factors. Only the p diagonal elements are parameters.

Measurement Model

$$y^* = v + \Lambda \eta + \varepsilon$$

Structural Regression model:

$$\eta = \alpha + B\eta + \Gamma x + \zeta$$

[Link to PDF handout version of this slide](#)

$$y = \nu + \Lambda\eta + Kx + \epsilon$$

$$\eta = \alpha + B\eta + \Gamma x + \zeta$$

$$V(y) = \Lambda\Psi\Lambda^T + \Theta$$

$$V(\epsilon) = \Theta$$

$$V(\zeta) = \Psi$$

$$V(\eta) = \ddot{B}^{-1}(\Gamma\Phi\Gamma^T + \Psi)\ddot{B}^T^{-1}$$

where

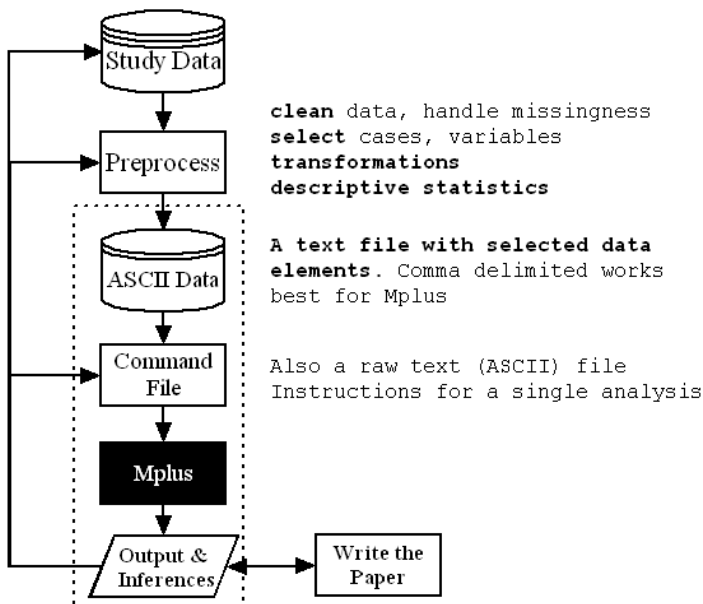
$$\ddot{B} = (I - B)$$

$$\Phi = COV(x)$$

Person-Machine Interface

How do I do this?

SAS, SPSS,
R/S-Plus
STATA



How to Write a Mplus Command File

- Read the Users Manual
 - (at least once all the way through)
- Keep it close to your computer
- Find a similar example
 - (Mplus provides input programs and data for all examples)
- Hack the example to suit your problem

Mplus Commands

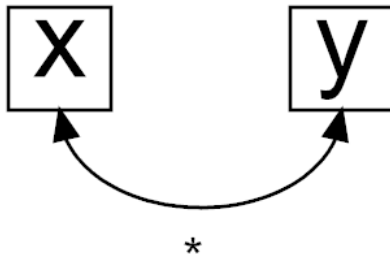
- Title
- Data
- Variable
- Analysis
- Model
- Output
- Define
- Plot
- Savedata
- Montecarlo

How to Write a Mplus Command File

- Read the Users Manual
 - ▶ (at least once all the way through)
- Keep it close to your computer
- Find a similar example
 - ▶ (Mplus provides input programs and data for all examples)
- Hack the example to suit your problem

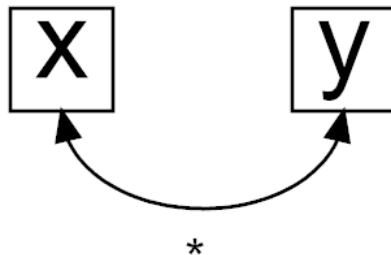
Understanding Mplus Syntax

Correlation



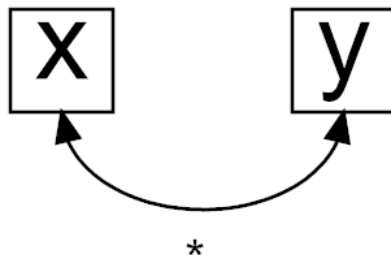
```
MODEL: x with y ;
```

Covariance



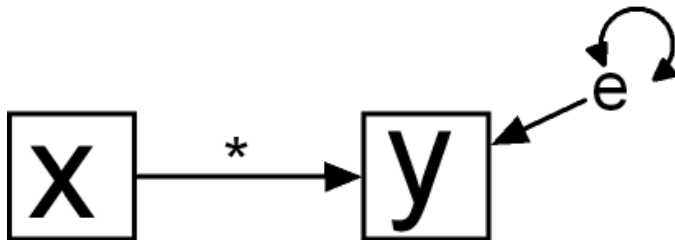
```
TITLE:      A Covariance
DATA:       FILE = temp.dat ;
VARIABLE:   NAMES = x y ;
MODEL:      x with y ;
```

Correlation



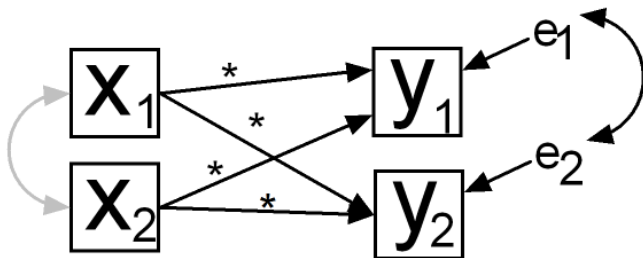
```
TITLE:      A Covariance
DATA:       FILE = temp.dat ;
VARIABLE:   NAMES = x y ;
OUTPUT:    STANDARDIZED ;
MODEL:      x with y ;
```

Simple Regression



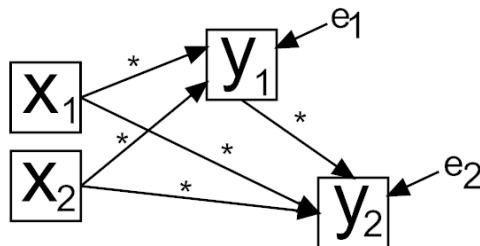
```
MODEL:  y on x * ;  
        y * ; ! error (residual) term  
        [y*] ; ! and intercept are  
               ! estimated by default  
               ! so "MODEL: y on x* ;" would be  
               ! sufficient
```

Multivariate Regression



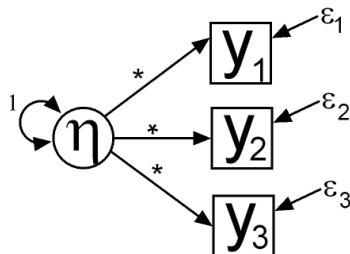
```
MODEL: y1 y2 on x1 * x2 * ;  
! default intercepts estimated -> [y1 * y2*] ;  
! default resid. var. estimated -> y1 * y2 * ;  
! default resid. cov. estimated -> y2 with y1 * ;  
! default covariance of x's -> x1 with x2 * ;
```

Path Analysis



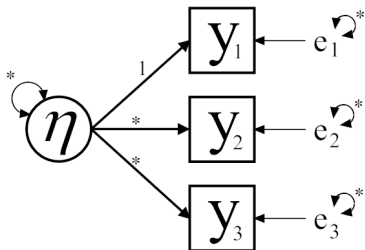
```
MODEL:  y1 y2 on x1 x2 * ;  
        y2 on y1 * ;
```

Confirmatory Factor Analysis



```
TITLE:      Confirmatory factor analysis
            Alternative specification
DATA:       FILE = trash.dat ;
VARIABLE:   NAMES = y1-y3 ;
OUTPUT:     STANDARDIZED ;
MODEL:      eta by y1-y3* ;
            eta@1 ;
```

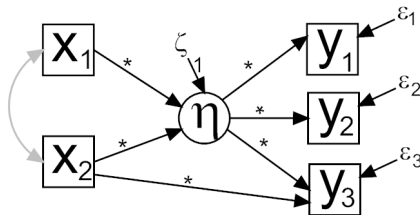
Confirmatory Factor Analysis



```
TITLE:      Confirmatory factor analysis
            Default specification
DATA:       FILE = trash.dat ;
VARIABLE:   NAMES = y1-y3 ;
OUTPUT:     STANDARDIZED ;
MODEL:      eta by y1-y3 ; ! NB: no *
```

MIMIC

Multiple Indicators Multiple Causes



```

TITLE:      MIMIC
DATA:       FILE = trash.dat ;
VARIABLE:   NAMES = y1-y3 x1 x2 ;
MODEL:      eta by y1-y3* ;
            eta@1 ;
            eta on x1-x2 ;    ! indirect effects
            y3 on x2 ;        ! direct effect
  
```

Getting Data into Mplus

What Mplus Eats

- Individual-level data
- Summary Data
 - ▶ correlations, covariances,
 - ▶ means, standard deviations
- ASCII Data
 - ▶ Raw text
 - ▶ Fixed, Free format

🔒 <https://stats.idre.ucla.edu/mplus/faq/>

UCLA Institute for Digital Research & Education

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ABOUT US

MPLUS FREQUENTLY ASKED QUESTIONS

- [How can I convert a Stata file to an Mplus file?](#)
- [How can I convert an SPSS file to an Mplus file?](#)
- [How can I convert a SAS file to an Mplus file?](#)

<https://stats.idre.ucla.edu/mplus/faq/>

General Idea

- ❶ Use a general purpose stat package for EDA and data preparation
- ❷ Always use scripts/syntax/programs
- ❸ Understand your data before you put move to Mplus
- ❹ Never overwrite original data
- ❺ Handle missing values
 - ▶ Replace with very large negative number (e.g., -9999)
 - ▶ Address missing values among INDEPENDENT variables
 - ★ Multiple Imputation (consider Mplus Bayes [Day 3])
 - ★ Declare independent variables dependent variables, assume a distribution, and use ML techniques
- ❻ Output selected variables to ASCII text file (or CSV file)

Converting Microsoft Excel File into raw data file for Mplus

- ➊ Click on "file", select "save as"
- ➋ Under "save as type" select CSV MS-DOS, click save
- ➌ A warning will pop up: click "OK"
 - ▶ "The selected file type does not support workbooks..."
- ➍ Then another warning will pop up: click "yes"
 - ▶ "*.csv may contain features..."
- ➎ Close out of Excel
- ➏ Find the *.csv file in your directory/file and right click
- ➐ Open with text editor (e.g., Notepad, TextWrangler) and examine
 - ▶ First row of variable names and
 - ▶ Commas delimiting the variables
- ➑ Copy variable names (for Mplus command file) and delete this line

Stata program

```
prepare__csv.do
```

```
use ms q1.dta, clear
```

```
summarize
```

```
* Missing values in dependent variables (-9999)
```

```
foreach var of varlist msq3-msq9 {
```

```
    replace 'var' = -9999 if inlist('var',0,1)~=1
```

```
}
```

```
foreach var of varlist male age educ {
```

```
    drop if 'var'==. // one way to handle missingness in x's
```

```
}
```

```
* assume anyone not identified as Black is White
```

```
generate white=black~=1
```

```
order msq3-msq9 site id male age white educ
```

```
keep msq3-msq9 site id male age white educ
```

```
outsheet using msq1.csv , nonames noquote nolabel comma
```

Selected lines from data file

msq1.csv

```
2,1,1,1,1,2,3,300065,0,67,1,1  
1,1,1,1,1,1,3,300066,0,67,0,4  
2,1,1,2,1,2,3,300067,1,87,0,1  
1,2,1,1,1,2,3,300068,1,82,0,1  
2,1,1,2,1,2,3,300069,1,82,0,2  
2,1,1,1,1,2,3,300070,0,77,0,2  
2,1,1,2,2,2,3,300071,0,72,0,2  
2,1,1,1,1,1,3,300072,1,72,0,4  
1,1,1,1,1,1,3,300073,1,67,0,4  
2,1,-9999,-9999,2,-9999,3,300074,1,67,0,1
```

SPSS program

prepare_csv.sps

```
get file= 'msq1.sav' .
execute .
compute n = 1 .
desc var all .
do repeat X = male age educ / .
    if missing(X) n=0 .
end repeat .
select if n=1 .
do repeat X = msq3 msq4 msq5 msq6 msq7 msq9 / .
    if missing(X) X=-9999 .
end repeat .
compute white=1 .
if black=1 white=0 .
save outfile ='trash.sav'
    /keep=msq3 msq4 msq5 msq6 msq7 msq9 site id male age white educ .
get file ='trash.sav' .
save translate outfile='msq_from_spss.dat'
    /type=tab /map /replace.
execute .
```

Selected lines from data file

msq_from_spss.dat

2	1	1	1	1	2	3	300065	0	67	1	1
1	1	1	1	1	1	3	300066	0	67	0	4
2	1	1	2	1	2	3	300067	1	87	0	1
1	2	1	1	1	2	3	300068	1	82	0	1
2	1	1	2	1	2	3	300069	1	82	0	2
2	1	1	1	1	2	3	300070	0	77	0	2
2	1	1	2	2	2	3	300071	0	72	0	2
2	1	1	1	1	1	3	300072	1	72	0	4
1	1	1	1	1	1	3	300073	1	67	0	4
2	1	-9999	-9999	2	-9999	3	300074	1	67	0	1

SAS procedure

from <http://www.ats.ucla.edu/stat/mplus/faq/sas2mplus.htm>

```
data sample2;  
set "sample";  
array allvars _numeric_ ;  
do over allvars;  
    if missing(allvars) then allvars = -9999 ;  
end;  
run;  
proc export data=sample2 outfile='sample.dat'  
    dbms=dlm replace ;  
run;
```

Make your life easy

- Stata

stata2mplus.ado - format and export data, start an inp. From Stata:

```
ssc install stata2mplus
```

runmplus.ado - format and export data, write inp, run Mplus, display results to Stata results window, save estimation results.

Visit <http://www.lvmworkshop.org/home/runmplus-stuff>

mplus.ado - similar to runmplus

- ▶ net from <http://www.soc-research.net/ado/mplus>

- R

- ▶ MplusAutomation package

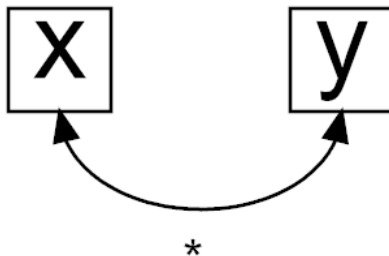
- SPSS, SAS

- ▶ Write your own macros

Stata/runmplus and R/MplusAutomation

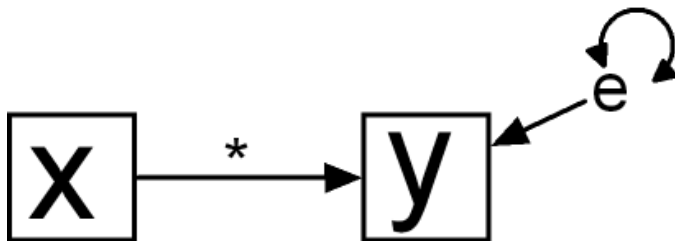
- Integration of Mplus with Stata and R
- Results stored as objects (estimates, fits)
- Facilitates
 - ▶ Reproducible data analysis
 - ▶ Simulation, bootstrap, multiple imputation
 - ▶ Automated model building
 - ▶ Learning about models
- Eliminates
 - ▶ Copying and pasting of output (and associated errors)
 - ▶ Other drudgery (making DAT or CSV files)
 - ▶ Clutter (inp, out, and dat, files of unknown origin)
- But,
 - ▶ Risk not learning about Mplus software
 - ▶ No Mplus plots

Correlation



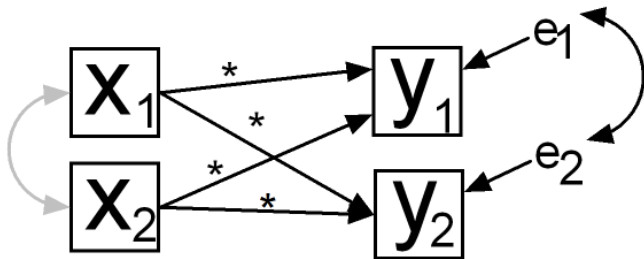
```
runmplus x y , model(x with y;) standardized
```

Simple Regression



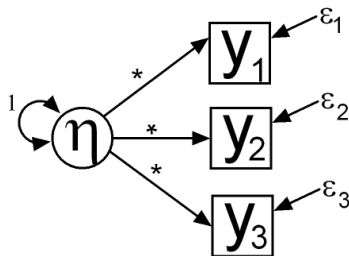
```
runmplus x y , model(y on x *;)
```

Multivariate Regression



```
runmplus y1-y2 x1-x2 , model(y1 y2 on x1 x2; )
```

Confirmatory Factor Analysis



```
runmplus y1-y3 , model(eta by y1-y3* ; eta@1;)
```

R/MplusAutomation

- Run Mplus from R
- Requires separately constructed INP file
- Good parameter and fit statistic extraction
- Nice utilities for SWEAVEing (output to LaTeX)
- Ideally suited for simulations (running lots of INPs sequentially)
- Can be tricked into being a general purpose Mplus launcher

Using Mplus via R

- MplusAutomation R package
- Designed to automate three major aspects of latent variable modeling in Mplus:
 - ▶ Creating related groups of models
 - ▶ Running Batches
 - ▶ Extracting and tabulating model parameters and test statistics
- <http://statmodel.com/usingmplusviar.shtml>
- <http://cran.r-project.org/web/packages/MplusAutomation/index.html>
- Written by Michael.Hallquist@gmail.com

Hallquist, M. N., & Wiley, J. F. (2018). Mplus Automation: An R package for facilitating large-scale latent variable analyses in Mplus. *Structural Equation Modeling*, 25(4), 621-638.

Step 1: Install R Packages

```
# Step 1
# Select your CRAN Mirror
# done already-> chooseCRANmirror()
# Install the MplusviaR package
# done already-> install.packages("MplusAutomation", .Library)
library(MplusAutomation)
```

Step 2: Set your work directory

```
# Step 2
# Set working directory
# This is where your data and input files should be
setwd("C:/work/shortcourse/posted/analysis/learnMPA")
```

Step 3: Getting Data into R

```
# Step 3
# Input data from a .dat file into R
EPESE_data <- read.table("ex01-02.dat",
  header=FALSE, sep=",")

# Declare column names
colnames(EPESE_data) <- c("sad", "blues", "depress",
  "happy", "enjoy", "hopeful", "age", "male")

# Attach dataframe (allows you to call variables by name)
attach(EPESE_data)
```

Optional Step 4: Create your variables

```
# Step 4
# Create a binary variable for age, which equals
# 0 when a participant's age is less than the mean
# age in the sample, and 1 if it is more
age_mean <- mean(age)
age_high <- ifelse(age < age_mean, 0, 1)

# Exchange binary age variable for continuous
# in EPESE dataset
EPESE_data_cat <- data.frame(sad, blues, depress,
                             happy, enjoy, hopeful, age_high, male)

# Export data from R to Mplus-friendly format
prepareMplusData(EPESE_data_cat,
                  "epesecfa.dat")
```

Step 5: Prepare your input file for Mplus

```
# Step 5
# Prepare Mplus INP file
cfa <- mplusObject(
  TITLE = "EX01-02 - CFA with covariates EPESE CESD;",
  VARIABLE = "CATEGORICAL = sad blues depress
             happy enjoy hopeful;",
  MODEL = "f by sad-hopeful;
          f on age_high male ;",
  OUTPUT = "STDYX;
           MODINDICES (ALL -0);",
  usevariables = c("sad", "blues", "depress",
                  "happy", "enjoy", "hopeful",
                  "age_high", "male"),
  rdata = EPESE_data_cat)
```

Step 6: Running the Model via Mplus

```
result <- mplusModeler(cfa,  
  "EPESE_data_cat.dat",  
  modelout = "ex0102.inp",  
  run = 1L)
```

Step 7a: Show output from Mplus in R

```
# Step 7a
# Show all output
model_output <- readModels(
  "C:/work/shortcourse/posted/analysis/learnMPA")
model_output

# Extract summary statistics
summaryStats <- extractModelSummaries(
  "C:/work/shortcourse/posted/analysis/learnMPA")
summaryStats
```

Step 7b: Show table of output from Mplus in R

```
# Step 7b
# Show relevant summary statistics in a table
summary_table <- cbind(summaryStats$Observations,
                        summaryStats$ChiSqM_Value,
                        summaryStats$ChiSqM_DF,
                        summaryStats$ChiSqM_PValue,
                        summaryStats$CFI,
                        summaryStats$TLI,
                        summaryStats$RMSEA_Estimate)

colnames(summary_table) <- cbind(
  "N", "Chi-square", "(DF)", "(p)", "CFI", "TLI", "RMSEA")
summary_table
```

Step 7c: Show table of other output from Mplus in R

```
# Step 7c
# Show modification indices
mod_indices <-extractModIndices(
  "C:/work/shortcourse/posted/analysis/learnMPA")
mod_indices

# Show parameter estimates
model_results <-extractModelParameters(
  "C:/work/shortcourse/posted/analysis/learnMPA")
model_results
```

R lavaan Package

lavaan

- R package for structural equation modeling and more
- Yves Rosseel, PhD, Ghent University, Belgium
- <http://lavaan.ugent.be/>
- Replicates Mplus estimators and output
- Supports some models and features (categorical data, sampling weights) but not all (multilevel, mixture, Bayesian) that Mplus does.
- GNU General Public License

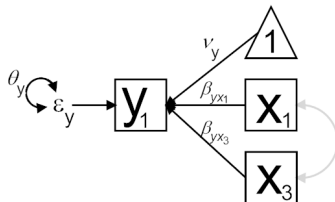
lavaan vs Stata/runmplus and R/MplusAutomation

- Stata/runmplus and R/MplusAutomation require Mplus
- lavaan is its own software
 - ▶ Have to learn new model syntax in lavaan
 - ▶ Have to learn something new regardless

lavaan vs. Mplus

	<i>Mplus</i>	<i>R/lavaan</i>
TITLE:	Multivariable regression	# Multivariable Regression library(lavaan) setwd("c:/whereverdatais ")
DATA:	FILE = ex3.1.dat ;	data <- read.csv("ex3.1.dat", head=FALSE)
VARIABLE:	NAMES = y1 x1 x3 ;	names(data) <- c("y1","x1","x3")
MODEL:	y1 on x1 x3 ;	fit <- sem('y1 ~ 1 + x1 + x3', std.lv=TRUE, data=data)
OUTPUT:	STANDARDIZED ;	standardizedsolution(fit)

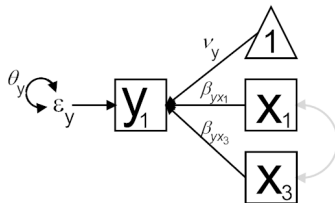
Mplus Users Guide Ex 3.1: Multivariable Regression



Mplus Users Guide Example 3.1

```
TITLE:      Multivariable regression
DATA:       FILE = ex3.1.dat ;
VARIABLE:   NAMES = y1 x1 x3 ;
MODEL:      y1 on x1 x3 ;
OUTPUT:     STANDARDIZED ;
```

Mplus Users Guide Ex 3.1: Multivariable Regression (lavaan)



Mplus Users Guide Example 3.1

```
# Mplus Users Guide Ex 3.1: Multivariable Regression
library(lavaan)
setwd("c:/work/shortcourse/private/rnj")
data <- read.csv("ex3.1.dat", head=FALSE)
names(data) <- c("y1", "x1", "x3")
fit <- sem('y1 ~ 1 + x1 + x3', std.lv=TRUE, data=data )
summary(fit, standardized=TRUE, fit.measures = TRUE)
standardizedsolution(fit)
```

Extra credit

Astute observers will note that Mplus and lavaan provide different loglikelihood and AIC values. This is because Mplus and lavaan allow different parts of the model into the likelihood function. To get Mplus to produce the same loglikelihood, use this model statement:

MODEL:

```
y1s by y1@1 ;  
x1s by x1@1 ;  
x3s by x3@1 ;  
y1@0 ;  
x1@0 ;  
x3@0 ;  
y1s on x1s x3s ;  
x1s with x3s ;
```

Refer back to the all-y model case (slide 85) to understand why.

Questions

Footnote on Data Management for LDA

Some approaches to LDA require data in "wide" format, others "long" format

LONG				WIDE						
Vertical				Horizontal						
Multiple record				Single record						
id	time	y1	x1	id	y1	y2	y3	x1	x2	x3
1	1	y11	x11	1	y11	y12	y13	x11	x12	x13
1	2	y12	x12	2	y21	y22	y23	x21	x22	x23
1	3	y13	x13	...						
2	1	y21	x21	n	yn1	yn2	yn3	xn1	xn2	xn3
2	2	y22	x22							
2	3	y23	x23							
3	1	y31	x31							
...										
n	p	ynp	xnp							

Most Mplus LDA models require WIDE format

Skills Challenge

- Invest time learning how to move from wide to long easily with your favorite general purpose statistics package
- Why
 - ▶ Wide
 - ★ Mplus LDA models want wide
 - ▶ Long
 - ★ Storage and data management more efficient
 - ★ Better for exploratory graphics
 - ★ Enables checking your growth models with mixed effects or HLM models

Resources

- UCLA IDRE Learning Modules on reshaping data
 - ▶ Stata <https://stats.idre.ucla.edu/stata/faq/how-can-i-reshape-doubly-or-triply-wide-data-to-long/>
 - ▶ SAS <https://stats.idre.ucla.edu/sas/modules/how-to-reshape-data-long-to-wide-using-proc-transpose/>
 - ▶ SPSS <https://stats.idre.ucla.edu/spss/modules/reshaping-data-wide-to-long/>
 - ▶ R <https://stats.idre.ucla.edu/r/faq/how-can-i-reshape-my-data-in-r/>
- Other Resources
 - ▶ R (S-plus) <https://stat.ethz.ch/pipermail/r-help/2009-February/189684.html>
- Mplus will do it with built-in functionality
 - ▶ But it is not the right tool for data management

Questions

Examples

Latent growth curve modeling is a highly constrained form of confirmatory factor analysis (CFA). The goal of this session is to provide examples of specifying and estimating a CFA model in Mplus, as a basic Mplus skills-building exercise.

Time permitting, we will look at growth modeling examples from Singer and Willet (2003).

Example Data Set

Established Populations for the Epidemiologic
Study of the Elderly

EPESE

- N=14,456
- General population survey of older adults
- Carried out at four US sites:
 - ▶ East Boston, MA (N=3,809)
 - ▶ Iowa and Washington Counties, IA (N=3,673)
 - ▶ New Haven, CT(N=2,812)
 - ▶ Durham-Piedmont area of NC (N=4,162)
- Field work began 1981, 1985 in NC
- Data are in the public domain (ICPSR)
- Results we derive have not been published

age	Freq.	Percent	Cum.
67: 65-69	4,457	30.83	30.83
72: 70-74	3,970	27.46	58.29
77: 75-79	2,872	19.87	78.16
82: 80-84	1,878	12.99	91.15
87: 85+	1,279	8.85	100.00
Total	14,456	100.00	

male	Freq.	Percent	Cum.
0	8,960	61.98	61.98
1	5,496	38.02	100.00
Total	14,456	100.00	

Black or African-American

site	0	1	.	Total
1 EB-MA	0	0	3,809	3,809
2 IW-IO	0	0	3,673	3,673
3 NH-CT	2,283	529	0	2,812
4 DP-NC	1,901	2,261	0	4,162
Total	4,184	2,790	7,482	14,456

educ	Freq.	Percent	Cum.
1 <8	4,056	28.06	28.06
2 8-11	5,686	39.33	67.39
3 12	2,552	17.65	85.04
4 >12	1,863	12.89	97.93
.	299	2.07	100.00
Total	14,456	100.00	

Cognition

Short Portable Mental Status Questionnaire

- Conventionally, 10 items
- We will use six items from EPESE
 - ▶ msq3 what is date
 - ▶ msq4 what is the day of week
 - ▶ msq5 who is the president
 - ▶ msq6 who was the prior president
 - ▶ msq7 what is your Mothers maiden name
 - ▶ msq9 Count back from 20 by 3s

EPESE

Mental Status Questionnaire

Wave 1 Item Response Frequency

		Response			
		Error	NoErr	miss	
item		0	1	.	Total
-----+-----					
msq3 what is the date...		3907	9788	761	14456
msq4 day of week.....		772	12916	768	14456
msq5 who is president...		1239	12436	781	14456
msq6 who was prior pres.		4302	9355	799	14456
msq7 what is M Maiden...		539	13075	842	14456
msq9 back from 20 by 3..		5194	7768	1494	14456

Depression

- CES-D Centers for Epidemiologic Studies Depression Scale
- We will use six items from New Haven site, which ask if *During the past week ...*

SAD: ...I felt sad

BLUES: ...I felt I could not
shake off the blues

DEPRESS: ... I felt depressed

0: Rarely or None of the time

1: Some of the time

2: Much of the time

3: Most or all of the time

HAPPY: ...I was happy

ENJOY: ...I enjoyed life

HOPEFUL: ...I felt hopeful
about the future

0: Most or all of the time

1: Much of the time

2: Some of the time

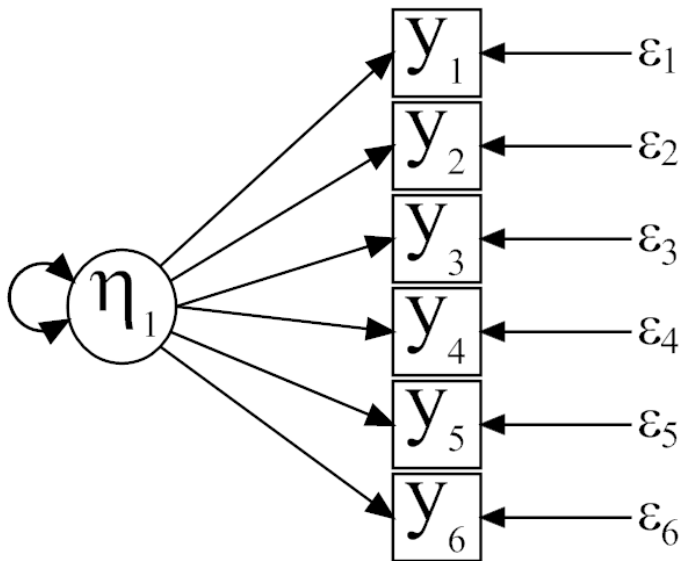
3: Rarely or None of the time

Item Response Frequency
(entries are percentages)

		Response				
item		0	1	2	3	N
-----+-----						
01 sad.....		66	27	4	4	2442
02 blues....		78	15	3	4	2442
03 depress..		64	26	4	6	2442
04 happy....		67	10	16	7	2442
05 enjoy....		76	8	11	6	2442
06 hopeful..		58	7	16	20	2442

Confirmatory Factor Analysis Example [ex0101.inp]

Confirmatory Factor Analysis
New Haven EPESE
Six-item CES-D
Continuous Factor Indicators



```
1 Mplus VERSION 8.8 DEMO (Mac)
2 MUTHEN & MUTHEN
3 07/11/2022    2:07 PM
4
5 INPUT INSTRUCTIONS
6
7   TITLE:      EX0101 CFA EPESE CESD ;
8
9   DATA:      FILE = ex0101.dat ;
10
11  VARIABLE: NAMES = sad blues depress happy enjoy hopeful age male ;
12             USEVARIABLES = sad blues depress happy enjoy hopeful ;
13             MISSING ARE ALL (-9999) ;
14
15  OUTPUT:     STANDARDIZED TECH1 ;
16
17  MODEL:      f1 by sad-hopeful;
18
19
20
21
22 INPUT READING TERMINATED NORMALLY
23
24
25
26 EX0101 CFA EPESE CESD ;
27
28 SUMMARY OF ANALYSIS
29
30 Number of groups                      1
31 Number of observations                2442
32
```

```

33 Number of dependent variables                6
34 Number of independent variables              0
35 Number of continuous latent variables        1
36
37 Observed dependent variables
38
39   Continuous
40   SAD          BLUES          DEPRESS      HAPPY          ENJOY          HOPEFUL
41
42 Continuous latent variables
43   F1
44
45
46 Estimator                                    ML
47 Information matrix                          OBSERVED
48 Maximum number of iterations                 1000
49 Convergence criterion                       0.500D-04
50 Maximum number of steepest descent iterations 20
51 Maximum number of iterations for H1          2000
52 Convergence criterion for H1                 0.100D-03
53
54 Input data file(s)
55   ex0101.dat
56
57 Input data format   FREE
58
59
60 SUMMARY OF DATA
61
62   Number of missing data patterns            1
63
64

```

65 COVARIANCE COVERAGE OF DATA

66

67 Minimum covariance coverage value 0.100

68

69

70 PROPORTION OF DATA PRESENT

71

72

73 Covariance Coverage

74	SAD	BLUES	DEPRESS	HAPPY	ENJOY
75	-----	-----	-----	-----	-----
76 SAD	1.000				
77 BLUES	1.000	1.000			
78 DEPRESS	1.000	1.000	1.000		
79 HAPPY	1.000	1.000	1.000	1.000	
80 ENJOY	1.000	1.000	1.000	1.000	1.000
81 HOPEFUL	1.000	1.000	1.000	1.000	1.000

82

83

84 Covariance Coverage

85 HOPEFUL

86 -----

87 HOPEFUL 1.000

88

89

90

91 UNIVARIATE SAMPLE STATISTICS

92

93

94 UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

95

96	Variable/ Percentiles	Mean/ Variance	Skewness/ Kurtosis	Minimum/ Maximum	% with Min/Max	20%/60%	40%/
97	Sample Size 80% Median						
98							
99	SAD	0.454	1.831	0.000	65.77%	0.000	
	0.000 0.000						
100	2442.000	0.548	3.185	3.000	3.81%	0.000	
	1.000						
101	BLUES	0.337	2.417	0.000	78.05%	0.000	
	0.000 0.000						
102	2442.000	0.545	5.259	3.000	4.34%	0.000	
	1.000						
103	DEPRESS	0.510	1.749	0.000	64.41%	0.000	
	0.000 0.000						
104	2442.000	0.674	2.457	3.000	5.81%	0.000	
	1.000						
105	HAPPY	0.630	1.241	0.000	67.08%	0.000	
	0.000 0.000						
106	2442.000	0.979	0.069	3.000	7.25%	0.000	
	2.000						
107	ENJOY	0.465	1.742	0.000	75.96%	0.000	
	0.000 0.000						
108	2442.000	0.816	1.624	3.000	5.94%	0.000	
	1.000						
109	HOPEFUL	0.973	0.681	0.000	57.70%	0.000	
	0.000 0.000						
110	2442.000	1.517	-1.263	3.000	19.57%	1.000	
	2.000						
111							
112							
113	THE MODEL ESTIMATION TERMINATED NORMALLY						

```

114
115
116
117 MODEL FIT INFORMATION
118
119 Number of Free Parameters          18
120
121 Loglikelihood
122
123           H0 Value          -17278.438
124           H1 Value          -17036.545
125
126 Information Criteria
127
128           Akaike (AIC)          34592.876
129           Bayesian (BIC)          34697.286
130           Sample-Size Adjusted BIC          34640.096
131           (n* = (n + 2) / 24)
132
133 Chi-Square Test of Model Fit
134
135           Value          483.785
136           Degrees of Freedom          9
137           P-Value          0.0000
138
139 RMSEA (Root Mean Square Error Of Approximation)
140
141           Estimate          0.147
142           90 Percent C.I.          0.136  0.158
143           Probability RMSEA <= .05          0.000
144
145 CFI/TLI

```

```

146
147             CFI                      0.883
148             TLI                      0.805
149
150 Chi-Square Test of Model Fit for the Baseline Model
151
152             Value                      4064.686
153             Degrees of Freedom          15
154             P-Value                    0.0000
155

```

```

156 SRMR (Standardized Root Mean Square Residual)
157

```

```

158             Value                      0.057
159
160
161

```

162 MODEL RESULTS

				Two-Tailed	
		Estimate	S.E.	Est./S.E.	P-Value
167	F1	BY			
168	SAD	1.000	0.000	999.000	999.000
169	BLUES	0.963	0.035	27.751	0.000
170	DEPRESS	1.193	0.040	30.075	0.000
171	HAPPY	1.230	0.049	24.975	0.000
172	ENJOY	1.075	0.045	23.950	0.000
173	HOPEFUL	0.883	0.057	15.489	0.000
174					
175	Intercepts				
176	SAD	0.454	0.015	30.325	0.000
177	BLUES	0.337	0.015	22.553	0.000

178	DEPRESS	0.510	0.017	30.685	0.000
179	HAPPY	0.630	0.020	31.451	0.000
180	ENJOY	0.465	0.018	25.421	0.000
181	HOPEFUL	0.973	0.025	39.018	0.000
182					
183	Variances				
184	F1	0.260	0.015	17.461	0.000
185					
186	Residual Variances				
187	SAD	0.288	0.011	27.263	0.000
188	BLUES	0.304	0.011	27.677	0.000
189	DEPRESS	0.304	0.013	23.883	0.000
190	HAPPY	0.586	0.021	27.902	0.000
191	ENJOY	0.516	0.018	28.700	0.000
192	HOPEFUL	1.315	0.039	33.536	0.000

193

194

195 QUALITY OF NUMERICAL RESULTS

196

197 Condition Number for the Information Matrix
 198 (ratio of smallest to largest eigenvalue)

0.141E-01

199

200

201 STANDARDIZED MODEL RESULTS

202

203

204 STDYX Standardization

205

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
--	----------	------	-----------	-----------------------

208

209 F1 BY

210	SAD	0.689	0.014	50.000	0.000
211	BLUES	0.665	0.015	45.558	0.000
212	DEPRESS	0.741	0.013	56.771	0.000
213	HAPPY	0.634	0.016	40.045	0.000
214	ENJOY	0.606	0.016	36.810	0.000
215	HOPEFUL	0.366	0.020	18.142	0.000
216					
217	Intercepts				
218	SAD	0.614	0.022	27.819	0.000
219	BLUES	0.456	0.021	21.463	0.000
220	DEPRESS	0.621	0.022	28.096	0.000
221	HAPPY	0.636	0.022	28.680	0.000
222	ENJOY	0.514	0.022	23.890	0.000
223	HOPEFUL	0.790	0.023	34.068	0.000
224					
225	Variances				
226	F1	1.000	0.000	999.000	999.000
227					
228	Residual Variances				
229	SAD	0.526	0.019	27.698	0.000
230	BLUES	0.558	0.019	28.739	0.000
231	DEPRESS	0.451	0.019	23.359	0.000
232	HAPPY	0.598	0.020	29.838	0.000
233	ENJOY	0.632	0.020	31.669	0.000
234	HOPEFUL	0.866	0.015	58.825	0.000
235					
236					
237	STDY Standardization				
238					
239					Two-Tailed
240		Estimate	S.E.	Est./S.E.	P-Value
241					

242	F1	BY				
243	SAD		0.689	0.014	50.000	0.000
244	BLUES		0.665	0.015	45.558	0.000
245	DEPRESS		0.741	0.013	56.771	0.000
246	HAPPY		0.634	0.016	40.045	0.000
247	ENJOY		0.606	0.016	36.810	0.000
248	HOPEFUL		0.366	0.020	18.142	0.000
249						
250	Intercepts					
251	SAD		0.614	0.022	27.819	0.000
252	BLUES		0.456	0.021	21.463	0.000
253	DEPRESS		0.621	0.022	28.096	0.000
254	HAPPY		0.636	0.022	28.680	0.000
255	ENJOY		0.514	0.022	23.890	0.000
256	HOPEFUL		0.790	0.023	34.068	0.000
257						
258	Variances					
259	F1		1.000	0.000	999.000	999.000
260						
261	Residual Variances					
262	SAD		0.526	0.019	27.698	0.000
263	BLUES		0.558	0.019	28.739	0.000
264	DEPRESS		0.451	0.019	23.359	0.000
265	HAPPY		0.598	0.020	29.838	0.000
266	ENJOY		0.632	0.020	31.669	0.000
267	HOPEFUL		0.866	0.015	58.825	0.000
268						
269						
270	STD Standardization					
271						
272					Two-Tailed	
273		Estimate	S.E.	Est./S.E.	P-Value	

274					
275	F1	BY			
276	SAD		0.510	0.015	34.922
277	BLUES		0.491	0.015	33.059
278	DEPRESS		0.608	0.016	37.799
279	HAPPY		0.627	0.021	30.522
280	ENJOY		0.548	0.019	28.846
281	HOPEFUL		0.450	0.027	16.672
282					
283	Intercepts				
284	SAD		0.454	0.015	30.325
285	BLUES		0.337	0.015	22.553
286	DEPRESS		0.510	0.017	30.685
287	HAPPY		0.630	0.020	31.451
288	ENJOY		0.465	0.018	25.421
289	HOPEFUL		0.973	0.025	39.018
290					
291	Variances				
292	F1		1.000	0.000	999.000
293					
294	Residual Variances				
295	SAD		0.288	0.011	27.263
296	BLUES		0.304	0.011	27.677
297	DEPRESS		0.304	0.013	23.883
298	HAPPY		0.586	0.021	27.902
299	ENJOY		0.516	0.018	28.700
300	HOPEFUL		1.315	0.039	33.536
301					
302					
303	R-SQUARE				
304					
305	Observed				Two-Tailed

Variable	Estimate	S.E.	Est./S.E.	P-Value
SAD	0.474	0.019	25.000	0.000
BLUES	0.442	0.019	22.779	0.000
DEPRESS	0.549	0.019	28.385	0.000
HAPPY	0.402	0.020	20.022	0.000
ENJOY	0.368	0.020	18.405	0.000
HOPEFUL	0.134	0.015	9.071	0.000

TECHNICAL 1 OUTPUT

PARAMETER SPECIFICATION

NU

SAD

BLUES

DEPRESS

HAPPY

ENJOY

1

2

3

4

5

NU

HOPEFUL

6

LAMBDA

F1

0

SAD

```

338 BLUES          7
339 DEPRESS        8
340 HAPPY           9
341 ENJOY          10
342 HOPEFUL        11
343
344
345          THETA
346          SAD          BLUES          DEPRESS          HAPPY          ENJOY
347          -----
348 SAD              12
349 BLUES              0          13
350 DEPRESS            0          0          14
351 HAPPY              0          0          0          15
352 ENJOY              0          0          0          0          16
353 HOPEFUL            0          0          0          0          0
354
355
356          THETA
357          HOPEFUL
358          -----
359 HOPEFUL            17
360
361
362          ALPHA
363          F1
364          -----
365              0
366
367
368          BETA
369          F1

```

```

370          -----
371 F1              0
372
373
374          PSI
375          F1
376          -----
377 F1              18
378
379
380          STARTING VALUES
381
382
383          NU
384          SAD          BLUES          DEPRESS          HAPPY          ENJOY
385          -----          -----          -----          -----          -----
386          0.454          0.337          0.510          0.630          0.465
387
388
389          NU
390          HOPEFUL
391          -----
392          0.973
393
394
395          LAMBDA
396          F1
397          -----
398 SAD          1.000
399 BLUES        0.991
400 DEPRESS      1.218
401 HAPPY        1.176

```

```

402 ENJOY          1.031
403 HOPEFUL        0.847
404
405
406          THETA
407          SAD          BLUES          DEPRESS          HAPPY          ENJOY
408          -----          -----          -----          -----          -----
409 SAD          0.274          0.000          0.000          0.000          0.000
410 BLUES        0.000          0.273          0.000          0.000          0.000
411 DEPRESS      0.000          0.000          0.337          0.000          0.000
412 HAPPY        0.000          0.000          0.000          0.490          0.000
413 ENJOY        0.000          0.000          0.000          0.000          0.408
414 HOPEFUL      0.000          0.000          0.000          0.000          0.000
415
416
417          THETA
418          HOPEFUL
419          -----
420 HOPEFUL      0.759
421
422
423          ALPHA
424          F1
425          -----
426          0.000
427
428
429          BETA
430          F1
431          -----
432 F1          0.000
433

```

```
434
435           PSI
436           F1
437           -----
438 F1           0.050
439
440
441 Beginning Time: 14:07:22
442 Ending Time: 14:07:22
443 Elapsed Time: 00:00:00
444
445
446 Mplus VERSION 8.8 DEMO (Mac) has the following limitations:
447 Maximum number of dependent variables: 6
448 Maximum number of independent variables: 2
449 Maximum number of between variables: 2
450 Maximum number of continuous latent variables in time series analysis: 2
451
452
453 MUTHEN & MUTHEN
454 3463 Stoner Ave.
455 Los Angeles, CA 90066
456
457 Tel: (310) 391-9971
458 Fax: (310) 391-8971
459 Web: www.StatModel.com
460 Support: Support@StatModel.com
461
462 Copyright (c) 1998-2022 Muthen & Muthen
```

ex0101.out

A Note Model Fit

- Model fit is based on how close the model-implied covariance matrix is to the observed covariance matrix
- Chi-Square should be low, P-value high
 - ▶ an *absolute fit* index
 - ▶ very sensitive to sample size
- CFI > 0.95
 - ▶ a *parsimony-adjusted relative fit* index
 - ▶ CFI ranges from 0 to 1
 - ▶ Bentler. *Psych Bull*, 1990; 107:238-46.
- RMSEA $< .05$
 - ▶ an *absolute fit* index
 - ▶ Hu & Bentler. *Psych Methods*, 1998; 4:424-53.

A note on Mplus Standardized Estimates

- Standardized with respect to...

STDYX ...the variances of the continuous latent variables, background, and outcome variables. Use this if the background variable is continuous (and it otherwise makes sense to do so).

STDY ...the variances of the continuous latent variables and the outcome variables. Use this one if the background variable is discrete.

STD ... the variances of the continuous latent variables.

- Standard errors and significance testing

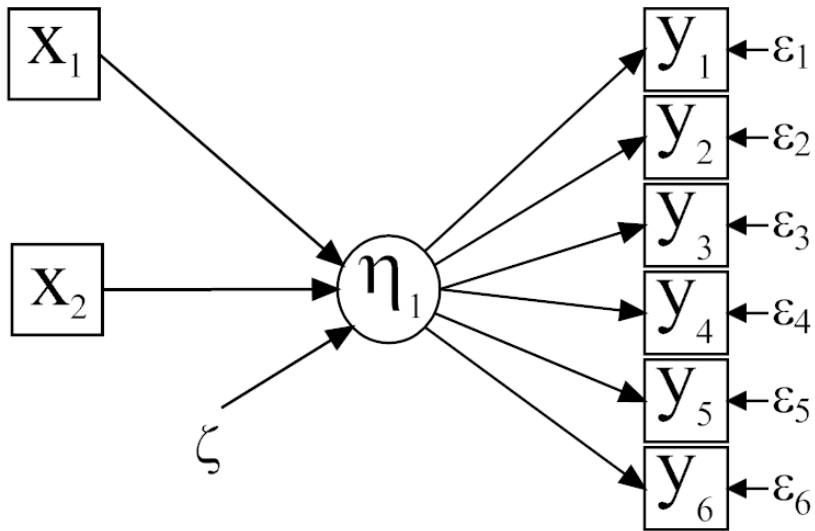
- ▶ Each standardization solution may come with standard errors, z-test, P-value (e.g., STDY not available for categorical dependent variables)
- ▶ Significance tests don't always agree with unstandardized estimates
- ▶ I recommend basing inferences on unstandardized estimates

Example 1.2

CFA with Covariates
CES-D Depression, Age, Sex
New Haven EPESE

CFA with Covariates Example 1.2 ex01-02.inp

MIMIC - Multiple Indicators Multiple Causes EPESE CES-D (six items)
Age, Gender



```
1 Mplus VERSION 8.8 DEMO (Mac)
2 MUTHEN & MUTHEN
3 07/11/2022    2:07 PM
4
5 INPUT INSTRUCTIONS
6
7   TITLE:      EX0102 CFA WITH COVARIATES EPESE CESD ;
8
9   DATA:      FILE = ex0101.dat ;
10
11  VARIABLE: NAMES =  sad blues depress happy enjoy hopeful age male ;
12             MISSING ARE ALL (-9999) ;
13
14  DEFINE:     CENTER age male (grandmean) ;
15
16  OUTPUT:     TECH1 ; standardized ;
17
18  MODEL:      f by sad-hopeful ;
19             f on age male ;
20
21
22
23
24
25
26 INPUT READING TERMINATED NORMALLY
27
28
29
30 EX0102 CFA WITH COVARIATES EPESE CESD ;
31
32 SUMMARY OF ANALYSIS
```

```

33
34 Number of groups                      1
35 Number of observations                2442
36
37 Number of dependent variables        6
38 Number of independent variables      2
39 Number of continuous latent variables 1
40
41 Observed dependent variables
42
43   Continuous
44   SAD          BLUES          DEPRESS    HAPPY          ENJOY          HOPEFUL
45
46 Observed independent variables
47   AGE          MALE
48
49 Continuous latent variables
50   F
51
52 Variables with special functions
53
54   Centering (GRANDMEAN)
55   AGE          MALE
56
57
58 Estimator                          ML
59 Information matrix                  OBSERVED
60 Maximum number of iterations        1000
61 Convergence criterion               0.500D-04
62 Maximum number of steepest descent iterations 20
63 Maximum number of iterations for H1 2000
64 Convergence criterion for H1        0.100D-03

```

```

65
66 Input data file(s)
67   ex0101.dat
68
69 Input data format   FREE
70
71
72 SUMMARY OF DATA
73
74       Number of missing data patterns           1
75
76
77 COVARIANCE COVERAGE OF DATA
78
79 Minimum covariance coverage value   0.100
80
81
82       PROPORTION OF DATA PRESENT
83
84
85             Covariance Coverage
86             SAD           BLUES           DEPRESS           HAPPY           ENJOY
87             -----
88 SAD           1.000
89 BLUES         1.000           1.000
90 DEPRESS       1.000           1.000           1.000
91 HAPPY         1.000           1.000           1.000           1.000
92 ENJOY         1.000           1.000           1.000           1.000           1.000
93 HOPEFUL       1.000           1.000           1.000           1.000           1.000
94 AGE           1.000           1.000           1.000           1.000           1.000
95 MALE          1.000           1.000           1.000           1.000           1.000
96

```

Covariance Coverage

HOPEFUL

AGE

MALE

HOPEFUL

1.000

AGE

1.000

1.000

MALE

1.000

1.000

1.000

UNIVARIATE SAMPLE STATISTICS

UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

Variable/
PercentilesMean/
VarianceSkewness/
KurtosisMinimum/ % with
Maximum Min/Max

20%/60%

40%/

Sample Size

80% Median

SAD

0.454

1.831

0.000

65.77%

0.000

0.000

0.000

2442.000

0.548

3.185

3.000

3.81%

0.000

1.000

BLUES

0.337

2.417

0.000

78.05%

0.000

0.000

0.000

2442.000

0.545

5.259

3.000

4.34%

0.000

1.000

DEPRESS

0.510

1.749

0.000

64.41%

0.000

0.000

0.000

2442.000

0.674

2.457

3.000

5.81%

0.000

1.000

121	HAPPY		0.630	1.241	0.000	67.08%	0.000
		0.000	0.000				
122		2442.000	0.979	0.069	3.000	7.25%	0.000
		2.000					
123	ENJOY		0.465	1.742	0.000	75.96%	0.000
		0.000	0.000				
124		2442.000	0.816	1.624	3.000	5.94%	0.000
		1.000					
125	HOPEFUL		0.973	0.681	0.000	57.70%	0.000
		0.000	0.000				
126		2442.000	1.517	-1.263	3.000	19.57%	1.000
		2.000					
127	AGE		0.000	0.497	-7.316	28.50%	-7.316
		-2.316	-2.316				
128		2442.000	40.696	-0.862	12.684	8.39%	2.684
		7.684					
129	MALE		0.000	0.307	-0.424	57.58%	-0.424
		-0.424	-0.424				
130		2442.000	0.244	-1.906	0.576	42.42%	0.576
		0.576					

131
132
133 THE MODEL ESTIMATION TERMINATED NORMALLY
134
135
136

137 MODEL FIT INFORMATION
138

139 Number of Free Parameters 20

140
141 Loglikelihood
142

```
143          H0 Value                -17255.177
144          H1 Value                -17006.684
145
146 Information Criteria
147
148          Akaike (AIC)              34550.354
149          Bayesian (BIC)            34666.365
150          Sample-Size Adjusted BIC  34602.820
151          (n* = (n + 2) / 24)
152
153 Chi-Square Test of Model Fit
154
155          Value                      496.985
156          Degrees of Freedom          19
157          P-Value                     0.0000
158
159 RMSEA (Root Mean Square Error Of Approximation)
160
161          Estimate                    0.101
162          90 Percent C.I.              0.094    0.109
163          Probability RMSEA <= .05     0.000
164
165 CFI/TLI
166
167          CFI                        0.883
168          TLI                        0.834
169
170 Chi-Square Test of Model Fit for the Baseline Model
171
172          Value                      4124.409
173          Degrees of Freedom          27
174          P-Value                     0.0000
```

175
 176 SRMR (Standardized Root Mean Square Residual)

177
 178 Value 0.048
 179
 180
 181

182 MODEL RESULTS

		Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
183					
184					
185					
186					
187	F BY				
188	SAD	1.000	0.000	999.000	999.000
189	BLUES	0.963	0.035	27.784	0.000
190	DEPRESS	1.190	0.040	30.103	0.000
191	HAPPY	1.231	0.049	25.035	0.000
192	ENJOY	1.075	0.045	24.007	0.000
193	HOPEFUL	0.884	0.057	15.518	0.000
194					
195	F ON				
196	AGE	0.006	0.002	3.234	0.001
197	MALE	-0.134	0.023	-5.778	0.000
198					
199	Intercepts				
200	SAD	0.454	0.015	30.494	0.000
201	BLUES	0.337	0.015	22.670	0.000
202	DEPRESS	0.510	0.017	30.881	0.000
203	HAPPY	0.630	0.020	31.599	0.000
204	ENJOY	0.465	0.018	25.530	0.000
205	HOPEFUL	0.973	0.025	39.078	0.000
206					

207 Residual Variances

208	SAD	0.288	0.011	27.278	0.000
209	BLUES	0.304	0.011	27.713	0.000
210	DEPRESS	0.306	0.013	24.034	0.000
211	HAPPY	0.585	0.021	27.942	0.000
212	ENJOY	0.516	0.018	28.740	0.000
213	HOPEFUL	1.314	0.039	33.541	0.000
214	F	0.254	0.015	17.456	0.000

215
216

217 QUALITY OF NUMERICAL RESULTS

218

219	Condition Number for the Information Matrix	0.568E-03
220	(ratio of smallest to largest eigenvalue)	

221
222

223 STANDARDIZED MODEL RESULTS

224
225

226 STDYX Standardization

227

				Two-Tailed
	Estimate	S.E.	Est./S.E.	P-Value

230

231 F BY

232	SAD	0.689	0.014	50.114	0.000
233	BLUES	0.665	0.015	45.608	0.000
234	DEPRESS	0.739	0.013	56.729	0.000
235	HAPPY	0.634	0.016	40.219	0.000
236	ENJOY	0.607	0.016	36.956	0.000
237	HOPEFUL	0.366	0.020	18.183	0.000

238

239	F	ON				
240		AGE	0.072	0.022	3.249	0.001
241		MALE	-0.130	0.022	-5.880	0.000
242						
243	Intercepts					
244		SAD	0.614	0.022	27.949	0.000
245		BLUES	0.456	0.021	21.564	0.000
246		DEPRESS	0.621	0.022	28.247	0.000
247		HAPPY	0.636	0.022	28.793	0.000
248		ENJOY	0.514	0.021	23.980	0.000
249		HOPEFUL	0.790	0.023	34.108	0.000
250						
251	Residual Variances					
252		SAD	0.525	0.019	27.705	0.000
253		BLUES	0.558	0.019	28.782	0.000
254		DEPRESS	0.453	0.019	23.521	0.000
255		HAPPY	0.598	0.020	29.868	0.000
256		ENJOY	0.632	0.020	31.705	0.000
257		HOPEFUL	0.866	0.015	58.794	0.000
258	F		0.977	0.007	145.590	0.000
259						
260						
261	STDY Standardization					
262						
263					Two-Tailed	
264			Estimate	S.E.	Est./S.E.	P-Value
265						
266	F	BY				
267		SAD	0.689	0.014	50.114	0.000
268		BLUES	0.665	0.015	45.608	0.000
269		DEPRESS	0.739	0.013	56.729	0.000
270		HAPPY	0.634	0.016	40.219	0.000

271	ENJOY	0.607	0.016	36.956	0.000
272	HOPEFUL	0.366	0.020	18.183	0.000
273					
274	F	ON			
275	AGE	0.011	0.003	3.252	0.001
276	MALE	-0.263	0.045	-5.900	0.000
277					
278	Intercepts				
279	SAD	0.614	0.022	27.949	0.000
280	BLUES	0.456	0.021	21.564	0.000
281	DEPRESS	0.621	0.022	28.247	0.000
282	HAPPY	0.636	0.022	28.793	0.000
283	ENJOY	0.514	0.021	23.980	0.000
284	HOPEFUL	0.790	0.023	34.108	0.000
285					
286	Residual Variances				
287	SAD	0.525	0.019	27.705	0.000
288	BLUES	0.558	0.019	28.782	0.000
289	DEPRESS	0.453	0.019	23.521	0.000
290	HAPPY	0.598	0.020	29.868	0.000
291	ENJOY	0.632	0.020	31.705	0.000
292	HOPEFUL	0.866	0.015	58.794	0.000
293	F	0.977	0.007	145.590	0.000
294					
295					
296	STD Standardization				
297					
298					Two-Tailed
299		Estimate	S.E.	Est./S.E.	P-Value
300					
301	F	BY			
302	SAD	0.510	0.015	34.965	0.000

303	BLUES	0.491	0.015	33.077	0.000
304	DEPRESS	0.607	0.016	37.766	0.000
305	HAPPY	0.628	0.021	30.604	0.000
306	ENJOY	0.548	0.019	28.920	0.000
307	HOPEFUL	0.451	0.027	16.703	0.000

308					
309	F	ON			
310	AGE	0.011	0.003	3.252	0.001
311	MALE	-0.263	0.045	-5.900	0.000

312					
313	Intercepts				
314	SAD	0.454	0.015	30.494	0.000
315	BLUES	0.337	0.015	22.670	0.000
316	DEPRESS	0.510	0.017	30.881	0.000
317	HAPPY	0.630	0.020	31.599	0.000
318	ENJOY	0.465	0.018	25.530	0.000
319	HOPEFUL	0.973	0.025	39.078	0.000

320					
321	Residual Variances				
322	SAD	0.288	0.011	27.278	0.000
323	BLUES	0.304	0.011	27.713	0.000
324	DEPRESS	0.306	0.013	24.034	0.000
325	HAPPY	0.585	0.021	27.942	0.000
326	ENJOY	0.516	0.018	28.740	0.000
327	HOPEFUL	1.314	0.039	33.541	0.000
328	F	0.977	0.007	145.590	0.000

329					
330					
331	R-SQUARE				

332					
333	Observed				Two-Tailed
334	Variable	Estimate	S.E.	Est./S.E.	P-Value

```

335
336 SAD                0.475      0.019      25.057      0.000
337 BLUES              0.442      0.019      22.804      0.000
338 DEPRESS            0.547      0.019      28.364      0.000
339 HAPPY              0.402      0.020      20.110      0.000
340 ENJOY              0.368      0.020      18.478      0.000
341 HOPEFUL            0.134      0.015       9.091      0.000
342
343 Latent
344 Variable            Estimate      S.E.  Est./S.E.  Two-Tailed
345                                     P-Value
346 F                    0.023      0.007      3.466      0.001
347
348

```

349 TECHNICAL 1 OUTPUT

350

351

352 PARAMETER SPECIFICATION

353

354

355

```

356      NU
357      SAD          BLUES          DEPRESS      HAPPY      ENJOY
358      -----
359      1            2            3            4            5

```

359

360

361

```

362      NU
363      HOPEFUL      AGE          MALE
364      -----
365      6            0            0

```

365

366

```

367          LAMBDA
368          F          AGE          MALE
369          -----
370 SAD              0              0              0
371 BLUES            7              0              0
372 DEPRESS          8              0              0
373 HAPPY            9              0              0
374 ENJOY           10              0              0
375 HOPEFUL          11              0              0
376 AGE              0              0              0
377 MALE             0              0              0
378
379
380          THETA
381          SAD          BLUES          DEPRESS          HAPPY          ENJOY
382          -----
383 SAD              12              13              14              15              16
384 BLUES            0              0              0              0              0
385 DEPRESS          0              0              0              0              0
386 HAPPY            0              0              0              0              0
387 ENJOY            0              0              0              0              0
388 HOPEFUL          0              0              0              0              0
389 AGE              0              0              0              0              0
390 MALE             0              0              0              0              0
391
392
393          THETA
394          HOPEFUL      AGE          MALE
395          -----
396 HOPEFUL          17              0              0
397 AGE              0              0              0
398 MALE             0              0              0

```

ALPHA

F

AGE

MALE

0

0

0

BETA

F

AGE

MALE

0

18

19

AGE

0

0

0

MALE

0

0

0

PSI

F

AGE

MALE

20

F

AGE

0

0

MALE

0

0

0

STARTING VALUES

NU

SAD

BLUES

DEPRESS

HAPPY

ENJOY

0.454

0.337

0.510

0.630

0.465

```

431
432      NU
433      HOPEFUL      AGE      MALE
434      -----
435      0.973      0.000      0.000
436
437
438      LAMBDA
439      F      AGE      MALE
440      -----
441  SAD      1.000      0.000      0.000
442  BLUES      0.991      0.000      0.000
443  DEPRESS      1.218      0.000      0.000
444  HAPPY      1.176      0.000      0.000
445  ENJOY      1.031      0.000      0.000
446  HOPEFUL      0.847      0.000      0.000
447  AGE      0.000      1.000      0.000
448  MALE      0.000      0.000      1.000
449
450
451      THETA
452      SAD      BLUES      DEPRESS      HAPPY      ENJOY
453      -----
454  SAD      0.274
455  BLUES      0.000      0.273
456  DEPRESS      0.000      0.000      0.337
457  HAPPY      0.000      0.000      0.000      0.490
458  ENJOY      0.000      0.000      0.000      0.000      0.408
459  HOPEFUL      0.000      0.000      0.000      0.000      0.000
460  AGE      0.000      0.000      0.000      0.000      0.000
461  MALE      0.000      0.000      0.000      0.000      0.000
462

```

```

463
464      THETA
465      HOPEFUL      AGE      MALE
466      -----
467 HOPEFUL      0.759
468 AGE      0.000      0.000
469 MALE      0.000      0.000      0.000
470
471
472      ALPHA
473      F      AGE      MALE
474      -----
475      0.000      0.000      0.000
476
477
478      BETA
479      F      AGE      MALE
480      -----
481 F      0.000      0.000      0.000
482 AGE      0.000      0.000      0.000
483 MALE      0.000      0.000      0.000
484
485
486      PSI
487      F      AGE      MALE
488      -----
489 F      0.050
490 AGE      0.000      40.696
491 MALE      0.000      -0.190      0.244
492
493

```

Beginning Time: 14:07:22

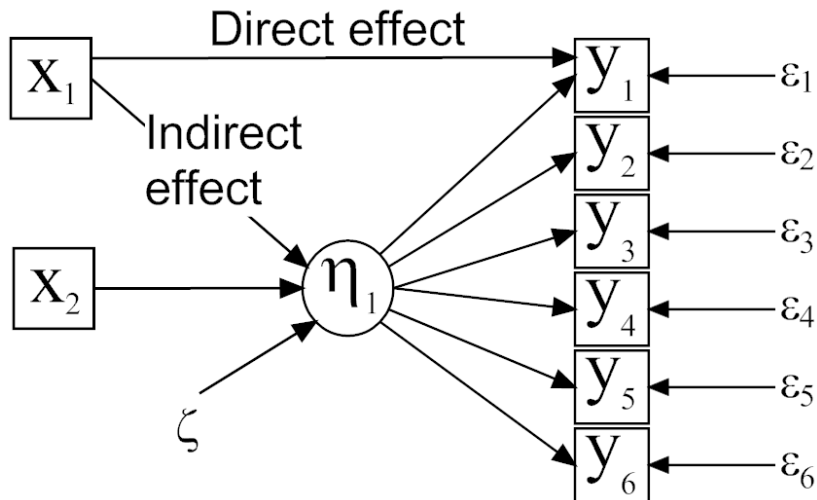
```
495           Ending Time:   14:07:22
496           Elapsed Time:   00:00:00
497
498
499 Mplus VERSION 8.8 DEMO (Mac) has the following limitations:
500   Maximum number of dependent variables: 6
501   Maximum number of independent variables: 2
502   Maximum number of between variables: 2
503   Maximum number of continuous latent variables in time series analysis: 2
504
505
506 MUTHEN & MUTHEN
507 3463 Stoner Ave.
508 Los Angeles, CA 90066
509
510 Tel: (310) 391-9971
511 Fax: (310) 391-8971
512 Web: www.StatModel.com
513 Support: Support@StatModel.com
514
515 Copyright (c) 1998-2022 Muthen & Muthen
```

ex0102.out

Example 1.3

CFA with Covariates and Direct
Effects

CES-D Depression, Age, Sex
New Haven EPESE



```
1 Mplus VERSION 8.8 DEMO (Mac)
2 MUTHEN & MUTHEN
3 07/11/2022    2:07 PM
4
5 INPUT INSTRUCTIONS
6
7   TITLE:      EX0103 CFA WITH COVARIATES AND DIRECT EFFECT EPESE CESD ;
8
9   DATA:      FILE = ex0101.dat ;
10
11  VARIABLE: NAMES = sad blues depress happy enjoy hopeful age male ;
12             MISSING ARE ALL (-9999) ;
13
14  OUTPUT:     TECH1 ;
15
16  MODEL:      f by sad-hopeful ;
17             f on age male ;
18             sad-hopeful ON age-male@0 ;
19             hopeful ON age ;
20
21
22
23
24
25 INPUT READING TERMINATED NORMALLY
26
27
28
29 EX0103 CFA WITH COVARIATES AND DIRECT EFFECT EPESE CESD ;
30
31 SUMMARY OF ANALYSIS
32
```

```

33 Number of groups                                1
34 Number of observations                          2442
35
36 Number of dependent variables                    6
37 Number of independent variables                  2
38 Number of continuous latent variables            1
39
40 Observed dependent variables
41
42   Continuous
43   SAD          BLUES          DEPRESS          HAPPY          ENJOY          HOPEFUL
44
45 Observed independent variables
46   AGE          MALE
47
48 Continuous latent variables
49   F
50
51
52 Estimator                                          ML
53 Information matrix                                OBSERVED
54 Maximum number of iterations                      1000
55 Convergence criterion                            0.500D-04
56 Maximum number of steepest descent iterations    20
57 Maximum number of iterations for H1              2000
58 Convergence criterion for H1                     0.100D-03
59
60 Input data file(s)
61   ex0101.dat
62
63 Input data format  FREE
64

```

```

65
66 SUMMARY OF DATA
67
68     Number of missing data patterns           1
69
70

```

```

71 COVARIANCE COVERAGE OF DATA
72

```

```

73 Minimum covariance coverage value    0.100
74
75

```

```

76     PROPORTION OF DATA PRESENT
77

```

```

78
79     Covariance Coverage
80     SAD          BLUES          DEPRESS          HAPPY          ENJOY
81     -----
82 SAD          1.000
83 BLUES        1.000          1.000
84 DEPRESS      1.000          1.000          1.000
85 HAPPY        1.000          1.000          1.000          1.000
86 ENJOY        1.000          1.000          1.000          1.000          1.000
87 HOPEFUL      1.000          1.000          1.000          1.000          1.000
88 AGE          1.000          1.000          1.000          1.000          1.000
89 MALE         1.000          1.000          1.000          1.000          1.000
90
91

```

```

92     Covariance Coverage
93     HOPEFUL      AGE          MALE
94     -----
95 HOPEFUL      1.000
96 AGE          1.000          1.000

```

```

97 MALE                1.000          1.000          1.000
98
99
100
101 UNIVARIATE SAMPLE STATISTICS
102
103
104     UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS
105
106           Variable/           Mean/           Skewness/           Minimum/ % with
107           Sample Size      Variance      Kurtosis      Maximum      Min/Max      20%/60%      40%/
108           80%      Median
109     SAD                0.454          1.831          0.000      65.77%          0.000
110           0.000          0.000
111           2442.000          0.548          3.185          3.000          3.81%          0.000
112           1.000
113     BLUES                0.337          2.417          0.000      78.05%          0.000
114           0.000          0.000
115           2442.000          0.545          5.259          3.000          4.34%          0.000
116           1.000
117     DEPRESS                0.510          1.749          0.000      64.41%          0.000
118           0.000          0.000
119           2442.000          0.674          2.457          3.000          5.81%          0.000
120           1.000
121     HAPPY                0.630          1.241          0.000      67.08%          0.000
122           0.000          0.000
123           2442.000          0.979          0.069          3.000          7.25%          0.000
124           2.000
125     ENJOY                0.465          1.742          0.000      75.96%          0.000
126           0.000          0.000

```

118		2442.000	0.816	1.624	3.000	5.94%	0.000
		1.000					
119	HOPEFUL		0.973	0.681	0.000	57.70%	0.000
		0.000	0.000				
120		2442.000	1.517	-1.263	3.000	19.57%	1.000
		2.000					
121	AGE		74.316	0.497	67.000	28.50%	67.000
		72.000	72.000				
122		2442.000	40.696	-0.862	87.000	8.39%	77.000
		82.000					
123	MALE		0.424	0.307	0.000	57.58%	0.000
		0.000	0.000				
124		2442.000	0.244	-1.906	1.000	42.42%	1.000
		1.000					

125
126
127 THE MODEL ESTIMATION TERMINATED NORMALLY

128
129
130
131 MODEL FIT INFORMATION

132
133 Number of Free Parameters 21

134
135 Loglikelihood

136
137 H0 Value -17251.875
138 H1 Value -17006.684

139
140 Information Criteria

141
142 Akaike (AIC) 34545.751

```
143          Bayesian (BIC)                34667.563
144          Sample-Size Adjusted BIC      34600.841
145          (n* = (n + 2) / 24)
146
147 Chi-Square Test of Model Fit
148
149          Value                          490.382
150          Degrees of Freedom              18
151          P-Value                        0.0000
152
153 RMSEA (Root Mean Square Error Of Approximation)
154
155          Estimate                       0.104
156          90 Percent C.I.                0.096    0.112
157          Probability RMSEA <= .05       0.000
158
159 CFI/TLI
160
161          CFI                            0.885
162          TLI                            0.827
163
164 Chi-Square Test of Model Fit for the Baseline Model
165
166          Value                          4124.409
167          Degrees of Freedom              27
168          P-Value                        0.0000
169
170 SRMR (Standardized Root Mean Square Residual)
171
172          Value                          0.048
173
174
```

175					
176	MODEL RESULTS				
177					
178				Two-Tailed	
179		Estimate	S.E.	Est./S.E.	P-Value
180					
181	F	BY			
182		SAD	1.000	0.000	999.000
183		BLUES	0.963	0.035	27.792
184		DEPRESS	1.191	0.040	30.113
185		HAPPY	1.230	0.049	25.034
186		ENJOY	1.074	0.045	24.002
187		HOPEFUL	0.872	0.057	15.319
188					
189	F	ON			
190		AGE	0.005	0.002	3.007
191		MALE	-0.134	0.023	-5.779
192					
193	SAD	ON			
194		AGE	0.000	0.000	999.000
195		MALE	0.000	0.000	999.000
196					
197	BLUES	ON			
198		AGE	0.000	0.000	999.000
199		MALE	0.000	0.000	999.000
200					
201	DEPRESS	ON			
202		AGE	0.000	0.000	999.000
203		MALE	0.000	0.000	999.000
204					
205	HAPPY	ON			
206		AGE	0.000	0.000	999.000

207	MALE	0.000	0.000	999.000	999.000
208					
209	ENJOY ON				
210	AGE	0.000	0.000	999.000	999.000
211	MALE	0.000	0.000	999.000	999.000
212					
213	HOPEFUL ON				
214	AGE	0.010	0.004	2.572	0.010
215	MALE	0.000	0.000	999.000	999.000
216					
217	Intercepts				
218	SAD	0.110	0.135	0.818	0.413
219	BLUES	0.006	0.130	0.047	0.963
220	DEPRESS	0.101	0.160	0.627	0.531
221	HAPPY	0.207	0.166	1.245	0.213
222	ENJOY	0.096	0.145	0.657	0.511
223	HOPEFUL	-0.037	0.291	-0.126	0.900
224					
225	Residual Variances				
226	SAD	0.287	0.011	27.267	0.000
227	BLUES	0.304	0.011	27.708	0.000
228	DEPRESS	0.305	0.013	24.009	0.000
229	HAPPY	0.586	0.021	27.941	0.000
230	ENJOY	0.516	0.018	28.747	0.000
231	HOPEFUL	1.312	0.039	33.574	0.000
232	F	0.254	0.015	17.467	0.000
233					
234					

235 QUALITY OF NUMERICAL RESULTS

236

237 Condition Number for the Information Matrix

238 (ratio of smallest to largest eigenvalue)

0.178E-06

239

240

241 TECHNICAL 1 OUTPUT

242

243

244

PARAMETER SPECIFICATION

245

246

247

NU

248

SAD

BLUES

DEPRESS

HAPPY

ENJOY

249

250

0

0

0

0

0

251

252

253

NU

254

HOPEFUL

AGE

MALE

255

256

0

0

0

257

258

259

LAMBDA

260

F

SAD

BLUES

DEPRESS

HAPPY

261

262

SAD

0

0

0

0

0

263

BLUES

0

0

0

0

0

264

DEPRESS

0

0

0

0

0

265

HAPPY

0

0

0

0

0

266

ENJOY

0

0

0

0

0

267

HOPEFUL

0

0

0

0

0

268

AGE

0

0

0

0

0

269

MALE

0

0

0

0

0

270

```

271
272          LAMBDA
273          ENJOY          HOPEFUL          AGE          MALE
274          -----
275 SAD          0          0          0          0
276 BLUES        0          0          0          0
277 DEPRESS      0          0          0          0
278 HAPPY        0          0          0          0
279 ENJOY        0          0          0          0
280 HOPEFUL      0          0          0          0
281 AGE          0          0          0          0
282 MALE         0          0          0          0
283
284
285          THETA
286          SAD          BLUES          DEPRESS          HAPPY          ENJOY
287          -----
288 SAD          0
289 BLUES        0          0
290 DEPRESS      0          0          0
291 HAPPY        0          0          0          0
292 ENJOY        0          0          0          0          0
293 HOPEFUL      0          0          0          0          0
294 AGE          0          0          0          0          0
295 MALE         0          0          0          0          0
296
297
298          THETA
299          HOPEFUL          AGE          MALE
300          -----
301 HOPEFUL      0
302 AGE          0          0

```

```

303 MALE                0                0                0
304
305
306     ALPHA
307     F                SAD                BLUES                DEPRESS                HAPPY
308     -----
309             0                1                2                3                4
310
311
312     ALPHA
313     ENJOY            HOPEFUL            AGE                MALE
314     -----
315             5                6                0                0
316
317
318     BETA
319     F                SAD                BLUES                DEPRESS                HAPPY
320     -----
321 F                0                0                0                0                0
322 SAD                0                0                0                0                0
323 BLUES              9                0                0                0                0
324 DEPRESS            10               0                0                0                0
325 HAPPY              11               0                0                0                0
326 ENJOY              12               0                0                0                0
327 HOPEFUL            13               0                0                0                0
328 AGE                0                0                0                0                0
329 MALE               0                0                0                0                0
330
331
332     BETA
333     ENJOY            HOPEFUL            AGE                MALE
334     -----

```

335	F	0	0	7	8
336	SAD	0	0	0	0
337	BLUES	0	0	0	0
338	DEPRESS	0	0	0	0
339	HAPPY	0	0	0	0
340	ENJOY	0	0	0	0
341	HOPEFUL	0	0	14	0
342	AGE	0	0	0	0
343	MALE	0	0	0	0
344					
345					
346	PSI				
347	F	SAD	BLUES	DEPRESS	HAPPY
348	-----	-----	-----	-----	-----
349	F	15			
350	SAD	0	16		
351	BLUES	0	0	17	
352	DEPRESS	0	0	0	18
353	HAPPY	0	0	0	0
354	ENJOY	0	0	0	0
355	HOPEFUL	0	0	0	0
356	AGE	0	0	0	0
357	MALE	0	0	0	0
358					
359					
360	PSI				
361	ENJOY	HOPEFUL	AGE	MALE	
362	-----	-----	-----	-----	
363	ENJOY	20			
364	HOPEFUL	0	21		
365	AGE	0	0		
366	MALE	0	0	0	

STARTING VALUES

NU

SAD

BLUES

DEPRESS

HAPPY

ENJOY

0.000

0.000

0.000

0.000

0.000

NU

HOPEFUL

AGE

MALE

0.000

0.000

0.000

LAMBDA

F

SAD

BLUES

DEPRESS

HAPPY

0.000

1.000

0.000

0.000

0.000

0.000

0.000

1.000

0.000

0.000

0.000

0.000

0.000

1.000

0.000

0.000

0.000

0.000

0.000

1.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

0.000

LAMBDA

ENJOY

HOPEFUL

AGE

MALE

```

399
400 SAD          0.000      0.000      0.000      0.000
401 BLUES        0.000      0.000      0.000      0.000
402 DEPRESS      0.000      0.000      0.000      0.000
403 HAPPY        0.000      0.000      0.000      0.000
404 ENJOY        1.000      0.000      0.000      0.000
405 HOPEFUL      0.000      1.000      0.000      0.000
406 AGE          0.000      0.000      1.000      0.000
407 MALE         0.000      0.000      0.000      1.000
408
409
410 THETA
411 SAD          BLUES      DEPRESS    HAPPY      ENJOY
412
413 SAD          0.000
414 BLUES        0.000      0.000
415 DEPRESS      0.000      0.000      0.000
416 HAPPY        0.000      0.000      0.000      0.000
417 ENJOY        0.000      0.000      0.000      0.000      0.000
418 HOPEFUL      0.000      0.000      0.000      0.000      0.000
419 AGE          0.000      0.000      0.000      0.000      0.000
420 MALE         0.000      0.000      0.000      0.000      0.000
421
422
423 THETA
424 HOPEFUL      AGE        MALE
425
426 HOPEFUL      0.000
427 AGE          0.000      0.000
428 MALE         0.000      0.000
429
430

```

431	ALPHA				
432	F	SAD	BLUES	DEPRESS	HAPPY
433	-----	-----	-----	-----	-----
434	0.000	0.454	0.337	0.510	0.630
435					
436					
437	ALPHA				
438	ENJOY	HOPEFUL	AGE	MALE	
439	-----	-----	-----	-----	
440	0.465	0.973	74.316	0.424	
441					
442					
443	BETA				
444	F	SAD	BLUES	DEPRESS	HAPPY
445	-----	-----	-----	-----	-----
446	F	0.000	0.000	0.000	0.000
447	SAD	1.000	0.000	0.000	0.000
448	BLUES	0.991	0.000	0.000	0.000
449	DEPRESS	1.218	0.000	0.000	0.000
450	HAPPY	1.176	0.000	0.000	0.000
451	ENJOY	1.031	0.000	0.000	0.000
452	HOPEFUL	0.847	0.000	0.000	0.000
453	AGE	0.000	0.000	0.000	0.000
454	MALE	0.000	0.000	0.000	0.000
455					
456					
457	BETA				
458	ENJOY	HOPEFUL	AGE	MALE	
459	-----	-----	-----	-----	
460	F	0.000	0.000	0.000	
461	SAD	0.000	0.000	0.000	
462	BLUES	0.000	0.000	0.000	

463	DEPRESS	0.000	0.000	0.000	0.000
464	HAPPY	0.000	0.000	0.000	0.000
465	ENJOY	0.000	0.000	0.000	0.000
466	HOPEFUL	0.000	0.000	0.000	0.000
467	AGE	0.000	0.000	0.000	0.000
468	MALE	0.000	0.000	0.000	0.000
469					
470					
471	PSI				
472	F	SAD	BLUES	DEPRESS	HAPPY
473					
474	F	0.050			
475	SAD	0.000	0.274		
476	BLUES	0.000	0.000	0.273	
477	DEPRESS	0.000	0.000	0.000	0.337
478	HAPPY	0.000	0.000	0.000	0.490
479	ENJOY	0.000	0.000	0.000	0.000
480	HOPEFUL	0.000	0.000	0.000	0.000
481	AGE	0.000	0.000	0.000	0.000
482	MALE	0.000	0.000	0.000	0.000
483					
484					
485	PSI				
486	ENJOY	HOPEFUL	AGE	MALE	
487					
488	ENJOY	0.408			
489	HOPEFUL	0.000	0.759		
490	AGE	0.000	0.000	40.696	
491	MALE	0.000	0.000	-0.190	0.244
492					
493					
494	Beginning Time:	14:07:23			

Example 1.4

Growth modeling

Alcohol Use from Singer and Willet
(2003)

Model a: Intercept only model

<https://stats.idre.ucla.edu/other/examples/alda/>

https://stats.idre.ucla.edu/other/examples/alda/



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TEXTBOOK EXAMPLES APPLIED LONGITUDINAL DATA ANALYSIS: MODELING CHANGE AND EVENT OCCURRENCE BY JUDITH D. SINGER AND JOHN B. WILLETT

This is one of the books available for loan from Academic Technology Services (see [Statistics Books for Loan](#) for other such books, and details about borrowing). We encourage you to obtain *Applied Longitudinal Data Analysis*, written by [Judith D. Singer and John B. Willett](#), published by the [Oxford University Press](#), to gain a deeper conceptual understanding of the analysis illustrated (see [Where to buy books](#) for tips on different places you can buy this book). We are very grateful to the authors for granting us permission to create these pages and to distribute the data files via our web pages. (In addition to the data formats below, you can also download the data files as [comma separated text files](#)).

Note: The files for MLwiN here were created for older version of MLwiN (version 2.0 and older). The newer version of MLwiN is compatible with these files. But the newer version of the same files created at MLwiN site will only work for newer version of MLwiN (version 2.10). Please visit [MLwiN site](#) to download the files and for other MLwiN related information.

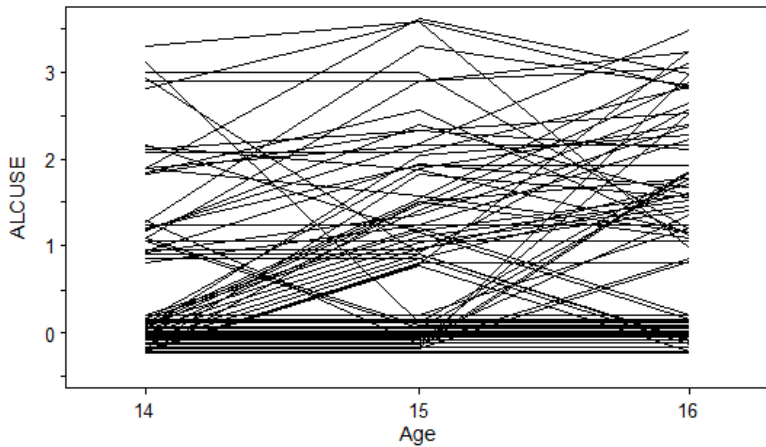
Note: The files for HLM consists of two types of files, the data files in SPSS format and the corresponding .mdmt (template) file for creating the binary HLM files. (Updated in January, 2012).

<https://stats.idre.ucla.edu/r/examples/alda/applied-longitudinal-data-analysis-chapter-15-r-textbook-examples/>

	Mplus	SAS	Stata	R	SPSS	Chapter Title
Download Data	Download	Download	Download	Download	Download	Table of Contents
Chapter 1						A framework for investigating change over time
Chapter 2		Chap 2	Chap 2	Chap 2	Chap 2	Exploring longitudinal data on change
Chapter 3		Chap 3	Chap 3	Chap 3	Chap 3	Introducing the multilevel model for change

Chapter 4. Doing Data Analysis with the Multilevel Model for Change. Singer and Willet Applied Longitudinal Data Analysis. Oxford. (2003)

- Adolescent Alcohol Use Example (Curran et al. 2002).
- 82 Adolescents, age 14, seen 3 times +1 year
- 4 item Questionnaire Alcohol use past year
 - ▶ (1) drank beer or wine, (2) drank liquor, (3) had 5+ drinks in a row, (4) got drunk
 - ▶ rated 0-7 (0=not at all, 7=every day)
 - ▶ $\text{alcuse} = (\text{Sy})^{0.5}$
- Covariates/predictors
 - ▶ coa : child of alcoholic (0/1)
 - ▶ peer : alcohol use (alcuse) among peers [0-2.53]



data from Singer and Willet (2003) Chapter 4

http://www.ats.ucla.edu/stat/stata/examples/alda/data/alcohol1_pp.dta

Random noise added to ALCUSE, between $\pm 0.5 \times \text{SD}(\text{ALCUSE}|\text{AGE}_{14})$

Table 4.1: Alcohol use data (alcohol1_pp): Fitted multilevel models for change

Model A

Covariance Parameter Estimates

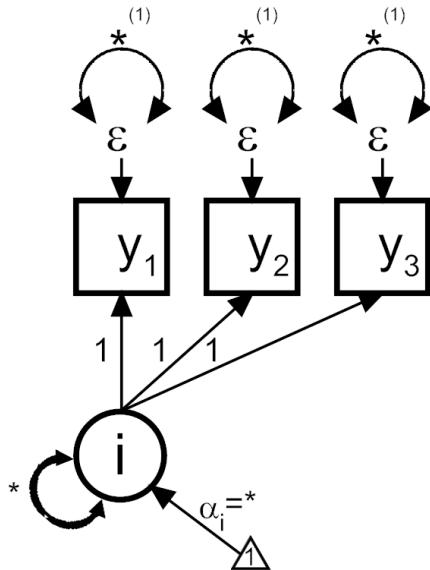
Cov Parm	Subject	Estimate	Standard Error	Z Value	Pr > Z
Intercept	ID	0.5639	0.1191	4.73	<.0001
Residual		0.5617	0.06203	9.06	<.0001

Fit Statistics

-2 Log Likelihood	670.2
AIC (smaller is better)	676.2
AICC (smaller is better)	676.3
BIC (smaller is better)	683.4

Solution for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	0.9220	0.09571	81	9.63	<.0001



```
1 Mplus VERSION 8.8 DEMO (Mac)
2 MUTHEN & MUTHEN
3 07/11/2022    2:07 PM
4
5 INPUT INSTRUCTIONS
6
7   TITLE:      EX0104 Singer and Willet (2003) Chapter 4 Model A ;
8
9   DATA:      FILE = swch4.dat ;
10
11  VARIABLE: NAMES =   alcuse1 alcuse2 alcuse3 ;
12
13  ANALYSIS: ESTIMATOR = mlr ;
14
15  MODEL:      i by alcuse1-alcuse3@1 ;
16              [alcuse1-alcuse3*] ; ! force mean to intercept
17              ![i*] ;              ! intercept
18              alcuse1-alcuse3 (1) ; ! homoscedasticity
19
20
21
22
23
24 INPUT READING TERMINATED NORMALLY
25
26
27
28 EX0104 Singer and Willet (2003) Chapter 4 Model A ;
29
30 SUMMARY OF ANALYSIS
31
32 Number of groups
```

```

33 Number of observations                        82
34
35 Number of dependent variables                3
36 Number of independent variables              0
37 Number of continuous latent variables        1
38
39 Observed dependent variables
40
41   Continuous
42   ALCUSE1      ALCUSE2      ALCUSE3
43
44 Continuous latent variables
45   I
46
47
48 Estimator                                MLR
49 Information matrix                        OBSERVED
50 Maximum number of iterations              1000
51 Convergence criterion                     0.500D-04
52 Maximum number of steepest descent iterations 20
53
54 Input data file(s)
55   swch4.dat
56
57 Input data format   FREE
58
59
60
61 UNIVARIATE SAMPLE STATISTICS
62
63
64   UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

```

Variable/ Percentiles	Mean/ Variance	Skewness/ Kurtosis	Minimum/ Maximum	% with Min/Max	20%/60%	40%/
Sample Size 80% Median						
ALCUSE1	0.630	1.320	0.000	62.20%	0.000	
0.000 0.000						
82.000	0.871	0.650	3.317	1.22%	0.000	
1.414						
ALCUSE2	0.964	0.831	0.000	43.90%	0.000	
0.000 1.000						
82.000	1.084	-0.187	3.606	2.44%	1.000	
2.000						
ALCUSE3	1.172	0.385	0.000	40.24%	0.000	
0.000 1.000						
82.000	1.273	-1.196	3.464	1.22%	1.414	
2.236						

77 THE MODEL ESTIMATION TERMINATED NORMALLY

81 MODEL FIT INFORMATION

83 Number of Free Parameters 5

85 Loglikelihood

87 H0 Value -323.402

88 H0 Scaling Correction Factor 1.0942

```

89           for MLR
90       H1 Value                               -316.186
91       H1 Scaling Correction Factor           1.2269
92           for MLR
93
94 Information Criteria
95
96       Akaike (AIC)                           656.804
97       Bayesian (BIC)                         668.837
98       Sample-Size Adjusted BIC               653.067
99       (n* = (n + 2) / 24)
100
101 Chi-Square Test of Model Fit
102
103       Value                                   10.361*
104       Degrees of Freedom                     4
105       P-Value                               0.0348
106       Scaling Correction Factor              1.3928
107           for MLR
108
109 *   The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used
110       for chi-square difference testing in the regular way.  MLM, MLR and WLSM
111       chi-square difference testing is described on the Mplus website.  MLMV, WLSMV,
112       and ULSMV difference testing is done using the DIFFTEST option.
113
114 RMSEA (Root Mean Square Error Of Approximation)
115
116       Estimate                               0.139
117       90 Percent C.I.                       0.034  0.247
118       Probability RMSEA <= .05               0.070
119
120 CFI/TLI

```

```

121
122             CFI                               0.848
123             TLI                               0.886
124
125 Chi-Square Test of Model Fit for the Baseline Model
126
127             Value                               44.816
128             Degrees of Freedom                   3
129             P-Value                             0.0000
130
131 SRMR (Standardized Root Mean Square Residual)
132
133             Value                               0.117
134
135
136
137 MODEL RESULTS
138
139
140             Estimate      S.E.   Est./S.E.   Two-Tailed
141                                     P-Value
142 I             BY
143   ALCUSE1      1.000      0.000    999.000    999.000
144   ALCUSE2      1.000      0.000    999.000    999.000
145   ALCUSE3      1.000      0.000    999.000    999.000
146
147 Intercepts
148   ALCUSE1      0.630      0.103     6.118     0.000
149   ALCUSE2      0.964      0.115     8.383     0.000
150   ALCUSE3      1.172      0.125     9.403     0.000
151
152 Variances

```

153 I 0.589 0.103 5.714 0.000

154

155 Residual Variances

156 ALCUSE1 0.487 0.071 6.845 0.000

157 ALCUSE2 0.487 0.071 6.845 0.000

158 ALCUSE3 0.487 0.071 6.845 0.000

159

160

161 QUALITY OF NUMERICAL RESULTS

162

163 Condition Number for the Information Matrix 0.102E+00

164 (ratio of smallest to largest eigenvalue)

165

166

167 Beginning Time: 14:07:23

168 Ending Time: 14:07:23

169 Elapsed Time: 00:00:00

170

171

172 Mplus VERSION 8.8 DEMO (Mac) has the following limitations:

173 Maximum number of dependent variables: 6

174 Maximum number of independent variables: 2

175 Maximum number of between variables: 2

176 Maximum number of continuous latent variables in time series analysis: 2

177

178

179 MUTHEN & MUTHEN

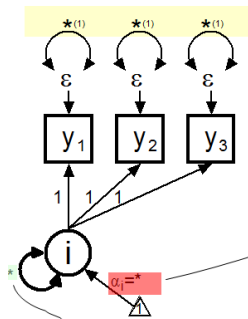
180 3463 Stoner Ave.

181 Los Angeles, CA 90066

182

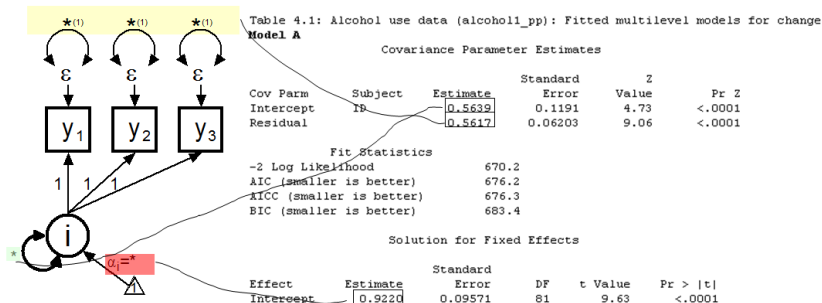
183 Tel: (310) 391-9971

184 Fax: (310) 391-8971



MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
I BY				
ALCUSE1	1.000	0.000	999.000	999.000
ALCUSE2	1.000	0.000	999.000	999.000
ALCUSE3	1.000	0.000	999.000	999.000
Means				
I	0.922	0.096	9.633	0.000
Intercepts				
ALCUSE1	0.000	0.000	999.000	999.000
ALCUSE2	0.000	0.000	999.000	999.000
ALCUSE3	0.000	0.000	999.000	999.000
Variances				
I	0.564	0.106	5.314	0.000
Residual Variances				
ALCUSE1	0.562	0.083	6.788	0.000
ALCUSE2	0.562	0.083	6.788	0.000
ALCUSE3	0.562	0.083	6.788	0.000



Note that the point estimates are the same. Standard errors of covariance parameters are slightly higher in PROC MIXED. It's a difference of N-1 (PROC Mixed) vs N in the denominator.

SAS Proc Mixed $\approx$ *Mplus*

intercept $\approx$ Means(I)

Cov parm intercept $\approx$ Variance(I)

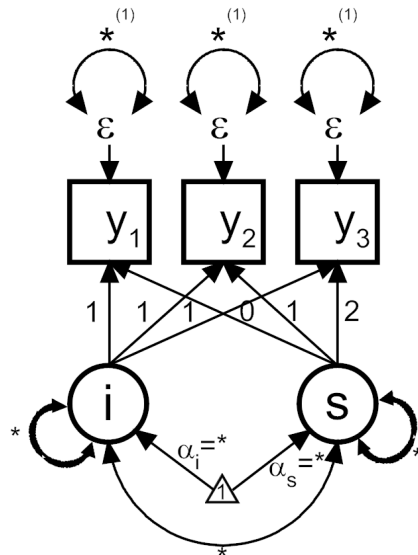
Cov parm residual $\approx$ Residual Variances(y)

Example 1.5

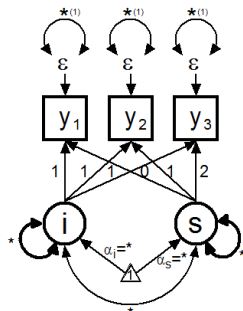
Growth modeling

Alcohol Use from Singer and Willet
(2003)

Model b: Intercept and Slope model



Model B (Table 4.1)



```

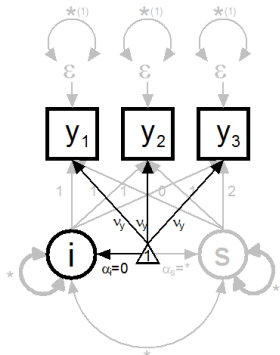
* SAS Example from http://www.ats.ucla.edu/stat/examples/alda.htm
*-----
* SAS
proc mixed data="c:\alda\alcohol1_pp" method=ml noclprint noinfo covtest;
  title2 "Model B";
  class id;
  model alcuse = age_14/solution notest;
  random intercept age_14/type=un sub=id;
*-----

* Mplus (short hand)
DATA:      FILE = C:\work\Shows\SHORTC~1\2009\data\swch4.dat ;
VARIABLE:  NAMES = alcuse1 alcuse2 alcuse3 ;
ANALYSIS:  ESTIMATOR = mlr ;
MODEL:     i s | alcuse1@0 alcuse2@1 alcuse3@2 ;
           alcuse1-alcuse3 (1) ;
*-----

* Mplus (long hand)
DATA:      FILE = C:\work\Shows\SHORTC~1\2009\data\swch4.dat ;
VARIABLE:  NAMES = alcuse1 alcuse2 alcuse3 ;
ANALYSIS:  ESTIMATOR = mlr ;
MODEL:     i by alcuse1-alcuse3@1 ;
           s by alcuse1@0 alcuse2@1 alcuse3@2 ;
           [alcuse1-alcuse3@0] ;
           [i* s*] ;
           alcuse1-alcuse3 (1) ;

```

Model B (Table 4.1) Alternative Mplus Specification



note: $v_y = v_{y(1)} = v_{y(2)} = v_{y(3)}$

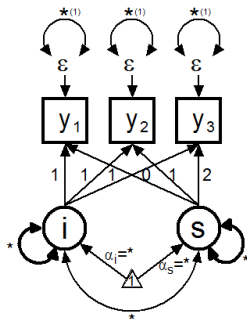
* Mplus (alternate specification - long hand)

DATA: FILE = swch4.dat ;

VARIABLE: NAMES = alcuse1 alcuse2 alcuse3 ;

ANALYSIS: ESTIMATOR = mlr ;

MODEL: alcuse1-alcuse3 (1) ;
 i by alcuse1-alcuse3@1 ;
 s by alcuse1@0 alcuse2@1 alcuse3@2 ;
 [alcuse1-alcuse3*0] (11) ;
 [i@0 s*] ;



$$V(y_{ti}) = V(i)_i + V(s)_i t^2 + 2COV(i,s) + V(\epsilon_{ti})$$

random slope contributes to increasing variance over time even if $\epsilon_t = \epsilon$

```
1 Mplus VERSION 8.8 DEMO (Mac)
2 MUTHEN & MUTHEN
3 07/11/2022    2:07 PM
4
5 INPUT INSTRUCTIONS
6
7 TITLE:      EX0105 Singer and Willet (2003) Chapter 4 Model B ;
8
9 DATA:      FILE = swch4.dat ;
10
11 VARIABLE: NAMES =  alcuse1 alcuse2 alcuse3 ;
12               missing are all (-9999) ;
13
14 ANALYSIS: ESTIMATOR = mlr ;
15
16 OUTPUT:    SAMPSTAT ; stdyx ;
17
18 MODEL:
19           ![alcuse1@0 alcuse2@0 alcuse3@0] ;
20           !i by alcuse1@1 alcuse2@1 alcuse3@1 ;
21           !s by alcuse1@0 alcuse2@1 alcuse3@2 ;
22           ![i*]; [s*];
23           i s | alcuse1@0 alcuse2@1 alcuse3@2 ;
24           alcuse1-alcuse3 (1) ; ! homoscedasticity
25
26
27
28
29
30 INPUT READING TERMINATED NORMALLY
31
32
```

```

33
34 EX0105 Singer and Willet (2003) Chapter 4 Model B ;
35
36 SUMMARY OF ANALYSIS
37
38 Number of groups                                1
39 Number of observations                          82
40
41 Number of dependent variables                    3
42 Number of independent variables                  0
43 Number of continuous latent variables            2
44
45 Observed dependent variables
46
47   Continuous
48   ALCUSE1      ALCUSE2      ALCUSE3
49
50 Continuous latent variables
51   I            S
52
53
54 Estimator                                MLR
55 Information matrix                        OBSERVED
56 Maximum number of iterations              1000
57 Convergence criterion                     0.500D-04
58 Maximum number of steepest descent iterations      20
59 Maximum number of iterations for H1              2000
60 Convergence criterion for H1                0.100D-03
61
62 Input data file(s)
63   swch4.dat
64

```

```

65 Input data format  FREE
66
67
68 SUMMARY OF DATA
69
70     Number of missing data patterns           1
71
72
73 COVARIANCE COVERAGE OF DATA
74
75 Minimum covariance coverage value    0.100
76
77
78     PROPORTION OF DATA PRESENT
79
80
81           Covariance Coverage
82           ALCUSE1      ALCUSE2      ALCUSE3
83           -----
84 ALCUSE1      1.000
85 ALCUSE2      1.000      1.000
86 ALCUSE3      1.000      1.000      1.000
87
88
89 SAMPLE STATISTICS
90
91
92     ESTIMATED SAMPLE STATISTICS
93
94
95           Means
96           ALCUSE1      ALCUSE2      ALCUSE3

```

97		-----	-----	-----
98		0.630	0.964	1.172
99				
100				
101		Covariances		
102		ALCUSE1	ALCUSE2	ALCUSE3
103		-----	-----	-----
104	ALCUSE1	0.871		
105	ALCUSE2	0.620	1.084	
106	ALCUSE3	0.432	0.714	1.273
107				
108				
109		Correlations		
110		ALCUSE1	ALCUSE2	ALCUSE3
111		-----	-----	-----
112	ALCUSE1	1.000		
113	ALCUSE2	0.638	1.000	
114	ALCUSE3	0.411	0.608	1.000
115				

MAXIMUM LOG-LIKELIHOOD VALUE FOR THE UNRESTRICTED (H1) MODEL IS -316.186

UNIVARIATE SAMPLE STATISTICS

UNIVARIATE HIGHER-ORDER MOMENT DESCRIPTIVE STATISTICS

Variable/ Percentiles	Mean/ Variance	Skewness/ Kurtosis	Minimum/ Maximum	% with Min/Max	20%/60%	40%/
Sample Size						
80% Median						

127							
128	ALCUSE1		0.630	1.320	0.000	62.20%	0.000
	0.000	0.000					
129		82.000	0.871	0.650	3.317	1.22%	0.000
		1.414					
130	ALCUSE2		0.964	0.831	0.000	43.90%	0.000
	0.000	1.000					
131		82.000	1.084	-0.187	3.606	2.44%	1.000
		2.000					
132	ALCUSE3		1.172	0.385	0.000	40.24%	0.000
	0.000	1.000					
133		82.000	1.273	-1.196	3.464	1.22%	1.414
		2.236					

134
135
136 THE MODEL ESTIMATION TERMINATED NORMALLY

137
138
139
140 MODEL FIT INFORMATION

141
142 Number of Free Parameters 6

143
144 Loglikelihood

145		
146	H0 Value	-318.306
147	H0 Scaling Correction Factor	1.1192
148	for MLR	
149	H1 Value	-316.186
150	H1 Scaling Correction Factor	1.2269
151	for MLR	

152

153 Information Criteria

154			
155	Akaike (AIC)	648.611	
156	Bayesian (BIC)	663.051	
157	Sample-Size Adjusted BIC	644.128	
158	(n* = (n + 2) / 24)		

159

160 Chi-Square Test of Model Fit

161			
162	Value	2.939*	
163	Degrees of Freedom	3	
164	P-Value	0.4011	
165	Scaling Correction Factor	1.4422	
166	for MLR		

167

168 * The chi-square value for MLM, MLMV, MLR, ULSMV, WLSM and WLSMV cannot be used
 169 for chi-square difference testing in the regular way. MLM, MLR and WLSM
 170 chi-square difference testing is described on the Mplus website. MLMV, WLSMV,
 171 and ULSMV difference testing is done using the DIFFTEST option.

172

173 RMSEA (Root Mean Square Error Of Approximation)

174			
175	Estimate	0.000	
176	90 Percent C.I.	0.000	0.185
177	Probability RMSEA <= .05	0.490	

178

179 CFI/TLI

180		
181	CFI	1.000
182	TLI	1.000

183

184 Chi-Square Test of Model Fit for the Baseline Model

```

185
186           Value                      44.816
187           Degrees of Freedom          3
188           P-Value                     0.0000
189
190 SRMR (Standardized Root Mean Square Residual)
191
192           Value                      0.055
193
194
195
196 MODEL RESULTS
197
198                                     Two-Tailed
199                                     P-Value
200           Estimate      S.E.  Est./S.E.
201 I           |
202   ALCUSE1      1.000      0.000    999.000    999.000
203   ALCUSE2      1.000      0.000    999.000    999.000
204   ALCUSE3      1.000      0.000    999.000    999.000
205
206 S           |
207   ALCUSE1      0.000      0.000    999.000    999.000
208   ALCUSE2      1.000      0.000    999.000    999.000
209   ALCUSE3      2.000      0.000    999.000    999.000
210
211 S           WITH
212   I      -0.068      0.074    -0.920      0.357
213
214 Means
215   I      0.651      0.105      6.198      0.000
216   S      0.271      0.062      4.334      0.000

```

```

217
218 Intercepts
219     ALCUSE1      0.000      0.000      999.000      999.000
220     ALCUSE2      0.000      0.000      999.000      999.000
221     ALCUSE3      0.000      0.000      999.000      999.000
222
223 Variances
224     I      0.624      0.162      3.861      0.000
225     S      0.151      0.056      2.721      0.007
226
227 Residual Variances
228     ALCUSE1      0.337      0.066      5.105      0.000
229     ALCUSE2      0.337      0.066      5.105      0.000
230     ALCUSE3      0.337      0.066      5.105      0.000
231
232

```

233 QUALITY OF NUMERICAL RESULTS

```

234
235     Condition Number for the Information Matrix      0.422E-01
236     (ratio of smallest to largest eigenvalue)
237
238

```

239 STANDARDIZED MODEL RESULTS

```

240
241
242 STDYX Standardization
243
244
245           Estimate      S.E.  Est./S.E.      Two-Tailed
246                               P-Value
247 I      |
248     ALCUSE1      0.806      0.046      17.355      0.000

```

249	ALCUSE2	0.800	0.083	9.686	0.000
250	ALCUSE3	0.695	0.092	7.552	0.000
251					
252	S				
253	ALCUSE1	0.000	0.000	999.000	999.000
254	ALCUSE2	0.394	0.080	4.938	0.000
255	ALCUSE3	0.684	0.120	5.679	0.000
256					
257	S				
258	I	-0.223	0.193	-1.154	0.249
259					
260	Means				
261	I	0.824	0.097	8.465	0.000
262	S	0.696	0.179	3.891	0.000
263					
264	Intercepts				
265	ALCUSE1	0.000	0.000	999.000	999.000
266	ALCUSE2	0.000	0.000	999.000	999.000
267	ALCUSE3	0.000	0.000	999.000	999.000
268					
269	Variances				
270	I	1.000	0.000	999.000	999.000
271	S	1.000	0.000	999.000	999.000
272					
273	Residual Variances				
274	ALCUSE1	0.351	0.075	4.688	0.000
275	ALCUSE2	0.346	0.062	5.588	0.000
276	ALCUSE3	0.261	0.058	4.486	0.000
277					
278					
279	R-SQUARE				
280					

Observed					Two-Tailed
Variable	Estimate	S.E.	Est./S.E.		P-Value
ALCUSE1	0.649	0.075	8.678		0.000
ALCUSE2	0.654	0.062	10.581		0.000
ALCUSE3	0.739	0.058	12.706		0.000

Beginning Time: 14:07:23
 Ending Time: 14:07:23
 Elapsed Time: 00:00:00

Mplus VERSION 8.8 DEMO (Mac) has the following limitations:

Maximum number of dependent variables: 6
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 Los Angeles, CA 90066

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 Web: www.StatModel.com
 Support: Support@StatModel.com

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Table 4.1: Alcohol use data (alcohol1_pp): Fitted multilevel models for change Model B

Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error	Z Value	Pr Z
UN(1,1)	ID	0.6244	0.1481	4.22	<.0001
UN(2,1)	ID	-0.06844	0.07008	-0.98	0.3288
UN(2,2)	ID	0.1512	0.05647	2.68	0.0037
Residual		0.3373	0.05268	6.40	<.0001

Fit Statistics

-2 Log Likelihood	636.6
AIC (smaller is better)	648.6
AICC (smaller is better)	649.0
BIC (smaller is better)	663.1

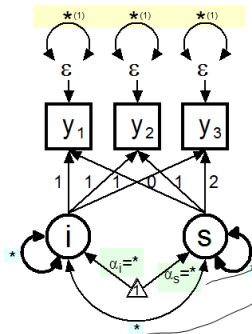
Null Model Likelihood Ratio Test

DF	Chi-Square	Pr > ChiSq
3	79.70	<.0001

Solution for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	0.6513	0.1051	81	6.20	<.0001
AGE_14	0.2707	0.06245	81	4.33	<.0001

MODEL RESULTS



	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
S				
I	-0.068	0.074	-0.920	0.357
Means				
I	0.651	0.105	6.198	0.000
S	0.271	0.062	4.334	0.000
Intercepts				
ALCUSE1	0.000	0.000	999.000	999.000
ALCUSE2	0.000	0.000	999.000	999.000
ALCUSE3	0.000	0.000	999.000	999.000
Variances				
I	0.624	0.162	3.861	0.000
S	0.151	0.056	2.721	0.007
Residual Variances				
ALCUSE1	0.337	0.066	5.105	0.000
ALCUSE2	0.337	0.066	5.105	0.000
ALCUSE3	0.337	0.066	5.105	0.000

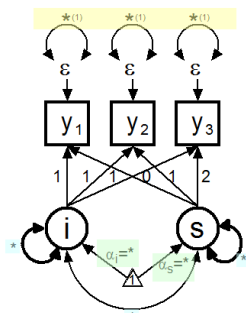


Table 4.1: Alcohol use data (alcohol1_pp): Fitted multilevel models for change
Model B

Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error	Z Value	Pr > Z
UN(1,1)	ID	0.6244	0.1481	4.22	<.0001
UN(2,1)	ID	-0.06844	0.07008	-0.98	0.3288
UN(2,2)	ID	0.1512	0.05647	2.68	0.0037
Residual		0.3373	0.05268	6.40	<.0001

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DF	Chi-Square	Pr > ChiSq
3	79.70	<.0001

Solution for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	0.6513	0.1051	81	6.20	<.0001
AGE_14	0.2707	0.06245	81	4.33	<.0001

SAS Proc Mixed $\approx$ *Mplus*

Intercept $\approx$ Means(I)

AGE_14 $\approx$ Means(S)

Cov parm UN(1,1) $\approx$ Variance(I)

Cov parm UN(2,1) $\approx$ S WITH I

Cov parm UN(2,2) $\approx$ Variance(S)

Cov parm residual $\approx$ Residual Variances(y)

LAB SESSION Alcohol Use LGC Models

Questions?

END OF DAY ONE