

An Introduction to Structural Equation Modeling

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A. What is SEM?

- It is a method to provide quantitative tests of theoretical models.
- The researcher hypothesizes certain **relationships** among a set of variables and then empirically tests those relationships.

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- A few quick and very basic examples from education, marketing and health care:

- Home environment → achievement
- Consumer trust → product sales
- Diet → Risk of heart attack
- Exercise →

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Bottom line question:

- Is the theoretical model supported by the sample data?
- The hypothesized model is based on theory and previous research.

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B. Types of variables

- 1. **Latent variables** (constructs / factors): these are variables that are not directly observed or measured; they are indirectly observed or measured, and thus are inferred from a set of observed variables that are actually measured using tests, surveys, observations, interviews, etc.
- examples – intelligence (psychological construct), consumer confidence (economic construct), fitness (health-related construct).

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- 2. **Observed variables** (indicators / measures): a set of measured variables that are used to define or infer the latent variables in a particular way.
- examples: WISC (measure of intelligence), Dow-Jones industrial average (measure of economy), blood pressure (measure of fitness).

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- 3. **Independent variables:** variables that **are not** influenced by any other variable in the model.

- 4. **Dependent variables:** variables that **are** influenced by some other variable in the model.

- Home environment → achievement

- Consumer trust → product sales

- Diet → Risk of heart attack
- Exercise →

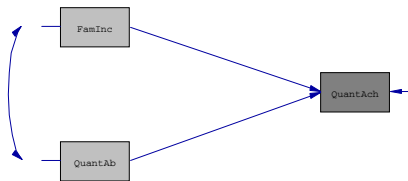
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C. Most basic types of models

- 1. Regression models
- 2. Path models
- 3. Confirmatory factor analysis models
- 4. Structural equation analysis models
- 5. More complex models briefly described later

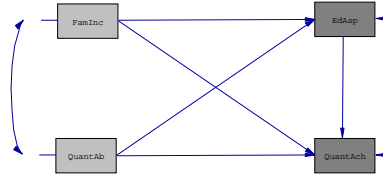
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- 1. **Regression models** – an observed variable model with a single equation; thus one dependent variable is explained or predicted by one or more independent variables; only simple theoretical models are possible, but does not deal with measurement error.



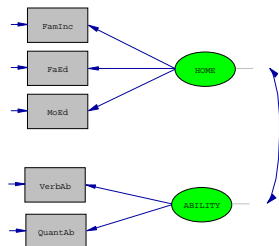
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- 2. **Path models** – an observed variable model with multiple equations; thus allows for multiple independent variables and multiple dependent variables; thus more complex theoretical models are possible, but does not deal with measurement error.



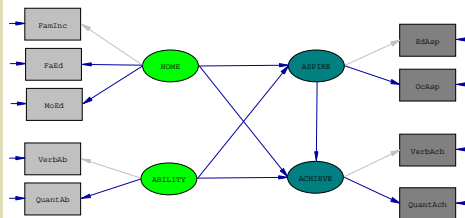
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- 3. **Confirmatory factor analysis (CFA) models** – a latent variable model with multiple observed variables measuring each latent variable or factor; should lead to better assessment of the latent variables, as it can deal with measurement error.



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- 4. **Structural equation analysis models** – a latent variable model which combines a confirmatory factor analysis model (latent variables assessed through multiple observed variables) and a path model (latent variables are related to one another); thus includes all 4 kinds of variables.



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D. Why conduct SEM?

- 1. Researchers need to use multiple variables to better understand an area of inquiry. As theories have become more sophisticated, basic methods that are limited to a small number of variables are no longer sufficient. SEM has become the preferred method for confirming (or disconfirming) theoretical models in a quantitative fashion.
- 2. Recognizes the effect of measurement error (which SEM explicitly takes into account). Greater recognition is given to the validity and reliability of the observed variable scores obtained from measurement instruments.

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- 3. More advanced SEM models are possible after 40 years of statistical and software development (e.g., models that involve multiple groups, multiple levels, and interactions).
- 4. SEM software more user-friendly than ever.

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E. Most popular SEM software

- AMOS (add-on to SPSS).
- EQS (stand-alone program).
- LISREL (stand-alone program).
- Mplus (stand-alone program).

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F. Steps in the SEM Process

1. Model Specification

- The researcher develops and hypothesizes a theoretical model (or models) from available theory and research.
- This is the “hard part” of SEM.
- This is done prior to any data analysis (and preferably prior to data collection and even prior to the selection of measures).

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- The components of model specification include:
- A) deciding which variables to include in the model and how these variables are related to one another.
- B) the sample covariance matrix is generated by true population model (which is unknown); the researcher wants to determine which model actually generated this matrix and thus wants to find that model which best reproduces it.

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- C) The tested model is properly specified when the true population model that generated this matrix is consistent with the model actually tested.
- D) The tested model is misspecified when the true population model that generated this matrix is not consistent with the model actually tested.
- E) Misspecified models can be due to errors of omission and/or inclusion of any variable and/or parameter.

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- F) Misspecified models can result in biased parameter estimates (known as misspecification error); also the misspecified model may not fit the data according to global model fit indices.
- G) All models are misspecified to some degree.

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2. Model Identification

- Often called solving “the identification problem.”
- Given the sample covariance matrix and the theoretical model to be tested, can a unique set of parameter estimates be found?
- In other words, can we obtain a unique value for every parameter estimate? For example, $X + Y = 75$.

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- Every potential parameter in a model is either:
- Free – unknown as you wish to estimate it;
- Fixed – not free, but fixed to a specific value (typically 0 or 1); or
- Constrained – unknown, but constrained to be equal to some other free parameter.

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- For a model to be identified, every parameter must be identified.
- Otherwise the results cannot be trusted.
- This is why identification is important.

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- Levels of identification can be assessed by looking at a models' degrees of freedom (df).
- A) under-identified model – not enough information (not identified); negative df .
- B) just-identified model – just enough information; zero df .
- C) over-identified model – more than enough information (preferred); positive df .

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3. Model Estimation

- Some commonly used estimation methods:
- OLS (ordinary least squares);
- GLS (generalized least squares);
- ML (maximum likelihood).

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4. Model Testing

- A) Use global fit indices to assess fit of an entire model.
- Unfortunately there is no single most powerful fit index (like the F in ANOVA or regression).
- Some global fit indices include chi-square, GFI (goodness-of-fit index), RMSEA (root mean square error of approximation), SRMR (standardized root-mean-square residual), and CFI (comparative fit index); it is recommended that you report multiple global fit indices.

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- B) To assess the fit of individual parameters, examine...
- (1) t values: $t = \text{estimate} / \text{standard error}$;
- (2) significance of t ;
- (3) sign of the estimate (in hypothesized direction);
- (4) whether the estimate makes sense (e.g., in expected range; no negative variances; no correlations beyond 1.0)

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5. Model Modification

- When global model fit is not acceptable, need to look for ways to modify the initial theoretical model to achieve a better fit.
- Searching for a more properly specified model is known as a “specification search.”

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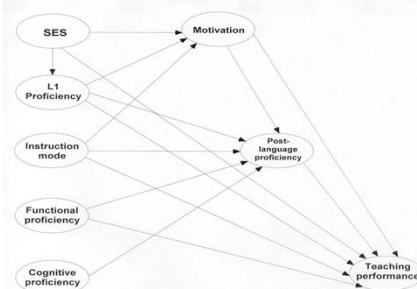
- True model \leftarrow ----- \rightarrow Tested model
- The less consistent these models, the more the tested model is misspecified.
- **In any specification search, substantive knowledge must be the #1 priority.**
- **In other words, do not add additional parameters to achieve better model fit if they do not make substantive sense.**

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G. Some examples from my work

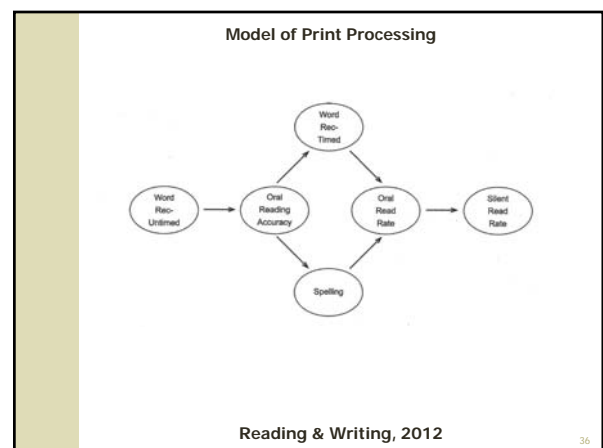
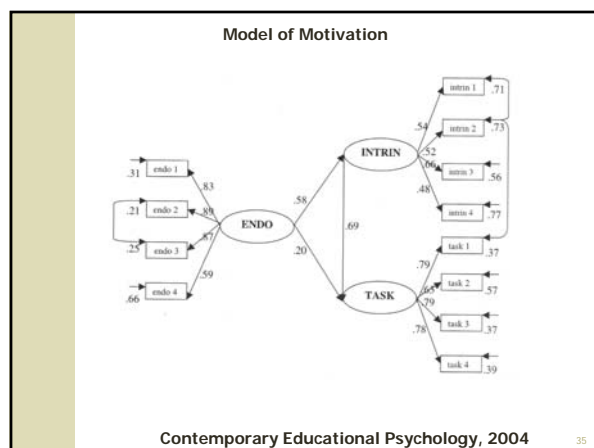
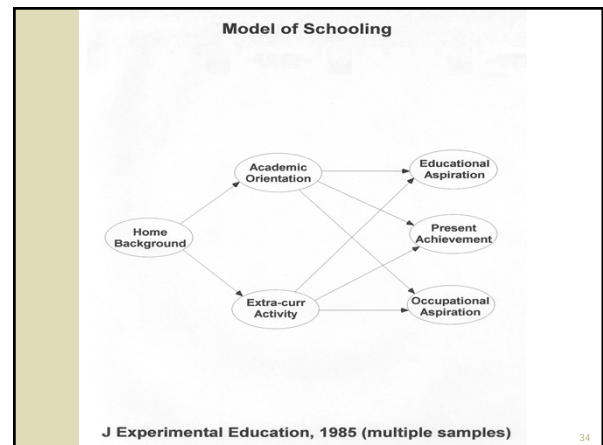
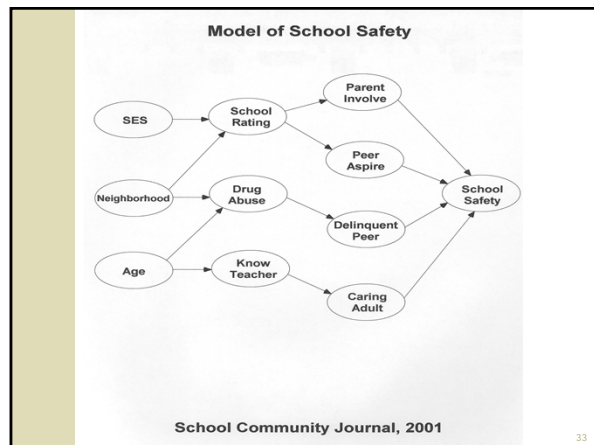
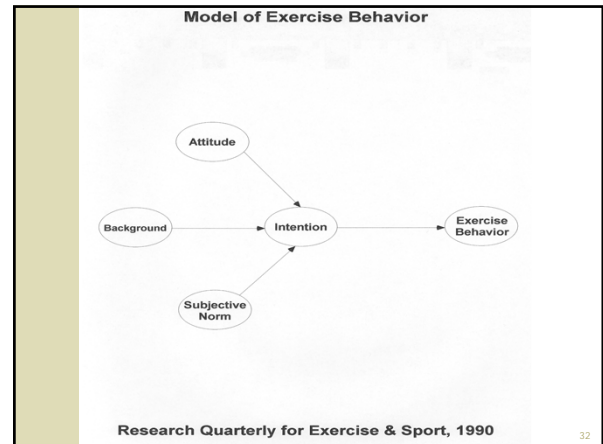
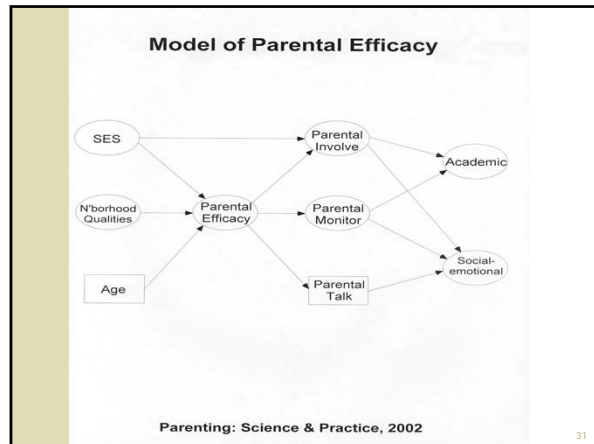
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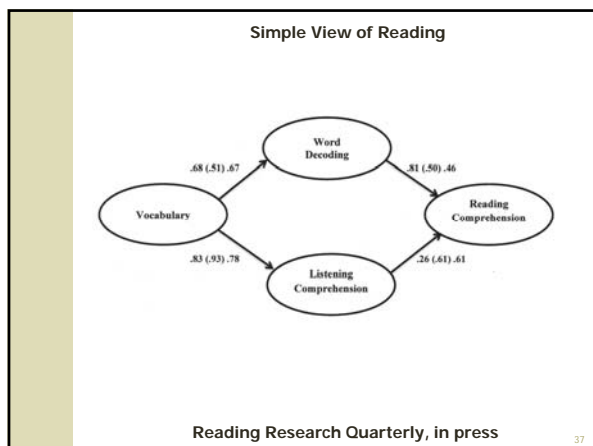
Model of Second Language Acquisition



Journal of Experimental Education, 1984.

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H. Advanced SEM Models

- Model validation (through multiple samples, cross-validation, or bootstrapping)
- Multiple group SEM (including structured means models)
- Second-order CFA
- Multitrait-multimethod models (MTMM)
- Multiple indicator-multiple cause models (MIMIC)

- Mixture models
- Multilevel models
- Interaction models
- Latent Growth models
- Monte Carlo methods

I. Further Information

- Schumacker, R. E. & Lomax, R. G. (2010). *A beginner's guide to structural equation modeling* (3rd ed.) New York: Routledge.
- Spring course in SEM (ESQREM 8659, Wednesdays 12:45-3:25)