

Writing about SEM/Best Practices

- Presenting results
- Best Practices
 - » specification
 - » data
 - » analysis
 - » interpretation

Writing about SEM

- Summary of recommendation from Hoyle & Panter, McDonald & Ho, Kline
 - » the model
 - » the data
 - » estimation and fit
 - » parameter estimates
 - » alternative models

Presenting the Model

- H&P suggest presenting a more abstract version of the model first ("conceptual model") followed by a concrete model specified in enough detail to allow the reader to reconstruct the analysis ("statistical model")
 - » reader should be able to compute observations and degrees of freedom from statistical model
 - » indicate clearly any parameters that were fixed
- M&H want to see more discussion of identifiability
- Both want readers to do more theoretical justification for presence and absence of paths

Data

- Check for violations of assumptions and present diagnostic information
 - » esp., provide information about kurtosis
 - » Mardia's coefficient gives information on multivariate normality
- Give information on missing data (how much?), and how this was handled (e.g., listwise deletion)
- Provide data
 - » covariance matrix that includes all observed variables OR
 - » correlation matrix and standard deviations
 - » M&H: if > 30 variables, put on web or state that the data are available from the author

Estimation and Fit

- State what method of estimation you used
 - » Maximum Likelihood best in most cases
- Present strategy for testing fit (H&P)
 - » state which indices will be presented and give justifications for choosing them
 - » give conceptual definition of each index used
 - » state the cutoff values you will be using
- Give χ^2 , df, sample size and p value
 - » can do this succinctly:
 $\approx \chi^2(48, N = 500) = 303.80, p < .001, TLI = .86, CFI = .90$

Global Fit Indices

- H&P recommend
 - » GFI because it's in the same metric as R
 - » NNFI or IFI
 - » CFI
- M&H recommend
 - » RMSEA, RMR, CFI, GFI
- Kline recommends
 - » RMSEA, CFI

More on Fit

- M&H and Kline suggest that discrepancy information should be presented
 - » M&H say to present it in the other half of the variance/covariance matrix of the observed variables
- M&H and Kline also both recommend a two-step testing strategy to insure that structural part of model fits well
 - » See M&H Table 2 (p. 74) for examples in which structural portion did not fit well but this was masked by the overall good fit of the model

Parameter Estimates

- Report all parameter estimates
 - » including variances
 - » report standard errors as well
 - » clearly indicate any paths that were fixed (e.g., to 1.0 to set the scale for a latent variable)
- M&H suggest presenting measurement model parameters in tabular form and leaving the observed variables out of the diagram, for clarity

Alternative Models

- Present and test alternative models
- If respecification is done, present this information clearly
 - » H&P recommend that results for the hypothesized model be presented first
 - » In a separate section, present the modified model
- Test equivalent models, if possible

Cross-Validation

- If you have a holdout sample, test your final model on them and present the results
- If you can't do this, provide estimates of the likelihood that your model will replicate
 - » Browne & Cudeck cross-validation statistic

Keeping Up

- Recommendations are changing, so look for people to continue writing on this topic in:
 - » Psychological Methods
 - » Specialty journals
 - Structural Equation Models
 - Multivariate Behavioral Research
 - Applied Psychological Measurement
 - Psychometrika

Best Practices: Specification

- Lay out your model before you collect data
 - » if it is not identified, you can add more measured variables
- Try to include all important causes that are already known
- When modeling latent variables, have enough indicators
 - » Kenny (1979): "Two *might be* fine, three is better, four is best and anything more is gravy."
 - » Number of indicators necessary for identification depends on the model

Best Practices: Specification

- Think hard about directionality
 - » does the logic of the study design and protocol rule out some causal orders?
 - » have some causal orders been confirmed or disconfirmed in other studies (especially longitudinal or experimental)?
 - » if not, you may want to consider (and test) other causal orders
- Don't use feedback loops (causal arrows going both ways) as a way to get around thinking hard about directionality

Best Practices: Specification

- Add correlations between error or disturbance terms only when conceptually justified
 - » try to work out ahead of time which error terms may need to be correlated
 - » avoid correlating error terms solely to improve fit
- Try for indicators that load on one factor (latent variable) only
 - » allow cross-loadings only if clearly justified theoretically

Best Practices: Data

- Implement quality control practices for examining data
- Minimize missing data
 - » If much data is missing, imputation may be the best method, if data are not missing at random
 - » watch for new developments
- Check for violations of assumptions
 - » normal distributions for endogenous variables
 - » linearity
 - » independence
- Screen for outliers
 - » As in regression and ANOVA, SEM is sensitive to outliers

Best Practices: Analysis

- Use theory and previous findings to guide respecification
 - » modification indices (e.g., Lagrange) can be useful, but do not rely on them blindly (unless you like making Type I errors)
- Double check your syntax
 - » make sure you are running the model you think you are
- Look carefully at your output for signs of problems
 - » error messages (or not "all is ok" message)
 - » negative variances (and other impossible things)
 - » huge standard errors (and other unlikely things)

Best Practices: Analysis

- Report unstandardized as well as standardized estimates
- Check for multicollinearity
 - » Kline says correlations $>.85$ may be problematic
- Check your sample size
 - » At least 100 cases AND
 - » 10:1 ratio for cases to parameters estimated (or, at an absolute minimum, 5:1)

Best Practices: Analysis

- Provide SEM program with start values, if it is having trouble
 - » If program doesn't converge or there are other signs of problems, but estimates are printed, use those as start values
 - » Kline has several appendices that give advice about providing start values
- Be aware of the possibility of empirical underidentification
- Evaluate the measurement and structural portions of the model separately

Best Practices: Interpretation

- Look at all of your output
 - » fit indices are important but they are only part of the picture
 - » be sure to look at matrix of residuals -- this lets you know if all of the model is fitting well or if there are some areas of misfit
- Do not assume, believe, or state that your model *must* be correct, because fit is good
 - » we can disprove models (state that they must be incorrect) but we can't prove that a model is correct

Best Practices: Interpretation

- Remember that good fit does not imply anything about how much variance in the endogenous variables is explained
 - » good fit means the variance in the variance-covariance matrix is well represented by the model
 - » if you care about being able to predict large amounts in the variance in some or all endogenous variables, you have to look at that separately
- Consider and test alternative models
- If possible, consider mathematically equivalent models
 - » nice if you can rule some out using theory or logic

Best Practices: Interpretation

- Remember that SEM is not a cure for poor theory or design
- Don't reify your factors
 - » you tried to choose indicators so as to create a latent factor that represents the construct of interest
 - » but you may not have succeeded
 - » when reading other people's work, don't rely just on their name for the latent variable -- look critically at the indicators
- Report enough information so that readers can reproduce your analysis and try alternative models