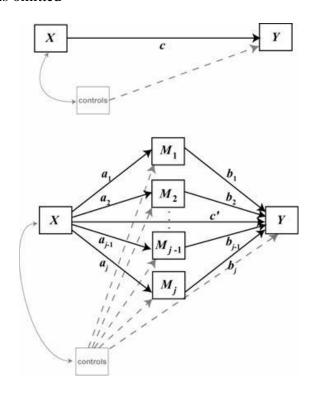
INDIRECT

Subcommands in braces are optional ** Default if subcommand is omitted



Overview

INDIRECT estimates the total, direct, and single-step indirect effects (specific and total) of causal variable xvar on outcome variable yvar through a proposed mediator variable or list of mediator variables mlist, controlling for one of more variables listed in covlist. It calculates the Sobel test for the total and specific indirect effect(s) as well as percentile-based, bias-corrected, and bias-corrected and accelerated bootstrap confidence intervals for the indirect effects. When more than one variable is listed in mvlist, it also calculates normal theory and bootstrap tests of the difference between the indirect effects. For details on the methods, see Preacher, K. J., & Hayes, A. F. (2008). Asymptotic and resampling methods for estimating and comparing indirect effects. *Behavior Research Methods*, 40, 879-891. Estimates of all paths are calculated using OLS regression.

Examples

```
%INDIRECT (DATA = example, Y = know, X = educ, M = attn elab, CONTRAST = 1, NORMAL = 1, BOOT = 5000);
```

- Estimates the total and direct effects of educ on know, as well as the total and specific indirect effects of educ on know through attn and elab
- Produces the Sobel test for the total and specific indirect effects
- Conducts a contrast between the two specific indirect effects
- Generates 95% bias-corrected and accelerated bootstrap confidence intervals for the indirect effects using 5000 bootstrap samples.
- The analysis is conducted using a SAS data file named "example"

```
%INDIRECT (DATA = example, Y = know, X = educ, M = attn elab sex age educ, C =
3, CONTRAST = 1, CONF = 99, PERCENT = 1, BOOT = 1000);
```

- Estimates the total and direct effects of educ on know, as well as the indirect effect of educ on know through attn and elab, while controlling for sex, age, and educ
- Conducts a contrast between the two specific indirect effects
- Generates 99% percentile and bias-corrected and accelerated bootstrap confidence intervals for the indirect effects using 1000 bootstrap samples.

Covariates

The direct, indirect, and total effects of xvar on yvar can be calculated with or without including a set of covariates which are partialled out of yvar and any and all variables in the mlist list. The covariates should be listed after the list of mediators in the M = subcommand, and then the number of covariates in covlist should be used as the argument for cov in the C = subcommand. For example, if four variables are provided in covlist, then specify C = 4. In the output, what is listed as the total effect of the independent variable is actually corrected for the effect of the covariates. To get an uncorrected total effect, remove the covariates from the model and rerun the macro.

Normal Theory Tests and Contrasts

By setting t to 1 in the NORMAL subcommand, the macro conducts Sobel tests for the total and specific indirect effects, defined as the effect divided by its standard error. A p-value is derived using the standard normal distribution. If covariates are listed, the Sobel tests are not conducted or printed.

Specifying n equal to 1 in the CONTRAST subcommand produces pairwise contrasts between all specific indirect effects by calculating the difference, dividing it by its standard error, and deriving a p-value from the standard normal distribution. When there are only two mediator variables in the model, the contrast between specific indirect effects is listed in the output as C1. With k mediators, the 0.5k(k-1) possible pairwaise contrasts are listed as C1, C2, C3, and so forth, and a key for interpreting which code corresponds to which contrast is provided at the bottom of the output.

Although Sobel tests are widely reported in the literature, recent thinking about mediation analysis argues against the use of the test. Bootstrap confidence intervals are preferred. For a discussion, see Hayes, A. F. (2009). Beyond Baron and Kenny: Statistical mediation analysis in the new millennium. *Communication Monographs*, 76, 408-420.

Bootstrapping

By default, the macro generates 95% bias-corrected and accelerated bootstrap confidence intervals for all indirect effects and contrasts of indirect effects using z=1000 bootstrap samples. The number of bootstrap samples can be changed by setting z in the BOOT subcommand to the desired number. The level of confidence for confidence intervals can be changed by setting ci to the desired number (such as 90, 99, and so forth) in the CONF subcommand. Percentile or bias-corrected confidence intervals can be requested by setting p and/or b to 1 in the PERCENT and/or BC subcommands, respectively. To turn off the printing of a particular form of bootstrap confidence interval, set its argument to 0 in the corresponding subcommand.

Multiple Independent Variables

In some cases the user might like to estimate a model that includes multiple independent variables each linked to the same set of mediators. The macro can be used to estimate the coefficients in such a model, although it provides no information that can be used to test a combined indirect effect involving all independent variables. Covariates are mathematically treated exactly like independent variables in the estimation, with paths to all mediators and the outcome, so if the desired model has k covariates, the macro can be run k times, each time listing one variable as the independent variable and the remaining k-1 independent variables as covariates. Each run of the macro will generate the desired indirect effect for the variable currently listed as the independent variable.

Notes

- mvlist and yvar must be quantitative variables and are assumed to have at least interval-level measurement properties. xvar and variables in covlist can be dichotomous or quantitative with interval-level properties.
- When bootstrapping is enabled, the bootstrap samples are saved to a temporary SAS data file called "indirect".
- A case will be deleted from the analysis if missing on any of the variables in the model.
- The macro is limited to the estimation of 10 specific indirect effects. If the user includes more than 10 mediators in the variable list, an error will result.
- All path coefficients in the output are unstandardized.