

semmcci: Monte Carlo Confidence Intervals

Ivan Jacob Agaloos Pesigan

Installation

You can install the CRAN release of `semmcci` with:

```
install.packages("semmcci")
```

You can install the development version of `semmcci` from [GitHub](#) with:

```
if (!require("remotes")) install.packages("remotes")
remotes::install_github("jeksterslab/semmcci")
```

Documentation

See [GitHub Pages](#) for package documentation.

Description

In the Monte Carlo method, a sampling distribution of parameter estimates is generated from the multivariate normal distribution using the parameter estimates and the sampling variance-covariance matrix. Confidence intervals for defined parameters are generated by obtaining percentiles corresponding to $100(1 - \alpha)\%$ from the generated sampling distribution, where α is the significance level.

Monte Carlo confidence intervals for free and defined parameters in models fitted in the structural equation modeling package `lavaan` can be generated using the `semmcci` package. The package has two main functions, namely, `MC()` and `MCStd()`. The output of `lavaan` is passed as the first argument to the `MC()` function to generate Monte Carlo confidence intervals. Monte Carlo confidence intervals for the standardized estimates can also be generated by passing the output of the `MC()` function to the `MCStd()` function. A description of the package and code examples are presented in Pesigan and Cheung (2024).

Example

A common application of the Monte Carlo method is to generate confidence intervals for the indirect effect. In the simple mediation model, variable **X** has an effect on variable **Y**, through a mediating variable **M**. This mediating or indirect effect is a product of path coefficients from the fitted model.

```
library(semmcci)
library(lavaan)
```

Data

```
n <- 1000
X <- rnorm(n = n)
M <- 0.50 * X + rnorm(n = n)
Y <- 0.25 * X + 0.50 * M + rnorm(n = n)
data <- data.frame(X, M, Y)
```

Model Specification

The indirect effect is defined by the product of the slopes of paths **X** to **M** labeled as **a** and **M** to **Y** labeled as **b**. In this example, we are interested in the confidence intervals of `indirect` defined as

the product of **a** and **b** using the `:=` operator in the **lavaan** model syntax.

```
model <- "  
  Y ~ cp * X + b * M  
  M ~ a * X  
  indirect := a * b  
  direct := cp  
  total := cp + (a * b)  
"
```

Model Fitting

We can now fit the model using the `sem()` function from **lavaan**.

```
fit <- sem(data = data, model = model)
```

Monte Carlo Confidence Intervals

The `fit` **lavaan** object can then be passed to the `MC()` function to generate Monte Carlo confidence intervals.

```
MC(fit, R = 20000L, alpha = 0.05)  
  
#> Monte Carlo Confidence Intervals  
#>      est      se      R  2.5% 97.5%  
#> cp      0.2258 0.0349 20000 0.1577 0.2946  
#> b      0.5437 0.0306 20000 0.4838 0.6044  
#> a      0.5257 0.0323 20000 0.4623 0.5894  
#> Y~~Y    1.0205 0.0454 20000 0.9314 1.1100  
#> M~~M    1.0994 0.0490 20000 1.0036 1.1944
```

```
#> X~~X      1.0483 0.0000 20000 1.0483 1.0483
#> indirect 0.2858 0.0240 20000 0.2402 0.3342
#> direct    0.2258 0.0349 20000 0.1577 0.2946
#> total     0.5116 0.0356 20000 0.4420 0.5817
```

Standardized Monte Carlo Confidence Intervals

Standardized Monte Carlo Confidence intervals can be generated by passing the result of the `MC()` function to `MCStd()`.

Note: We recommend setting `fixed.x = FALSE` when generating standardized estimates and confidence intervals to model the variances and covariances of the predictors if they are assumed to be random.

```
fit <- sem(data = data, model = model, fixed.x = FALSE)
unstd <- MC(fit, R = 20000L, alpha = 0.05)
vcov(unstd)
```

#>	cp	b	a	Y~~Y	M~~M
#> cp	1.221749e-03	-4.902496e-04	-1.214125e-05	-3.284633e-07	-2.023017e-05
#> b	-4.902496e-04	9.224239e-04	1.195221e-05	-8.635838e-06	-1.755196e-06
#> a	-1.214125e-05	1.195221e-05	1.032803e-03	1.306815e-05	-2.055988e-05
#> Y~~Y	-3.284633e-07	-8.635838e-06	1.306815e-05	2.080198e-03	-2.717524e-05
#> M~~M	-2.023017e-05	-1.755196e-06	-2.055988e-05	-2.717524e-05	2.400729e-03
#> X~~X	1.220898e-05	5.572485e-07	2.287230e-06	-2.036124e-06	8.422999e-06
#> indirect	-2.642937e-04	4.912222e-04	5.679065e-04	2.659857e-06	-1.168587e-05
#> direct	1.221749e-03	-4.902496e-04	-1.214125e-05	-3.284633e-07	-2.023017e-05
#> total	9.574551e-04	9.725343e-07	5.557652e-04	2.331394e-06	-3.191604e-05
#>	X~~X	indirect	direct	total	
#> cp	1.220898e-05	-2.642937e-04	1.221749e-03	9.574551e-04	

```
#> b      5.572485e-07  4.912222e-04 -4.902496e-04  9.725343e-07
#> a      2.287230e-06  5.679065e-04 -1.214125e-05  5.557652e-04
#> Y~~Y   -2.036124e-06  2.659857e-06 -3.284633e-07  2.331394e-06
#> M~~M    8.422999e-06 -1.168587e-05 -2.023017e-05 -3.191604e-05
#> X~~X    2.208937e-03  1.209539e-06  1.220898e-05  1.341852e-05
#> indirect 1.209539e-06  5.679071e-04 -2.642937e-04  3.036134e-04
#> direct   1.220898e-05 -2.642937e-04  1.221749e-03  9.574551e-04
#> total    1.341852e-05  3.036134e-04  9.574551e-04  1.261069e-03
```

MCStd(unstd)

```
#> Standardized Monte Carlo Confidence Intervals
#>      est      se      R  0.05%   0.5%   2.5%  97.5%  99.5%  99.95%
#> cp      0.1816 0.0279 20000 0.0884 0.1100 0.1273 0.2360 0.2536 0.2721
#> b      0.5035 0.0254 20000 0.4248 0.4369 0.4524 0.5522 0.5677 0.5850
#> a      0.4567 0.0250 20000 0.3773 0.3914 0.4063 0.5043 0.5194 0.5374
#> Y~~Y    0.6300 0.0241 20000 0.5495 0.5673 0.5820 0.6768 0.6903 0.7060
#> M~~M    0.7915 0.0228 20000 0.7112 0.7303 0.7457 0.8349 0.8468 0.8576
#> X~~X    1.0000 0.0000 20000 1.0000 1.0000 1.0000 1.0000 1.0000 1.0000
#> indirect 0.2299 0.0175 20000 0.1742 0.1854 0.1960 0.2641 0.2756 0.2879
#> direct   0.1816 0.0279 20000 0.0884 0.1100 0.1273 0.2360 0.2536 0.2721
#> total    0.4116 0.0261 20000 0.3282 0.3428 0.3594 0.4617 0.4757 0.4935
```

References

Pesigan, I. J. A., & Cheung, S. F. (2024). Monte Carlo confidence intervals for the indirect effect with missing data. *Behavior Research Methods*, 56(3), 1678–1696. <https://doi.org/10.3758/s13428-023-02114-4>

R Core Team. (2025). *R: A language and environment for statistical computing*. R Foundation for Statistical Computing. Vienna, Austria. <https://www.R-project.org/>