

Synthesizing Indirect Effects in Mediation Models with Meta-Analytic Methods: Supplementary Materials 1

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• This file demonstrates how to compute effect sizes and their sampling covariance matrix with two approaches using the delta method. The first one uses a numeric approach with the structural equation modeling (SEM) framework. The second approach computes the sampling covariance matrix with the symbolic calculations.	

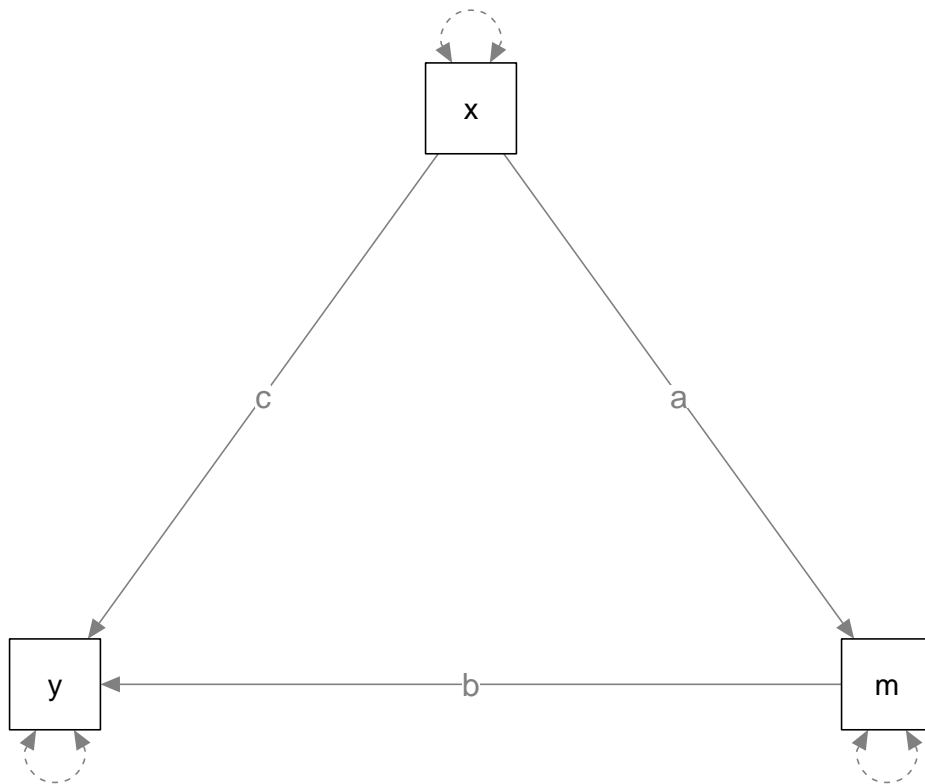
Numeric calculations with the SEM approach

One mediator

```
library(metaSEM)

## Model with one mediator: x -> m -> y
model1 <- "y ~ c*x + b*m
          m ~ a*x
          # Define indirect and direct effects
          Indirect := a*b
          Direct := c"

## Display the model
plot(model1)
```



```

## Sample correlation matrix
my.cor <- matrix(c(1, .5, .3,
                  .5, 1, .4,
                  .3, .4, 1),
                nrow = 3,
                ncol = 3,
                dimnames = list(c("y", "m", "x"),
                               c("y", "m", "x"))))

```

```
my.cor
```

```

##      y    m    x
## y 1.0 0.5 0.3
## m 0.5 1.0 0.4
## x 0.3 0.4 1.0

```

```

## Calculate the indirect and direct effects and their sampling covariance matrix
calEffSizes(model=model1, n=300, Cov=my.cor)

```

```

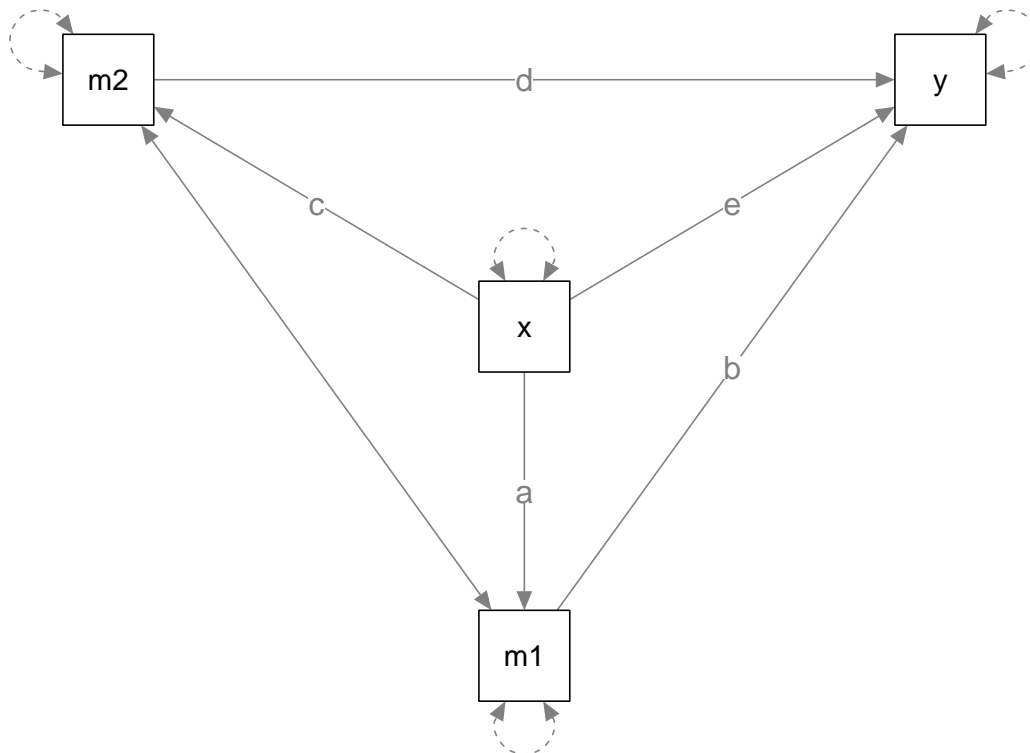
## $ES
## Indirect Direct
## 0.1809524 0.1190476
##
## $VCOV
## Indirect Direct
## Indirect 0.0010416478 -0.0004686319
## Direct -0.0004686319 0.0029289494

```

Two parallel mediators

```
## Model with two specific mediators: x -> m1 -> y and x -> m2 -> y
model2 <- "y ~ e*x + b*m1 + d*m2
          m1 ~ a*x
          m2 ~ c*x
          # m1 and m2 are correlated
          m1 ~~ m2
          # Define indirect and direct effects
          Ind_m1 := a*b
          Ind_m2 := c*d
          Direct := e"
```

```
plot(model2, layout="circle")
```



```
## Sample correlation matrix
my.cor <- matrix(c(1, .5, .6, .3,
                  .5, 1, .4, .2,
                  .6, .4, 1, .3,
                  .3, .2, .3, 1),
                nrow = 4,
                ncol = 4,
                dimnames = list(c("y", "m1", "m2", "x"),
                              c("y", "m1", "m2", "x"))))
my.cor
```

```
##      y  m1  m2  x
## y  1.0 0.5 0.6 0.3
## m1 0.5 1.0 0.4 0.2
## m2 0.6 0.4 1.0 0.3
```

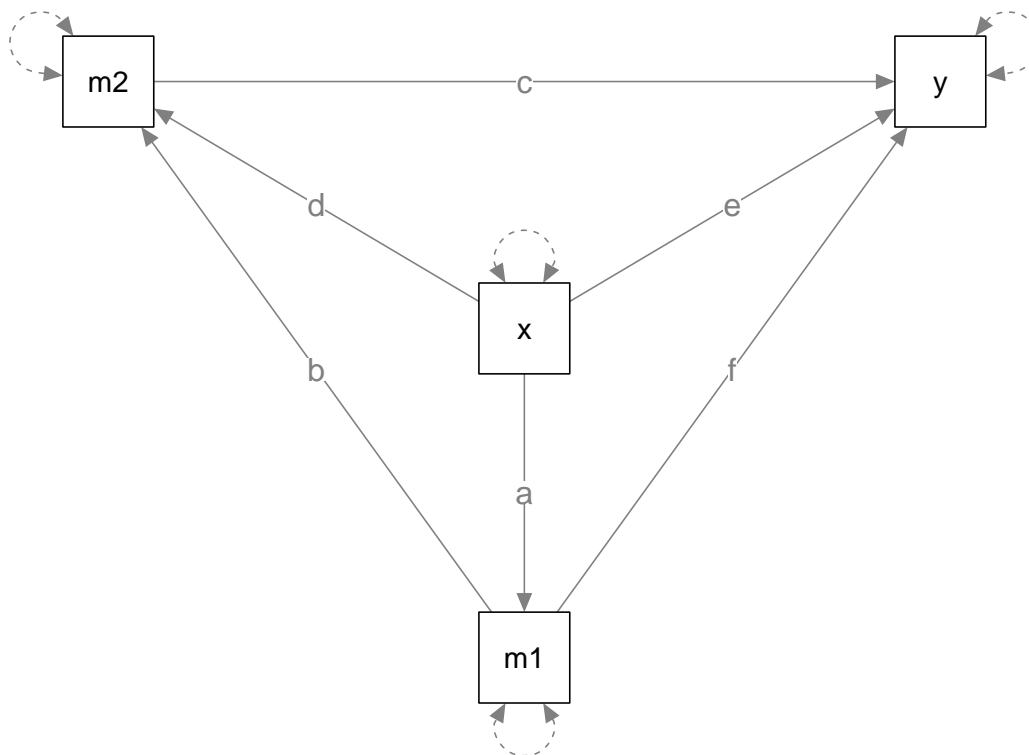
```
## x 0.3 0.2 0.3 1.0
## Calculate the indirect and direct effects and their sampling covariance matrix
calEffSizes(model=model2, n=300, Cov=my.cor)

## $ES
##      Ind_m1      Ind_m2      Direct
## 0.05989446 0.13456464 0.10554090
##
## $VCOV
##           Ind_m1      Ind_m2      Direct
## Ind_m1 0.0003749419 0.0001029453 -0.0000386612
## Ind_m2 0.0001029453 0.0008190651 -0.0001594774
## Direct -0.0000386612 -0.0001594774 0.0020297130
```

Two serial mediators

```
## Model with two intermediate mediators: x -> m1 -> m2 -> y
model3 <- "y ~ e*x + f*m1 + c*m2
          m1 ~ a*x
          m2 ~ b*m1
          m2 ~ d*x
          # Define indirect and direct effects
          Ind_m1m2 := a*b*c
          Ind_m1 := a*f
          Ind_m2 := d*c
          Direct := e"

plot(model3, layout="circle")
```



```
## Calculate the indirect and direct effects and their sampling covariance matrix
calEffSizes(model=model3, n=300, Cov=my.cor)
```

```
## $ES
##   Ind_m1m2   Ind_m1   Ind_m2   Direct
## 0.03177221 0.05989446 0.10279244 0.10554090
##
## $VCOV
##           Ind_m1m2   Ind_m1   Ind_m2   Direct
## Ind_m1m2 1.144605e-04 0.0001405997 1.559037e-05 -0.0000376544
## Ind_m1    1.405997e-04 0.0003749419 -3.765440e-05 -0.0000386612
## Ind_m2    1.559037e-05 -0.0000376544 6.734239e-04 -0.0001218231
## Direct   -3.765440e-05 -0.0000386612 -1.218231e-04 0.0020297130
```

Symbolic calculations

One mediator

```
library(symSEM)

## fn: The effect sizes
## Covfn: Sampling covariance matrix of fn: "b^2*Va+2*b*a*Cba+a^2*Vb"
## Va: Sampling variance of a
## Vb: Sampling variance of b
## Cba: Sampling covariance of a and b
deltamethod(fn="a*b")

## $fn
##      [,1]
## fn1 "a*b"
##
## $Covfn
##      fn1
## fn1 "b^2*Va+2*b*a*Cba+a^2*Vb"
##
## $vars
## [1] "a" "b"
##
## $Covvars
##      a      b
## a "Va" "Cba"
## b "Cba" "Vb"
##
## $Jmatrix
##      a      b
## fn1 "b" "a"
```

Two parallel mediators

```
deltamethod(fn=c("a*b", "c*d"))
```

```
## $fn
##      [,1]
```

```
## fn1 "a*b"
## fn2 "c*d"
##
## $Covfn
##      fn1                      fn2
## fn1 "b^2*Va+2*b*a*Cba+a^2*Vb"      "b*Cca*d+b*Cda*c+a*Ccb*d+a*Cdb*c"
## fn2 "d*Cca*b+d*Ccb*a+c*Cda*b+c*Cdb*a" "d^2*Vc+2*d*c*Cdc+c^2*Vd"
##
## $vars
## [1] "a" "b" "c" "d"
##
## $Covvars
##      a      b      c      d
## a "Va"  "Cba" "Cca" "Cda"
## b "Cba" "Vb"  "Ccb" "Cdb"
## c "Cca" "Ccb" "Vc"  "Cdc"
## d "Cda" "Cdb" "Cdc" "Vd"
##
## $Jmatrix
##      a      b      c      d
## fn1 "b"  "a"  "0"  "0"
## fn2 "0"  "0"  "d"  "c"
```

Two serial mediators

```
deltamethod(fn="a*b*c")
```

```
## $fn
##      [,1]
## fn1 "a*b*c"
##
## $Covfn
##      fn1
## fn1 "b^2*c^2*Va+2*b^2*c*a*Cca+b^2*a^2*Vc+2*b*c^2*a*Cba+2*b*c*a^2*Ccb+c^2*a^2*Vb"
##
## $vars
## [1] "a" "b" "c"
##
## $Covvars
##      a      b      c
## a "Va"  "Cba" "Cca"
## b "Cba" "Vb"  "Ccb"
## c "Cca" "Ccb" "Vc"
##
## $Jmatrix
##      a      b      c
## fn1 "b*c" "a*c" "a*b"
```

```
sessionInfo()
```

```
## R version 4.0.3 (2020-10-10)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 20.04.2 LTS
##
## Matrix products: default
```

```

## BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.9.0
## LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.9.0
##
## locale:
## [1] LC_CTYPE=en_SG.UTF-8 LC_NUMERIC=C
## [3] LC_TIME=en_SG.UTF-8 LC_COLLATE=en_SG.UTF-8
## [5] LC_MONETARY=en_SG.UTF-8 LC_MESSAGES=en_SG.UTF-8
## [7] LC_PAPER=en_SG.UTF-8 LC_NAME=C
## [9] LC_ADDRESS=C LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_SG.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats graphics grDevices utils datasets methods base
##
## other attached packages:
## [1] symSEM_0.1.1 metaSEM_1.2.5.1 OpenMx_2.19.5
##
## loaded via a namespace (and not attached):
## [1] nlme_3.1-152 RColorBrewer_1.1-2 rprojroot_1.3-2
## [4] mi_1.0 tools_4.0.3 backports_1.2.1
## [7] R6_2.5.0 rpart_4.1-15 Hmisc_4.4-1
## [10] colorspace_1.4-1 nnet_7.3-14 withr_2.3.0
## [13] tidyselect_1.1.0 gridExtra_2.3 mnormt_2.0.2
## [16] compiler_4.0.3 fdrtool_1.2.15 qgraph_1.6.9
## [19] htmlTable_2.1.0 regsem_1.6.2 desc_1.2.0
## [22] scales_1.1.1 checkmate_2.0.0 psych_2.0.9
## [25] mvtnorm_1.1-1 pbapply_1.4-3 sem_3.1-11
## [28] stringr_1.4.0 digest_0.6.27 pbivnorm_0.6.0
## [31] foreign_0.8-80 minqa_1.2.4 rmarkdown_2.7
## [34] base64enc_0.1-3 jpeg_0.1-8.1 pkgconfig_2.0.3
## [37] htmltools_0.5.0 lme4_1.1-26 lisrelToR_0.1.4
## [40] htmlwidgets_1.5.2 rlang_0.4.10 rstudioapi_0.11
## [43] generics_0.0.2 gtools_3.8.2 dplyr_1.0.2
## [46] zip_2.1.1 magrittr_1.5 Formula_1.2-3
## [49] Matrix_1.2-18 Rcpp_1.0.5 munsell_0.5.0
## [52] abind_1.4-5 rockchalk_1.8.144 lifecycle_0.2.0
## [55] stringi_1.5.3 yaml_2.2.1 carData_3.0-4
## [58] MASS_7.3-53 plyr_1.8.6 matrixcalc_1.0-3
## [61] lavaan_0.6-8 grid_4.0.3 parallel_4.0.3
## [64] crayon_1.3.4 lattice_0.20-41 semPlot_1.1.2
## [67] kutils_1.70 splines_4.0.3 Ryacas_1.1.3.1
## [70] tmvnsim_1.0-2 knitr_1.30 pillar_1.4.6
## [73] igraph_1.2.6 boot_1.3-25 corpcor_1.6.9
## [76] pkgload_1.1.0 reshape2_1.4.4 stats4_4.0.3
## [79] XML_3.99-0.5 glue_1.4.2 evaluate_0.14
## [82] latticeExtra_0.6-29 data.table_1.13.2 png_0.1-7
## [85] vctrs_0.3.4 nloptr_1.2.2.2 testthat_3.0.2
## [88] gtable_0.3.0 purrr_0.3.4 assertthat_0.2.1
## [91] ggplot2_3.3.2 xfun_0.19 openxlsx_4.2.2
## [94] xtable_1.8-4 coda_0.19-4 Rsolnp_1.16
## [97] glasso_1.11 survival_3.2-7 truncnorm_1.0-8
## [100] tibble_3.0.4 arm_1.11-2 ellipse_0.4.2
## [103] cluster_2.1.0 statmod_1.4.35 ellipsis_0.3.1

```