Package: modsem (via r-universe)

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Type Package

Title Latent Interaction (and Moderation) Analysis in Structural Equation Models (SEM)

Version 1.0.3

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Description Estimation of interaction (i.e., moderation) effects between latent variables in structural equation models (SEM). The supported methods are: The constrained approach (Algina & Moulder, 2001). The unconstrained approach (Marsh et al., 2004). The residual centering approach (Little et al., 2006). The double centering approach (Lin et al., 2010). The latent moderated structural equations (LMS) approach (Klein & Moosbrugger, 2000). The quasi-maximum likelihood (QML) approach (Klein & Muthén, 2007) (temporarily unavailable) The constrained- unconstrained, residual- and double centering-approaches are estimated via 'lavaan' (Rosseel, 2012), whilst the LMS- and QML- approaches are estimated via by modsem it self. Alternatively model can be estimated via 'Mplus' (Muthén & Muthén, 1998-2017). References: Algina, J., & Moulder, B. C. (2001). <doi:10.1207/S15328007SEM0801_3>. ``A note on estimating the Jöreskog-Yang model for latent variable interaction using 'LISREL' 8.3." Klein, A., & Moosbrugger, H. (2000). <doi:10.1007/BF02296338>. ``Maximum likelihood estimation of latent interaction effects with the LMS method." Klein, A. G., & Muthén, B. O. (2007). <doi:10.1080/00273170701710205>. ``Quasi-maximum likelihood estimation of structural equation models with multiple interaction and quadratic effects." Lin, G. C., Wen, Z., Marsh, H. W., & Lin, H. S. (2010). <doi:10.1080/10705511.2010.488999>. ``Structural equation models of latent interactions: Clarification of orthogonalizing and double-mean-centering strategies." Little, T. D., Bovaird, J. A., & Widaman, K. F. (2006). <doi:10.1207/s15328007sem1304_1>. ``On the merits of orthogonalizing powered and product terms: Implications for modeling interactions among latent variables." Marsh, H. W., Wen, Z., & Hau, K. T. (2004).

<doi:10.1037/1082-989X.9.3.275>. ``Structural equation models of latent interactions: evaluation of alternative estimation strategies and indicator construction." Muthén, L.K. and Muthén, B.O. (1998-2017). ``'Mplus' User's Guide. Eighth Edition." <https://www.statmodel.com/>. Rosseel Y (2012). <doi:10.18637/jss.v048.i02>. ``'lavaan': An R Package for Structural Equation Modeling."

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Encoding UTF-8

LazyData true

RoxygenNote 7.3.2

LinkingTo Rcpp, RcppArmadillo

Imports Rcpp, purrr, stringr, lavaan, rlang, MplusAutomation, nlme, dplyr, mvnfast, stats, fastGHQuad, mvtnorm, ggplot2, parallel

Depends R (>= 3.50)

URL https://github.com/Kss2k/modsem

Suggests knitr, rmarkdown

VignetteBuilder knitr

NeedsCompilation yes

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Repository CRAN

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Contents

`coef_modsem_da` *Wrapper for coef*

Description

wrapper for coef, to be used with modsem::coef_modsem_da, since coef is not in the namespace of modsem, but stats

Usage

```
coef_modsem_da(object, ...)
```

Arguments

`object`	fittet model to inspect
`...`	additional arguments

`compare_fit` *compare model fit for qml and lms models*

Description

Compare the fit of two models using the likelihood ratio test. ‘estH0‘ representing the null hypothesis model, and ‘estH1‘ the alternative hypothesis model. Importantly, the function assumes that ‘estH0‘ does not have more free parameters (i.e., degrees of freedom) than ‘estH1‘. alternative hypothesis model

Usage

```
compare_fit(estH0, estH1)
```

Arguments

`estH0`	object of class ‘modsem_da‘ representing the null hypothesis model
`estH1`	object of class ‘modsem_da‘ representing the

Examples

```
## Not run:
H0 <- "
 # Outer Model
 X =~ x1 + x2 + x3
 Y =~ y1 + y2 + y3
 Z =~ z1 + z2 + z3

 # Inner model
 Y ~ X + Z
"

estH0 <- modsem(m1, oneInt, "lms")

H1 <- "
 # Outer Model
 X =~ x1 + x2 + x3
 Y =~ y1 + y2 + y3
 Z =~ z1 + z2 + z3

 # Inner model
 Y ~ X + Z + X:Z
"

estH1 <- modsem(m1, oneInt, "lms")
compare_fit(estH0, estH1)

## End(Not run)
```

`default_settings_da` *default arguments fro LMS and QML approach*

Description

This function returns the default settings for the LMS and QML approach.

Usage

```
default_settings_da(method = c("lms", "qml"))
```

Arguments

`method`	which method to get the settings for

Value

list

Examples

```
library(modsem)
default_settings_da()
```

`default_settings_pi` *default arguments for product indicator approaches*

Description

This function returns the default settings for the product indicator approaches

Usage

```
default_settings_pi(method = c("rca", "uca", "pind", "dblcent", "ca"))
```

Arguments

`method`	which method to get the settings for

Value

list

Examples

```
library(modsem)
default_settings_pi()
```

`extract_lavaan` *extract lavaan object from modsem object estimated using product indicators*

Description

extract lavaan object from modsem object estimated using product indicators

Usage

```
extract_lavaan(object)
```

Arguments

`object`	modsem object

Value

lavaan object

Examples

```
library(modsem)
m1 <- '
  # Outer Model
  X =~ x1 + x2 + x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'
est <- modsem_pi(m1, oneInt)
lav_est <- extract_lavaan(est)
```

---

fit_modsem_da *Fit measures for QML and LMS models*

Description

Calculates chi-sq test and p-value, as well as RMSEA for the LMS and QML models. Note that the Chi-Square based fit measures should be calculated for the baseline model, i.e., the model without the interaction effect

Usage

```
fit_modsem_da(model, chisq = TRUE)
```

Arguments

model	fitted model. Thereafter, you can use 'compare_fit()' to assess the comparative fit of the models. If the interaction effect makes the model better, and e.g., the RMSEA is good for the baseline model, the interaction model likely has a good RMSEA as well.
chisq	should Chi-Square based fit-measures be calculated?

---

get_pi_data *Get data with product indicators for different approaches*

Description

`get_pi_syntax()` is a function for creating the `lavaan` syntax used for estimating latent interaction models using one of the product indicators in `lavaan`.

Usage

```
get_pi_data(model.syntax, data, method = "dblcent", match = FALSE, ...)
```

Arguments

model.syntax	lavaan syntax
data	data to create product indicators from
method	method to use: "rca" = residual centering approach, "uca" = unconstrained approach, "dblcent" = double centering approach, "pind" = prod ind approach, with no constraints or centering, "custom" = use parameters specified in the function call
match	should the product indicators be created by using the match-strategy
...	arguments passed to other functions (e.g., modsem_pi)

Value

data.frame

Examples

```
library(modsem)
library(lavaan)
m1 <- '
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'
syntax <- get_pi_syntax(m1)
data <- get_pi_data(m1, oneInt)
est <- sem(syntax, data)
```

---

get_pi_syntax *Get* lavaan *syntax for product indicator approaches*

Description

get_pi_syntax() is a function for creating the lavaan syntax used for estimating latent interaction models using one of the product indicators in lavaan.

Usage

```
get_pi_syntax(model.syntax, method = "dblcent", match = FALSE, ...)
```

Arguments

model.syntax	lavaan syntax
method	method to use: "rca" = residual centering approach, "uca" = unconstrained approach, "dblcent" = double centering approach, "pind" = prod ind approach, with no constraints or centering, "custom" = use parameters specified in the function call
match	should the product indicators be created by using the match-strategy
...	arguments passed to other functions (e.g., modsem_pi)

Value

character vector

Examples

```
library(modsem)
library(lavaan)
m1 <- '
  # Outer Model
  X =~ x1 + x2 + x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'
syntax <- get_pi_syntax(m1)
data <- get_pi_data(m1, oneInt)
est <- sem(syntax, data)
```

---

jordan *Jordan subset of PISA 2006 data*

Description

The data stem from the large-scale assessment study PISA 2006 (Organisation for Economic Co-Operation and Development, 2009) where competencies of 15-year-old students in reading, mathematics, and science are assessed using nationally representative samples in 3-year cycles. In this eacademicample, data from the student background questionnaire from the Jordan sample of PISA 2006 were used. Only data of students with complete responses to all 15 items (N = 6,038) were considered.

Format

A data frame of fifteen variables and 6,038 observations:

enjoy1 indicator for enjoyment of science, item ST16Q01: I generally have fun when I am learning <broad science> topics.

enjoy2 indicator for enjoyment of science, item ST16Q02: I like reading about <broad science>.

enjoy3 indicator for enjoyment of science, item ST16Q03: I am happy doing <broad science> problems.

enjoy4 indicator for enjoyment of science, item ST16Q04: I enjoy acquiring new knowledge in <broad science>.

enjoy5 indicator for enjoyment of science, item ST16Q05: I am interested in learning about <broad science>.

academic1 indicator for academic self-concept in science, item ST37Q01: I can easily understand new ideas in <school science>.

academic2 indicator for academic self-concept in science, item ST37Q02: Learning advanced <school science> topics would be easy for me.

academic3 indicator for academic self-concept in science, item ST37Q03: I can usually give good answers to <test questions> on <school science> topics.

academic4 indicator for academic self-concept in science, item ST37Q04: I learn <school science> topics quickly.

academic5 indicator for academic self-concept in science, item ST37Q05: <School science> topics are easy for me.

academic6 indicator for academic self-concept in science, item ST37Q06: When I am being taught <school science>, I can understand the concepts very well.

career1 indicator for career aspirations in science, item ST29Q01: I would like to work in a career involving <broad science>.

career2 indicator for career aspirations in science, item ST29Q02: I would like to study <broad science> after <secondary school>.

career3 indicator for career aspirations in science, item ST29Q03: I would like to spend my life doing advanced <broad science>.

career4 indicator for career aspirations in science, item ST29Q04: I would like to work on <broad science> projects as an adult.

Source

This version of the dataset, as well as the description was gathered from the documentation of the 'nlsem' package (https://cran.r-project.org/package=nlsem), where the only difference is that the names of the variables were changed

Originally the dataset was gathered by the Organisation for Economic Co-Operation and Development (2009). Pisa 2006: Science competencies for tomorrow's world (Tech. Rep.). Paris, France. Obtained from: https://www.oecd.org/pisa/pisaproducts/database-pisa2006.htm

Examples

```
## Not run:
m1 <- '
  ENJ =~ enjoy1 + enjoy2 + enjoy3 + enjoy4 + enjoy5
  CAREER =~ career1 + career2 + career3 + career4
  SC =~ academic1 + academic2 + academic3 + academic4 + academic5 + academic6
  CAREER ~ ENJ + SC + ENJ:ENJ + SC:SC + ENJ:SC
'

est <- modsem(m1, data = jordan)

## End(Not run)
```

`modsem` *Estimate interaction effects in structural equation models (SEMs)*

Description

`modsem()` is a function for estimating interaction effects between latent variables in structural equation models (SEMs). Methods for estimating interaction effects in SEMs can basically be split into two frameworks: 1. Product Indicator-based approaches (`"dblcent"`, `"rca"`, `"uca"`, `"ca"`, `"pind"`) 2. Distributionally based approaches (`"lms"`, `"qml"`).

For the product indicator-based approaches, `modsem()` is essentially a fancy wrapper for `lavaan::sem()` which generates the necessary syntax and variables for the estimation of models with latent product indicators.

The distributionally based approaches are implemented separately and are not estimated using `lavaan::sem()`, but rather using custom functions (largely written in `C++` for performance reasons). For greater control, it is advised that you use one of the sub-functions (modsem_pi, modsem_da, modsem_mplus) directly, as passing additional arguments to them via `modsem()` can lead to unexpected behavior.

Usage

```
modsem(model.syntax = NULL, data = NULL, method = "dblcent", ...)
```

Arguments

`model.syntax`	`lavaan` syntax
`data`	dataframe
`method`	method to use: `"rca"` = residual centering approach (passed to `lavaan`), `"uca"` = unconstrained approach (passed to `lavaan`), `"dblcent"` = double centering approach (passed to `lavaan`), `"pind"` = prod ind approach, with no constraints or centering (passed to `lavaan`), `"lms"` = latent model structural equations (not passed to `lavaan`), `"qml"` = quasi maximum likelihood estimation of latent model structural equations (not passed to `lavaan`), `"custom"` = use parameters specified in the function call (passed to `lavaan`).
`...`	arguments passed to other functions depending on the method (see `modsem_pi`, `modsem_da`, and `modsem_mplus`)

Value

modsem object with class modsem_pi, modsem_da, or modsem_mplus

Examples

```
library(modsem)
# For more examples, check README and/or GitHub.
# One interaction
m1 <- '
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'

# Double centering approach
est1 <- modsem(m1, oneInt)
summary(est1)

## Not run:
# The Constrained Approach
est1_ca <- modsem(m1, oneInt, method = "ca")
summary(est1_ca)

# LMS approach
est1_lms <- modsem(m1, oneInt, method = "lms")
summary(est1_lms)

# QML approach
est1_qml <- modsem(m1, oneInt, method = "qml")
summary(est1_qml)

## End(Not run)

# Theory Of Planned Behavior
tpb <- '
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC
  BEH ~ INT:PBC
'
```

```
# Double centering approach
est_tpb <- modsem(tpb, data = TPB)
summary(est_tpb)

## Not run:
# The Constrained Approach
est_tpb_ca <- modsem(tpb, data = TPB, method = "ca")
summary(est_tpb_ca)

# LMS approach
est_tpb_lms <- modsem(tpb, data = TPB, method = "lms")
summary(est_tpb_lms)

# QML approach
est_tpb_qml <- modsem(tpb, data = TPB, method = "qml")
summary(est_tpb_qml)

## End(Not run)
```

`modsemify` *Generate parameter table for* `lavaan` *syntax*

Description

Generate parameter table for `lavaan` syntax

Usage

```
modsemify(syntax)
```

Arguments

`syntax`	model syntax

Value

`data.frame` with columns `lhs`, `op`, `rhs`, `mod`

Examples

```
library(modsem)
m1 <- '
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'
modsemify(m1)
```

modsem_da *Interaction between latent variables using lms and qml approaches*

Description

modsem_da() is a function for estimating interaction effects between latent variables in structural equation models (SEMs) using distributional analytic (DA) approaches. Methods for estimating interaction effects in SEMs can basically be split into two frameworks: 1. Product Indicator-based approaches ("dblcent", "rca", "uca", "ca", "pind") 2. Distributionally based approaches ("lms", "qml").

modsem_da() handles the latter and can estimate models using both QML and LMS, necessary syntax, and variables for the estimation of models with latent product indicators.

NOTE: Run default_settings_da to see default arguments.

Usage

```
modsem_da(
  model.syntax = NULL,
  data = NULL,
  method = "lms",
  verbose = NULL,
  optimize = NULL,
  nodes = NULL,
  convergence = NULL,
  optimizer = NULL,
  center.data = NULL,
  standardize.data = NULL,
  standardize.out = NULL,
  standardize = NULL,
  mean.observed = NULL,
  cov.syntax = NULL,
  double = NULL,
  calc.se = NULL,
  FIM = NULL,
  EFIM.S = NULL,
  OFIM.hessian = NULL,
  EFIM.parametric = NULL,
  robust.se = NULL,
  max.iter = NULL,
  max.step = NULL,
  fix.estep = NULL,
  start = NULL,
  epsilon = NULL,
  quad.range = NULL,
  n.threads = NULL,
  ...
)
```

Arguments

model.syntax	lavaan syntax
data	dataframe
method	method to use: "lms" = latent model structural equations (not passed to lavaan). "qml" = quasi maximum likelihood estimation of latent model structural equations (not passed to lavaan).
verbose	should estimation progress be shown
optimize	should starting parameters be optimized
nodes	number of quadrature nodes (points of integration) used in lms, increased number gives better estimates but slower computation. How many are needed depends on the complexity of the model. For simple models, somewhere between 16-24 nodes should be enough; for more complex models, higher numbers may be needed. For models where there is an interaction effect between an endogenous and exogenous variable, the number of nodes should be at least 32, but practically (e.g., ordinal/skewed data), more than 32 is recommended. In cases where data is non-normal, it might be better to use the qml approach instead. For large numbers of nodes, you might want to change the 'quad.range' argument.
convergence	convergence criterion. Lower values give better estimates but slower computation.
optimizer	optimizer to use, can be either "nlminb" or "L-BFGS-B". For LMS, "nlminb" is recommended. For QML, "L-BFGS-B" may be faster if there is a large number of iterations, but slower if there are few iterations.
center.data	should data be centered before fitting model
standardize.data	should data be scaled before fitting model, will be overridden by standardize if standardize is set to TRUE. **NOTE**: It is recommended that you estimate the model normally and then standardize the output using standardized_estimates.
standardize.out	should output be standardized (note will alter the relationships of parameter constraints since parameters are scaled unevenly, even if they have the same label). This does not alter the estimation of the model, only the output. **NOTE**: It is recommended that you estimate the model normally and then standardize the output using standardized_estimates.
standardize	will standardize the data before fitting the model, remove the mean structure of the observed variables, and standardize the output. Note that standardize.data, mean.observed, and standardize.out will be overridden by standardize if standardize is set to TRUE. **NOTE**: It is recommended that you estimate the model normally and then standardize the output using standardized_estimates.
mean.observed	should the mean structure of the observed variables be estimated? This will be overridden by standardize if standardize is set to TRUE. **NOTE**: Not recommended unless you know what you are doing.

cov.syntax	model syntax for implied covariance matrix (see vignette("interaction_two_etas", "modsem"))
double	try to double the number of dimensions of integration used in LMS, this will be extremely slow but should be more similar to mplus.
calc.se	should standard errors be computed? **NOTE**: If FALSE, the information matrix will not be computed either.
FIM	should the Fisher information matrix be calculated using the observed or expected values? Must be either "observed" or "expected".
EFIM.S	if the expected Fisher information matrix is computed, EFIM.S selects the sample size of the generated data.
OFIM.hessian	should the observed Fisher information be computed using the Hessian? If FALSE, it is computed using the gradient.
EFIM.parametric	should data for calculating the expected Fisher information matrix be simulated parametrically (simulated based on the assumptions and implied parameters from the model), or non-parametrically (stochastically sampled)? If you believe that normality assumptions are violated, EFIM.parametric = FALSE might be the better option.
robust.se	should robust standard errors be computed? Meant to be used for QML, can be unreliable with the LMS approach.
max.iter	maximum number of iterations.
max.step	maximum steps for the M-step in the EM algorithm (LMS).
fix.estep	if TRUE, the E-step will be fixed, and the prior probabilities will be set to the best prior probabilities, if the log-likelihood decreases for more than 30 iterations.
start	starting parameters.
epsilon	finite difference for numerical derivatives.
quad.range	range in z-scores to perform numerical integration in LMS using Gaussian-Hermite Quadratures. By default Inf, such that f(t) is integrated from -Inf to Inf, but this will likely be inefficient and pointless at a large number of nodes. Nodes outside +/- quad.range will be ignored.
n.threads	number of cores to use for parallel processing. If NULL, it will use <= 2 threads. If an integer is specified, it will use that number of threads (e.g., n.threads = 4 will use 4 threads). If "default", it will use the default number of threads (2). If "max", it will use all available threads, "min" will use 1 thread.
...	additional arguments to be passed to the estimation function.

Value

modsem_da object

Examples

```
library(modsem)
# For more examples, check README and/or GitHub.
# One interaction
```

```
m1 <- "
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
"

## Not run:
# QML Approach
est1 <- modsem_da(m1, oneInt, method = "qml")
summary(est1)

# Theory Of Planned Behavior
tpb <- "
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  # Covariances
  ATT ~~ SN + PBC
  PBC ~~ SN
  # Causal Relationships
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC
  BEH ~ INT:PBC
"

# LMS Approach
estTpb <- modsem_da(tpb, data = TPB, method = lms)
summary(estTpb)

## End(Not run)
```

`modsem_inspect` *Inspect model information*

Description

function used to inspect fittet object. similar to 'lavInspect()' argument 'what' decides what to inspect

Usage

```
modsem_inspect(object, what = NULL, ...)
```

Arguments

object	fittet model to inspect
what	what to inspect
...	Additional arguments passed to other functions

Details

for ‘modsem_da‘, and ‘modsem_lavaan‘ for ‘modsem_lavaan‘, it is just a wrapper for ‘lavInspect()‘ for ‘modsem_da‘ and “ what can either be "all", "matrices", "optim", or just the name of what to extract.

---

modsem_mplus *Estimation latent interactions through mplus*

Description

Estimation latent interactions through mplus

Usage

```
modsem_mplus(
  model.syntax,
  data,
  estimator = "ml",
  type = "random",
  algorithm = "integration",
  process = "8",
  ...
)
```

Arguments

model.syntax	lavaan/modsem syntax
data	dataset
estimator	estimator argument passed to mplus
type	type argument passed to mplus
algorithm	algorithm argument passed to mplus
process	process argument passed to mplus
...	arguments passed to other functions

Value

modsem_mplus object

Examples

```
# Theory Of Planned Behavior
tpb <- '
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  # Covariances
  ATT ~~ SN + PBC
  PBC ~~ SN
  # Causal Relationsships
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC
  BEH ~ INT:PBC
'

## Not run:
estTpbMplus <- modsem_mplus(tpb, data = TPB)
summary(estTpbLMS)

## End(Not run)
```

`modsem_pi` *Interaction between latent variables using product indicators*

Description

`modsem_pi()` is a function for estimating interaction effects between latent variables, in structural equation models (SEMs), using product indicators. Methods for estimating interaction effects in SEMs can basically be split into two frameworks: 1. Product Indicator based approaches (`"dblcent"`, `"rca"`, `"uca"`, `"ca"`, `"pind"`), and 2. Distributionally based approaches (`"lms"`, `"qml"`). `modsem_pi()` is essentially a fancy wrapper for `lavaan::sem()` which generates the necessary syntax and variables for the estimation of models with latent product indicators. Use `default_settings_pi()` to get the default settings for the different methods.

Usage

```
modsem_pi(
  model.syntax = NULL,
  data = NULL,
  method = "dblcent",
  match = NULL,
  standardize.data = FALSE,
```

```
  center.data = FALSE,
  first.loading.fixed = TRUE,
  center.before = NULL,
  center.after = NULL,
  residuals.prods = NULL,
  residual.cov.syntax = NULL,
  constrained.prod.mean = NULL,
  constrained.loadings = NULL,
  constrained.var = NULL,
  constrained.res.cov.method = NULL,
  auto.scale = "none",
  auto.center = "none",
  estimator = "ML",
  group = NULL,
  run = TRUE,
  suppress.warnings.lavaan = FALSE,
  ...
)
```

Arguments

model.syntax	lavaan syntax
data	dataframe
method	method to use: "rca" = residual centering approach (passed to lavaan), "uca" = unconstrained approach (passed to lavaan), "dblcent" = double centering approach (passed to lavaan), "pind" = prod ind approach, with no constraints or centering (passed to lavaan), "custom" = use parameters specified in the function call (passed to lavaan)
match	should the product indicators be created by using the match-strategy
standardize.data	should data be scaled before fitting model
center.data	should data be centered before fitting model
first.loading.fixed	Should the first factor loading in the latent product be fixed to one?
center.before	should indicators in products be centered before computing products (overwritten by method, if method != NULL)
center.after	should indicator products be centered after they have been computed?
residuals.prods	should indicator products be centered using residuals (overwritten by method, if method != NULL)
residual.cov.syntax	should syntax for residual covariances be produced (overwritten by method, if method != NULL)
constrained.prod.mean	should syntax for product mean be produced (overwritten by method, if method != NULL)

constrained.loadings
: should syntax for constrained loadings be produced (overwritten by method, if method != NULL)

constrained.var
: should syntax for constrained variances be produced (overwritten by method, if method != NULL)

constrained.res.cov.method
: method for constraining residual covariances

auto.scale
: methods which should be scaled automatically (usually not useful)

auto.center
: methods which should be centered automatically (usually not useful)

estimator
: estimator to use in lavaan

group
: group variable for multigroup analysis

run
: should the model be run via lavaan, if FALSE only modified syntax and data is returned

suppress.warnings.lavaan
: should warnings from lavaan be suppressed?

...
: arguments passed to other functions, e.g., lavaan

Value

modsem object

Examples

```
library(modsem)
# For more examples, check README and/or GitHub.
# One interaction
m1 <- '
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'

# Double centering approach
est1 <- modsem_pi(m1, oneInt)
summary(est1)

## Not run:
# The Constrained Approach
est1Constrained <- modsem_pi(m1, oneInt, method = "ca")
summary(est1Constrained)

## End(Not run)

# Theory Of Planned Behavior
```

```
tpb <- '
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  # Covariances
  ATT ~~ SN + PBC
  PBC ~~ SN
  # Causal Relationships
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC
  BEH ~ INT:PBC
'

# Double centering approach
estTpb <- modsem_pi(tpb, data = TPB)
summary(estTpb)

## Not run:
# The Constrained Approach
estTpbConstrained <- modsem_pi(tpb, data = TPB, method = "ca")
summary(estTpbConstrained)

## End(Not run)
```

`multiplyIndicatorsCpp` *Multiply indicators*

Description

Multiply indicators

Usage

```
multiplyIndicatorsCpp(df)
```

Arguments

`df`	A data DataFrame

Value

A NumericVector

oneInt *oneInt*

Description

A simulated dataset with one interaction effect

parameter_estimates *Extract parameterEstimates from an estimated model*

Description

Extract parameterEstimates from an estimated model

Usage

```
parameter_estimates(object, ...)
```

Arguments

object	An object of class modsem_pi, modsem_da, or modsem_mplus
...	Additional arguments passed to other functions

plot_interaction *Plot Interaction Effects*

Description

Plot Interaction Effects

Usage

```
plot_interaction(
  x,
  z,
  y,
  xz = NULL,
  vals_x = seq(-3, 3, 0.001),
  vals_z,
  model,
  alpha_se = 0.15,
  ...
)
```

Arguments

x	The name of the variable on the x-axis
z	The name of the moderator variable
y	The name of the outcome variable
xz	The name of the interaction term. If the interaction term is not specified, it will be created using x and z.
vals_x	The values of the x variable to plot, the more values the smoother the std.error-area will be
vals_z	The values of the moderator variable to plot. A separate regression line (y ~ x | z) will be plotted for each value of the moderator variable
model	An object of class modsem_pi, modsem_da, or modsem_mplus
alpha_se	The alpha level for the std.error area
...	Additional arguments passed to other functions

Value

A ggplot object

Examples

```
library(modsem)
## Not run:
m1 <- "
# Outer Model
  X =~ x1
  X =~ x2 + x3
  Z =~ z1 + z2 + z3
  Y =~ y1 + y2 + y3

# Inner model
  Y ~ X + Z + X:Z
"
est1 <- modsem(m1, data = oneInt)
plot_interaction("X", "Z", "Y", "X:Z", -3:3, c(-0.2, 0), est1)

tpb <- "
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  # Causal Relationsships
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC
  # BEH ~ ATT:PBC
```

```
  BEH ~ PBC:INT
  # BEH ~ PBC:PBC
"

est2 <- modsem(tpb, TPB, method = "lms")
plot_interaction(x = "INT", z = "PBC", y = "BEH", xz = "PBC:INT",
                 vals_z = c(-0.5, 0.5), model = est2)

## End(Not run)
```

`standardized_estimates` *Get standardized estimates*

Description

Get standardized estimates

Usage

```
standardized_estimates(object, ...)
```

Arguments

`object`	An object of class `modsem_da`, `modsem_mplus`, or a `parTable` of class `data.frame`
`...`	Additional arguments passed to other functions

Details

For `modsem_da`, and `modsem_mplus` objects, the interaction term is not standardized such that `var(xz) = 1`. The interaction term is not an actual variable in the model, meaning that it does not have a variance. It must therefore be calculated from the other parameters in the model. Assuming normality and zero-means, the variance is calculated as `var(xz) = var(x) * var(z) + cov(x, z)^2`. Thus setting the variance of the interaction term to 1 would only be 'correct' if the correlation between `x` and `z` is zero. This means that the standardized estimates for the interaction term will be different from those using `lavaan`, since there the interaction term is an actual latent variable in the model, with a standardized variance of 1.

summary.modsem_da *summary for modsem objects*

Description

summary for modsem objects

summary for modsem objects

summary for modsem objects

Usage

```
## S3 method for class 'modsem_da'
summary(
  object,
  H0 = TRUE,
  verbose = TRUE,
  r.squared = TRUE,
  adjusted.stat = FALSE,
  digits = 3,
  scientific = FALSE,
  ci = FALSE,
  standardized = FALSE,
  loadings = TRUE,
  regressions = TRUE,
  covariances = TRUE,
  intercepts = TRUE,
  variances = TRUE,
  var.interaction = FALSE,
  ...
)

## S3 method for class 'modsem_mplus'
summary(
  object,
  scientific = FALSE,
  standardize = FALSE,
  ci = FALSE,
  digits = 3,
  loadings = TRUE,
  regressions = TRUE,
  covariances = TRUE,
  intercepts = TRUE,
  variances = TRUE,
  ...
)
```

```
## S3 method for class 'modsem_pi'
summary(object, ...)
```

Arguments

object	modsem object to summarized
H0	should a null model be estimated (used for comparison)
verbose	print progress for the estimation of null model
r.squared	calculate R-squared
adjusted.stat	should sample size corrected/adjustes AIC and BIC be reported?
digits	number of digits to print
scientific	print p-values in scientific notation
ci	print confidence intervals
standardized	print standardized estimates
loadings	print loadings
regressions	print regressions
covariances	print covariances
intercepts	print intercepts
variances	print variances
var.interaction	if FALSE (default) variances for interaction terms will be removed (if present)
...	arguments passed to lavaan::summary()
standardize	standardize estimates

Examples

```
## Not run:
m1 <- "
 # Outer Model
 X =~ x1 + x2 + x3
 Y =~ y1 + y2 + y3
 Z =~ z1 + z2 + z3

 # Inner model
 Y ~ X + Z + X:Z
"

est1 <- modsem(m1, oneInt, "qml")
summary(est1, ci = TRUE, scientific = TRUE)

## End(Not run)
```

TPB *TPB*

Description

A simulated dataset based on the Theory of Planned Behaviour

Examples

```
tpb <- '
# Outer Model (Based on Hagger et al., 2007)
  ATT =~ att1 + att2 + att3 + att4 + att5
  SN =~ sn1 + sn2
  PBC =~ pbc1 + pbc2 + pbc3
  INT =~ int1 + int2 + int3
  BEH =~ b1 + b2

# Inner Model (Based on Steinmetz et al., 2011)
  INT ~ ATT + SN + PBC
  BEH ~ INT + PBC + INT:PBC
'

est <- modsem(tpb, data = TPB)
```

TPB_UK *TPB_UK*

Description

A dataset based on the Theory of Planned Behaviour from a UK sample. 4 variables with high communality were selected for each latent variable (ATT, SN, PBC, INT, BEH), from two time points (t1 and t2).

Source

Gathered from a replciation study of the original by Hagger et al. (2023). Obtained from https://doi.org/10.23668/psycharchiv

Examples

```
tpb_uk <- '
# Outer Model (Based on Hagger et al., 2007)
 ATT =~ att3 + att2 + att1 + att4
 SN =~ sn4 + sn2 + sn3 + sn1
 PBC =~ pbc2 + pbc1 + pbc3 + pbc4
 INT =~ int2 + int1 + int3 + int4
 BEH =~ beh3 + beh2 + beh1 + beh4
```

```
# Inner Model (Based on Steinmetz et al., 2011)
 # Causal Relationsships
 INT ~ ATT + SN + PBC
 BEH ~ INT + PBC
 BEH ~ INT:PBC
'

est <- modsem(tpb_uk, data = TPB_UK)
```

`trace_path` *Estimate formulas for (co-)variance paths using Wright's path tracing rules*

Description

This function estimates the path from `x` to `y` using the path tracing rules. Note that it only works with structural parameters, so `"=~"` are ignored, unless `measurement.model = TRUE`. If you want to use the measurement model, `"~"` should be in the `mod` column of `pt`.

Usage

```
trace_path(
  pt,
  x,
  y,
  parenthesis = TRUE,
  missing.cov = FALSE,
  measurement.model = FALSE,
  maxlen = 100,
  ...
)
```

Arguments

`pt`	A data frame with columns `lhs`, `op`, `rhs`, and `mod`, from modsemify
`x`	Source variable
`y`	Destination variable
`parenthesis`	If `TRUE`, the output will be enclosed in parenthesis
`missing.cov`	If `TRUE`, covariances missing from the model syntax will be added
`measurement.model`	If `TRUE`, the function will use the measurement model
`maxlen`	Maximum length of a path before aborting
`...`	Additional arguments passed to trace_path

Value

A string with the estimated path (simplified if possible)

Examples

```
library(modsem)
m1 <- '
  # Outer Model
  X =~ x1 + x2 +x3
  Y =~ y1 + y2 + y3
  Z =~ z1 + z2 + z3

  # Inner model
  Y ~ X + Z + X:Z
'
pt <- modsemify(m1)
trace_path(pt, x = "Y", y = "Y", missing.cov = TRUE) # variance of Y
```

`var_interactions` *Extract or modify parTable from an estimated model with estimated variances of interaction terms*

Description

Extract or modify parTable from an estimated model with estimated variances of interaction terms

Usage

```
var_interactions(object, ...)
```

Arguments

`object`	An object of class `modsem_da`, `modsem_mplus`, or a parTable of class `data.frame`
`...`	Additional arguments passed to other functions

`vcov_modsem_da` *Wrapper for vcov*

Description

wrapper for vcov, to be used with modsem::vcov_modsem_da, since vcov is not in the namespace of modsem, but stats

Usage

```
vcov_modsem_da(object, ...)
```

Arguments

`object`	fittet model to inspect
`...`	additional arguments

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