Package 'metaSEM'

October 10, 2010

Type Package

Title Meta-Analysis using Structural Equation Modeling

Version 0.5-1										
Date 2010-10-10										
Depends R (>= $2.11.1$), OpenMx										
Imports Matrix										
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Description The metaSEM package conducts univariate and multivariate meta-analyses using a structural equation modeling (SEM) approach via the OpenMx package. It also implements the two-stage SEM approach to conducting meta-analytic structural equation modeling on correlation/covariance matrices.										
License GPL (>=2)										
LazyLoad yes										
LazyData yes										
<pre>URL http://courses.nus.edu.sg/course/psycwlm/Internet/metaSEM/</pre>										
R topics documented:										
metaSEM-package 2 as.mxMatrix 3 asyCov 4 bdiagMat 5 bdiagRep 6 Becker83 8 Berkey98 6 Cheung09 6 coef 1 Digman97 10 HedgesOlkin85 1										

2 metaSEM-package

ndex																							29
	wls		•		•		•	•	•	 •			•	•	•	 	•	•	•		•		25
	vec2symMat																						
	vcov																						
	tssem1																						
	summary .																						
	readData																						
	print																						
	meta																						
	matrix2bdiag																						
	list2matrix .																						
	is.pd																						
	Hox02																						

metaSEM-package

Meta-Analysis using Structural Equation Modeling

Description

The metaSEM package conducts univariate and multivariate meta-analyses using a structural equation modeling (SEM) approach via the OpenMx package. It also implements the two-stage SEM approach to conducting meta-analytic structural equation modeling on correlation/covariance matrices.

Details

Package: metaSEM
Type: Package
Version: 0.5-1
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License: GPL (>=2)
LazyLoad: yes

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

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References

Cheung, M. W. L. (2008). A model for integrating fixed-, random-, and mixed-effects meta-analyses into structural equation modeling. *Psychological Methods*, **13**, 182-202.

Cheung, M. W. L. (2009). Constructing approximate confidence intervals for parameters with structural equation models. *Structural Equation Modeling*, **16**, 267-294.

Cheung, M. W. L. (2010). Modeling multivariate effect sizes with structural equation models. *Manuscript submitted for publication*.

as.mxMatrix 3

Cheung, M. W. L., & Chan, W. (2004). Testing dependent correlation coefficients via structural equation modeling. *Organizational Research Methods*, 7, 206-223.

Cheung, M. W. L., & Chan, W. (2005). Meta-analytic structural equation modeling: A two-stage approach. *Psychological Methods*, **10**, 40-64.

Cheung, M. W. L., & Chan, W. (2009). A two-stage approach to synthesizing covariance matrices in meta-analytic structural equation modeling. *Structural Equation Modeling*, **16**, 28-53.

as.mxMatrix

Convert a Matrix into MxMatrix-class

Description

It converts a matrix into MxMatrix-class via mxMatrix.

Usage

```
as.mxMatrix(x, ...)
```

Arguments

x A character or numeric matrix

Further arguments to be passed to mxMatrix. Please note that type, nrow, ncol, values, free and possibly labels will be created automatically. Thus, these arguments should not be avoided in ...

Details

The dimensions are the same as x. If there are non-numeric values, it will be treated as a free parameter. If an "*" is present, the numeric value on the left hand side will be treated as either a fixed value for a fixed parameter or a starting value for a free parameter. For example, "1" for a fixed parameter with "1" as the value and "5*beta" for a free parameter with "5" as the starting value and "beta" as the label. If it is a matrix of numeric values, there is no free parameters in the output matrix.

Value

A MxMatrix-class object with the same dimensions as x

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
mxMatrix
```

4 asyCov

Examples

```
\# a and b are free parameters with starting values and labels
a1 <- matrix(c(1:4, "5*a", 6, "7*b", 8, 9), ncol=3, nrow=3)
a1 <- as.mxMatrix(a1, name="a1")</pre>
# Fixed parameters without any labels
a2 <- matrix(1:9, ncol=3, nrow=3)
a2 <- as.mxMatrix(a2, name="a2")</pre>
# Free parameters without starting values
a3 <- matrix(c(1:4, "a", 6, "b", 8, 9), ncol=3, nrow=3)
a3 <- as.mxMatrix(a3, name="a3")
# A free parameter without label
a4 \leftarrow matrix(c(1:4, "5*", 6, "7*b", 8, 9), ncol=3, nrow=3)
a4 <- as.mxMatrix(a4, name="a4")
```

asyCov

Asymptotic Covariance Matrix of a Correlation/Covariance Matrix

Description

It estimates asymptotic covariance matrices of the correlation/covariance matrix by assuming multivariate normality.

Usage

```
asyCov(x, n, cor.analysis = TRUE, dropNA = TRUE, as.matrix = TRUE,
       silent = TRUE, suppressWarnings = TRUE, ...)
```

Arguments

Х

A correlation/covariance matrix or a list of correlation/covariance matrices. NA on the variables or other values defined in na.strings will be removed before the analysis. Note that it only checks the diagonal elements of the matrices. If there are missing values, make sure that the diagonals are coded with NA or values defined in na.string.

Sample size or a vector of sample sizes

cor.analysis Logical. The output is either a correlation or covariance matrix.

dropNA

Logical. If it is TRUE, the resultant dimensions will be reduced by dropping the missing variables. If it is FALSE, the resultant dimensions are the same as the input by keeping the missing variables.

as.matrix

Logical. If it is \mathtt{TRUE} and \mathtt{x} is a list of correlation/covariance matrices with the same dimensions, the asymptotic covariance matrices will be column vectorized and stacked together. If it is FALSE, the output will be a list of asymptotic covariance matrices. Note that if it is TRUE, dropNA will be FALSE automatically. This option is useful when passing the asymptotic covariance matrices to

meta

Logical. Argument to be passed to mxRun silent suppressWarnings

Logical. If TRUE, warnings are suppressed. Argument to be passed to mxRun.

Futher arguments to be passed to mxRun . . .

bdiagMat 5

Value

An asymptotic covariance matrix of the vectorized correlation/covariance matrix or a list of these matrices. If as .matrix=TRUE and x is a list of matrices, the output is a stacked matrix.

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

References

Cheung, M. W. L., & Chan, W. (2004). Testing dependent correlation coefficients via structural equation modeling. *Organizational Research Methods*, **7**, 206-223.

Examples

```
C1 <- matrix(c(1,0.5,0.4,0.5,1,0.2,0.4,0.2,1), ncol=3)
asyCov(C1, n=100)

# Data with missing values
C2 <- matrix(c(1,0.4,NA,0.4,1,NA,NA,NA,NA), ncol=3)
C3 <- matrix(c(1,0.2,0.2,1), ncol=2)

# Output is a list of asymptotic covariance matrices
asyCov(list(C1,C2,C3), n=c(100,50,50), as.matrix=FALSE)

# Output is a stacked matrix of asymptotic covariance matrices
asyCov(list(C1,C2), n=c(100,50), as.matrix=TRUE)</pre>
```

bdiagMat

Create a Block Diagonal Matrix

Description

It creates a block diagonal matrix from a list of numeric or character matrices.

Usage

```
bdiagMat(x)
```

Arguments

Х

A list of numeric or character matrices (or values)

Value

A numeric or character block diagonal matrix

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

6 bdiagRep

References

It was based on a function posted by Scott Chasalow at http://www.math.yorku.ca/Who/Faculty/Monette/pub/stmp/0827.html.

See Also

```
bdiagRep, matrix2bdiag
```

Examples

bdiagRep

Create a Block Diagonal Matrix by Repeating the Input

Description

It creates a block diagonal matrix by repeating the input matrix several times.

Usage

```
bdiagRep(x, times)
```

Arguments

x A numeric or character matrix (or values)
times Number of times of x to be repeated

Value

A numeric or character block diagonal matrix

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
bdiagMat, matrix2bdiag
```

```
# Block diagonal matrix of numerics
bdiagRep( matrix(1:4,nrow=2,ncol=2), 2 )

# Block diagonal matrix of characters
bdiagRep( matrix(letters[1:4],nrow=2,ncol=2), 2 )
```

Becker83

Becker83

Studies on Sex Differences in Conformity Reported by Becker (1983)

Description

Studies on sex differences in conformity using the fictitious norm group paradigm reported by Becker (1983).

Usage

```
data(Becker83)
```

Details

The variables are:

study study number

di Standardized mean difference

vi Sampling variance of the effect size

percentage Percentage of male authors

items Number of items

Source

Becker, B. J. (1983, April). Influence again: A comparison of methods for meta-analysis. *Paper presented at the annual meeting of the American Educational Research Association, Montreal.*

Hedges, L. V., & Olkin, I. (1985). Statistical methods for meta-analysis. Orlando, FL: Academic Press.

References

Cheung, M. W. L. (2010). Fixed-effects meta-analyses as multiple-group structural equation models. *Structural Equation Modeling*, **17**, 481-509.

```
data(Becker83)
## maybe str(Becker83); plot(Becker83) ...
```

8 Cheung09

Berkey98

Five Published Trails from Berkey et al. (1998)

Description

Five published trials, reported by Berkey et al. (1998), comparing surgical and non-surgical treatments for medium-severity periodontal disease, one year after treatment.

Usage

```
data(Berkey98)
```

Details

The variables are:

trial Trial number

pub_year Publication year

no_of_patients Number of patients

PD Patient improvements (mm) in probing depth

AL Patient improvements (mm) in attachment level

var_PD Sampling variance of PD

cov_PD_AL Sampling covariance between PD and AD

var_AL Sampling variance of AL

Source

Berkey, C. S., Hoaglin, D. C., Antczak-Bouckoms, A., Mosteller, F, & Colditz, G. A. (1998). Meta-analysis of multiple outcomes by regression with random effects. *Statistics in Medicine*, **17**, 2537-2550.

Examples

```
data(Berkey98)
## maybe str(Berkey98); plot(Berkey98) ...
```

Cheung09

Data Set from TSSEM User's Guide Version 1.11 by Cheung (2009)

Description

Four studies were selected from the data set used by Cheung and Chan (2005; 2009). Some variables were randomly deleted to illustrate the analysis with missing data.

Usage

```
data(Cheung09)
```

coef 9

Details

A list of data with the following structure:

data A list of 4 studies of correlation matrices

n A vector of sample sizes

Source

Cheung, M. W. L. (2009). TSSEM: A LISREL syntax generator for two-stage structural equation modeling (Version 1.11) [Computer software]. Retrieved from http://courses.nus.edu.sg/course/psycwlm/internet/tssem.zip.

References

Cheung, M. W. L., & Chan, W. (2005). Meta-analytic structural equation modeling: A two-stage approach. *Psychological Methods*, **10**, 40-64.

Cheung, M. W. L., & Chan, W. (2009). A two-stage approach to synthesizing covariance matrices in meta-analytic structural equation modeling. *Structural Equation Modeling*, **16**, 28-53.

Examples

```
data(Cheung09)
## maybe str(Cheung09)
```

coef

Extract Parameter Estimates from tssem1, wls and meta Objects

Description

It extracts the parameter estimates from either tssem1, wls or meta objects.

Usage

```
## S3 method for class 'tssem1':
coef(object, ...)
## S3 method for class 'wls':
coef(object, ...)
## S3 method for class 'meta':
coef(object, ...)
```

Arguments

```
object An object returned from either class tssem1, class wls or class meta
... Further arguments; currently none is used)
```

Value

Parameter estimates for both fixed- and random-effects (if any)

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

Digman97

See Also

```
tssem1, wls, meta
```

Examples

```
## Random-effects meta-analysis
model1 <- with( Hox02, meta(y=yi, v=vi) )
coef(model1)</pre>
```

Digman97

Factor Correlation Matrices of Big Five Model from Digman (1997)

Description

Fourteen studies of the factor correlation matrices of the Big Five Model of personality reported by Digman (1997).

Usage

```
data(Digman97)
```

Details

A list of data with the following structure:

data A list of 14 studies of correlation matrices. The variables are Extraversion (E), Agreeableness (A), Conscientiousness (C), Emotional Stability (ES) and Intellect (I)

n A vector of sample sizes

cluster Types of participants of the studies

Source

Digman, J.M. (1997). Higher-order factors of the Big Five. *Journal of Personality and Social Psychology*, **73**, 1246-1256.

References

Cheung, M. W. L., & Chan, W. (2005). Classifying correlation matrices into relatively homogeneous subgroups: A cluster analytic approach. *Educational and Psychological Measurement*, **65**, 954-979.

```
data(Digman97)
## maybe str(Digman97) ...
```

HedgesOlkin85

HedgesOlkin85

Effects of Open Education Reported by Hedges and Olkin (1985)

Description

Effects of open education on attitude toward school and on reading achievement reported by Hedges and Olkin (1985).

Usage

```
data(HedgesOlkin85)
```

Details

The variables are:

study Study number

d_att Standardized mean difference on attitude

d_ach Standardized mean difference on achievement

var_att Sampling variance of the effect size of attitude

cov_att_ach Sampling covariance between the effect sizes

var_ach Sampling variance of the effect size of achievement

Source

Hedges, L. V., & Olkin, I. (1985). *Statistical methods for meta-analysis*. Orlando, FL: Academic Press.

References

Cheung, M. W. L. (2010). Fixed-effects meta-analyses as multiple-group structural equation models. *Structural Equation Modeling*, **17**, 481-509.

```
data(HedgesOlkin85)
## maybe str(HedgesOlkin85); plot(HedgesOlkin85)...
```

12 homoStat

homoSt.at.

Test Statistic on Homogeneity of Effect Sizes

Description

A test statistic on the homogeneity of univariate and multivariate effect sizes.

Usage

```
homoStat(y, v)
```

Arguments

A vector of effect size for univariate meta-analysis or a $k \times p$ matrix of effect sizes for multivariate meta-analysis where k is the number of studies and p is the number of effect sizes.

A vector of the sampling variance of the effect size for univarite meta-analysis or a $k \times p*$ matrix of the sampling covariance matrix of the effect sizes for multivariate meta-analysis where p*=p(p+1)/2. It is arranged by column major as used by vech. It is assumed that there is no missing value on v. If there are missing values on v due to the missing values on y, the missing values on v will be removed automatically.

Value

A list of

Q statistic on the null hypothesis of homogeneity of effect sizes. It has an approximate chi-square distribution under the null hypothesis. If fact, it is a badness-of-fit index.

Q.df Degrees of freedom of the Q statistic

pval p value on the test of homogeneity of effect sizes

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

References

Becker, B. J. (1992). Using results from replicated studies to estimate linear models. *Journal of Educational Statistics*, **17**, 341-362.

Cheung, M. W. L. (2010). Fixed-effects meta-analyses as multiple-group structural equation models. *Structural Equation Modeling*, **17**, 481-509.

Cochran, W. G. (1954). The combination of estimates from different experiments. *Biometrics*, **10**, 101-129.

See Also

meta

Hox02

Examples

Hox02

Simulated Effect Sizes Reported by Hox (2002)

Description

Twenty stimulated studies on standardized mean difference and one continuous study characteristic reported by Hox (2002).

Usage

```
data(Hox02)
```

Details

The variables are:

study Study number

yi Effect size (standardized mean difference)

vi Sampling variance of the effect size

weeks Duration of the experimental intervention in terms of weeks

Source

Hox, J. J. (2002). *Multilevel analysis: Techniques and applications*. Mahwah, N.J.: Lawrence Erlbaum Associates.

References

Cheung, M. W. L. (2008). A model for integrating fixed-, random-, and mixed-effects meta-analyses into structural equation modeling. *Psychological Methods*, **13**, 182-202.

```
data(Hox02) ## maybe str(Hox02); plot(Hox02) ...
```

14 list2matrix

is.pd

Test Positive Definiteness of a List of Square Matrices

Description

It tests the positive definiteness of a square matrix or a list of square matrices. It returns FALSE if the matrix is not symmetric. Variables with NA in the diagonals will be removed before the test.

Usage

```
is.pd(x, tol = 1e-06)
```

Arguments

x A square matrix or a list of square matrices

tol

Relative tolerance of positiveness of smallest eigenvalue compared to largest eigenvalue. The matrix is considered positive definite if the ratio of the smallest eigenvalue to the largest eigenvalue is larger than tol. See nearPD

Value

TRUE or FALSE or a list of it.

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

Examples

```
A <- diag(1,3)
is.pd(A)
# TRUE

B <- matrix(c(1,2,2,1), ncol=2)
is.pd(B)
# FALSE

is.pd(list(A, B))
# TRUE FALSE</pre>
```

list2matrix

Convert a List of Symmetric Matrices into a Stacked Matrix

Description

It converts a list of symmetric matrices into a stacked matrix. Dimensions of the symmetric matrices have to be the same. It tries to preserve the dimension names if possible. Dimension names will be created if there is no dimension names in the first symmetric matrices.

matrix2bdiag 15

Usage

```
list2matrix(x, diag = FALSE)
```

Arguments

 \mathbf{x} A list of $k p \mathbf{x} p$ symmetric matrices.

diag Logical. If it is TRUE, vech is used to vectorize the matrices. If it is FALSE,

vechs is used to vectorize the matrices.

Value

A k x p* stacked matrix where p*=p(p-1)/2 for diag=FALSE or p*=p(p+1)/2 for diag=TRUE.

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

Examples

```
C1 <- matrix(c(1,0.5,0.4,0.5,1,0.2,0.4,0.2,1), ncol=3) asyCov(C1, n=100)  
C2 <- matrix(c(1,0.4,NA,0.4,1,NA,NA,NA,NA), ncol=3) list2matrix(list(C1, C2))  
dimnames(C1) <- list( c("x","y","z"), c("x","y","z") ) dimnames(C2) <- list( c("x","y","z"), c("x","y","z") ) list2matrix(list(C1, C2))
```

matrix2bdiag

Convert a Matrix into a Block Diagonal Matrix

Description

It converts a matrix into a block diagonal matrix.

Usage

```
matrix2bdiag(x, ...)
```

Arguments

 \mathbf{x} A $k \mathbf{x} p$ matrix of numerics or characters.

... Further arguments to be passed to vec2symMat

Details

Each row of x is converted into a symmetric matrix via vec2symMat. Then the list of the symmetric matrices is converted into a block diagonal matrix via a function written by Scott Chasalow posted at http://www.math.yorku.ca/Who/Faculty/Monette/pub/stmp/0827.html.

16 meta

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
vec2symMat
```

Examples

```
matrix2bdiag( matrix(1:18, ncol=6, byrow=TRUE) )
matrix2bdiag( matrix(letters[1:18], ncol=6, byrow=TRUE) )
```

meta

Univariate and Multivariate Meta-Analysis

Description

It conducts univariate and multivariate meta-analysis with maximum likelihood estimation method. Mixed-effects meta-analysis can be conducted by including study characteristics as predictors. Equality constraints on intercepts, regression coefficients and variance components can be easily imposed.

Usage

```
meta(y, v, x, intercept.constraints, coeff.constraints, RE.constraints, RE.startvalues=0.1, RE.lbound = 1e-10, intervals.type = c("z", "LB"), suppressWarnings = TRUE, ...)
```

Arguments

V

A vector of effect size for univariate meta-analysis or a $k \times p$ matrix of effect sizes for multivariate meta-analysis where k is the number of studies and p is the number of effect sizes.

A vector of the sampling variance of the effect size for univarite meta-analysis or a $k \times p*$ matrix of the sampling covariance matrix of the effect sizes for multivariate meta-analysis where p*=p(p+1)/2. It is arranged by column major as used by vech.

 ${\bf x}$ A predictor or a k x m matrix of predictors where m is the number of predictors. intercept.constraints

A $1 \times p$ matrix specifying whether the intercepts of the effect sizes are fixed or free. The default is that the intercepts are free. When there is no predictor, these intercepts are the same as the pooled effect sizes. The format of this matrix follows as.mxMatrix. The parameter estimates will be constrained equally if the labels are the same.

coeff.constraints

A $p \times m$ matrix specifying how the predictors predict the effect sizes. The default is that all m predictors predict all p effect sizes. The format of this matrix follows as .mxMatrix. The parameter estimates will be constrained equally if the labels are the same.

meta 17

RE.constraints

A $p \times p$ matrix specifying the variance componets of the random effects. The default is that all covariance/variance components are free. The format of this matrix follows as.mxMatrix. The parameter estimates will be constrained equally if the labels are the same. If a zero matrix is specified, it becomes a fixed-effects meta-analysis.

RE.startvalues

A vector of p starting values on the diagonals of the variance component of the random effects. If only one scalar is given, it will be duplicated across the diagonals. Starting values for the off-diagonals of the variance component are all 0. A $p \times p$ symmetric matrix of starting values is also acceptable.

RE.lbound

A vector of p lower bounds on the diagonals of the variance component of the random effects. If only one scalar is given, it will be duplicated across the diagonals. Lower bounds for the off-diagonals of the variance component are set at NA. A $p \times p$ symmetric matrix of the lower bounds is also acceptable

intervals.type

Either z (default if missing) or LB. If it is z, it calculates 95% Wald confidence intervals (CIs) based on the z statistic. If it is LB, it calculates 95% likelihood-based CIs on the parameter estimates. Note that the z values and their associated p values are based on the z statistic. They are not related to the likelihood-based CIs.

suppressWarnings

Logical. If TRUE, warnings are suppressed. Argument to be passed to mxRun.

Futher arguments to be passed to mxRun

Value

An object of class meta with a list of

call	Object returned by match.call
data	A data matrix of y, v and x
no.y	No. of effect sizes
no.x	No. of predictors
miss.x	A vector indicating whether the predictors are missing. Studies will be removed before the analysis if they are ${\tt TRUE}$
meta.fit	A fitted object returned from mxRun

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

References

Cheung, M. W. L. (2008). A model for integrating fixed-, random-, and mixed-effects meta-analyses into structural equation modeling. *Psychological Methods*, **13**, 182-202.

Cheung, M. W. L. (2009). Constructing approximate confidence intervals for parameters with structural equation models. *Structural Equation Modeling*, **16**, 267-294.

Cheung, M. W. L. (2010). Modeling multivariate effect sizes with structural equation models. *Manuscript submitted for publication.*

Hardy, R. J., & Thompson, S. G. (1996). A likelihood approach to meta-analysis with random effects. *Statistics in Medicine*, **15**, 619-629.

18 print

Neale, M. C., & Miller, M. B. (1997). The use of likelihood-based confidence intervals in genetic models. *Behavior Genetics*, **27**, 113-120.

Examples

```
## Random-effects meta-analysis
attach (Hox02)
summary( meta(y=yi, v=vi) )
## Fixed-effects meta-analysis
summary( meta(y=yi, v=vi, RE.constraints=matrix(0, ncol=1, nrow=1)) )
## Mixed-effects meta-analysis with "weeks" as a predictor
## Use likelihood-based CI
summary( meta(y=yi, v=vi, x=weeks, intervals.type="LB") )
detach (Hox02)
## Multivariate meta-analysis
attach (Berkey98)
summary( meta(y=cbind(PD, AL), v=cbind(var_PD, cov_PD_AL, var_AL)) )
## Multivariate meta-analysis with "publication year-1979" as the predictor
summary( meta(y=cbind(PD, AL), v=cbind(var_PD, cov_PD_AL, var_AL),
         x=scale(pub_year, center=1979)) )
## Multivariate meta-analysis with an equality constraint on regression coefficients
summary( meta(y=cbind(PD, AL), v=cbind(var_PD, cov_PD_AL, var_AL),
         x=scale(pub_year, center=1979),
         coeff.constraints=matrix(c("0.3*Eq_slope", "0.3*Eq_slope"), nrow=2)) )
detach (Berkey98)
```

print

Print Methods for tssem1, wls and meta Objects

Description

Print methods for either tssem1, wls or meta objects.

Usage

```
## S3 method for class 'tssem1':
print(x, ...)
## S3 method for class 'wls':
print(x, ...)
## S3 method for class 'meta':
print(x, ...)
```

Arguments

```
x An object returned from either class tssem1, class wls or class meta... Further arguments to be passed to summary.default
```

readData 19

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
tssem1, wls, meta
```

readData

Read External Correlation/Covariance Matrices

Description

It reads full/lower triangle/stacked vectors of correlation/covariance data into a list of correlation/covariance matrices.

Usage

```
readFullMat(file, ...)
readStackVec(file, ...)
readLowTriMat(file, no.var, ...)
```

Arguments

```
file File name of the data.
no.var Number of variables in the data.
... Further arguments to be passed to scan for readLowTriMat and to read.table
for readFullMat and readStackVec.
```

Value

A list of correlation/covariance matrices.

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

```
## Not run:
# Write two full correlation matrices into a file named "fullmat.dat".
# x2 is missing in the second matrix.
# The content of "fullmat.dat" is
#1.0 0.3 0.4
#0.3 1.0 0.5
#0.4 0.5 1.0
#1.0 NA 0.4
#NA NA NA
#0.4 NA 1.0
cat("1.0 0.3 0.4\n0.3 1.0 0.5\n0.4 0.5 1.0\n1.0 NA 0.4\nNA NA NA\n0.4 NA 1.0",
file="fullmat.dat", sep="")
```

20 readData

```
# Read the correlation matrices
my.full <- readFullMat("fullmat.dat")</pre>
my.full
#$`1`
# x1 x2 x3
#x1 1.0 0.3 0.4
#x2 0.3 1.0 0.5
#x3 0.4 0.5 1.0
#$`2`
    x1 x2 x3
#x1 1.0 NA 0.4
#x2 NA NA NA
#x3 0.4 NA 1.0
# Write two lower triangle correlation matrices into a file named "lowertriangle.dat".
\# x2 is missing in the second matrix.
# The content of "lowertriangle.dat" is
#1.0
#0.3 1.0
#0.4 0.5 1.0
#1.0
#NA NA
#0.4 NA 1.0
cat("1.0\n0.3 1.0\n0.4 0.5 1.0\n1.0\nNA NA\n0.4 NA 1.0",
    file="lowertriangle.dat", sep="")
# Read the lower triangle correlation matrices
my.lowertri <- readLowTriMat(file = "lowertriangle.dat", no.var = 3)</pre>
my.lowertri
#$`1`
    x1 x2 x3
#x1 1.0 0.3 0.4
#x2 0.3 1.0 0.5
#x3 0.4 0.5 1.0
#$`2`
# x1 x2 x3
#x1 1.0 NA 0.4
#x2 NA NA NA
#x3 0.4 NA 1.0
# Write two vectors of correlation coefficients based on
# column major into a file named "stackvec.dat".
# x2 is missing in the second matrix.
# The content of "stackvec.dat" is
#1.0 0.3 0.4 1.0 0.5 1.0
#1.0 NA 0.4 NA NA 1.0
cat("1.0 0.3 0.4 1.0 0.5 1.0\n1.0 NA 0.4 NA NA 1.0",
    file="stackvec.dat", sep="")
my.vec <- readStackVec("stackvec.dat")</pre>
my.vec
#$`1`
```

summary 21

```
# x1 x2 x3

#x1 1.0 0.3 0.4

#x2 0.3 1.0 0.5

#x3 0.4 0.5 1.0

#

#$`2`

# x1 x2 x3

#x1 1.0 NA 0.4

#x2 NA NA NA

#x3 0.4 NA 1.0

### End(Not run)
```

summary

Summary Method for tssem1, wls and meta Objects

Description

It summaries results for either tssem1, wls or meta objects.

Usage

```
## S3 method for class 'tssem1':
summary(object, ...)
## S3 method for class 'wls':
summary(object, ...)
## S3 method for class 'meta':
summary(object, ...)
## S3 method for class 'tssem1':
print.summary(x, ...)
## S3 method for class 'wls':
print.summary(x, ...)
## S3 method for class 'meta':
print.summary(x, ...)
```

Arguments

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
tssem1, wls, meta
```

22 tssem1

tssem1

First Stage of the Two-Stage Structural Equation Modeling (TSSEM)

Description

It conducts the first stage analysis of TSSEM by pooling correlation/covariance matrices with a fixed-effets model. The function expects that there is no missing data in the first group.

Usage

Arguments

my .df A list of correlation/covariance matrices

n A vector of sample sizes

start.values A vector of starting values for the pooled correlation/covariance matrix based

on column major. If it is missing, .startValues will be used to generate the

starting values.

cor.analysis Logical. The output is either a pooled correlation or a covariance matrix.

suppressWarnings

Logical. If TRUE, warnings are suppressed. Argument to be passed to mxRun.

... Further arguments to be passed to mxRun

Value

call The matched call

data A list of correlation/covariance matrices from input

pooledS The pooled correlation/covariance matrix

acovS The asymptotic sampling covariance matrix of the pooled correlation/covariance

matrix

total.n Total sample size of all studies

modelMinus2LL

-2LogLikelihood of the model

independentMinus2LL

-2LogLikelihood of the independent model returned by .minus2LL

saturatedMinus2LL

-2LogLikelihood of the saturated model returned by .minus2LL

tssem1.fit A fitted object returned from mxRun

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

vcov 23

References

Cheung, M. W. L., & Chan, W. (2005). Meta-analytic structural equation modeling: A two-stage approach. *Psychological Methods*, **10**, 40-64.

Cheung, M. W. L., & Chan, W. (2009). A two-stage approach to synthesizing covariance matrices in meta-analytic structural equation modeling. *Structural Equation Modeling*, **16**, 28-53.

See Also

wls

Examples

```
digman1 <- tssem1(Digman97$data, Digman97$n)
summary(digman1)</pre>
```

vcov

Extract Variance-Covariance Matrix of Parameter Estimates from tssem1, wls and meta Objects

Description

It extracts the variance-covariance matrix of the parameter estimates from either tssem1, wls or meta objects.

Usage

```
## S3 method for class 'tssem1':
vcov(object, ...)
## S3 method for class 'wls':
vcov(object, ...)
## S3 method for class 'meta':
vcov(object, ...)
```

Arguments

```
object An object returned from either class tssem1, class wls or class meta
... Further arguments; currently none is used
```

Value

A variance-covariance matrix of the parameter estimates for both fixed- and random-effects (if any)

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
tssem1, wls, meta
```

24 vec2symMat

Examples

```
## Random-effects meta-analysis
model1 <- with( Hox02, meta(y=yi, v=vi) )
vcov(model1)</pre>
```

vec2symMat

Convert a Vector into a Symmetric Matrix

Description

It converts a vector into a symmetric matrix.

Usage

```
vec2symMat(x, diag = TRUE)
```

Arguments

x A vector of numerics or characters

diag

Logical. If it is TRUE (the default), the diagonals of the created matrix are replaced by elements of x; otherwise, the diagonals of the created matrix are replaced by "1".

Value

A symmetric square matrix based on column major

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

See Also

```
matrix2bdiag
```

```
vec2symMat(1:6)
vec2symMat(letters[1:6])
vec2symMat(1:6, diag=FALSE)
```

wls	Conduct a Correlation/Covariance Structure Analysis with Weighte	d
	Least Squares	

Description

It fits a correlation or covariance structure with weighted least squares (WLS) where the inverse of the asymptotic covariance matrix is used as the weight matrix. tssem2 conducts the second stage analysis of the two-stage strutural equation modeling (TSSEM). tssem2 is a wrapper of wls.

Usage

```
wls(S, acovS, n, impliedS, matrices, cor.analysis = TRUE,
    intervals.type = c("z", "LB"), suppressWarnings = TRUE, ...)
tssem2(tssem1.obj, impliedS, matrices, intervals.type = c("z", "LB"),
    suppressWarnings = TRUE, ...)
```

Arguments

tssem1.obj	An object returned from tssem1
S	A $p \times p$ sample correlation/covariance matrix where p is the number of variables
acovS	A $p*$ x $p*$ asymptotic sampling covariance matrix of either vechs (S) or vech (S) where $p*=p(p-1)/2$ for covariance matrix and $p*=p(p+1)/2$ for covariance matrix
n	Sample size
impliedS	Model implied correlation/covariance matrix of an object of either MxMatrix-class or MxAlgebra-class
matrices	A list of matrices used to calculate impliedS. They are objects of either MxMatrix-class or MxAlgebra-class
cor.analysis	Logical. Analysis of correlation or covariance structure. If cor.analysis=TRUE, vechs is used to vectorize S; otherwise, vech is used to vectorize S.
intervals.typ	pe
	Either z (default if missing) or LB. If it is z, it calculates 95% Wald confidence intervals (CIs) based on the z statistic. If it is LB, it calculates 95% likelihood-based CIs on the parameter estimates. Please note that the z values and their associated p values are based on the z statistic. They are not related to the likelihood-based CIs.
suppressWarn	
	Logical. If TRUE, warnings are suppressed. Argument to be passed to mxRun.
	Futher arguments to be passed to mxRun

Value

n

```
An object of class wls with a list of

call The matched call
noObservedStat
Number of observed statistics
```

Sample size

indepModelChisq

 $\label{lem:chi-square statistic} Chi-square statistic of the independent model returned by .indepwlsChisq \\ indepModelDf \ Degrees of freedom of the independent model returned by .indepwlsChisq \\ wls.fit \qquad A fitted object returned from mxRun$

Author(s)

Mike W.-L. Cheung <mikewlcheung@nus.edu.sg>

References

Bentler, P.M., & Savalei, V. (2010). Analysis of correlation structures: current status and open problems. In Kolenikov, S., Thombs, L., & Steinley, D. (Eds.). *Recent Methodological Developments in Social Science Statistics* (pp. 1-36). Hoboken, NJ: Wiley.

Cheung, M. W. L., & Chan, W. (2005). Meta-analytic structural equation modeling: A two-stage approach. *Psychological Methods*, **10**, 40-64.

Cheung, M. W. L., & Chan, W. (2009). A two-stage approach to synthesizing covariance matrices in meta-analytic structural equation modeling. *Structural Equation Modeling*, **16**, 28-53.

Joreskog, K. G., Sorbom, D., Du Toit, S., & Du Toit, M. (1999). LISREL 8: New Statistical Features. Chicago: Scientific Software International.

See Also

tssem1

```
#### Analysis of correlation structure
R1 \leftarrow matrix(c(1.00, 0.22, 0.24, 0.18,
               0.22, 1.00, 0.30, 0.22,
                0.24, 0.30, 1.00, 0.24,
                0.18, 0.22, 0.24, 1.00), ncol=4, nrow=4)
n < -1000
acovR1 <- asyCov(R1, n)</pre>
## One-factor CFA model- P1: Factor variance; L1: Factor loadings
P1 <- mxMatrix("Full", ncol=1, nrow=1, value=1, free=FALSE, name="P1")
L1 <- mxMatrix("Full", ncol=1, nrow=4, free=TRUE, name="L1")
impliedR1 <- mxAlgebra(L1 %&% P1, name="impliedR1")</pre>
wls.fit1 <- wls(S=R1, acovS=acovR1, n=n, impliedS=impliedR1,
                matrices=c(P1, L1), cor.analysis=TRUE)
summary(wls.fit1)
#### Multiple regression analysis with RAM specification
## Variables in R2: y, x1, x2
R2 \leftarrow matrix(c(1.00, 0.22, 0.24,
                0.22, 1.00, 0.30,
               0.24, 0.30, 1.00,
               0.18, 0.22, 0.24), ncol=3, nrow=3)
acovR2 <- asyCov(R2, n)</pre>
## A2: Regression coefficents
# y x1 x2
#y FT T
```

```
#x1 F F F
#x2 F F F
A2 <- mxMatrix("Full", ncol=3, nrow=3, byrow=TRUE,
               free=c(FALSE, rep(TRUE, 2), rep(FALSE, 6)), name="A2")
## S2: Covariance matrix of free parameters
## Note that the diagonal elements are not involved in
## the analysis of correlation structure
  y x1 x2
#v F F F
#x1 F F F
#x2 F T F
S2 <- mxMatrix("Stand", ncol=3, nrow=3, free=c(FALSE, FALSE, TRUE), name="S2")
## Identity matrix
Id <- mxMatrix("Iden", ncol=3, nrow=3, name="Id")</pre>
## Model implied correlation matrix: (Id-A2)^-1 %*% S2 %*% ((Id-A2)^-1)'
impliedR2 <- mxAlgebra( solve(Id-A2) %&% S2, name="impliedR2")</pre>
wls.fit2 <- wls(S=R2, acovS=acovR2, n=n, impliedS=impliedR2,
                matrices=c(A2, S2, Id), cor.analysis=TRUE)
summary(wls.fit2)
#### Analysis of covariance structure
S3 \leftarrow matrix(c(1.50, 0.22, 0.24, 0.18,
               0.22, 1.60, 0.30, 0.22,
               0.24, 0.30, 1.80, 0.24,
               0.18, 0.22, 0.24, 1.30), ncol=4, nrow=4)
n < -1000
acovS3 <- asyCov(S3, n, cor.analysis=FALSE)</pre>
P3 <- mxMatrix("Full", ncol=1, nrow=1, value=1, free=FALSE, name="P3")
L3 <- mxMatrix("Full", ncol=1, nrow=4, value=c(0.3, 0.4, 0.5, 0.4),
               free=TRUE, name="L3")
E3 <- mxMatrix("Diag", ncol=4, nrow=4, value=0.2, free=TRUE, name="E3")
impliedS3 <- mxAlgebra(L3 %&% P3 + E3, name="impliedS3")</pre>
## Use likelihood-based CI
wls.fit3 <- wls(S=S3, acovS=acovS3, n=n, impliedS=impliedS3,
                matrices=c(P3, L3, E3), cor.analysis=FALSE,
                intervals.type="LB")
summary(wls.fit3)
##### Example of tssem2
digman1 <- tssem1(Digman97$data, Digman97$n)</pre>
P4 <- mxMatrix("Stand", ncol=2, nrow=2, value=.2, free=TRUE, name="P4")
L4 <- mxMatrix("Full", ncol=2, nrow=5, value=c(0,.3,.3,.3,0,.3,0,0,0,0,.3),
free=c(FALSE, TRUE, TRUE, TRUE, FALSE, TRUE, FALSE, FALSE, FALSE, TRUE), name="L4")
impliedR4 <- mxAlgebra(L4 %&% P4, name="impliedR4")</pre>
digman2 <- tssem2(digman1, impliedS=impliedR4, matrices=c(P4, L4))</pre>
summary(digman2)
```

Index

*Topic datasets	homoStat, 11
Becker83, 6	Hox02, 12
Berkey98,7	110,002, 12
Cheung09,8	is.pd, 13
Digman 97, 9	
	list2matrix, 14
HedgesOlkin85, 10	
Hox02, 12	${\tt match.call}, 16$
*Topic meta-analysis	matrix2bdiag, 5 , 6 , 14 , 23
asyCov, 3	meta, 4, 9, 12, 15, 18, 21-23
homoStat, 11	metaSEM(metaSEM-package),1
meta, 15	metaSEM-package, 1
*Topic package	MxAlgebra-class,24
metaSEM-package, 1	mxMatrix, 2, 3
*Topic summary	MxMatrix-class, 3, 24
coef,9	mxRun, 4, 16, 21, 22, 24, 25
print, 17	
summary, 20	nearPD, <i>13</i>
vcov, 22	
*Topic tssem	print, 17
tssem1, 21	print.summary.meta(summary), 20
wls, 24	print.summary.tssem1(summary),20
*Topic utilities	print.summary.wls(summary),20
as. $mxMatrix, 2$	printCoefmat,21
bdiagMat, 5	1
bdiagRep, 6	read.table, 18
is.pd, 13	readData, 18
list2matrix, 14	readFullMat (readData), 18
matrix2bdiag, 14	readLowTriMat(readData), 18
readData, 18	readStackVec(<i>readData</i>), 18
vec2symMat, 23	scan, 18
	summary, 20
as.mxMatrix, 2, 16	summary, 20 summary), 20
asyCov, 3	
	summary.tssem1 (summary), 20
bdiagMat, 5, 6	summary.wls(summary),20
bdiagRep, 5, 6	tssem1, 9, 18, 21, 21-25
Becker83,6	tssem2 (wls), 24
Berkey98,7	C33Cm2 (W13), 24
Choung() 8	vcov, 22
Cheung09,8	vec2symMat, 15, 23
coef,9	vech, 11, 14, 15, 24
Digman97,9	vechs, 14, 24
HedgesOlkin85, 10	wls, 9, 18, 21-23, 24