

Package ‘nda’

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Type Package

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Description Non-parametric dimensionality reduction function. Reduction with and without feature selection. Plot functions. Automated feature selections. Kosztyan et. al. (2022) <[doi:10.1016/j.knosys.2022.109180](https://doi.org/10.1016/j.knosys.2022.109180)>.

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nda-package	<i>Package of Generalized Network-based Dimensionality Reduction and Analyses</i>
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Description

The package of Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

Author(s)

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References

Kosztyan, Z. T., Kurbucz, M. T., & Katona, A. I. (2022). Network-based dimensionality reduction of high-dimensional, low-sample-size datasets. *Knowledge-Based Systems*, 109180.

Kurbucz, M. T., Katona, A. I., Lantos, Z., & Kosztyan, Z. T. (2021). The Role of Societal Aspects in the Formation of Official COVID-19 Reports: A Data-Driven Analysis. *International journal of environmental research and public health*, 18(4), 1505.

See Also

[ndr](#), [plot](#), [biplot](#), [summary](#), [dCor](#).

biplot.nda	<i>Biplot function for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)</i>
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Description

Biplot function for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Usage

```
## S3 method for class 'nda'  
biplot(x, main=NULL,...)
```

Arguments

x	an object of class 'NDA'.
main	main title of biplot.
...	other graphical parameters.

Author(s)

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See Also

[plot](#), [summary](#), [ndr](#), [data_gen](#).

Examples

```
# Biplot function without feature selection  
  
# Generate 200 x 50 random block matrix with 3 blocks and lambda=0 parameter  
  
df<-data_gen(200,50,3,0)  
p<-ndr(df)  
biplot(p)
```

COVID19_2020	<i>Covid' 19 case datasets of countries (2020), where the data frame has 138 observations of 18 variables.</i>
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Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA) Covid' 19 of countries (2020), where the data frame has 138 observations of 18 variables.

Usage

```
data("COVID19_2020")
```

Format

A data frame with 138 observations 18 variables.

Source

Kurbucz, M. T. (2020). A joint dataset of official COVID-19 reports and the governance, trade and competitiveness indicators of World Bank group platforms. Data in brief, 31, 105881.

Examples

```
data(COVID19_2020)
```

CrimesUSA1990.X	<i>Crimes in USA cities in 1990. Independent variables (X)</i>
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Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA) Crimes in USA cities in 1990. Independent variables (X)

Usage

```
data("CrimesUSA1990.X")
```

Format

A data frame with 1994 observations 123 variables.

Source

UCI - Machine Learning Repository: <https://archive.ics.uci.edu/ml/datasets/communities+and+crime>

Examples

```
data(CrimesUSA1990.X)
```

CrimesUSA1990.Y	<i>Crimes in USA cities in 1990. Dependent variable (Y)</i>
-----------------	---

Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)
Crimes in USA cities in 1990. Dependent variable (Y)

Usage

```
data("CrimesUSA1990.Y")
```

Format

A data frame with 1994 observations 1 variables.

Source

UCI - Machine Learning Repository: <https://archive.ics.uci.edu/ml/datasets/communities+and+crime>

Examples

```
data(CrimesUSA1990.Y)
```

CWTS_2020	<i>CWTS Leiden's University Ranking 2020 for all scientific fields, within the period of 2016-2019. 1176 observations (i.e., universities), and 42 variables (i.e., indicators).</i>
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Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)
CWTS Leiden's 2020 dataset, where the data frame has 1176 observations of 42 variables.

Usage

```
data("CWTS_2020")
```

Format

A data frame with 1176 observations of 42 variables.

Source

CWTS Leiden Ranking 2020: <https://www.leidenranking.com/ranking/2020/list>

Examples

```
data(CWTS_2020)
```

data_gen

Generate random block matrix for GNDA

Description

Generate random block matrix for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Usage

```
data_gen(n, m, nfactors=2, lambda=1)
```

Arguments

n	number of rows
m	number of columns
nfactors	number of blocks (factors, where the default value is 2)
lambda	exponential smoothing, where the default value is 1

Details

n, m, nfactors must be integers, and they are not less than 1; lambda should be a positive real number.

Value

M a dataframe of a block matrix

Author(s)

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Examples

```
# Specification 30 by 10 random block matrices with 2 blocks/factors
df<-data_gen(30,10)
library(psych)
scree(df)
biplot(ndr(df))
# Specification 40 by 20 random block matrices with 3 blocks/factors
df<-data_gen(40,20,3)
library(psych)
scree(df)
biplot(ndr(df))
plot(ndr(df))

# Specification 50 by 20 random block matrices with 4 blocks/factors
# lambda=0.1
df<-data_gen(50,15,4,0.1)
scree(df)
biplot(ndr(df))
plot(ndr(df))
```

dCor

Calculating distance correlation of two vectors or columns of a matrix

Description

Calculating distance correlation of two vectors or columns of a matrix for Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

The calculation is very slow for large matrices!

Usage

```
dCor(x, y=NULL)
```

Arguments

x	a numeric vector, a numeric matrix (in this case y=NULL), or a numeric data frame (in this case y=NULL)
y	a numeric vector (optional)

Details

If x is a numeric vector, y must be specified. If x is a numeric matrix or numeric data frame, y must be ignored from the parameters.

Value

Either a distance correlation value of vectors x and y, or a distance correlation matrix of x.

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References

Rizzo M, Szekely G (2021). *_energy: E-Statistics: Multivariate Inference via the Energy of Data_*. R package version 1.7-8, <URL: <https://CRAN.R-project.org/package=energy>>.

Examples

```
# Specification of distance correlation value of vectors x and y.
x<-rnorm(36)
y<-rnorm(36)
dCor(x,y)
# Specification of distance correlation matrix.
x<-matrix(rnorm(36),nrow=6)
dCor(x)
```

dCov

Calculating distance covariance of two vectors or columns of a matrix

Description

Calculating distance covariance of two vectors or columns of a matrix for Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

The calculation is very slow for large matrices!

Usage

```
dCov(x,y=NULL)
```

Arguments

x	a numeric vector, a numeric matrix (in this case y=NULL), or a numeric data frame (in this case y=NULL)
y	a numeric vector (optional)

Details

If x is a numeric vector, y must be specified. If x is a numeric matrix or numeric data frame, y must be ignored from the parameters.

Value

Either a distance covariance value of vectors x and y, or a distance covariance matrix of x.

Author(s)

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References

Rizzo M, Szekely G (2021). *_energy: E-Statistics: Multivariate Inference via the Energy of Data_*. R package version 1.7-8, <URL: <https://CRAN.R-project.org/package=energy>>.

Examples

```
# Specification of distance covariance value of vectors x and y.
x<-rnorm(36)
y<-rnorm(36)
dCov(x,y)
# Specification of distance covariance matrix.
x<-matrix(rnorm(36),nrow=6)
dCov(x)
```

fs.dimred

Feature selection for PCA, FA, and (G)NDA

Description

This function drops variables that have low communality values and/or are common indicators (i.e., correlates more than one latent variables).

Usage

```
fs.dimred(fn,DF,min_comm=0.25,com_comm=0.25)
```

Arguments

fn	It is a list variable of the output of a principal (PCA), a fa (FA), or an ndr (NDA) function.
DF	Numeric data frame, or a numeric matrix of the data table
min_comm	Scalar between 0 to 1. Minimal communality value, which a variable has to be achieved. The default value is 0.25.
com_comm	Scalar between 0 to 1. The minimal difference value between loadings. The default value is 0.25.

Details

This function only works with `principal`, `fa`, and `ndr` functions.

This function drops each variable that has a low communality value (under `min_comm` value). In other words, that variable does not fit enough of any latent variable.

This function also drops so-called common indicators, which correlate highly with more than one latent variable. And the difference in the correlation is either lower than the `com_comm` value or the greatest absolute factor loading value is not twice greater than the second greatest factor loading.

Value

<code>dropped_low</code>	Numeric data frame or numeric matrix. Set of indicators (i.e. variables), which are dropped by their low communalities. This value is NULL if a correlation matrix is used as an input or there is no dropped indicator.
<code>dropped_com</code>	Numeric data frame or numeric matrix. Set of dropped common indicators (i.e. common variables). This value is NULL if a correlation matrix is used as an input or there is no dropped indicator.
<code>remain_DF</code>	Numeric data frame or numeric matrix. Set of retained indicators
<code>...</code>	Other outputs came from principal , fa , or in ndr

Author(s)

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References

Abonyi, J., Czvetkó, T., Kosztyán, Z. T., & Héberger, K. (2022). Factor analysis, sparse PCA, and Sum of Ranking Differences-based improvements of the Promethee-GAIA multicriteria decision support technique. *Plos one*, 17(2), e0264277. doi:10.1371/journal.pone.0264277

See Also

[principal](#), [fa](#), [ndr](#).

Examples

```
data<-I40_2020

library(psych)

# Principal Component Analysis (PCA)

pca<-principal(data,nfactors=2,covar=TRUE)
pca

# Feature selection with default values
```

```

PCA<-fs.dimred(pca,data)
PCA

# List of dropped, low communality value indicators
print(colnames(PCA$dropped_low))

# List of dropped, common communality value indicators
print(colnames(PCA$dropped_com))

# List of retained indicators
print(colnames(PCA$retained_DF))

# Principal Component Analysis (PCA) of correlation matrix

pca<-principal(cor(data,method="spearman"),nfactors=2,covar=TRUE)
pca

# Feature selection
min_comm<-0.25 # Minimal communality value
com_comm<-0.20 # Minimal common communality value

PCA<-fs.dimred(pca,cor(data,method="spearman"),min_comm,com_comm)
PCA

```

fs.KMO

Feature selection for KMO

Description

Drop variables if their MSA_i value is lower than a threshold, in order to increase the overall KMO (MSA) value.

Usage

```
fs.KMO(data,min_MSA=0.5,cor.mtx=FALSE)
```

Arguments

data	A numeric data frame
min_MSA	A numeric value. Minimal MSA value for variable i
cor.mtx	Boolean value. The input is either a correlation matrix (cor.mtx=TRUE), or not (cor.mtx=FALSE)

Details

Low Kaiser-Meyer-Olkin (KMO) measure of sampling adequacy does not suggest using principal component or factor analysis. Therefore, this function drop variables with low KMO/MSA values.

Value

data Cleaned data or the cleaned correlation matrix.

Author(s)

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References

Abonyi, J., Czvetkó, T., Kosztyán, Z. T., & Héberger, K. (2022). Factor analysis, sparse PCA, and Sum of Ranking Differences-based improvements of the Promethee-GAIA multicriteria decision support technique. *Plos one*, 17(2), e0264277. doi:10.1371/journal.pone.0264277

See Also

[summary](#).

Examples

```
library(psych)
data(I40_2020)
data<-I40_2020
KMO(fs.KMO(data,min_MSA=0.7,cor.mtx=FALSE))
```

GOVDB2020

Governmental and economic data of countries (2020), where the data frame has 138 observations of 2161 variables.

Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA) Governmental and economic data of countries (2020), where the data frame has 138 observations of 2161 variables.

Usage

```
data("GOVDB2020")
```

Format

A data frame with 138 observations of 2161 variables.

Source

Kurbucz, M. T. (2020). A joint dataset of official COVID-19 reports and the governance, trade and competitiveness indicators of World Bank group platforms. *Data in brief*, 31, 105881.

Examples

```
data(GOVDB2020)
```

I40_2020

NUTS2 regional development data (2020) of I4.0 readiness, where the data frame has 414 observations of 101 variables.

Description

Sample datasets for Generalized Network-based Dimensionality Reduction and Analysis (GNDA) NUTS2 regional development data (2020), where the data frame has 414 observations of 101 variables.

Usage

```
data("COVID19_2020")
```

Format

A data frame with 414 observations of 101 variables.

Source

Honti, G., Czvetkó, T., & Abonyi, J. (2020). Data describing the regional Industry 4.0 readiness index. Data in Brief, 33, 106464.

Examples

```
data(I40_2020)
```

ndr

Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Description

The main function of Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

Usage

```
ndr(r, covar=FALSE, cor_method=1, cor_type=1, min_R=0, min_comm=2, Gamma=1, null_model_type=4, mod_mode=6, min_evalue=0, min_communality=0, com_communalities=0, use_rotation=FALSE)
```

Arguments

<code>r</code>	A numeric data frame
<code>covar</code>	If this value is FALSE (default), it finds the correlation matrix from the raw data. If this value is TRUE, it uses the matrix <code>r</code> as a correlation/similarity matrix.
<code>cor_method</code>	Correlation method (optional). '1' Pearson's correlation (default), '2' Spearman's correlation, '3' Kendall's correlation, '4' Distance correlation
<code>cor_type</code>	Correlation type (optional). '1' Bivariate correlation (default), '2' partial correlation, '3' semi-partial correlation
<code>min_R</code>	Minimal square correlation between indicators (default: 0).
<code>min_comm</code>	Minimal number of indicators per community (default: 2).
<code>Gamma</code>	Gamma parameter in multiresolution null model (default: 1).
<code>null_model_type</code>	'1' Differential Newman-Grivan's null model, '2' The null model is the mean of square correlations between indicators, '3' The null model is the specified minimal square correlation, '4' Newman-Grivan's model (default)
<code>mod_mode</code>	Community-based modularity calculation mode: '1' Louvain modularity, '2' Fast-greedy modularity, '3' Leading Eigen modularity, '4' Infomap modularity, '5' Walktrap modularity, '6' Leiden modularity (default)
<code>min_evalue</code>	Minimal eigenvector centrality value (default: 0)
<code>min_communality</code>	Minimal communality value of indicators (default: 0)
<code>com_communalities</code>	Minimal common communalities (default: 0)
<code>use_rotation</code>	FALSE no rotation (default), TRUE varimax rotation

Details

NDA both works on low and high simple size datasets. If `min_evalue=min_communality=com_communalities=0` than there is no feature selection.

Value

<code>communality</code>	Communality estimates for each item. These are merely the sum of squared factor loadings for that item. It can be interpreted in correlation matrices.
<code>loadings</code>	A standard loading matrix of class "loadings".
<code>uniqueness</code>	Uniqueness values of indicators.
<code>factors</code>	Number of found factors.
<code>scores</code>	Estimates of the factor scores are reported (if <code>covar=FALSE</code>).
<code>n.obs</code>	Number of observations specified or found.
<code>fn</code>	Factor name: NDA
<code>Call</code>	Callback function

Author(s)

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References

Kosztyan, Z. T., Kurbucz, M. T., & Katona, A. I. (2022). Network-based dimensionality reduction of high-dimensional, low-sample-size datasets. *Knowledge-Based Systems*, 109180. doi:10.1016/j.knosys.2022.109180

See Also

[plot](#), [biplot](#), [summary](#).

Examples

```
# Dimension reduction

data(swiss)
df<-swiss
p<-ndr(df)
summary(p)
plot(p)
biplot(p)

# Data reduction
# Distance is Euclidean's distance
# covar=TRUE means only the distance matrix is considered.

q<-ndr(1-normalize(as.matrix(dist(df))),covar=TRUE)
summary(q)
plot(q)
```

normalize	<i>Min-max normalization</i>
-----------	------------------------------

Description

Min-max normalization for data matrices and data frames

Usage

```
normalize(x, type="all")
```

Arguments

x	a data frame or data matrix
type	type of normalizaion. "row" normalization row by row, "col" normalization column by colum, "all" normalization for the entire data frame/matrix (default)

Value

Returns a normalized data.frame/matrix.

Author(s)

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Examples

```
mtx<-matrix(rnorm(20),5,4)
n_mtx<-normalize(mtx) # Fully normalized matrix
r_mtx<-normalize(mtx,type="row") # Normalize row by row
c_mtx<-normalize(mtx,type="col") # Normalize col by col
print(n_mtx) # Print fully normalized matrix
```

pdCor

Calculating partial distance correlation of columns of a matrix

Description

Calculating partial distance correlation of two columns of a matrix for Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

The calculation is very slow for large matrices!

Usage

```
pdCor(x)
```

Arguments

x a numeric matrix, or a numeric data frame

Value

Partial distance correlation matrix of x.

Author(s)

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References

Rizzo M, Szekely G (2021). *_energy: E-Statistics: Multivariate Inference via the Energy of Data_*. R package version 1.7-8, <URL: <https://CRAN.R-project.org/package=energy>>.

Examples

```
# Specification of partial distance correlation matrix.
x<-matrix(rnorm(36),nrow=6)
pdCor(x)
```

plot.nda

Plot function for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Description

Plot variable network graph

Usage

```
## S3 method for class 'nda'
plot(x, cuts=0.3, interactive=TRUE, edgescale=1.0, labeldist=-1.5, ...)
```

Arguments

x	an object of class 'NDA'.
cuts	minimal square correlation value for an edge in the correlation network graph (default 0.3).
interactive	Plot interactive visNetwork graph or non-interactive igraph plot (default TRUE).
edgescale	Proportion scale value of edge width.
labeldist	Vertex label distance in non-interactive igraph plot (default value ==-1.5).
...	other graphical parameters.

Author(s)

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See Also

[biplot](#), [summary](#), [ndr](#).

Examples

```
# Plot function with feature selection

data("CrimesUSA1990.X")
df<-CrimesUSA1990.X
p<-ndr(df)
biplot(p,main="Biplot of CrimesUSA1990 without feature selection")
```

```
# Plot function with feature selection
# minimal eigen values (min_evalue) is 0.0065
# minimal communality value (min_communality) is 0.1
# minimal common communality value (com_communalities) is 0.1

p<-ndr(df,min_evalue = 0.0065,min_communality = 0.1,com_communalities = 0.1)

# Plot with default (cuts=0.3)
plot(p)

# Plot with higher cuts
plot(p,cuts=0.6)
```

spdCor

Calculating semi-partial distance correlation of columns of a matrix

Description

Calculating semi-partial distance correlation of two columns of a matrix for Generalized Network-based Dimensionality Reduction and Analysis (GNDA).

The calculation is very slow for large matrices!

Usage

```
spdCor(x)
```

Arguments

x a numeric matrix, or a numeric data frame

Value

Semi-partial distance correlation matrix of x.

Author(s)

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References

Rizzo M, Szekely G (2021). *_energy: E-Statistics: Multivariate Inference via the Energy of Data_*. R package version 1.7-8, <URL: <https://CRAN.R-project.org/package=energy>>.

Examples

```
# Specification of semi-partial distance correlation matrix.
x<-matrix(rnorm(36),nrow=6)
spdCor(x)
```

summary.nda

Summary function of Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Description

Print summary of Generalized Network-based Dimensionality Reduction and Analysis (GNDA)

Usage

```
## S3 method for class 'nda'
summary(object, digits = getOption("digits"), ...)
```

Arguments

```
object      an object of class 'nda'.
digits      the number of significant digits to use when add.stats = TRUE.
...         additional arguments affecting the summary produced.
```

Author(s)

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See Also

[biplot](#), [plot](#), [ndr](#).

Examples

```
# Example of summary function of NDA without feature selection

data("CrimesUSA1990.X")
df<-CrimesUSA1990.X
p<-ndr(df)
summary(p)

# Example of summary function of NDA with feature selection
# minimal eigen values (min_evalue) is 0.0065
# minimal communality value (min_communality) is 0.1
# minimal common communality value (com_communalities) is 0.1

p<-ndr(df,min_evalue = 0.0065,min_communality = 0.1,com_communalities = 0.1)
```

summary(p)

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