

Package ‘nda’

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Type Package

Title Generalized Network-Based Dimensionality Reduction and Analysis

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Description Non-parametric dimensionality reduction function. Reduction with and without feature selection. Plot functions. Automated feature selections. Kosztyan et. al. (2024) <[doi:10.1016/j.eswa.2023.121779](https://doi.org/10.1016/j.eswa.2023.121779)>.

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LazyData true

URL <https://github.com/kzst/nda>

Depends R (>= 4.00)

Imports energy, psych, stats, igraph, Matrix, methods, Rfast, MASS, ppcor, leidenAlg, visNetwork

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Contents

| | |
|---------------------------|---|
| nda-package | 2 |
| biplot.nda | 3 |
| COVID19_2020 | 4 |
| CrimesUSA1990.X | 4 |
| CrimesUSA1990.Y | 5 |

| | |
|-----------------------|----|
| CWTS_2020 | 5 |
| data_gen | 6 |
| dCor | 7 |
| dCov | 8 |
| fs.dimred | 9 |
| fs.KMO | 11 |
| GOVDB2020 | 12 |
| I40_2020 | 13 |
| ndr | 13 |
| normalize | 15 |
| pdCor | 16 |
| plot.nda | 17 |
| print.nda | 18 |
| spdCor | 19 |
| summary.nda | 20 |

| | |
|--------------|-----------|
| Index | 22 |
|--------------|-----------|

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

Description

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

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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.

Kosztyán, Z. T., Katona, A. I., Kurbucz, M. T., & Lantos, Z. (2024). Generalized network-based dimensionality analysis. *Expert Systems with Applications*, 238, 121779. <URL: <https://doi.org/10.1016/j.eswa.2023.121779>>

See Also

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

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

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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References

Kosztyán, Z. T., Katona, A. I., Kurbucz, M. T., & Lantos, Z. (2024). Generalized network-based dimensionality analysis. *Expert Systems with Applications*, 238, 121779. <URL: <https://doi.org/10.1016/j.eswa.2023.121779>>

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> |
|--------------|--|

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> |
|-----------------|--|

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> |
|-----------|--|

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, matrix or data frame. |
| y | NULL (default) or a vector, matrix or data frame with compatible dimensions to x. The default is equivalent to y = x (but more efficient). |

Details

If x is a numeric vector, y must be specified. If x is a numeric matrix or numeric data frame, y will be neglected.

Value

Either a distance correlation coefficient of vectors x and y, or a distance correlation matrix of x if x is a matrix or a dataframe.

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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, matrix or data frame. |
| y | NULL (default) or a vector, matrix or data frame with compatible dimensions to x. The default is equivalent to y = x (but more efficient). |

Details

If x is a numeric vector, y must be specified. If x is a numeric matrix or numeric data frame, y will be neglected.

Value

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

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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, and 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

| | |
|-------------|--|
| dropped_low | 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. |
| dropped_com | 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. |
| remain_DF | Numeric data frame or numeric matrix. Set of retained indicators |
| ... | Other outputs came from |

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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

[psych::principal](#), [psych::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.

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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 | <i>NUTS2 regional development data (2020) of I4.0 readiness, where the data frame has 414 observations of 101 variables.</i> |
|----------|--|

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 | <i>Generalized Network-based Dimensionality Reduction and Analysis (GNDA)</i> |
|-----|---|

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,
    rotation="oblimin")
```

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 modell (default: 1). |
| <code>null_modell_type</code> | '1' Differential Newmann-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' Newmann-Grivan's modell (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 the rotation is used. |
| <code>rotation</code> | "none", "varimax", "quartimax", "promax", "oblimin", "simplimax", and "cluster" are possible rotations/transformations of the solution. "oblimin" is the default, if <code>use_rotation</code> is TRUE. |

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 |

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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 | The type of normalization. "row" normalization row by row, "col" normalization column by column, and "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 | <i>Plot function for Generalized Network-based Dimensionality Reduction and Analysis (GNDA)</i> |
|----------|---|

Description

Plot variable network graph

Usage

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

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). |
| show_weights | Show edge weights (default FALSE)). |
| ... | other graphical parameters. |

Author(s)

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References

Kosztyán, Z. T., Katona, A. I., Kurbucz, M. T., & Lantos, Z. (2024). Generalized network-based dimensionality analysis. *Expert Systems with Applications*, 238, 121779. <URL: <https://doi.org/10.1016/j.eswa.2023.121779>>

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)

# GNDA is used for clustering, where the similarity function is the 1-Euclidean distance
# Data is the swiss data

SIM<-1-normalize(as.matrix(dist(swiss)))
q<-ndr(SIM,covar = TRUE)
plot(q,interactive = FALSE)

```

print.nda

Print 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'
print(x, digits = getOption("digits"), ...)

```

Arguments

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

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References

Kosztyán, Z. T., Katona, A. I., Kurbucz, M. T., & Lantos, Z. (2024). Generalized network-based dimensionality analysis. *Expert Systems with Applications*, 238, 121779. <URL: <https://doi.org/10.1016/j.eswa.2023.121779>>

See Also

[biplot](#), [plot](#), [summary](#), [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)
print(p)
```

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 a numeric matrix, or a numeric data frame

Value

Semi-partial 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 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 <code>add.stats = TRUE</code> . |
| ... | additional arguments affecting the summary produced. |

Value

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

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References

Kosztyán, Z. T., Katona, A. I., Kurbucz, M. T., & Lantos, Z. (2024). Generalized network-based dimensionality analysis. *Expert Systems with Applications*, 238, 121779. <URL: <https://doi.org/10.1016/j.eswa.2023.121779>>

See Also

[biplot](#), [plot](#), [print](#), [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)
```

Index

- * **array**
 - data_gen, 6
 - dCor, 7
 - dCov, 8
 - pdCor, 16
 - spdCor, 19
 - * **control chart**
 - plot.nda, 17
 - * **correlation matrix**
 - dCor, 7
 - dCov, 8
 - pdCor, 16
 - spdCor, 19
 - * **datasets**
 - COVID19_2020, 4
 - CrimesUSA1990.X, 4
 - CrimesUSA1990.Y, 5
 - CWTS_2020, 5
 - GOVDB2020, 12
 - I40_2020, 13
 - * **dimensionality**
 - fs.dimred, 9
 - fs.KMO, 11
 - ndr, 13
 - * **distance correlation**
 - dCor, 7
 - dCov, 8
 - pdCor, 16
 - spdCor, 19
 - * **matrix**
 - normalize, 15
 - * **multivariate**
 - data_gen, 6
 - dCor, 7
 - dCov, 8
 - fs.dimred, 9
 - fs.KMO, 11
 - ndr, 13
 - pdCor, 16
 - plot.nda, 17
 - print.nda, 18
 - spdCor, 19
 - summary.nda, 20
 - * **nonparametric**
 - ndr, 13
 - * **package**
 - nda-package, 2
 - * **plot**
 - biplot.nda, 3
 - * **random block matrix**
 - data_gen, 6
 - * **reduction**
 - fs.dimred, 9
 - fs.KMO, 11
 - ndr, 13
- biplot, 2, 15, 17, 19, 21
- biplot.nda, 3
- COVID19_2020, 4
- CrimesUSA1990.X, 4
- CrimesUSA1990.Y, 5
- CWTS_2020, 5
- data_gen, 3, 6
- dCor, 2, 7
- dCov, 8
- fs.dimred, 9
- fs.KMO, 11
- GOVDB2020, 12
- I40_2020, 13
- nda (nda-package), 2
- nda-package, 2
- ndr, 2, 3, 10, 13, 17, 19, 21
- normalize, 15

pdCor, [16](#)
plot, [2](#), [3](#), [15](#), [19](#), [21](#)
plot.nda, [17](#)
print, [21](#)
print.nda, [18](#)
psych::fa, [10](#)
psych::principal, [10](#)

spdCor, [19](#)
summary, [2](#), [3](#), [12](#), [15](#), [17](#), [19](#)
summary.nda, [20](#)