Package: NetworkComparisonTest (via r-universe)

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Title Statistical Comparison of Two Networks Based on Several Invariance Measures
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Description This permutation based hypothesis test, suited for several types of data supported by the estimateNetwork function of the bootnet package (Epskamp & Fried, 2018), assesses the difference between two networks based on several invariance measures (network structure invariance, global strength invariance, edge invariance, several centrality measures, etc.). Network structures are estimated with 11-regularization. The Network Comparison Test is suited for comparison of independent (e.g., two different groups) and dependent samples (e.g., one group that is measured twice). See van Borkulo et al. (2021, in press; the final article will be available, upon publication, via its DOI: 10.1037/met0000476).
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NetworkComparisonTest-package

Statistical Comparison of Two Networks Based on Several Invariance Measures

Description

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Details

Package: NetworkComparisonTest

Type: Package Version: 2.2.2 License: GPL-2

Author(s)

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References

Epskamp, S., & Fried, E. I. (2018). A tutorial on regularized partial correlation networks. Psychological Methods, 23(4), 617-634.

Ernst, M.D. Permutation methods: A basis for exact inference. Stat Sci. 2004;19(4):676-685.

Good, P.I. Permutation, parametric and bootstrap tests of hypotheses. Vol. 3. New York:: Springer, 2005.

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NCT-methods 3

Methods for NCT objects
V

Description

Print method, prints the NCT output, plot method plots the output, summary method returns a summary of the output.

Usage

```
## S3 method for class 'NCT'
print(x,...)
## S3 method for class 'NCT'
summary(object,...)
## S3 method for class 'NCT'
plot(x, what = c("strength","network","edge","centrality"),...)
```

Arguments

x output of NCT object output of NCT

what defines what has to be plotted: results pertaining to test on invariance of global

strength ("strength"), network structure ("network"), edge strength ("edge"), or

specific centrality measure ("centrality")

... Arguments only used in plot method

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NetworkComparisonTest Statistical Comparison of Two Networks Based on Several Invariance Measures

Description

This permutation based hypothesis test, suited for several types of data supported by the estimateNetwork function of the bootnet package (Epskamp & Fried, 2018), assesses the difference between two networks based on several invariance measures (network structure invariance, global strength invariance, edge invariance, several centrality measures, etc.). Network structures are estimated with 11-regularization. The Network Comparison Test is suited for comparison of independent (e.g., two different groups) and dependent samples (e.g., one group that is measured twice).

Usage

Arguments

guments	
data1	One of two datasets. The dimension of the matrix is nobs x nvars; each row is a vector of observations of the variables. Must be cross-sectional data. Can also be the result of estimateNetwork from the bootnet package.
data2	The other of two datasets. The dimension of the matrix is nobs x nvars; each row is a vector of observations of the variables. Must be cross-sectional data. Can also be the result of estimateNetwork from the bootnet package.
gamma	A single value between 0 and 1. When not entered, gamma is set to 0.25 for binary data and 0.50 for gaussian data. Networks are estimated with this value for hyperparameter gamma in the extended BIC.
it	The number of iterations (permutations).
binary.data	Logical. Can be TRUE or FALSE to indicate whether the data is binary or not. If binary.data is FALSE, the data is regarded gaussian. This argument is ignored when using estimateNetwork() output as input for NCT.
paired	Logical. Can be TRUE of FALSE to indicate whether the samples are dependent

or not. If paired is TRUE, relabeling is performed within each pair of observations. If paired is FALSE, relabeling is not restricted to pairs of observations. Note that, currently, dependent data is assumed to entail one group measured twice.

weighted Logical. Can be TRUE of FALSE to indicate whether the networks to be com-

pared should be weighted of not. If not, the estimated networks are dichotomized.

Defaults to TRUE.

AND Logical. Can be TRUE of FALSE to indicate whether the AND-rule or the OR-

rule should be used to define the edges in the network. Defaults to TRUE. Only

necessary for binary data.

abs Logical. Should global strength consider the absolute value of edge weights, or

the raw value (i.e., global expected influence)?

test.edges Logical. Can be TRUE of FALSE to indicate whether or not differences in

individual edges should be tested.

edges Character or list. When 'all', differences between all individual edges are tested.

When provided a list with one or more pairs of indices referring to variables, the

provided edges are tested.

progressbar Logical. Should the pbar be plotted in order to see the progress of the estimation

procedure? Defaults to TRUE.

make.positive.definite

If make.positive.definite = TRUE, the covariance matrices used for the glasso are projected to the nearest positive definite matrices, if they are not yet positive definite. This is useful for small n, for which it is very likely that at least one of the bootstrap comparisons involves a covariance matrix that is not positive

definite.

p.adjust.methods

Character. Can be one of "holm", "hochberg", "hommel", "bonferroni", "BH", "BY", "fdr", or "none". To control (or not) for testing of multiple edges. Defaults to "none".

test.centrality

Logical. Should centrality metrics be compared across networks?

centrality Type of centrality metrics to test. Can be any of c("all", "closeness", "be-

tweenness", "strength", "expectedInfluence", "bridgeStrength", "bridgeClose-

ness", "bridgeBetweenness", "bridgeExpectedInfluence")

nodes Specific nodes for centrality tests. Can be character names or index numbers.

Only used if test.centrality=TRUE

communities Passed to bridge() if computing bridge centrality useCommunities Passed to bridge() if computing bridge centrality

estimator A function that takes data as input and returns a network structure. This can be

used for custom estimation algorithms. Note, supplying this function will overwrite the arguments binary.data, AND, gamma and make.positive.definite.

estimatorArgs Arguments to the estimator function

verbose Logical: Should some warnings and notes be printed?

Value

NCT returns a 'NCT' object that contains the following items:

glstrinv.real The difference in global strength between the networks of the observed data sets.

glstrinv.perm
The difference in global strength between the networks of the permutated data

sets.

glstrinv.sep The global strength values of the individual networks

glstrinv.pval The p value resulting from the permutation test concerning difference in global

strength.

nwinv.real The value of the maximum difference in edge weights of the observed networks

nwinv.perm The values of the maximum difference in edge weights of the permuted networks

nwinv.pval The p value resulting from the permutation test concerning the maximum differ-

ence in edge weights.

einv.pvals	p-values (corrected for multiple testing or not according to 'p.adjust.methods') per edge from the permutation test concerning differences in edges weights. Only returned if test.edges = TRUE.
einv.real	The value of the difference in edge weight of the observed networks (multiple values if more edges are called to test). Only if test.edges = TRUE.
einv.perm	The values of the difference in edge weight of the permuted networks. Only if test.edges = TRUE.
diffcen.real	The values of the difference in centralities of the observed networks. Only if test.centrality = TRUE.
diffcen.perm	The values of the difference in centralities of the permuted networks. Only if test.centrality = TRUE.
diffcen.pval	p-values(corrected for multiple testing or not according to 'p.adjust.methods') per node from the permutation test concerning differences in centralities. Only if test.centrality = TRUE.

Note

See also my website: http://cvborkulo.com

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Ernst, M.D. Permutation methods: A basis for exact inference. Stat Sci. 2004;19(4):676-685.

Good, P.I. Permutation, parametric and bootstrap tests of hypotheses. Vol. 3. New York:: Springer, 2005

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Examples

```
library("IsingSampler")
library("IsingFit")
library("bootnet")

### Simulate binary datasets under null hypothesis:
### underlying network structures are similar
# Input:
N <- 6 # Number of nodes</pre>
```

```
nSample <- 500 # Number of samples
# Ising parameters:
set.seed(123)
Graph <- matrix(sample(0:1,N^2,TRUE,prob = c(0.8, 0.2)),N,N) * runif(N^2,0.5,2)
Graph <- pmax(Graph,t(Graph))</pre>
Graph[4,1] <- Graph[4,1]*-1
Graph[1,4] <- Graph[1,4]*-1
Graph[5,1] <- Graph[5,1]*-1
Graph[1,5] <- Graph[1,5]*-1
Graph[6,1] <- Graph[6,1]*-1
Graph[1,6] <- Graph[1,6]*-1
diag(Graph) <- 0
Thresh <- -rowSums(Graph) / 2
# Simulate:
data1 <- IsingSampler(nSample, Graph, Thresh)</pre>
data2 <- IsingSampler(nSample, Graph, Thresh)</pre>
colnames(data1) <- colnames(data2) <- c('V1', 'V2', 'V3', 'V4', 'V5', 'V6')</pre>
### Compare networks of data sets using NCT ###
## Networks can be compared by either (1) feeding the data directly into NCT (whereby
## you need to specify arguments such as "gamma" and "binary.data") or (2) by using
## estimateNetwork() (bootnet package) and feeding that output into NCT. For the latter
## option, we refer to the help file of estimateNetwork() for its usage. Below, both
## options are illustrated. We recommend using estimateNetwork(), since this function
## has implemented many network estimation methods.
## gamma = 0 (in estimateNetwork this hyperparameter is called "tuning"; to illustrate
# how to specify a different value than the default)
## iterations (it) set to 10 to save time
## Note: Low number of iterations can give unreliable results; should be 1000 at least
## Testing whether there are differences in the three aspects that are validated
# (network invariance, global strength, edge weight)
## 2 edges are tested here: between variable 1 and 2, and between 3 and 6 (can be
# "list(c(2,1),c(6,3))" as well)
## (1) Feeding data directly into NCT
set.seed(123)
NCT_a <- NCT(data1, data2, gamma=0, it=10, binary.data = TRUE,
             test.edges=TRUE, edges=list(c(1,2),c(3,6)))
summary(NCT_a)
## Plot results of global strength invariance test (not reliable with only 10
# permutations!)
plot(NCT_a, what="strength")
## (2) Feeding the estimateNetwork() output into NCT
est_1 <- estimateNetwork(data1, default = "IsingFit", tuning = 0)</pre>
est_2 <- estimateNetwork(data2, default = "IsingFit", tuning = 0)</pre>
## When using estimateNetwork() output, there is no need to specify gamma and binary.data
## This yields similar output as NCT_a
set.seed(123)
```

```
NCT_b <- NCT(est_1, est_2, it=10, test.edges=TRUE,</pre>
edges=list(c(1,2),c(3,6)))
summary(NCT_b)
## Next, an example of testing whether there are differences in node strength
# when data is paired (e.g., a group which is measured pre- and post-treatement).
# Also, here you can see how to specify that you want to take the sign of node strength
# into account (by default, the absolute value is taken and, therefore, the sign is
# ignored).
## abs = FALSE
set.seed(123)
NCT_c = NCT(est_1, est_2, paired = TRUE, abs = FALSE, test.edges = TRUE,
edges = list(c(1,2),c(3,6)), test.centrality = TRUE,
centrality = c("strength"), nodes = "all", it=10)
summary(NCT_c)
## Finally, an example how to test for differences in centrality (e.g., expectedInfluence)
NCT_d = NCT(est_1, est_2, paired = TRUE, abs = FALSE, test.edges = TRUE,
edges = list(c(1,2),c(3,6)), test.centrality = TRUE,
centrality = c("expectedInfluence"), nodes = "all", it=10)
summary(NCT_d)
```

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