# Package 'hdbinseg'

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Type Package

<b>Title</b> Change-Point Analysis of High-Dimensional Time Series via Binary Segmentation				
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<b>Description</b> Binary segmentation methods for detecting and estimating multiple change-points in the mean or second-order structure of high-dimensional time series as described in Cho and Fry-zlewicz (2014) <doi:10.1111 rssb.12079=""> and Cho (2016) <doi:10.1214 16-ejs1155="">.</doi:10.1214></doi:10.1111>				
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dcbs.alg

Double CUSUM Binary Segmentation

# Description

Perform the Double CUSUM Binary Segmentation algorithm detecting change points in the mean or second-order structure of the data.

## Usage

```
dcbs.alg(
   x,
   cp.type = c(1, 2)[1],
   phi = 0.5,
   thr = NULL,
   trim = NULL,
   height = NULL,
   tau = NULL,
   temporal = TRUE,
   scales = NULL,
   diag = FALSE,
   B = 1000,
   q = 0.01,
   do.parallel = 4
)
```

## Arguments

X	input data matrix, with each row representing the component time series
cp.type	cp.type = 1 specifies change points in the mean, cp.type = 2 specifies change points in the second-order structure
phi	choice of parameter for weights in Double CUSUM statistic; $0 \le phi \le 1$ or $phi = -1$ allowed with the latter leading to the DC statistic combining $phi = 0$ and $phi = 1/2$ , see Section 4.1 of Cho (2016) for further details
thr	pre-defined threshold values; when thr = NULL, bootstrap procedure is employed for the threshold selection; when thr != NULL and cp.type = 1, length(thr) should be one, if cp.type = 2, length(thr) should match length(scales)
trim	<pre>length of the intervals trimmed off around the change point candidates; trim = NULL activates the default choice (trim = round(log(dim(x)[2])))</pre>
height	maximum height of the binary tree; height = NULL activates the default choice (height = floor(log(dim(x)[2], 2)/2))
tau	a vector containing the scaling constant for each row of x; if tau = NULL, a data- driven choice is made which takes into account the presence of possibly multiple mean shifts and temporal dependence when temporal = TRUE

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temporal	used when cp.type = 1; if temporal = FALSE, rows of x are scaled by mad estimates, if temporal = TRUE, their long-run variance estimates are used
scales	used when cp.type = 2; negative integers representing Haar wavelet scales to be used for computing $nrow(x)*(nrow(x) + 1)/2$ dimensional wavelet transformation of x; a small negative integer represents a fine scale
diag	used when cp.type = 2; if diag = TRUE, only changes in the diagonal elements of the autocovariance matrices are searched for
В	used when is.null(thr); number of bootstrap samples for threshold selection
q	used when is.null(thr); indicates the quantile of bootstrap test statistics to be used for threshold selection
do.parallel	used when is.null(thr); number of copies of R running in parallel, if do.parallel = 0, %do% operator is used, see also foreach

#### Value

S3 bin. tree object, which contains the following fields:

tree	a list object containing information about the nodes at which change points are detected
mat	matrix concatenation of the nodes of tree
еср	estimated change points
thr	threshold used to construct the tree

## References

H. Cho (2016) change point detection in panel data via double CUSUM statistic. *Electronic Journal of Statistics*, vol. 10, pp. 2000–2038.

## **Examples**

```
x <- matrix(rnorm(10*100), nrow = 10)
dcbs.alg(x, cp.type = 1, phi=.5, temporal = FALSE, do.parallel = 0)$ecp
x <- matrix(rnorm(100*300), nrow = 100)
x[1:10, 151:300] <- x[1:10, 151:300] + 1
dcbs.alg(x, cp.type = 1, phi=-1, temporal = FALSE, do.parallel = 0)$ecp</pre>
```

dcbs.thr

Bootstrapping for threshold selection in DCBS algorithm

# Description

Generate thresholds for DCBS algorithm via bootstrapping

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## Usage

```
dcbs.thr(
  z,
  interval = c(1, dim(z)[2]),
  phi = 0.5,
  cp.type = 1,
  do.clean.cp = FALSE,
  temporal = TRUE,
  scales = NULL,
  diag = FALSE,
  sgn = NULL,
  B = 1000,
  q = 0.01,
  do.parallel = 4
)
```

## **Arguments**

```
input data matrix, with each row representing the component time series

a vector of two containing the start and the end points of the interval from which
the bootstrap test statistics are to be calculated

phi, cp.type, temporal, scales, diag, B, q, do.parallel
see dcbs.alg

do.clean.cp if do.clean.cp = TRUE pre-change point cleaning is performed

sgn if diag = FALSE, wavelet transformations of the cross-covariances are computed
with the matching signs
```

#### Value

a numeric value for the threshold

sbs.alg

Sparsified Binary Segmentation

## **Description**

Perform the Sparsified Binary Segmentation algorithm detecting change-points in the mean or second-order structure of the data.

# Usage

```
sbs.alg(
   x,
   cp.type = c(1, 2)[1],
   thr = NULL,
   trim = NULL,
```

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```
height = NULL,
  tau = NULL,
  temporal = TRUE,
  scales = NULL,
 diag = FALSE,
 B = 1000,
 q = 0.01,
 do.parallel = 4
)
```

## **Arguments**

Х	input data matrix, with each row representing the component time series
cp.type	cp. type = 1 specifies change-points in the mean, cp. type = 2 specifies change- points in the second-order structure
thr	<pre>pre-defined threshold values; when thr = NULL, bootstrap procedure is employed for the threshold selection; when thr != NULL and cp.type = 1, length(thr) should match nrow(x), if cp.type = 2, length(thr) should match nrow(x)*(nrow(x)+1)/2*length(s)</pre>
trim	<pre>length of the intervals trimmed off around the change-point candidates; trim = NULL activates the default choice (trim = round(log(dim(x)[2])))</pre>
height	maximum height of the binary tree; height = NULL activates the default choice $(\text{height} = \text{floor}(\log(\dim(x)[2], 2)/2))$
tau	a vector containing the scaling constant for each row of x; if tau = NULL, a data- driven choice is made which takes into account the presence of possibly multiple mean shifts and temporal dependence when temporal = TRUE
temporal	used when cp.type = 1; if temporal = FALSE, rows of x are scaled by mad estimates, if temporal = TRUE, their long-run variance estimates are used
scales	used when cp. type = 2; negative integers representing Haar wavelet scales to be used for computing $nrow(x)*(nrow(x)+1)/2$ dimensional wavelet transformation of x; a small negative integer represents a fine scale
diag	used when cp.type = 2; if diag = TRUE, only changes in the diagonal elements of the autocovariance matrices are searched for
В	used when is.null(thr); number of bootstrap samples for threshold selection
q	used when is.null(thr); quantile of bootstrap test statistics to be used for threshold selection
do.parallel	used when is.null(thr); number of copies of R running in parallel, if do.parallel = 0, %do% operator is used, see also foreach

## Value

S3 bin. tree object, which contains the following fields:

tree	a list object containing information about the nodes at which change-points are detected
mat	matrix concatenation of the nodes of tree
еср	estimated change-points
thr	threshold used to construct the tree

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## References

H. Cho and P. Fryzlewicz (2014) Multiple-change-point detection for high dimensional time series via sparsified binary segmentation. *JRSSB*, vol. 77, pp. 475–507.

## **Examples**

```
x <- matrix(rnorm(20*300), nrow = 20)
sbs.alg(x, cp.type = 2, scales = -1, diag = TRUE, do.parallel = 0)$ecp
x <- matrix(rnorm(100*300), nrow = 100)
x[1:10, 151:300] <- x[1:10, 151:300]*sqrt(2)
sbs.alg(x, cp.type = 2, scales = -1, diag = TRUE, do.parallel = 0)$ecp</pre>
```

sbs.thr

Bootstrapping for threshold selection in SBS algorithm

## **Description**

Generate thresholds for SBS algorithm via bootstrapping

## Usage

```
sbs.thr(
  z,
  interval = c(1, dim(z)[2]),
  cp.type = 1,
  do.clean.cp = TRUE,
  scales = NULL,
  diag = FALSE,
  sgn = NULL,
  B = 1000,
  q = 0.01,
  do.parallel = 4
)
```

## **Arguments**

```
z input data matrix, with each row representing the component time series

a vector of two containing the start and the end points of the interval from which
the bootstrap test statistics are to be calculated

cp.type, scales, diag, B, q, do.parallel
see sbs.alg

do.clean.cp if do.clean.cp = TRUE pre-change-point cleaning is performed

sgn if diag = FALSE, wavelet transformations of the cross-covariances are computed
with the matching signs
```

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## Value

if cp. type = 1, a vector of length nrow(z), each containing the threshold applied to the CUSUM statistics from the corresponding coordinate of z if cp. type = 2, a vector of length length(scales)\*nrow(z) (when diag = TRUE) or length(scales)\*nrow(z)\*(nrow(z)+1)/2 (when diag = FALSE), each containing the threshold applied to the CUSUM statistics of the corresponding wavelet transformation of z

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