Package ‘offlineChange’

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Title Detect Multiple Change Points from Time Series

Version 0.0.4

Description Detect the number and locations of change points. The locations can be either exact or in terms of ranges, depending on the available computational resource. The method is based on Jie Ding, Yu Xiang, Lu Shen, Vahid Tarokh (2017) <doi:10.1109/TSP.2017.2711558>.

Depends R (>= 3.5.0)

License GPL-3

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LinkingTo Rcpp

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URL

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ChangePoints

Detect Number and Location of Change Points of Independent Data

Description

Detect the number and locations of change points based on minimizing within segment quadratic loss and applying penalized model selection approach with restriction of largest candidate number of change points.

Usage

ChangePoints(
  x,
  point_max = 5,
  penalty = "bic",
  seg_min = 1,
  num_init = NULL,
  cpp = TRUE
)

Arguments

- **x**: The data to find change points.
- **point_max**: The largest candidate number of change points.
- **penalty**: Penalty type term. Default is "bic". Users can use other penalty term.
- **seg_min**: Minimal segment size between change points at transformed scale, must be positive integer.
- **num_init**: The number of repetition times, in order to avoid local minimum. Default is squared root of number of observations. Must be integer.
- **cpp**: Option to accelerate using rcpp. Default is TRUE.

Details

The K change points form K+1 segments (1 2 ... change_point(1)) ... (change_point(K) ... N).
**Value**

- `num_change_point` optimal number of change points.
- `change_point` location of change points.

**References**


**Examples**

```r
a <- matrix(rnorm(40, mean=-1, sd=1), nrow=20, ncol=2)
b <- matrix(rnorm(120, mean=0, sd=1), nrow=60, ncol=2)
c <- matrix(rnorm(40, mean=1, sd=1), nrow=20, ncol=2)
x <- rbind(a, b, c)
ChangePoints(x, point_max=5)
ChangePoints(x, point_max=5, penalty="hq")
```

**Description**

Plot the peak ranges of change points produced by `MultiWindow`. The blue solid line is the start of a peak range and the red dashed line is the end of that peak range.

**Usage**

`ChangePointsPlot(y, result, ...)`

**Arguments**

- `y` The original data to find change points. Must be one dimensional data.
- `result` The result of function `MultiWindow`.
- `...` Arguments to be passed to plot, such as `main`, `xlab`, `ylab`.

**Value**

A plot of original data and peak ranges of change points.

**References**

### Examples

```r
N <- 1000
N1 <- floor(0.1*N)
N2 <- floor(0.3*N)
a1 <- c(0.8, -0.3); c1 <- 0
a2 <- c(-0.5, 0.1); c2 <- 0
a3 <- c(0.5, -0.5); c3 <- 0
y <- rep(0, N)
L <- 2
y[1:L] <- rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] <- y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] <- y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] <- y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
result <- MultiWindow(y, window_list=c(100, 50, 20, 10, 5), point_max=5)
ChangePointsPlot(y, result)
```

### Description

For a series of one dimensional data, get the log likelihood.

### Usage

`GetLogLik(y, left, right)`

### Arguments

- `y`: The data to calculate log likelihood. The data must be one dimensional.
- `left`: The left end of the data that users want to use.
- `right`: The right end of the data that users want to use.

### Value

`log_lik`
GetMle

Description
Transform N dependent data into approximated independent data (N/window_size) x (L+1). Computes the estimated coefficients of each window of original data.

Usage
GetMle(y, window_size)

Arguments

y                The original data to find change points.
window_size      The number of observations each window contains.

Value
x                The transformed data, which are the estimated coefficients of original data.

References

Examples
N <- 1000
N1 <- floor(0.1*N)
N2 <- floor(0.3*N)
a1 <- c(0.8, -0.3); c1 <- 0
a2 <- c(-0.5, 0.1); c2 <- 0
a3 <- c(0.5, -0.5); c3 <- 0
y <- rep(0,N)
L<-2
y[1:L] <- rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] <- y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] <- y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] <- y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
GetMle(y,window_size=100)
GetMleAr

GetMleAr

Description

Transform N dependent data into approximated independent data (N/window_size) x (L+1). Computes the estimated coefficients of each window of original data.

Usage

GetMleAr(y, window_size)

Arguments

y
The original data to find change points.

window_size
The number of observations each window contains.

Value

x
The transformed data, which are the estimated coefficients of original data.

References


Examples

N = 1000
N1 = floor(0.1*N)
N2 = floor(0.3*N)
a1 = c(0.8, -0.3); c1 = 0
a2 = c(-0.5, 0.1); c2 = 0
a3 = c(0.5, -0.5); c3 = 0
y = rep(0,N)
L=2
y[1:L] = rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] = y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] = y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] = y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
GetMleAr(y,window_size=100)
**MultiWindow**

**Multi-window Change Points Detection**

**Description**

Use a sequence of window sizes to capture ranges of change points.

**Usage**

```r
MultiWindow(
  y,
  window_list = c(100, 50, 20, 10, 5),
  point_max = 5,
  prior_range = NULL,
  get_mle = GetMle,
  penalty = "bic",
  seg_min = 1,
  num_init = NULL,
  tolerance = 1,
  cpp = TRUE,
  ret_score = FALSE
)
```

**Arguments**

- **y**: The original data to find change points. Must be one dimensional data.
- **window_list**: The list of window sizes, must be in form c(100, 50, 20, 10, 5), in descending order and each window_size > 2L. L is the lag order of the dataset.
- **point_max**: The largest candidate number of change points.
- **prior_range**: The prior ranges that considered to contain change points. Each prior range contains one change point. example: prior_range=list(c(30, 200), c(220, 400))
- **get_mle**: The method used to transform dependent data to independent data.
- **penalty**: Penalty type term. Default is "bic". Users can use other penalty term.
- **seg_min**: Minimal segment size between change points at transformed scale, must be positive integer.
- **num_init**: The number of repetition times, in order to avoid local minimum. Default is the squared root of number of transformed data.
- **tolerance**: The tolerance level. The maximal difference between the score of selected peak ranges and highest score.
- **cpp**: Logical value indicating whether to accelerate using rcpp. Default is TRUE.
- **ret_score**: Logical value indicating whether to return score. Default is FALSE.
Details

Given time series data $y_1, y_2, \ldots, y_N$, a sequence of window sizes $w_1 > \ldots > w_R$ can be used to capture any true segment of small size. For each $w_r$, the original data is turned into a sequence of $L+1$ dimensional data that can be approximated as independent. Then the change points of independent data can be detected by minimizing penalized quadratic loss. By further mapping these change points back to the original scale, several short ranges (each of size $2w_r$) that probably contain the desired change points are obtained. After repeating the above procedure for different $w_r$, the detected ranges of change points from each window size are scored by one. The scores are aggregated, and the ranges with highest score or around the highest score (determined by the tolerance parameter) are finally selected.

Value

- `n_peak_range`: The number of peak ranges.
- `peak_ranges`: The location of peak ranges.
- `score`: score matrix. (only when `ret_score` is `TRUE`)

References


Examples

```r
N <- 1000
N1 <- floor(0.1*N)
N2 <- floor(0.3*N)
a1 <- c(0.8, -0.3); c1 <- 0
a2 <- c(-0.5, 0.1); c2 <- 0
a3 <- c(0.5, -0.5); c3 <- 0
y <- rep(0,N)
L<-2
y[1:L] <- rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] <- y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] <- y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] <- y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
MultiWindow(y,window_list=c(100,50,20,10,5),point_max=5)
MultiWindow(y,window_list=c(100,50,20,10,5),prior_range=list(c(30,200),c(220,400)))
```
OrderKmeans  

*Detect Location of Change Points of Independent Data*

**Description**

Detect the location of change points based on minimizing within segment quadratic loss with fixed number of change points.

**Usage**

```r
OrderKmeans(x, K = 4, num_init = 10)
```

**Arguments**

- `x`: The data to find change points with dimension N x D, must be matrix
- `K`: The number of change points.
- `num_init`: The number of repetition times, in order to avoid local minimum. Default is 10. Must be integer.

**Details**

The K change points form K+1 segments (1 2 ... change_point(1)) ... (change_point(K) ... N).

**Value**

- `wgss_sum`: total within segment sum of squared distances to the segment mean (wgss) of all segments.
- `num_each`: number of observations of each segment
- `wgss`: total wgss of each segment.
- `change_point`: location of optimal change points.

**References**


**Examples**

```r
a<-matrix(rnorm(40,mean=-1,sd=1),nrow=20,ncol=2)
b<-matrix(rnorm(120,mean=0,sd=1),nrow=60,ncol=2)
c<-matrix(rnorm(40,mean=1,sd=1),nrow=20,ncol=2)
x<-rbind(a,b,c)
OrderKmeans(x,K=3)
OrderKmeans(x,K=3,num_init=8)
```
OrderKmeansCpp

Detect Location of Change Points of Independent Data using Rcpp

Description

Detect the location of change points based on minimizing within segment quadratic loss with fixed number of change points.

Usage

OrderKmeansCpp(x, K = 4, num_init = 10)

Arguments

x The data to find change points with dimension N x D, must be matrix
K The number of change points.
num_init The number of repetition times, in order to avoid local minimal. Default is 10. Must be integer.

Details

The K change points form K+1 segments (1 2 ... change_point(1)) ... (change_point(K) ... N).

Value

wgss_sum total within segment sum of squared distances to the segment mean (wgss) of all segments.
num_each number of observations of each segment
wgss total wgss of each segment.
change_point location of optimal change points.

References


Examples

a<-matrix(rnorm(40,mean=-1,sd=1),nrow=20,ncol=2)
b<-matrix(rnorm(120,mean=0,sd=1),nrow=60,ncol=2)
c<-matrix(rnorm(40,mean=1,sd=1),nrow=20,ncol=2)
x<-rbind(a,b,c)
OrderKmeansCpp(x,K=3)
OrderKmeansCpp(x,K=3,num_init=8)
PeakRange

**Peak Ranges Selection**

**Description**
Select the narrow peak ranges.

**Usage**
\[\text{PeakRange(score, tolerance = 1, point_max = 5)}\]

**Arguments**
- `score`: The score data to peak ranges.
- `tolerance`: The tolerance level, the selected narrow ranges are with score at least S-tolerance
- `point_max`: The largest candidate number of change points.

**Details**
For each column(window type), find the union of all the peak ranges whose associated scores are no less than S-tolerance, where S is highest score, then choose the largest window type with that the number of peak ranges meet the restriction.

**Value**
- `n_peak_range`: The number of peak ranges.
- `peak_range`: The location of peak ranges.

**References**

PriorRangeOrderKmeans

**Detect Number and Location of Change Points of Independent Data with Prior Ranges**

**Description**
Detect the number and locations of change points based on minimizing within segment quadratic loss with restriction of prior ranges that contain change points.

**Usage**
\[\text{PriorRangeOrderKmeans(x, prior_range_x, num_init = 10)}\]
Arguments

x The data to find change points.
prior_range_x The prior ranges that contain change points.
num_init The number of repetition times, in order to avoid local minimal. Default is 10. Must be integer.

Details

The K prior ranges contain K change points, each prior range contains one change point.

Value

num_change_point optimal number of change points.
change_point location of change points.

References


Examples

```r
a<-matrix(rnorm(40,mean=-1,sd=1),nrow=20,ncol=2)
b<-matrix(rnorm(120,mean=0,sd=1),nrow=60,ncol=2)
c<-matrix(rnorm(40,mean=1,sd=1),nrow=20,ncol=2)
x<-rbind(a,b,c)
l1<-c(15,25)
l2<-c(75,100)
prior_range_x<-list(l1,l2)
PriorRangeOrderKmeans(x,prior_range_x=list(l1,l2))
```

PriorRangeOrderKmeansCpp

*Detect Location of Change Points of Independent Data with Prior Ranges using Rcpp*

Description

Detect the location of change points based on minimizing within segment quadratic loss with restriction of prior ranges that contain change points.

Usage

`PriorRangeOrderKmeansCpp(x, prior_range_x, num_init = 10)`
RangeToPoint

Arguments

x           The data to find change points with dimension N x D, must be matrix
prior_range_x  The prior ranges that contain change points.
num_init     The number of repetition times, in order to avoid local minimal. Default is 10. Must be integer.

Details

The K change points form K+1 segments (1 2 ... change_point(1)) ... (change_point(K) ... N).

Value

num_change_point
  optimal number of change points.
change_point
  location of change points.

References


Examples

```r
a<-matrix(rnorm(40,mean=-1,sd=1),nrow=20,ncol=2)
b<-matrix(rnorm(120,mean=0,sd=1),nrow=60,ncol=2)
c<-matrix(rnorm(40,mean=1,sd=1),nrow=20,ncol=2)
x<-rbind(a,b,c)
l1<-c(15,25)
l2<-c(75,100)
prior_range_x<-list(l1,l2)
PriorRangeOrderKmeansCpp(x,prior_range_x=list(l1,l2))
```

---

RangeToPoint  Get Change Points from Peak Ranges

Description

Transform the peak ranges of change points to exact change points.

Usage

```r
RangeToPoint(y, n_peak_range, peak_range, get_loglik = GetLogLik)
```
ScorePlot

Arguments

- **y**: The original data to find change points. Must be one dimensional data.
- **n_peak_range**: The number of peak ranges of change points
- **peak_range**: The location of ranges of change points
- **get_loglik**: The method to get

Details

Find the exact change points with peak ranges based on log likelihood comparison.

Value

- **change_point**

Examples

```r
N <- 1000
N1 <- floor(0.1*N)
N2 <- floor(0.3*N)
a1 <- c(0.8, -0.3); c1 <- 0
a2 <- c(-0.5, 0.1); c2 <- 0
a3 <- c(0.5, -0.5); c3 <- 0
y <- rep(0,N)
L<-2
y[1:L] <- rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] <- y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] <- y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] <- y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
RangeToPoint(y,n_peak_range=2,peak_range=list(seq(70,105),seq(395,420)))
```

ScorePlot

Plot score

Description

Plot the score of each range, which represents how likely the range contains change points.

Usage

ScorePlot(result, ...)
Arguments

result The result of function MultiWindow. The argument ret_score of MultiWindow must be TRUE.

... Arguments to be passed to plot, such as main, xlab, ylab.

Value

A stair plot of score.

References


Examples

```r
N <- 1000
N1 <- floor(0.1*N)
N2 <- floor(0.3*N)
a1 <- c(0.8, -0.3); c1 <- 0
a2 <- c(-0.5, 0.1); c2 <- 0
a3 <- c(0.5, -0.5); c3 <- 0
y <- rep(0,N)
L<-2
y[1:L] <- rnorm(L)
for (n in (L+1):N){
  if (n <= N1) {
    y[n] <- y[(n-1):(n-L)] %*% a1 + c1 + rnorm(1)
  } else if (n <= (N1+N2)) {
    y[n] <- y[(n-1):(n-L)] %*% a2 + c2 + rnorm(1)
  } else {
    y[n] <- y[(n-1):(n-L)] %*% a3 + c3 + rnorm(1)
  }
}
result <- MultiWindow(y,window_list=c(100,50,20,10,5),point_max=5,ret_score=TRUE)
ScorePlot(result, main="score plot")
```
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