Package ‘segmentr’

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Type Package
Title Segment Data With Maximum Likelihood
Version 0.2.0
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Description Given a likelihood provided by the user, this package applies it to a given matrix dataset in order to find change points in the data that maximize the sum of the likelihoods of all the segments. This package provides a handful of algorithms with different time complexities and assumption compromises so the user is able to choose the best one for the problem at hand. The implementation of the segmentation algorithms in this package are based on the paper by Bruno M. de Castro, Florencia Leonardi (2018) <arXiv:1501.01756>. The Berlin weather sample dataset was provided by Deutscher Wetterdienst <https://dwd.de/>. You can find all the references in the Acknowledgments section of this package's repository via the URL below.
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auto_penalize

Penalize a likelihood function with a guessed penalty function

Description

Given a dataset, a likelihood function and penalty parameters on how to penalize big and small segments, this function makes an educated guess on a penalty function for the likelihood function.

Usage

auto_penalize(data, likelihood, big_segment_penalty = 10, small_segment_penalty = 10)

Arguments

data
dataset to be segmented by the segment() function

likelihood
function to be maximized using the segment() function. It’s used to find out the scale of the values in the segment function

big_segment_penalty
penalty factor for big segments. The bigger it is, the bigger the penalty on big segments. Must be greater than or equal to 1. Penalty on big segments is constant when it’s equal to 1. Default: 10

small_segment_penalty
penalty factor for small segments. The bigger it is, the bigger the penalty on small segments. Must be greater than or equal to 1. Penalty on small segments is constant when it’s equal to 1. Default: 10
berlin

Details

This function tries to fit a sum of two exponential functions to values inferred from the dataset and the likelihood function. The model for the penalty function we try to fit is in the form:

\[ C_1 \exp(s_1(x - L/2)) + C_2 \exp(s_2(-x + L/2)) \]

In the equation, \(C_1\) and \(s_1\) are, respectively, a multiplier constant and an exponential scale modifier for small segments, whereas \(C_2\) and \(s_2\) are the equivalent ones for big segments. \(L\) is the number of columns in the data matrix.

Assuming the penalty function to be as such, the parameters are estimated considering the scale of values yielded by the likelihood function for small and big segments, also taking into account the \texttt{big_segment_penalty} and \texttt{small_segment_penalty} tuning parameters, which can be used to adjust the effect of the penalty function over big and small segments, respectively.

Value

the likelihood function with the guessed penalty function applied

Examples

```r
## Not run:
penalized_likelihood <- auto_penalize(berlin, multivariate)
## End(Not run)
```

berlin

Daily temperatures from weather stations in Berlin

Description

Contains weather daily weather data from many Deutscher Wetterdienst weather stations in Berlin from the years of 2010 and 2011. Data was obtained using the package \texttt{rdwd} and reformatted to a format appropriate to be used for analysis in this object.

Usage

berlin

Format

A matrix containing daily temperatures, with each column representing a date and each column representing a weather station in Berlin

rows

columns dates from the years 2010 and 2011 ...

Source

https://www.dwd.de/DE/Home/home_node.html
**calculate_likelihood**  
*Calculate a dataset’s likelihood using change points of segmentr object*

**Description**

Given the change points in a segmentr object, this function splits a new dataset into segments and then calculates the total likelihood using the likelihood function of the segmentr object.

**Usage**

```r
calculate_likelihood(results, newdata, likelihood)
```

**Arguments**

- `results`: a segmentr object, which contains the definition of the change points to be applied
- `newdata`: a dataset for which we wish to calculate the total likelihood
- `likelihood`: a likelihood function to be used to calculate the likelihood of each segment

**Details**

This function splits a newdata dataset into segments according to the change points in the results segmentr object. It then uses the likelihood function of the segmentr object to calculate the total likelihood of the new object.

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**exactalg**  
*Segment data into exact change points*

**Description**

Find changes points in data calculating the penalized likelihood for all possible segment combinations.

**Usage**

```r
exactalg(data, likelihood, max_segments = ncol(data), allow_parallel = TRUE)
```
Arguments

- **data**: matrix for which to find the change points
- **likelihood**: a function receives the segment matrix as argument and returns a likelihood estimation. This function is used to calculate the change points that maximize the total likelihood. Depending on the algorithm being used, this function is likely to be executed many times, in which case it’s also likely to be the bottleneck of the function execution, so it’s advised that this function should have fast implementation.
- **max_segments**: an integer that defines the maximum amount of segments to split the data into.
- **allow_parallel**: allows parallel execution to take place using the registered cluster. Assumes a cluster is registered with the `foreach` package. Defaults to TRUE.

Details

Function that implements the dynamic programming algorithm, with the intent of finding points of independent change points for which the likelihood function is maximized. It analyzes all possible combinations, returning the change points that are guaranteed to segment the data matrix in the maximum likelihood independent change points. Because it analyzes all possible combinations of change points, it has a $O^2$ algorithm complexity, meaning it works in an acceptable computation time for small datasets, but it takes quite longer for datasets with many columns. For big datasets, `hieralg()` might be more adequate.

Value

A list of type `segmentr`, which has the two attributes:

- **changepoints**: a vector with the first index of each identified change point
- **segments**: a list of vectors, in which each vector corresponds to the indices that identifies a segment.

**hieralg**

*Segment data into change points assuming hierarchical structure*

Description

By assuming change points follow an hierarchical architecture, this architecture manages to run faster by not searching all possible branches

Usage

```r
hieralg(data, likelihood, max_segments = ncol(data),
   allow_parallel = TRUE)
```
Arguments

data matrix for which to find the change points

likelihood a function receives the segment matrix as argument and returns a likelihood estimation. This function is used to calculate the change points that maximize the total likelihood. Depending on the algorithm being used, this function is likely to be executed many times, in which case it’s also likely to be the bottleneck of the function execution, so it’s advised that this function should have fast implementation.

max_segments an integer that defines the maximum amount of segments to split the data into.

allow_parallel allows parallel execution to take place using the registered cluster. Assumes a cluster is registered with the foreach package. Defaults to TRUE.

Details

Fast algorithm that segments data into change points, and it does so by simplifying by reducing the search possibilities by assuming data split in an hierarchical structure, i.e. a segment found in a first trial is assumed to contain only segments independent of the rest of the data. This algorithm usually runs very fast, but is known to yield less accurate results, possibly not finding the exact change points that would maximize likelihood.

Value

a list of type segmentr, which has the two attributes:

• changepoints: a vector with the first index of each identified change point

• segments: a list of vectors, in which each vector corresponds to the indices that identifies a segment.

hybridalg Segment data into change points using a mixed hierarchical-exact approach

Description

For the larger datasets, assume the data is hierarchical, but calculate the exact segments when they’re smaller than a threshold

Usage

hybridalg(data, likelihood, allow_parallel = TRUE,
max_segments = ncol(data), threshold = 50)
Argument

- **data**: Matrix for which to find the change points
- **likelihood**: A function receives the segment matrix as argument and returns a likelihood estimation. This function is used to calculate the change points that maximize the total likelihood. Depending on the algorithm being used, this function is likely to be executed many times, in which case it’s also likely to be the bottleneck of the function execution, so it’s advised that this function should have fast implementation.
- **allow_parallel**: Allows parallel execution to take place using the registered cluster. Assumes a cluster is registered with the foreach package. Defaults to TRUE.
- **max_segments**: An integer that defines the maximum amount of segments to split the data into.
- **threshold**: The threshold for which the exact algorithm will be used, i.e. when the number of columns in the segment is less than or equal to the threshold.

Details

This algorithm implements an approach mixing the hierarchical and exact algorithms. It uses the hierarchical algorithms when the size of the segment is bigger than the threshold, and then goes on to use the exact algorithm when the size of the segment is less than or equal to the threshold.

Value

A list of type `segmentr`, which has the two attributes:

- **changepoints**: A vector with the first index of each identified change point
- **segments**: A list of vectors, in which each vector corresponds to the indices that identifies a segment.

Description

Estimate the likelihood of a given segment using the discrete multivariate estimation, implemented efficiently in C++

Usage

```r
multivariate(data, na_action = function(d) d[, colSums(is.na(d)) == 0, drop = FALSE])
```

Arguments

- **data**: Matrix to estimate the multivariate of. Each row is considered to be an observation, and each column is considered to be a different variable.
- **na_action**: A function that is applied to the data parameter. Defaults to removing columns with NA.
Details

Calculates the discrete log likelihood multivariate estimation of a data matrix using an algorithm implemented in C++ for performance. This is intended to be used in conjunction with `segment()`, as the log likelihood function is executed multiple times, which makes it the bottleneck of the computation. Because the multivariate is so commonly used, this efficient implementation is provided.

Value

the estimate of the Discrete Maximum Likelihood for the dataframe provided.

Description

Prints a short description of the segments found in the `segmentr` object

Usage

```r
## S3 method for class 'segmentr'
print(x, ...)
```

Arguments

- `x`: an object of type `segmentr`, containing change point information
- `...`: further arguments to be passed down to other methods

Details

A short representation of the segments is printed on the screen, using the `start:end` range notation.

Examples

```r
make_segment <- function(n, p) matrix(rbinom(100 * n, 1, p), nrow = 100)
data <- cbind(make_segment(5, 0.1), make_segment(10, 0.9), make_segment(2, 0.1))
mean_lik <- function(X) abs(mean(X) - 0.5) * ncol(X)^2
x <- segment(data, likelihood = mean_lik, algorithm = "hieralg")
print(x)
```
r_multivariate

Logarithmic Discrete Multivariate Likelihood estimation function implemented in R

Description

Estimate the likelihood of a given segment using the discrete multivariate estimation, but code runs more slowly due to R implementation

Usage

r_multivariate(data, na.omit = TRUE)

Arguments

data Matrix to estimate the multivariate of. Each row is considered to be an observation, and each column is considered to be a different variable.

na.omit If true, omits NAs from the dataset.

Details

This log likelihood function is implemented in R in order to be used to benchmark against the multivariate() version implemented in C++ for performance.

Value

The estimate of the Discrete Maximum Likelihood for the dataframe provided.

segment

Segment data into change points

Description

Generic function to segment data into separate change points according to specified algorithm

Usage

segment(data, likelihood, max_segments = ncol(data), allow_parallel = TRUE, algorithm = "exact", ...)

Arguments

- **data**: matrix for which to find the change points
- **likelihood**: a function receives the segment matrix as argument and returns a likelihood estimation. This function is used to calculate the change points that maximize the total likelihood. Depending on the algorithm being used, this function is likely to be executed many times, in which case it's also likely to be the bottleneck of the function execution, so it's advised that this function should have fast implementation.
- **max_segments**: an integer that defines the maximum amount of segments to split the data into.
- **allow_parallel**: allows parallel execution to take place using the registered cluster. Assumes a cluster is registered with the foreach package. Defaults to TRUE.
- **algorithm**: can be of type exact, hierarchical or hybrid. Default: exact
- ... other parameters to be passed to the underlying function

Details

This function can be used as a generic function to call any of the algorithms implemented by the package. Depending on the type of data the user wants to segment, one algorithm might be more adequate than the others.

Value

a list of type segmentr, which has the two attributes:

- **changepoints**: a vector with the first index of each identified change point
- **segments**: a list of vectors, in which each vector corresponds to the indices that identifies a segment.

See Also

- `exactalg()` for the exact algorithm, `hieralg()` for the hierarchical algorithm implementation, `hybridalg()` for the hybrid algorithm implementation.

Examples

```r
make_segment <- function(n, p) matrix(rbinom(100 * n, 1, p), nrow = 100)
data <- cbind(make_segment(5, 0.1), make_segment(10, 0.9), make_segment(2, 0.1))
mean_lik <- function(X) abs(mean(X) - 0.5) * ncol(X)^2
segment(data, likelihood = mean_lik, algorithm = "hieralg")
```
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