

# Package ‘multiscale’

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**Type** Package

**Title** Multiscale Inference for Nonparametric Regression(s) with Time Series Errors

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**Description** This package performs a multiscale analysis of a nonparametric regression or nonparametric regressions with time series errors. In case of one regression, it is possible to detect where the trend function is increasing or decreasing. In case of multiple regression, the test identifies regions where the trend functions are different from each other. See Khismatullina and Vogt (2019) and Khismatullina and Vogt (2020) for more information and theory.

**License** GPL ( $\geq 2$ )

**Imports** Rcpp ( $\geq 1.0.4$ )

**LinkingTo** Rcpp

**RoxygenNote** 7.1.0

**Encoding** UTF-8

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multiscale-package	<i>Multiscale Inference for Nonparametric Regression(s)</i>
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## Description

This package performs a multiscale analysis of a nonparametric regression or nonparametric regressions with time series errors.

In case of a single nonparametric regression, the multiscale method to test qualitative hypotheses about the nonparametric time trend  $m$  in the model  $Y_t = m(t/T) + \epsilon_t$  with time series errors  $\epsilon_t$  is provided. The method was first proposed in Khismatullina and Vogt (2019). It allows to test for shape properties (areas of monotonic decrease or increase) of the trend  $m$ .

This method require an estimator of the long-run error variance  $\sigma^2 = \sum_{l=-\infty}^{\infty} Cov(\epsilon_0, \epsilon_l)$ . Hence, the package also provides the difference-based estimator for the case that the errors belong to the class of  $AR(\infty)$  processes. The estimator was first proposed in Khismatullina and Vogt (2019).

In case of multiple nonparametric regressions, the multiscale method to test qualitative hypotheses about the nonparametric time trends in the context of epidemic modelling is provided. Specifically, we assume that the we observe a sample of the count data  $\{\mathcal{X}_i = \{X_{it} : 1 \leq t \leq T\}\}$ , where  $X_{it}$  are quasi-Poisson distributed with time-varying intensity parameter  $\lambda_i(t/T)$ . The multiscale method allows to test whether intensisty parameters are different or not, and if they are, it detects the regions where the trends significantly differ from each other. The method was first proposed in Khismatullina and Vogt (2020) and can be used for comparing the rates of infection of COVID-19 across countries.

## Details

This package performs a multiscale analysis of a nonparametric regression or nonparametric regressions with time series errors.

## References

Khismatullina M., Vogt M. Multiscale inference and long-run variance estimation in non-parametric regression with time series errors //Journal of the Royal Statistical Society: Series B (Statistical Methodology). - 2019.

## See Also

<https://rss.onlinelibrary.wiley.com/doi/full/10.1111/rssb.12347>

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compute_minimal_intervals
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*Computes the set of minimal intervals as described in Duembgen (2002)*

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**Description**

Given a set of intervals, this function computes the corresponding subset of minimal intervals which are defined as follows. For a given set of intervals  $\mathcal{K}$ , all intervals  $\mathcal{I}_k \in \mathcal{K}$  such that  $\mathcal{K}$  does not contain a proper subset of  $\mathcal{I}_k$  are called minimal.

This function is needed for illustrative purposes. The set of all the intervals where our test rejects the null hypothesis may be quite large, hence, we would like to focus our attention on the smaller subset, for which we are still able to make simultaneous confidence intervals. This subset is the subset of minimal intervals, and it helps us to precisely locate the intervals of further interest.

More details can be found in Duembgen (2002) and Khismatullina and Vogt (2019, 2020)

**Usage**

```
compute_minimal_intervals(dataset)
```

**Arguments**

dataset	Set of the intervals. It needs to contain the following columns: "startpoint" - left end of the interval; "endpoint" - right end of the interval.
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**Value**

Subset of minimal intervals

**Examples**

```
startpoint <- c(0, 0.5, 1)
endpoint   <- c(2, 2, 2)
dataset    <- data.frame(startpoint, endpoint)
minimal_ints <- compute_minimal_intervals(dataset)
```

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compute_quantiles	<i>Computes quantiles of the gaussian multiscale statistics.</i>
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**Description**

Quantiles from the gaussian version of the test statistics which are used to approximate the critical values for the multiscale test.

**Usage**

```
compute_quantiles(
  t_len,
  grid = NULL,
  n_ts = 1,
  ijset = NULL,
  sigma = 1,
  deriv_order = 0,
  sim_runs = 1000,
  probs = seq(0.5, 0.995, by = 0.005)
)
```

**Arguments**

t_len	An integer. Sample size.
grid	Grid of location-bandwidth points as produced by the function <a href="#">construct_grid</a> or <a href="#">construct_weekly_grid</a> , list with the elements 'gset', 'bws', 'gtype'. If not provided, then the default grid is produced and used. For the construction of the default grid, see <a href="#">construct_grid</a> .
n_ts	An integer. Number of time series analyzed. Default is 1.
ijset	A matrix of integers. In case of multiple time series, we need to know which pairwise comparisons to perform. This matrix consists of all pairs of indices $(i, j)$ that we want to compare. If not provided, then all possible pairwise comparison are performed.
sigma	A double that is equal to $\sqrt{\text{long} - \text{runvaraince}}$ in case of $n\_ts = 1$ , or the overdispersion in case of $n\_ts > 1$ . If not given, then the default is 1.
deriv_order	An integer. Order of the derivative of the trend that is being investigated. Default is 0.
sim_runs	Number of simulation runs to produce quantiles. Default is 1000.
probs	A numeric vector of probability levels (1-alpha) for which the quantiles are computed. Default is probs=seq(0.5,0.995,by=0.005).

**Value**

quant Matrix with 2 rows where the first row contains the vector of probabilities and the second contains corresponding quantile of the gaussian statistics distribution.

**Examples**

```
compute_quantiles(100)
```

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compute_statistics	<i>Calculates the value of the test statistics both for single time series analysis and multiple time series analysis.</i>
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**Description**

Calculates the value of the test statistics both for single time series analysis and multiple time series analysis.

**Usage**

```
compute_statistics(
  data,
  sigma,
  n_ts = 1,
  grid = NULL,
  ijset = NULL,
  deriv_order = 0
)
```

**Arguments**

data	Vector (in case of $n_{ts} = 1$ ) or matrix (in case of $n_{ts} > 1$ ) that contains (a number of) time series that needs to be analyzed. In the latter case, each column of the matrix must contain one time series.
sigma	A double. Estimator of the square root of the long-run variance in case of $n_{ts} = 1$ , or estimator of the overdispersion parameter in case of $n_{ts} > 1$ .
n_ts	Number of time series analysed. Default is 1.
grid	Grid of location-bandwidth points as produced by the function <a href="#">construct_grid</a> or <a href="#">construct_weekly_grid</a> , it is a list with the elements 'gset', 'bws', 'gtype'. If not provided, then the default grid is produced and used. For the construction of the default grid, see <a href="#">construct_grid</a> .
ijset	In case of multiple time series ( $n_{ts} > 1$ ), we need to know which pairs of time series to compare. This matrix consists of all pairs of indices $(i, j)$ that we want to compare. If not provided, then all possible pairwise comparison are performed.
deriv_order	In case of time series, this denotes the order of the derivative of the trend that we estimate. Default is 0.

**Value**

In case of  $n_{ts} = 1$ :

stat Value of the multiscale statistics.

gset\_with\_vals A matrix (in case of  $n_{ts} = 1$ ) that contains the values of the individual test statistics for each pair of location-bandwidth together with the corresponding location and bandwidth.

In case of  $n_{ts} > 1$ :

stat Value of the multiscale statistics.

stat\_pairwise Matrix of the values of the statistics of the pairwise comparisons.

test\_matrices Return in case of  $n_{ts} > 1$ . List of matrices, each matrix contains test results for the pairwise comparison between time series. Each matrix is coded exactly as in case of  $n_{ts} = 1$ .

gset\_with\_vals A list of matrices that contains the values of the each individual test statistics for each pair of location-bandwidth together with the corresponding location and bandwidth.

ijset The matrix that consists of all pairs of indices  $(i, j)$  that we compared. The order of these pairs corresponds to the order in the list gset\_with\_vals.

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construct\_grid

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*Computes the location-bandwidth grid for the multiscale test.*


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**Description**

Computes the location-bandwidth grid for the multiscale test.

**Usage**

```
construct_grid(t, u_grid = NULL, h_grid = NULL, deletions = NULL)
```

**Arguments**

t	Sample size.
u_grid	Vector of location points in the unit interval [0,1]. If u.grid=NULL, a default grid is used.
h_grid	Vector of bandwidths, each bandwidth is supposed to lie in (0,0.5). If h.grid=NULL, a default grid is used.
deletions	Logical vector of the length $\text{len}(\text{u.grid}) * \text{len}(\text{h.grid})$ . Each element is either TRUE, which means that the corresponding location-bandwidth point (u, h) is NOT deleted from the grid, or FALSE, which means that the corresponding location-bandwidth point (u, h) IS deleted from the grid. Default is deletions = NULL in which case nothing is deleted. See vignette for the use.

**Value**

gset Matrix of location-bandwidth points (u,h) that remains after deletions, the i-th row gset[i,] corresponds to the i-th point (u,h).

bws Vector of bandwidths (after deletions).

lens Vector of length=length(bws), lens[i] gives the number of locations in the grid for the i-th bandwidth level.

gtype Type of grid that is used, either 'default' or 'non-default'.

gset\_full Matrix of all location-bandwidth pairs (u,h) including deleted ones.

pos\_full Logical vector indicating which points (u,h) have been deleted.

**Examples**

```
construct_grid(100)
construct_grid(100, u_grid = seq(from = 0.05, to = 1, by = 0.05),
               h_grid = c(0.1, 0.2, 0.3, 0.4))
```

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`construct_weekly_grid` *Computes the location-bandwidth weekly grid for the multiscale test.*

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**Description**

Computes the location-bandwidth weekly grid for the multiscale test.

**Usage**

```
construct_weekly_grid(t, min_len = 7, nmbr_of_wks = 4)
```

**Arguments**

t	Sample size.
min_len	Integer, equal to the minimal length of the interval considered. Default is 7, i.e. a week.
nmbr_of_wks	Integer, equal to the numbers of wks considered as maximal interval possible. Default is 4

**Value**

gset Matrix of location-bandwidth points (u, h), the i-th row gset[i, ] corresponds to the i-th point (u, h).

bws Vector of bandwidths.

lens Vector of length=length(bws), lens[i] gives the number of locations in the grid for the i-th bandwidth level.

gtype Type of grid that is used, either 'default' or 'non-default'.

gset\_full Matrix of all possible location-bandwidth pairs (u,h).

pos\_full Logical vector indicating which points (u,h) have been deleted.

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estimate_lrv	<i>Computes estimator of the long-run variance of the error terms.</i>
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**Description**

A difference based estimator for the coefficients and long-run variance in case of a nonparametric regression model  $Y(t) = m(t/T) + \epsilon(t)$  where the errors are AR(p). The procedure was first introduced in Khismatullina and Vogt (2019).

**Usage**

```
estimate_lrv(data, q, r_bar, p)
```

**Arguments**

data A vector of Y(1), Y(2), ... Y(T).

q, r\_bar Integers, tuning parameters.

p AR order of the error terms.

**Value**

lrv Estimator of the long run variance of the error terms.

ahat Vector of length p of estimated AR coefficients.

vareta Estimator of the variance of the innovation term

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multiscale_test	<i>Carries out the multiscale test given that the values the estimates of long-run variance have already been computed.</i>
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## Description

Carries out the multiscale test given that the values the estimates of long-run variance have already been computed.

## Usage

```
multiscale_test(
  data,
  sigma,
  n_ts = 1,
  grid = NULL,
  ijset = NULL,
  alpha = 0.05,
  sim_runs = 1000,
  deriv_order = 0
)
```

## Arguments

data	Vector (in case of $n\_ts = 1$ ) or matrix (in case of $n\_ts > 1$ ) that contains (a number of) time series that needs to be analyzed.
sigma	A double. Estimator of the square root of the long-run variance in case of one time series, or estimator of the overdispersion parameter in case of multiple time series.
n_ts	Number of time series analysed. Default is 1.
grid	Grid of location-bandwidth points as produced by the function <a href="#">construct_grid</a> .
ijset	Matrix of all pairs of indices (i, j) that we want to compare.
alpha	Significance level. Default is 0.05.
sim_runs	Number of simulation runs to produce quantiles. Default is 1000.
deriv_order	An integer. Order of the derivative of the trend that is being investigated. Default is 0.

## Value

quant Quantile that was used for testing calculated from the gaussian distribution.

statistics Value of the multiscale statistics.

test\_matrix Return in case of  $n\_ts = 1$ . Matrix of test results for the multiscale test defined in Khismatullina and Vogt (2019).  $test\_matrix[i,j] = -1$ : test rejects the null for the j-th location u and the i-th bandwidth h and indicates a decrease in the trend  $test\_matrix[i,j] = 0$ : test does not reject the null for the j-th location u and the i-th bandwidth h  $test\_matrix[i,j] = 1$ : test rejects the null for the j-th location u and the i-th bandwidth h and indicates an increase in the trend  $test\_matrix[i,j] = 2$ : no test is carried out at j-th location u and i-th bandwidth h (because the point (u, h) is excluded from the grid as specified by the 'deletions' option in the function [construct\\_grid](#))



`test_matrices` Return in case of `n_ts > 1`. List of matrices, each matrix contains test results for the pairwise comparison between time series. Each matrix is coded exactly as in case of `n_ts = 1`.

`gset_with_vals` Either a matrix (in case of `n_ts = 1`) or a list of matrices (in case of `n_ts > 1`) that contains test results together with location-bandwidth points.

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<code>plot_sizer_map</code>	<i>Plots SiZer map from the test results of the multiscale testing procedure.</i>
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### Description

Plots SiZer map from the test results of the multiscale testing procedure.

### Usage

```
plot_sizer_map(
  u_grid,
  h_grid,
  test_results,
  plot_title = NA,
  greyscale = FALSE,
  ...
)
```

### Arguments

<code>u_grid</code>	Vector of location points in the unit interval [0,1].
<code>h_grid</code>	Vector of bandwidths from (0,0.5).
<code>test_results</code>	Matrix of test results created by <a href="#">multiscale_test</a> .
<code>plot_title</code>	Title of the plot. Default is NA and no title is written.
<code>greyscale</code>	Whether SiZer map is plotted in grey scale. Default is FALSE.
<code>...</code>	Any further options to be passed to the image function.

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<code>select_order</code>	<i>Calculates different information criterions for the number of time series based on the long-run variance estimator (defined in Khismatulina and Vogt (2019)) for a range of tuning parameters.</i>
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### Description

Tries to fit AR(1), ... AR(9) models for all given time series and calculates different information criterions (fpe, aic, aicc, sic, hq) for each of this fits.

### Usage

```
select_order(data, q = NULL, r = 5:15)
```

**Arguments**

data	One or a number of time series in a matrix. Column names of the matrix should be reasonable
q	A vector of integers that consists of different tuning parameters to analyse. If not supplied, q is taken to be $[2 \log T] : ([2\sqrt{T}] + 1)$ .
r	A vector of integers that consists of different tuning parameters $r\_bar$ to analyse. If not supplied, $r = 5:15$ .

**Value**

A list with a number of elements. orders A vector of chosen orders of length equal to the number of time series. For each time series the order is calculated as  $\max(which.min(fpe), \dots, which.min(hq))$ . The rest of the elements of the list are matrices that contain selected orders (among 1, ..., 9) for each information criterion. One matrix for each time series.

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