# Package 'breaktest'

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ACF

Calculating ACF values

## **Description**

A simple auxiliary function providing the estimates of autocorrelations of orders from 0 to N-1.

#### Usage

```
ACF(y)
```

#### **Arguments**

У

An input time series of interest.

## **Details**

The function is not intended to be used directly so it's not exported.

## Value

The vector of ACF values of orders from 0 to N-1. Due to the R's way of indexing the 1-st element is the autocorrelation of order 0, and the N-th value is the autocorrelation of order N-1.

 ${\sf ADF.test}$ 

A simple implementation of ADF test

# Description

A function for ADF test with the ability to select the number of lags. Lags are selected by informational criterions which can be modified as in Ng and Perron (2001) and Cavaliere et al. (2015).

# Usage

```
ADF.test(
  y,
  const = TRUE,
  trend = FALSE,
  max.lag = 0,
  criterion = NULL,
  modified.criterion = FALSE,
  rescale.criterion = FALSE
)
```

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#### **Arguments**

y The input time series of interest.

const, trend Whether a constand and trend are to be included.

max.lag Maximum lag number

criterion A criterion used to select number of lags. If lag selection is not needed keep this

NULL.

modified.criterion

Whether the unit-root test modification is needed.

rescale.criterion

Whether the rescaling informational criterion is needed. Designed to cope with

heteroscedasticity in residuals.

#### **Details**

Due to the Frisch-Waugh-Lovell theorem we first detrend y and then apply the test to the detrended series.

#### Value

List containing:

- y,
- · const,
- · trend,
- residuals,
- · coefficient estimates.
- · t-statistic value,
- · critical value,
- · Number of lags,
- indicator of stationarity.

#### References

Cavaliere, Giuseppe, Peter C. B. Phillips, Stephan Smeekes, and A. M. Robert Taylor. "Lag Length Selection for Unit Root Tests in the Presence of Nonstationary Volatility." Econometric Reviews 34, no. 4 (April 21, 2015): 512–36. https://doi.org/10.1080/07474938.2013.808065.

Ng, Serena, and Pierre Perron. "Lag Length Selection and the Construction of Unit Root Tests with Good Size and Power." Econometrica 69, no. 6 (2001): 1519–54. https://doi.org/10.1111/1468-0262.00256.

ADF.test.S 5

ADF. test. S Detrending bootstrap test by Smeekes (2013)
--

# Description

This bootstrap test is based on the recursive detrending procedure of Taylor (2002). The main idea is to apply the standard ADF test to the series with nuissanse parameters eliminated.

# Usage

```
ADF.test.S(
   y,
   const = TRUE,
   trend = FALSE,
   c = 0,
   gamma = 0,
   trim = 0.15,
   max.lag = 0,
   criterion = NULL,
   modified.criterion = FALSE,
   iter = 999
)
```

## **Arguments**

У	A series of interest.
const, trend	Whether the constant and trend are to be included.
С	A filtration parameter used to construct an autocorrelation coefficient.
gamma	Detrending type selection parameter. If 0 the OLS detrending is applied, if 1 the GLS detrending is applied, otherwise the autocorrelation coefficient is calculated as $1+c^{\gamma}T^{-\gamma}$ .
trim	A trimming parameter.
max.lag	The maximum lag for inner ADF testing.
criterion	A criterion used to select number of lags. If lag selection is not needed keep this NULL.
modified.criter	rion
	Whether the unit-root test modification is needed.
iter	The number of bootstrap iterations.

## **Details**

Critical values are calculated via a bootstrapping using MacKinnon-like regressions. For each number of observations and each number of variables obtained were 1999 values of test statistics. After that 1st, 2.5-th, 5-th, 10-th, and 97.5-th percentiles were calculated and saved along with the corresponding number of observations. This step was repeated 5 times to cope with possible biases. After that MacKunnon-like regressions were estimated.

AR

#### References

Taylor, A. M. Robert. "Regression-Based Unit Root Tests With Recursive Mean Adjustment for Seasonal and Nonseasonal Time Series." Journal of Business & Economic Statistics 20, no. 2 (April 2002): 269–81. https://doi.org/10.1198/073500102317352001.

MacKinnon, James G. "Critical Values for Cointegration Tests." Working Paper. Economics Department, Queen's University, January 2010. https://ideas.repec.org/p/qed/wpaper/1227.html.

Smeekes, Stephan. "Detrending Bootstrap Unit Root Tests." Econometric Reviews 32, no. 8 (July 2013): 869–91. https://doi.org/10.1080/07474938.2012.690693.

AR

Custom AR with extra information

# Description

Custom AR with extra information

#### Usage

```
AR(y, x, max.lag, criterion = "aic")
```

## **Arguments**

y Dependent variable.

x Exogenous explanatory variables.

max.lag The maximum number of lags.

criterion A criterion for lag number estimation.

#### **Details**

The function is not intended to be used directly so it's not exported.

#### Value

#### A list of:

• beta: estimates of coefficients,

• residuals: estimated residuals,

• predict: forecasted values,

ullet t.beta: t-statistics for beta,

• lag: estimated number of lags.

coint.test.GH 7

coint.test.GH	Gregory-Hansen test for the absense of cointegration
COTITE CCS C. OIT	Gregory Hansen test for the absense of connegration

#### **Description**

Gregory and Hansen (1996) test for the null hypothesis of no cointegration under a possible structural break at the unknown moment of time.

The authors proposed ADF- and Z-type tests, slightly modified to allow the presence of a possible regime shift. Three type of shifts are allowed:

- a shift in the constant,
- a shift in the constand with the trend included,
- and a shift in the constant and the cointegrating vector.

Critical values are calculated via the adopted MacKinnon procedure of estimating the model for the response surface.

# Usage

```
coint.test.GH(
    ...,
    shift = "level",
    trim = 0.15,
    max.lag = 10,
    criterion = "aic",
    add.criticals = TRUE
)
```

# Arguments

	Variables of interest.
shift	Expected break type.
trim	The trimming parameter to calculate break moment bounds.
max.lag	The maximum number of lags for the internal ADF testing.
criterion	The criterion for lag selection.
add.criticals	Whether critical values are to be returned. This argument is needed to suppress the calculation of critical values during the precalculation of tables needed for the p-values estimating.

## Value

An object of type cointGH. It's a list of

- shift: shift type,
- Za:  $MZ_{\alpha}$  statistic and c.v.,
- Zt:  $MZ_t$  statistic and c.v.,
- ADF: ADF statistic and c.v..

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#### References

MacKinnon, James G. "Critical Values for Cointegration Tests." Working Paper. Working Paper. Economics Department, Queen's University, January 2010. https://ideas.repec.org/p/qed/wpaper/1227.html.

Gregory, Allan W., and Bruce E. Hansen. "Residual-Based Tests for Cointegration in Models with Regime Shifts." Journal of Econometrics 70, no. 1 (January 1, 1996): 99–126. https://doi.org/10.1016/0304-4076(69)41685-7.

CPST.rescale

Generating rescaled series as in Cavaliere et al. (2015)

#### **Description**

This rescaling procedure is needed to cope with possible heteroscedasticity in the data. Simply it's achieved by taking a cumulative sum of the first difference normalized by the non-parametric local estimate of the variance.

# Usage

```
CPST.rescale(d.y, x, deter, k, max.lag)
```

## **Arguments**

d.y A series of first differences.

x A matrix of ADF RHS variables.

deter A matrix of deterministic variables for detrending.

k A lag of the corresponding ADF model.

max.lag The maximum possible lag.

# Value

A rescaled series.

#### References

Cavaliere, Giuseppe, Peter C. B. Phillips, Stephan Smeekes, and A. M. Robert Taylor. "Lag Length Selection for Unit Root Tests in the Presence of Nonstationary Volatility." Econometric Reviews 34, no. 4 (April 21, 2015): 512–36. https://doi.org/10.1080/07474938.2013.808065.

```
critical.values.KPSS.1p
```

Critical values for KPSS test with 1 break

# Description

Auxiliary function returning pre-calculated critical values for KPSS.1.break

# Usage

```
critical.values.KPSS.1p(model, break.point, N, k)
```

## **Arguments**

model	A scalar equal to	)

• 1: for model An,

• 2: for model A,

• 3: for model B,

• 4: for model C,

• 5: for model D,

• 6: for model E.

break.point Position of the break point.

N Number of observations.

k Number of RHS variables.

## **Details**

The function is not intended to be used directly so it's not exported.

```
critical.values.KPSS.2p
```

Critical values for KPSS test with 2 breaks

# Description

Auxiliary function returning pre-calculated critical values for KPSS.2.breaks

# Usage

```
critical.values.KPSS.2p(model, break.point, N)
```

# **Arguments**

model A scalar equal to

1: for the AA (without trend) model,2: for the AA (with trend) model,

3: for the BB model, 4: for the CC model,

• 5: for the AC-CA model.

break.point Position of the break point.

N Number of observations.

#### **Details**

The function is not intended to be used directly so it's not exported.

determinants.KPSS.1.break

Construct determinant variables for KPSS.1.break

# **Description**

Construct determinant variables for KPSS.1.break

# Usage

```
determinants.KPSS.1.break(model, N, break.point)
```

#### **Arguments**

model A scalar equal to

• 1: Model with trend, break in const,

2: Model with const and trend, break in const,3: Model with const and trend, break in trend,

• 4: Model with const and trend, break in const and trend.

N Number of observations.

break.point Break point.

## **Details**

Procedure to compute deterministic terms for KPSS with 1 structural break.

The function is not intended to be used directly so it's not exported.

#### Value

Matrix of determinant variables.

determinants.KPSS.2.breaks

Construct determinant variables for KPSS.2.breaks

## **Description**

Construct determinant variables for KPSS.2.breaks

## Usage

```
determinants.KPSS.2.breaks(model, N, break.point)
```

## **Arguments**

model A scalar equal to

- 1: for the AA (without trend) model,
- 2: for the AA (with trend) model,
- 3: for the BB model,
- 4: for the CC model,
- 5: for the AC-CA model,
- 6: for the AC-CA model,
- 7: for the AC-CA model.

N Number of observations.

break.point Positions for the first and second structural breaks (respective to the origin which

is 1).

# **Details**

Procedure to compute deterministic terms for KPSS with 2 structural breaks.

The function is not intended to be used directly so it's not exported.

# Value

Matrix of deterministic terms.

```
determinants.KPSS.N.breaks
```

Deterministic terms for KPSS.N.breaks

# Description

Procedure to compute deterministic terms for KPSS with m structural breaks.

## Usage

```
determinants.KPSS.N.breaks(model, N, break.point, const = FALSE, trend = FALSE)
```

DOLS DOLS

#### **Arguments**

model A scalar or vector of

1: for the break in const,2: for the break in trend,

• 3: for the break in const and trend.

N Number of observations. break.point Array of structural breaks.

const, trend Include constant and trend if TRUE.

#### **Details**

**model** should be either a scalar or a vector of the same size as the **break.point**. If scalar **model** will be repeated till the length of **break.point** is achieved.

## Value

Matrix of deterministic terms.

DOLS

Estimating DOLS regression for multiple known break points

## **Description**

Estimating DOLS regression for multiple known break points

## Usage

```
DOLS(y, x, model, break.point, k.lags, k.leads)
```

# **Arguments**

y A dependent (LHS) variable.

x A matrix of explanatory (RHS) variables.model See Carrion-i-Silvestre and Sansó (2006)

• 1: for model An,

• 2: for model A,

• 3: for model B,

• 4: for model C,

• 5: for model D,

• 6: for model E.

break.point A position of the break point.

k.lags, k.leads

A number of lags and leads in DOLS regression.

## **Details**

The function is not intended to be used directly so it's not exported.

DOLS.N.breaks 13

#### Value

A list of:

- Estimates of coefficients,
- · Estimates of residuals.
- · A value of BIC,
- t-statistics for the estimates of coefficients.

#### References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

DOLS.N.breaks

Estimating DOLS regression for multiple known break points

## **Description**

Estimating DOLS regression for multiple known break points

## Usage

```
DOLS.N.breaks(
   y,
   x,
   model,
   break.point,
   const = FALSE,
   trend = FALSE,
   k.lags,
   k.leads
)
```

# **Arguments**

y A dependent (LHS) variable.

x A matrix of explanatory (RHS) variables.

model A scalar or vector of break types:

• 1: for the break in const.

• 2: for the break in trend.

• 3: for the break in const and trend.

break.point An array of moments of structural breaks.

Whether a constant or trend are to be included.

k.lags, k.leads

const, trend

A number of lags and leads in DOLS regression.

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#### **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of:

- Estimates of coefficients,
- Estimates of residuals,
- A set of informational criterions values,
- t-statistics for the estimates of coefficients.

DOLS.vars.N.breaks

Preparing variables for DOLS regression with multiple known break points

## **Description**

Preparing variables for DOLS regression with multiple known break points

#### Usage

```
DOLS.vars.N.breaks(
   y,
   x,
   model,
   break.point,
   const = FALSE,
   trend = FALSE,
   k.lags,
   k.leads
)
```

## **Arguments**

y A dependent (LHS) variable.

x A matrix of explanatory (RHS) variables.

model A scalar or vector of

• 1: for the break in const.

• 2: for the break in trend.

• 3: for the break in const and trend.

break.point An array of moments of structural breaks.

whether a constant or trend are to be included.

k.lags, k.leads

A number of lags and leads in DOLS regression.

## **Details**

The function is not intended to be used directly so it's not exported.

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#### Value

A list of LHS and RHS variables.

eos.break.test

Andrews-Kim (2006) test

## **Description**

Test for structural break at the end of the sample.

# Usage

```
eos.break.test(eq, m, dataset)
```

## **Arguments**

eq Base model formula. At the moment all the variables included should be defined

explicitly, dynamic regressors (i.e. functions etc.) are not supported.

m Post-break period length.

dataset Source of the data.

#### **Details**

See Andrews and Kim (2006) for the detailed description.

# Value

A list of

- m,
- estimated values of P- and R-tests,
- sequences of auxiliary statistics  $P_j$  and  $R_j$ ,
- the corresponding p-values.

#### References

Andrews, D. W. K. "End-of-Sample Instability Tests." Econometrica 71, no. 6 (2003): 1661–94. https://doi.org/10.1111/1468-0262.00466.

Andrews, Donald W. K, and Jae-Young Kim. "Tests for Cointegration Breakdown Over a Short Time Period." Journal of Business & Economic Statistics 24, no. 4 (2006): 379–94. https://doi.org/10.1198/07350010600

16 GLS.bt

GLS

Custom GLS with extra information

# Description

Getting GLS estimates of betas, residuals, forecasted values and t-values.

# Usage

```
GLS(y, z, c)
```

# **Arguments**

y Dependent variable.

z Explanatory variables.

c Coefficient for  $\rho$  calculation.

## **Details**

The function is not intended to be used directly so it's not exported.

## Value

The list of betas, residuals, forecasted values and t-values.

GLS.bt GLS fitering

# Description

GLS fitering

# Usage

# Arguments

trim Trimming parameter.

c Filtering parameter.

info.criterion 17

rions	
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# Description

Information criterions

# Usage

```
info.criterion(resids, extra, modification = FALSE, alpha = 0, y = NULL)
```

# Arguments

resids	Input residuals needed for estimating the values of information criterions.
extra	Number of extra parameters needed for estimating the punishment term.
modification	Whether the unit-root test modificaton is needed. See Ng and Perron (2001) for further information.
alpha	The coefficient $\alpha$ of $y_{t-1}$ in ADF model. Needed only for criterion modification purposes.
У	The vector of $y_{t-1}$ in ADF model. Needed only for criterion modification purposes.

#### **Details**

Calculating the value of the following informational criterions:

- Akaike,
- Schwarz (Bayesian),
- Hannan-Quinn,
- Liu et al.

## Value

The list of information criterions values.

# References

Ng, Serena, and Pierre Perron. "Lag Length Selection and the Construction of Unit Root Tests with Good Size and Power." Econometrica 69, no. 6 (2001): 1519–54. https://doi.org/10.1111/1468-0262.00256.

18 KPSS

ΚP

Kejrival-Perron procedure of breaks number detection

## **Description**

Kejrival-Perron procedure of breaks number detection

# Usage

```
KP(y, const = FALSE, breaks = 1, criterion = "aic", trim = 0.15)
```

#### **Arguments**

y An input series of interest.

const Whether the break in constant is allowed.

breaks Number of breaks.

criterion Needed information criterion: aic, bic, hq or lwz.
trim A trimming value for a possible break date bounds.

#### Value

The estimated optimal break point.

## References

Kejriwal, Mohitosh, and Pierre Perron. "A Sequential Procedure to Determine the Number of Breaks in Trend with an Integrated or Stationary Noise Component: Determination of Number of Breaks in Trend." Journal of Time Series Analysis 31, no. 5 (September 2010): 305–28. https://doi.org/10.1111/j.1467-9892.2010.00666.x.

**KPSS** 

Auxiliary function returning KPSS statistic value.

# **Description**

Auxiliary function returning KPSS statistic value.

#### Usage

```
KPSS(resids, variance)
```

# **Arguments**

resids The series of residuals.

variance The value of the long-run variance.

# **Details**

The function is not intended to be used directly so it's not exported.

KPSS.1.break

KPSS.1.break

KPSS-test with known structural break

#### **Description**

Computes the cointegration test with one known structural break.

#### Usage

```
KPSS.1.break(y, x, model, break.point, weakly.exog = TRUE, 11.init)
```

## **Arguments**

y An input (LHS) time series of interest.

x A matrix of (RHS) explanatory stochastic regressors.

model A scalar equal to

• 1: for model An,

• 2: for model A,

• 3: for model B,

• 4: for model C,

• 5: for model D,

• 6: for model E.

break.point Posit

Position of the break point.

weakly.exog

Exogeneity of the stochastic regressors

• TRUE: if the regressors are weakly exogenous,

• FALSE: if the regressors are not weakly exogenous (DOLS is used in this case).

ll.init

Scalar, defines the initial number of leads and lags for DOLS.

## **Details**

The code provided is the original GAUSS code ported to R.

See Carrion-i-Silvestre and Sansó (2006) for further details.

#### Value

A list of:

- beta: DOLS estimates of the coefficients regressors,
- tests: SC test (coinKPSS-test),
- resid: Residuals of the model,
- t.beta: Individual significance t-statistics,
- break\_point: Break points.

#### References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

20 KPSS.1.break.unknown

KPSS.1.break.unknown KPSS-test of cointegration

#### **Description**

Procedure for testing the null of cointegration in the possible presence of structural breaks.

#### Usage

```
KPSS.1.break.unknown(y, x, model, weakly.exog, ll.init)
```

## **Arguments**

y (Tx1)-vector of the dependent variable

x (Txk)-matrix of explanatory stochastic regressors

model A scalar equal to

• 1: for model An,

• 2: for model A,

• 3: for model B,

• 4: for model C,

• 5: for model D,

• 6: for model E.

weakly.exog Exogeneity of the stochastic regressors

• TRUE: if the regressors are weakly exogenous,

• FALSE: if the regressors are not weakly exogenous (DOLS is used in this case).

11. init Scalar, defines the initial number of leads and lags for DOLS.

## **Details**

Computes the cointegration test with one unknown structural break where the break point is estimated either minimizing the value of the statistic or the sum of the squared residuals. The estimation of the cointegrating relationship bases on DOLS.

The code provided is the original GAUSS code ported to R.

See Carrion-i-Silvestre and Sansó (2006) for further details.

# Value

(2x2)-matrix, where the first rows gives the value of the min(SC) test and the estimated break point; the second row gives the value of the SC statistic, where the break point is estimated as min(SSR).

#### References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

KPSS.2.breaks 21

KPSS.2.breaks

KPSS-test with 2 known structural breaks

## **Description**

Procedure to compute the KPSS test with two structural breaks

#### Usage

```
KPSS.2.breaks(y, model, break.point, max.lag, kernel)
```

#### **Arguments**

y An input (LHS) time series of interest.

model A scalar equal to

1: for the AA (without trend) model,2: for the AA (with trend) model,

3: for the BB model, 4: for the CC model,

• 5: for the AC-CA model.

break.point Positions for the first and second structural breaks (respective to the origin which

is 1).

max.lag scalar, with the maximum order of the parametric correction. The final order of

the parametric correction is selected using the BIC information criterion.

kernel Kernel for calculating long-run variance

• bartlett: for Bartlett kernel,

• quadratic: for Quadratic Spectral kernel,

• NULL for the Kurozumi's proposal, using Bartlett kernel.

#### **Details**

The break points are known

The code provided is the original GAUSS code ported to R.

See Carrion-i-Silvestre and Sansó (2007) for further details.

#### Value

#### A list of:

• beta: DOLS estimates of the coefficients regressors,

• tests: SC test (coinKPSS-test),

• resid: Residuals of the model.

• t.beta: t-statistics for beta,

• break\_point: Break points.

# References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "The KPSS Test with Two Structural Breaks." Spanish Economic Review 9, no. 2 (May 16, 2007): 105–27. https://doi.org/10.1007/s10108-006-9017-8.

22 KPSS.2.breaks.unknown

KPSS.2.breaks.unknown KPSS-test with 2 unknown structural breaks

## **Description**

Procedure to compute the KPSS test with two structural breaks

## Usage

```
KPSS.2.breaks.unknown(y, model, max.lag = 0, kernel = "bartlett")
```

## **Arguments**

y (Tx1)-vector of time series.

model A scalar equal to

- 1: for the AA (without trend) model,
- 2: for the AA (with trend) model,
- 3: for the BB model,
- 4: for the CC model,
- 5: for the AC-CA model.

max.lag scalar, with the maximum order of the parametric correction. The final order of

the parametric correction is selected using the BIC information criterion.

kernel Kernel for calculating long-run variance

- bartlett: for Bartlett kernel,
- quadratic: for Quadratic Spectral kernel,
- NULL for the Kurozumi's proposal, using Bartlett kernel.

## **Details**

The break points are known

The code provided is the original GAUSS code ported to R.

See Carrion-i-Silvestre and Sansó (2007) for further details.

## Value

Value of test statistic.

#### References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

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KPSS.HLT

Unit root testing procedure under a single structural break.

## **Description**

Unit root testing procedure under a single structural break.

#### Usage

```
KPSS.HLT(y, const = FALSE, trim = 0.15)
```

#### **Arguments**

y A series of interest.

const Whether a constant should be included.

trim The trimming parameter to find the lower and upper bounds of possible break

dates.

#### Value

The value of test statistic.

#### References

Harvey, David I., Stephen J. Leybourne, and A. M. Robert Taylor. "Unit Root Testing under a Local Break in Trend." Journal of Econometrics 167, no. 1 (2012): 140–67.

KPSS.N.breaks

KPSS-test with multiple known structural breaks

## **Description**

Procedure to compute the KPSS test with multiple known structural breaks

## Usage

```
KPSS.N.breaks(
  y,
  x,
  model,
  break.point,
  const = FALSE,
  trend = FALSE,
  weakly.exog = TRUE,
  lags.init,
  leads.init,
  max.lag,
  kernel,
  criterion = "bic"
)
```

24 KPSS.N.breaks

#### **Arguments**

y An input (LHS) time series of interest.

x A matrix of (RHS) explanatory stochastic regressors.

model A scalar or vector of

1: for the break in const,2: for the break in trend,

• 3: for the break in const and trend.

break.point Array of structural breaks.

const, trend Whether a constant or trend should be included.

weakly.exog Boolean where we specify whether the stochastic regressors are exogenous or

not

• TRUE: if the regressors are weakly exogenous,

FALSE: if the regressors are not weakly exogenous (DOLS is used in this
case).

lags.init, leads.init

Scalars defining the initial number of lags and leads for DOLS.

max.lag scalar, with the maximum order of the parametric correction. The final order of

the parametric correction is selected using the BIC information criterion.

kernel Kernel for calculating long-run variance

• bartlett: for Bartlett kernel,

• quadratic: for Quadratic Spectral kernel,

• NULL for the Kurozumi's proposal, using Bartlett kernel.

criterion Information criterion for DOLS lags and leads selection: aic, bic, hq, or lwz.

# Value

#### A list of

• beta: DOLS estimates of the coefficients regressors,

• tests: SC test (coinKPSS-test),

• resid: Residuals of the model.

• t.beta: t-statistics for beta,

• DOLS. lags: The estimated number of lags and leads in DOLS,

• break\_point: Break points.

#### References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "The KPSS Test with Two Structural Breaks." Spanish Economic Review 9, no. 2 (May 16, 2007): 105–27. https://doi.org/10.1007/s10108-006-9017-8.

```
KPSS.N.breaks.bootstrap
```

KPSS-test with multiple unknown structural breaks

# Description

Procedure to compute the KPSS test with multiple unknown structural breaks

## Usage

```
KPSS.N.breaks.bootstrap(
  у,
  Х,
  model,
  break.point,
  const = FALSE,
  trend = FALSE,
  weakly.exog = TRUE,
  lags.init,
  leads.init,
  max.lag,
  kernel,
  iter = 9999,
  bootstrap = "sample",
  criterion = "bic"
)
```

## **Arguments**

y An input (LHS) time series of interest.

x A matrix of (RHS) explanatory stochastic regressors.

model A scalar or vector of

- 1: for the break in const,
- 2: for the break in trend,
- 3: for the break in const and trend.

break.point Array of structural breaks.

const Include constant if **TRUE**.

trend Include trend if **TRUE**.

Boolean where we specify whether the stochastic regressors are exogenous or

- TRUE: if the regressors are weakly exogenous,
- FALSE: if the regressors are not weakly exogenous (DOLS is used in this case).

lags.init, leads.init

weakly.exog

Scalars defining the initial number of lags and leads for DOLS.

max.lag scalar, with the maximum order of the parametric correction. The final order of the parametric correction is selected using the BIC information criterion.

26 lagn

kernel Kernel for calculating long-run variance

• bartlett: for Bartlett kernel,

• quadratic: for Quadratic Spectral kernel,

• NULL for the Kurozumi's proposal, using Bartlett kernel.

iter Number of bootstrap iterations.

bootstrap Type of bootstrapping:

• "sample": sampling from residuals with replacement,

- "Cavaliere-Taylor": multiplying residuals by  ${\cal N}(0,1)$ -distributed variable,

• "Rademacher": multiplying residuals by Rademacher-distributed variable.

criterion Information criterion for DOLS lags and leads selection: aic, bic or lwz.

#### Value

A list of:

• test: The value of KPSS test statistic,

• p.value: The estimates p-value,

• bootstrapped: Bootstrapped auxiliary statistics.

lagn

Produce a vector lagged backward of forward

# Description

Produce a vector lagged backward of forward

#### Usage

$$lagn(x, i, na = NA)$$

#### **Arguments**

x Initial vector.

i Size of lag (lead if negative).

na Value to fill missing observations, NA by default.

# **Details**

The function is not intended to be used directly so it's not exported.

#### Value

Lagged or leaded vector.

Ir.var.bartlett 27

lr.var.bartlett

Calculating long-run variance with Bartlett kernel

## **Description**

Calculating long-run variance with Bartlett kernel

# Usage

```
lr.var.bartlett(y, 1 = NULL)
```

## **Arguments**

y A series of interest.

Number of lags/leads. If NULL then it's estimated.

## **Details**

The function is not intended to be used directly so it's not exported.

#### Value

Long-run variance.

lr.var.bartlett.AK

Calculating long-run variance

# Description

Procedure ALRVR to estimate the long-run variance as in Andrews (1991) and Kurozumi (2002).

# Usage

```
lr.var.bartlett.AK(y)
```

# **Arguments**

У

(Tx1) vector or residuals.

# Value

Long-run variance.

#### References

Andrews, Donald W. K. "Heteroskedasticity and Autocorrelation Consistent Covariance Matrix Estimation." Econometrica 59, no. 3 (1991): 817–58. https://doi.org/10.2307/2938229.

Kurozumi, Eiji. "Testing for Stationarity with a Break." Journal of Econometrics 108, no. 1 (May 1, 2002): 63–99. https://doi.org/10.1016/S0304-4076(01)00106-3.

28 lr.var.SPC

lr.var.quadratic

Calculating long-run variance with quadratic kernel

## **Description**

Calculating long-run variance with quadratic kernel

## Usage

```
lr.var.quadratic(y)
```

#### **Arguments**

У

A series of interest.

#### **Details**

The function is not intended to be used directly so it's not exported.

#### References

Andrews, Donald W. K. "Heteroskedasticity and Autocorrelation Consistent Covariance Matrix Estimation." Econometrica 59, no. 3 (1991): 817–58. https://doi.org/10.2307/2938229.

lr.var.SPC

Calculating long-run variance

# **Description**

Procedure ALRVR to estimate the long-run variance as in Sul, Phillips and Choi (2003).

# Usage

```
lr.var.SPC(y, max.lag = 0, kernel = "bartlett", criterion = "bic")
```

## **Arguments**

3

(Tx1) vector or residuals.

max.lag

Maximum number of lags. The exact number is selected by information criteri-

ons.

kernel

Kernel for calculating long-run variance

• "bartlett"": for Bartlett kernel,

• "quadratic": for Quadratic Spectral kernel,

• "NULL": for the Kurozumi's proposal, using Bartlett kernel.

criterion

The information crietreion: bic, aic or lwz.

# Details

Used are Quadratic Spectral and Bartlett kernels.

MDF.CHLT 29

#### Value

Long-run variance.

#### References

Sul, Donggyu, Peter C. B. Phillips, and Chi-Young Choi. "Prewhitening Bias in HAC Estimation." Oxford Bulletin of Economics and Statistics 67, no. 4 (August 2005): 517–46. https://doi.org/10.1111/j.1468-0084.2005.00130.x.

MDF.CHLT

MDF test for a single break and possible heterscedasticity

# Description

MDF test for a single break and possible heterscedasticity

## Usage

```
MDF.CHLT(y, max.lag = 10, trim = 0.15, iter = 499)
```

## **Arguments**

y A series of interest

max.lag The maximum possible lag.

trim Trimming parameter for lag selection

iter Number of bootstrap iterations.

## Value

An object of type mdfCHLT. It's a list of four sublists each containing:

- The value of  $MZ_{\alpha}$ , MSB,  $MZ_{t}$ , or ADF,
- The asymptotic c.v.,
- The bootstrapped c.v.

# References

Cavaliere, Giuseppe, David I. Harvey, Stephen J. Leybourne, and A.M. Robert Taylor. "Testing for Unit Roots in the Presence of a Possible Break in Trend and Nonstationary Volatility." Econometric Theory 27, no. 5 (October 2011): 957–91. https://doi.org/10.1017/S0266466610000605.

30 MDF.single

MDF.multiple

MDF procedure for multiple unknown breaks.

## **Description**

MDF procedure for multiple unknown breaks.

# Usage

```
MDF.multiple(
   y,
   const = FALSE,
   breaks = 1,
   breaks.star = 1,
   trim = 0.15,
   ZA = FALSE
)
```

## **Arguments**

y A series of interest.

const Whether the constant term should be included.

breaks Number of breaks.

breaks.star Number of breaks got from the Kejrival-Perron procedure.

trim Trimming value for a possible break date bounds.

ZA Whether ZA variant should be used.

# Value

A list of sublists each containing

- The value of statistic: MDF GLS, MDF OLS,
- ullet The asymptotic critical values. UR values are included as well.

MDF.single

MDF procedure for a single unknown break.

# Description

MDF procedure for a single unknown break.

# Usage

```
MDF.single(y, const = FALSE, trend = FALSE, trim = 0.15)
```

MZ.statistic 31

#### **Arguments**

y A series of interest.

const Whether the constant term should be included.
trend Whether the trend term should be included.

trim Trimming value for a possible break date bounds.

#### Value

A list of sublists each containing

- The value of statistic: MDF GLS, MDF OLS,
- The asymptotic critical values. UR values are included as well.

MZ.statistic

Calculating M-statistics by Stock (1990) and Perron and Ng (1996).

# **Description**

Calculating M-statistics by Stock (1990) and Perron and Ng (1996).

#### Usage

```
MZ.statistic(y, 1, const = FALSE, trend = FALSE)
```

# Arguments

y A series of interest.

1 Number of lags for inner ADF test.

const, trend Whether a constand and trend are to be included.

#### **Details**

The function is not intended to be used directly so it's not exported.

# Value

List of values of  $MZ_{\alpha}$ ,  $MZ_{t}$  and MSB statistics.

#### References

Perron, Pierre, and Serena Ng. "Useful Modifications to Some Unit Root Tests with Dependent Errors and Their Local Asymptotic Properties." The Review of Economic Studies 63, no. 3 (July 1, 1996): 435–63. https://doi.org/10.2307/2297890.

Stock, James H. "A Class of Tests for Integration and Cointegration." Kennedy School of Government, Harvard University, 1990.

NW.kernel

NW.estimation	Nadaraya-Watson kernel regression.
iiii.cotimation	Tradura ya Traibon kernet regression.

# Description

Nadaraya-Watson kernel regression.

#### Usage

```
NW.estimation(y, x, h, kernel = "unif")
```

## **Arguments**

y LHS dependent variable.x RHS explanation variable.

h Bandwidth.

kernel Needed kernel, currently only unif and gauss.

#### **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of arguments as well as the estimated coefficient vector and residuals.

# References

Harvey, David I., S. Leybourne, Stephen J., and Yang Zu. "Nonparametric Estimation of the Variance Function in an Explosive Autoregression Model." School of Economics. University of Nottingham, 2022.

NW.kernel

Nadaraya-Watson kernel series

# **Description**

Nadaraya-Watson kernel series

## Usage

```
NW.kernel(i, x, h, kernel = "unif")
```

## **Arguments**

i Current index.x Kernel variable.h Bandwidth.

kernel Needed kernel, currently only unif and gauss.

NW.loocv 33

#### **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A series of kernel function values.

#### References

Harvey, David I., S. Leybourne, Stephen J., and Yang Zu. "Nonparametric Estimation of the Variance Function in an Explosive Autoregression Model." School of Economics. University of Nottingham, 2022.

NW.loocy

LOO-CV for h in Nadaraya-Watson kernel regression.

# Description

LOO-CV for h in Nadaraya-Watson kernel regression.

## Usage

```
NW.loocv(y, x, kernel = "unif")
```

# Arguments

y LHS dependent variable.

x RHS explanation variable.

kernel Needed kernel, currently only unif and gauss.

# **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of arguments as well as the estimated bandwidth h.

## References

Harvey, David I., S. Leybourne, Stephen J., and Yang Zu. "Nonparametric Estimation of the Variance Function in an Explosive Autoregression Model." School of Economics. University of Nottingham, 2022.

OLS OLS

NW.volatility

NW.volatility - Nadaraya-Watson kernel volatility estimation

## **Description**

NW.volatility - Nadaraya-Watson kernel volatility estimation

#### Usage

```
NW.volatility(e, h, kernel = "unif")
```

## **Arguments**

e The series of interest.

h Bandwidth.

kernel Needed kernel, currently only unif and gauss.

#### **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of arguments as well as the estimated omega and s.e.

#### References

Cavaliere, Giuseppe, Peter C. B. Phillips, Stephan Smeekes, and A. M. Robert Taylor. "Lag Length Selection for Unit Root Tests in the Presence of Nonstationary Volatility." Econometric Reviews 34, no. 4 (April 21, 2015): 512–36. https://doi.org/10.1080/07474938.2013.808065.

0LS

Custom OLS with extra information

## **Description**

Getting OLS estimates of betas, residuals, forecasted values and t-values.

## Usage

```
OLS(y, x)
```

# Arguments

y Dependent variable.x Explanatory variables.

# Details

The function is not intended to be used directly so it's not exported.

p.values.SADF 35

#### Value

The list of:

• beta: estimates of coefficients,

• resid: estimated residuals,

• predict: forecasted values,

• t.beta: t-statistics for beta.

p.values.SADF

Critical values for SADF-type tests

# Description

Interpolating p-value for intermediate observation numbers for SADF-type tests.

# Usage

```
p.values.SADF(statistic, N.obs, cr.values)
```

# **Arguments**

statistic The statistic value.

N. obs The number of observations.cr.values The set of precalculated tables.

#### **Details**

The function is not intended to be used directly so it's not exported.

print.sadf

Custom functions for printing results in a nice way.

## **Description**

Custom functions for printing results in a nice way.

## Usage

```
## S3 method for class 'sadf'
print(x, ...)
## S3 method for class 'mdfHLT'
print(x, ...)
## S3 method for class 'mdfHLTN'
print(x, ...)
## S3 method for class 'mdfCHLT'
print(x, ...)
## S3 method for class 'cointGH'
print(x, ...)
```

#### **Arguments**

x Object containing results.

... Any additional arguments for print function.

PY.sequential

Sequential Perron-Yabu (2009) statistic for breaks at unknown date.

## **Description**

Sequential Perron-Yabu (2009) statistic for breaks at unknown date.

## Usage

```
PY.sequential(
   y,
   const = FALSE,
   breaks = 1,
   criterion = "aic",
   trim = 0.15,
   max.lag = 1
)
```

## **Arguments**

y The input series of interest.

const Allowing the break in constant.

breaks Number of breaks.

criterion Needed information criterion: aic, bic, hq or lwz.

trim A trimming value for a possible break date bounds.

max.lag The maximum possible lag in the model.

#### Value

The estimated Wald statistic.

#### References

Kejriwal, Mohitosh, and Pierre Perron. "A Sequential Procedure to Determine the Number of Breaks in Trend with an Integrated or Stationary Noise Component: Determination of Number of Breaks in Trend." Journal of Time Series Analysis 31, no. 5 (September 2010): 305–28. https://doi.org/10.1111/j.1467-9892.2010.00666.x.

PY.single 37

PY.single

Perron-Yabu (2009) statistic for break at unknown date.

# **Description**

Perron-Yabu (2009) statistic for break at unknown date.

# Usage

```
PY.single(
   y,
   const = FALSE,
   trend = FALSE,
   criterion = "aic",
   trim = 0.15,
   max.lag
)
```

## Arguments

y The input series of interest.

const, trend Allowing the break in constant or trend.

criterion Needed information criterion: aic, bic, hq or lwz.
trim A trimming value for a possible break date bounds.

max.lag The maximum possible lag in the model.

## Value

A list of the estimated Wald statistic as well as its c.v.

#### References

Perron, Pierre, and Tomoyoshi Yabu. "Testing for Shifts in Trend With an Integrated or Stationary Noise Component." Journal of Business & Economic Statistics 27, no. 3 (July 2009): 369–96. https://doi.org/10.1198/jbes.2009.07268.

recursive.detrend

Detrending the data recursively

## **Description**

This procedure is aimed to provide a recursively detrended series. More or less classical approach of full-sample detrending may lead to the regressors correlated with the error term.

```
recursive.detrend(y, x, c, gamma, trim)
```

38 reindex

# **Arguments**

У	The dependent (LHS) variable.
Х	The matrix of explanatory (RHS) variables.
С	A filtration parameter used to construct an autocorrelation coefficient.
gamma	A detrending type selection parameter. If 0 the OLS detrending is applied, if 1 the GLS detrending is applied, otherwise the autocorrelation coefficient is calculated as $1+c^{\gamma}T^{-\gamma}$ .
trim	The trimming parameter. It's used to find the minimum size of subsamples while calculating recursive estimates. The ending point of the subsample for the $t$ is $max(t, trim \times T)$ .

## **Details**

Elliott et al (1996) recommend using c=-7 for the model with only an intercept, and c=-13.5 for the model with a linear trend.

The function is not intended to be used directly so it's not exported.

#### Value

A detrended series.

# References

Elliott, Graham, Thomas J. Rothenberg, and James H. Stock. "Efficient Tests for an Autoregressive Unit Root." Econometrica 64, no. 4 (1996): 813–36. https://doi.org/10.2307/2171846.

Taylor, A. M. Robert. "Regression-Based Unit Root Tests With Recursive Mean Adjustment for Seasonal and Nonseasonal Time Series." Journal of Business & Economic Statistics 20, no. 2 (April 2002): 269–81. https://doi.org/10.1198/073500102317352001.

reindex A function that makes reindexing
--

# Description

The function is aimed to calculate the sequence of indices providing a new "time transformed" time series as in Cavaliere and Taylor (2008).

# Usage

reindex(u)

# Arguments

u The residuals series for reindexing.

# **Details**

The function is not intended to be used directly so it's not exported.

robust.tests.multiple 39

#### References

Cavaliere, Giuseppe, and A. M. Robert Taylor. "Time-Transformed Unit Root Tests for Models with Non-Stationary Volatility." Journal of Time Series Analysis 29, no. 2 (March 2008): 300–330. https://doi.org/10.1111/j.1467-9892.2007.00557.x.

Kurozumi, Eiji, Anton Skrobotov, and Alexey Tsarev. "Time-Transformed Test for Bubbles under Non-Stationary Volatility." Journal of Financial Econometrics, April 23, 2022. https://doi.org/10.1093/jjfinec/nbac004.

robust.tests.multiple *A wrapping function around KP and MDF.multiple*.

# **Description**

A wrapping function around KP and MDF.multiple.

## Usage

```
robust.tests.multiple(
   y,
   const = FALSE,
   season = FALSE,
   breaks = 2,
   trim = 0.15
)
```

# **Arguments**

y A series of interest.

const Whether the constant term should be included. season Whether the seasonal adjustment is needed.

breaks Number of breaks.

trim Trimming value for a possible break date bounds.

robust.tests.single *A wrapping function around MDF.single*.

# Description

A wrapping function around MDF.single.

```
robust.tests.single(
  y,
  const = FALSE,
  trend = FALSE,
  season = FALSE,
  trim = 0.15
)
```

40 SADF.bootstrap.test

# **Arguments**

у	A series of interest.
const	Whether the constant term should be included.
trend	Whether the trend term should be included.
season	Whether the seasonal adjustment is needed.
trim	Trimming value for a possible break date bounds.

SADF.bootstrap.test Supremum ADF tests with wild bootstrap.

# Description

SADF.bootstrap.test is a wild bootstrapping procedure for estimating critical and p-values for SADF test

 ${\tt GSADF.bootstrap.test}\ is\ the\ same\ procedure\ but\ for\ {\tt GSADF.test.}$ 

# Usage

```
SADF.bootstrap.test(
   y,
   trim = 0.01 + 1.8/sqrt(length(y)),
   const = TRUE,
   alpha = 0.05,
   iter = 999,
   seed = round(10^4 * sd(y))
)

GSADF.bootstrap.test(
   y,
   trim = 0.01 + 1.8/sqrt(length(y)),
   const = TRUE,
   alpha = 0.05,
   iter = 4 * 200,
   seed = round(10^4 * sd(y))
)
```

# **Arguments**

У	An input time series of interest.
trim	Trimming parameter to determine the lower and upper bounds.
const	Whether the constant needs to be included.
alpha	The significance level of interest.
iter	The number of iterations.
seed	The seed parameter for the random number generator.

SADF.test 41

#### Value

An object of type sadf. It's a list of:

- y,
- trim,
- const,
- alpha,
- iter,
- seed,
- vector of t-values,
- the value of the corresponding test statistic,
- series of bootstrapped test statistics,
- · bootstrapped critical values,
- p-value.

## References

Kurozumi, Eiji, Anton Skrobotov, and Alexey Tsarev. "Time-Transformed Test for Bubbles under Non-Stationary Volatility." Journal of Financial Econometrics, April 23, 2022. https://doi.org/10.1093/jjfinec/nbac004.

SADF.test

Supremum ADF tests

# **Description**

SADF. test is a test statistic equal to the minimum value of ADF.test for subsamples starting at t=1.

GSADF.test is a generalized version of SADF.test. Subsamples are allowed to start at any point between 1 and T(1-trim).

```
SADF.test(
   y,
   trim = 0.01 + 1.8/sqrt(length(y)),
   const = TRUE,
   add.p.value = TRUE
)

GSADF.test(
   y,
   trim = 0.01 + 1.8/sqrt(length(y)),
   const = TRUE,
   add.p.value = TRUE
)
```

42 sb.GSADF.test

# **Arguments**

y The input time series of interest.

trim Trimming parameter to determine the lower and upper bounds.

const Whether the constant needs to be included.

add.p.value Whether the p-value is to be returned. This argument is needed to suppress

the calculation of p-values during the precalculation of tables needed for the

p-values estimating.

## Value

An object of type sadf. It's a list of:

- y,
- trim,
- · const,
- vector of t-values,
- the value of the corresponding test statistic,
- p-value if it was asked for.

# References

Kurozumi, Eiji, Anton Skrobotov, and Alexey Tsarev. "Time-Transformed Test for Bubbles under Non-Stationary Volatility." Journal of Financial Econometrics, April 23, 2022. https://doi.org/10.1093/jjfinec/nbac004.

sb.GSADF.test

Sign-based SADF test

# **Description**

Sign-based SADF test

```
sb.GSADF.test(
   y,
   trim = 0.01 + 1.8/sqrt(length(y)),
   const = TRUE,
   alpha = 0.05,
   iter = 999,
   urs = TRUE,
   seed = round(10^4 * sd(y))
)
```

seasonal.dummies 43

# **Arguments**

У	A series of interest.
trim	Trimming parameter to determine the lower and upper bounds.
const	Whether the constant needs to be included.
alpha	Needed level of significance.
iter	Number of bootstrapping iterations.
urs	Use union of rejections strategy if TRUE.
seed	The seed parameter for the random number generator.

## References

Harvey, David I., Stephen J. Leybourne, and Yang Zu. "Sign-Based Unit Root Tests for Explosive Financial Bubbles in the Presence of Deterministically Time-Varying Volatility." Econometric Theory 36, no. 1 (February 2020): 122–69. https://doi.org/10.1017/S0266466619000057.

seasonal.dummies

Generating monthly seasonal dummy variables

# Description

Generating monthly seasonal dummy variables

# Usage

```
seasonal.dummies(N)
```

# **Arguments**

N number of observations.

# **Details**

The function is not intended to be used directly so it's not exported.

# Value

The matrix of values od seasonal dummies.

44 segments.GLS

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Procedure to minimize the GLS-SSR for 1 break point

# Description

Procedure to minimize the GLS-SSR for 1 break point

# Usage

```
segments.GLS(
  y,
  const = FALSE,
  trend = FALSE,
  breaks = 1,
  first.break = NULL,
  last.break = NULL,
  trim = 0.15
)
```

## **Arguments**

y Variable of interest.

const Whether there is a break in the constant.

trend Whether there is a break in the trend.

breaks Number of breaks.

first.break First possible break point.

last.break Last possible break point.

trim Trim value to calculate first.break and last.break if not provided.

## **Details**

The function is not intended to be used directly so it's not exported.

# Value

The point of possible break.

# References

Skrobotov, Anton. "On Trend Breaks and Initial Condition in Unit Root Testing." Journal of Time Series Econometrics 10, no. 1 (2018): 1–15. https://doi.org/10.1515/jtse-2016-0014.

segments.OLS 45

segments	ΩI	S
JUSTINCTIUS.	. 🔾	

Find m+1 optimal partitions

# Description

Find m+1 optimal partitions

## Usage

```
segments.OLS(y, x, m = 1, width = 2, SSR.data = NULL)
```

# Arguments

y (Tx1)-vector of the dependent variable.

x (Txk)-vector of the explanatory stochastic regressors.

m Number of breaks.

width Minimum spacing between the breaks.

SSR.data Optional matrix of recursive SSR's.

## **Details**

The function is not intended to be used directly so it's not exported.

# Value

A list of:

- · optimal SSR,
- the vector of break points.

## References

Bai, Jushan, and Pierre Perron. "Computation and Analysis of Multiple Structural Change Models." Journal of Applied Econometrics 18, no. 1 (2003): 1–22. https://doi.org/10.1002/jae.659.

segments.OLS.double

Procedure to minimize the SSR for 2 break points

# **Description**

Procedure to minimize the SSR for 2 break points

```
segments.OLS.double(y, model)
```

# **Arguments**

y (Tx1)-vector of time series

model A scalar equal to

- 1: for the AA (without trend) model,
- 2: for the AA (with trend) model,
- 3: for the BB model,
- 4: for the CC model,
- 5: for the AC-CA model.

# **Details**

The function is not intended to be used directly so it's not exported.

# Value

A list of

- resid: (Tx1) vector of estimated OLS residuals,
- tb1: The first break point,
- tb2: The second break point.

## References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "The KPSS Test with Two Structural Breaks." Spanish Economic Review 9, no. 2 (May 16, 2007): 105–27. https://doi.org/10.1007/s10108-006-9017-8.

segments.OLS.single

Procedure to minimize the SSR for 1 break point

# Description

Procedure to minimize the SSR for 1 break point

# Usage

```
segments.OLS.single(beg, end, first.break, last.break, len, SSR.data)
```

# Arguments

beg Sample begin. end Sample end.

first.break First possible break point.
last.break Last possible break point.
len Total number of observations.

SSR. data The matrix of recursive SSR values.

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## **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of:

- SSR: Optimal SSR value,
- break.point: The point of possible break.

# References

Carrion-i-Silvestre, Josep Lluís, and Andreu Sansó. "Testing the Null of Cointegration with Structural Breaks." Oxford Bulletin of Economics and Statistics 68, no. 5 (October 2006): 623–46. https://doi.org/10.1111/j.1468-0084.2006.00180.x.

SSR.matrix

Pre-calculate matrix of recursive SSR values.

# Description

Pre-calculate matrix of recursive SSR values.

# Usage

```
SSR.matrix(y, x, width = 2)
```

# **Arguments**

y Dependent variable.

x Explanatory variables.

width Minimum spacing between the breaks.

## **Details**

The function is not intended to be used directly so it's not exported.

# Value

The matrix of recursive SSR values.

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CCD	recu	ırci	

Calculate SSR recursively

# **Description**

Calculate SSR recursively

# Usage

```
SSR.recursive(y, x, beg, end, width = 2)
```

# **Arguments**

y (Tx1)-vector of the dependent variable.

x (Txk)-vector of the explanatory stochastic regressors.

beg, end The start and the end of SSR calculating period.

width Minimum spacing between the breaks.

# **Details**

The function is not intended to be used directly so it's not exported.

# Value

The vector of calculated recursive SSR.

# References

Brown, R. L., J. Durbin, and J. M. Evans. "Techniques for Testing the Constancy of Regression Relationships over Time." Journal of the Royal Statistical Society. Series B (Methodological) 37, no. 2 (1975): 149–92.

STADF.test

Supremum ADF tests with time transformation

# Description

See SADF.test. Tests with time transformation are the modified versions of the ordinary SADF and GSADF tests using Nadaraya-Watson residuals and reindexing procedure by Cavaliere-Taylor (2008).

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# Usage

```
STADF.test(
  у,
  trim = 0.01 + 1.8/sqrt(length(y)),
  const = FALSE,
  omega.est = TRUE,
  truncated = TRUE,
  is.reindex = TRUE,
  ksi.input = "auto",
  hc = 1,
  pc = 1,
  add.p.value = TRUE
GSTADF.test(
  у,
  trim = 0.01 + 1.8/sqrt(length(y)),
  const = FALSE,
  omega.est = TRUE,
  truncated = TRUE,
  is.reindex = TRUE,
  ksi.input = "auto",
  hc = 1,
  pc = 1,
  add.p.value = TRUE
```

# **Arguments**

у	An input time series of interest.
trim	Trimming parameter to determine the lower and upper bounds.
const	Whether the constant needs to be included.
omega.est	Whether the variance of Nadaraya-Watson residuals should be used.
truncated	Whether the truncation of Nadaraya-Watson residuals is needed.
is.reindex	Whether the Cavaliere and Taylor (2008) time transformation is needed.
ksi.input	The value of the truncation parameter. Can be either auto or the explicit numerical value. In the former case the numeric value is estimated.
hc	The scaling parameter for Nadaraya-Watson bandwidth.
рс	The scaling parameter for the estimated truncation parameter value.
add.p.value	Whether the p-value is to be returned. This argument is needed to suppress the calculation of p-values during the precalculation of tables needed for the p-values estimating.

# Value

An object of type sadf. It's a list of:

- y,
- N: Number of observations,

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- trim,
- const,
- omega.est,
- truncated,
- is.reindex,
- new.index: the vector of new indices,
- ksi.input,
- hc,
- · h.est,
- u.hat,
- pc,
- w.sq,
- t.values: vector of t-values,
- the value of the corresponding test statistic,
- u.hat.truncated: truncated residuals if truncation was asked for,
- ksi, sigma: estimated values of the truncation parameter and resulting s.e. if ksi.input equals auto,
- eta.hat: the values of reindexing function if reindexing was asked for,
- p-value if it was asked for.

## References

Cavaliere, Giuseppe, and A. M. Robert Taylor. "Time-Transformed Unit Root Tests for Models with Non-Stationary Volatility." Journal of Time Series Analysis 29, no. 2 (March 2008): 300–330. https://doi.org/10.1111/j.1467-9892.2007.00557.x.

Kurozumi, Eiji, Anton Skrobotov, and Alexey Tsarev. "Time-Transformed Test for Bubbles under Non-Stationary Volatility." Journal of Financial Econometrics, April 23, 2022. https://doi.org/10.1093/jjfinec/nbac004.

supBZ.statistic

Calculate supBZ statistic

## **Description**

Calculate supBZ statistic

```
supBZ.statistic(
  y,
  trim = 0.01 + 1.8/sqrt(length(y)),
  sigma.sq = NULL,
  generalized = FALSE
)
```

supSBADF.statistic 51

# **Arguments**

y The series of interest.

trim Trimming parameter to determine the lower and upper bounds.

sigma.sq Local non-parametric estimates of variance. If NULL they will be estimated via

Nadaraya-Watson procedure.

generalized Whether to calculate generalized statistic value.

## **Details**

The function is not intended to be used directly so it's not exported.

#### Value

A list of:

- y,
- trim,
- sigma.sq,
- BZ. values: a series of BZ-statistic,
- supBZ.value: the maximum of supBZ.values,
- h.est: the estimated value of bandwidth if sigma.sq is NULL.

## References

Harvey, David I., Stephen J. Leybourne, and Yang Zu. "Testing Explosive Bubbles with Time-Varying Volatility." Econometric Reviews 38, no. 10 (November 26, 2019): 1131–51. https://doi.org/10.1080/07474938.

supSBADF.statistic

Calculate superior sign-based SADF statistic.

# Description

Calculate superior sign-based SADF statistic.

# Usage

```
supSBADF.statistic(y, trim = 0.01 + 1.8/sqrt(length(y)), generalized = FALSE)
```

# **Arguments**

y The series of interest.

trim Trimming parameter to determine the lower and upper bounds.

generalized Whether to calculate generalized statistic value.

# Details

The function is not intended to be used directly so it's not exported.

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#### Value

A list of

- y,
- trim,
- C. t: the cumulative sum of "signs" (1 or -1) of the first difference of y,
- SBADF. values: series of sign-based ADF statistics,
- supSBADF.value: the maximum of SBADF.values.

## References

Harvey, David I., Stephen J. Leybourne, and Yang Zu. "Sign-Based Unit Root Tests for Explosive Financial Bubbles in the Presence of Deterministically Time-Varying Volatility." Econometric Theory 36, no. 1 (February 2020): 122–69. https://doi.org/10.1017/S0266466619000057.

weighted.SADF.test

Weighted supremum ADF test

# **Description**

Weighted supremum ADF test

# Usage

```
weighted.SADF.test(
  у,
  trim = 0.01 + 1.8/sqrt(length(y)),
  const = TRUE,
  alpha = 0.05,
  iter = 4 * 200,
  urs = TRUE,
  seed = round(10^4 * sd(y))
)
weighted.GSADF.test(
  trim = 0.01 + 1.8/sqrt(length(y)),
  const = TRUE,
  alpha = 0.05,
  iter = 4 * 200,
  urs = TRUE,
  seed = round(10^4 * sd(y))
)
```

## **Arguments**

y The input time series of interest.

trim Trimming parameter to determine the lower and upper bounds.

const Whether the constant needs to be included.

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alpha	The significance level of interest.
iter	The number of iterations.
urs	Use union of rejections strategy.
seed	The seed parameter for the random number generator.

## Value

An object of type sadf. It's a list of:

- y,
- trim,
- const,
- alpha,
- iter,
- urs,
- · seed.
- sigma.sq: the estimated variance,
- BZ. values: a series of BZ-statistic,
- supBZ.value: the maximum of supBZ.values,
- supBZ.bootstsrap.values: bootstrapped supremum BZ values,
- supBZ.cr.value: supremum BZ  $\alpha$  critical value,
- p.value,
- is.explosive: 1 if supBZ.value is greater than supBZ.cr.value.

if urs is TRUE the following items are also included:

- vector of t-values,
- the value of the SADF test statistic,
- $\bullet \ \ \mathsf{SADF}. bootstrap. values: bootstrapped \ \mathsf{SADF} \ values,$
- U. value: union test statistic value,
- U.bootstrap.values: bootstrapped series of U.value,
- U.cr.value: critical value of U.value.

#### References

Harvey, David I., Stephen J. Leybourne, and Yang Zu. "Testing Explosive Bubbles with Time-Varying Volatility." Econometric Reviews 38, no. 10 (November 26, 2019): 1131–51. https://doi.org/10.1080/07474938.2

Kurozumi, Eiji, Anton Skrobotov, and Alexey Tsarev. "Time-Transformed Test for Bubbles under Non-Stationary Volatility." Journal of Financial Econometrics, April 23, 2022. https://doi.org/10.1093/jjfinec/nbac004.

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