Package 'slm'

October 24, 2019

Type Package

Title Stationary Linear Models

Version 1.1.0
Author Emmanuel Caron, Jérôme Dedecker, Bertrand Michel
Maintainer Emmanuel Caron <emmanuelcaron3@gmail.com></emmanuelcaron3@gmail.com>
Description Provides statistical procedures for linear regression in the general context where the errors are assumed to be correlated. Different ways to estimate the asymptotic covariance matrix of the least squares estimators are available. Starting from this estimation of the covariance matrix, the confidence intervals and the usual tests on the parameters are modified. The functions of this package are very similar to those of 'lm': it contains methods such as summary(), plot(), confint() and predict(). The 'slm' package is described in the paper by E. Caron, J. Dedecker and B. Michel (2019), "Linear regression with stationary errors: the R package slm", arXiv preprint <arxiv:1906.06583>.</arxiv:1906.06583>
License GPL-3
Encoding UTF-8
LazyData true
RoxygenNote 6.1.1
Depends R (>= 2.10)
Collate 'slm-main.R' 'slm.R' 'generative.R' 'auxiliary-fun.R' 'slm-method.R' 'data.R'
Imports ltsa, methods, stats, capushe, sandwich, expm
NeedsCompilation no
Repository CRAN
Date/Publication 2019-10-24 14:10:02 UTC
R topics documented:
slm-package 2 confint.slm 3 cov_AR 4 cov_efromovich 5

2 slm-package

cov_kernel	6
cov_matrix_estimator	7
cov_method	8
cov_select	9
cov_spectralproj	10
generative_model	11
generative_process	12
plot.slm	13
predict.slm	13
Rboot	15
rectangle	16
shan	16
slm	17
slm-class	19
summary.slm	20
trapeze	22
triangle	22
vcov.slm	23
	2 4

slm-package

slm: A package for stationary linear models

Description

Index

The slm package enables to fit linear models on datasets considering the dependence between the observations. Most of the functions are based on the functions and methods of lm, with the same arguments and the same format for the outputs.

slm function, in "slm-main.R"

The slm function is the main function of this package. Its architecture is the same as the lm function but it takes into account the possible correlation between the observations. To estimate the asymptotic covariance matrix of the least squares estimator, several approaches are available: "fitAR" calls the cov_AR function, "spectralproj" the cov_spectralproj function, "kernel" the cov_kernel function, "efromovich" the cov_efromovich function and "select" the cov_select function. The "hac" method uses the sandwich package, and more precisely, the method described by Andrews (1991) and Zeileis (2004).

Methods for slm, in "slm-method.R"

The slm function has several associated methods, which are the same as for the lm function. The available methods are: summary, confint, predict, plot and vcov.

confint.slm 3

Others functions, in "auxiliary-fun.R"

The package has some auxiliary functions, in particular some predefined kernels for the kernel method of slm function: the trapeze kernel, the triangle kernel and the rectangular kernel. The user can also define his own kernel and put it in the argument kernel_fonc in the slm function.

Generative functions, in "generative.R"

The generative_process function generates some stationary processes. The generative_model function generates some designs.

Data

The package contains a dataset "shan". This dataset comes from a study about fine particle pollution in the city of Shanghai. The data are available on the following website https://archive.ics.uci.edu/ml/datasets/PM2.5+Data+of+Five+Chinese+Cities#.

References

D. Andrews (1991). Heteroskedasticity and autocorrelation consistent covariant matrix estimation. *Econometrica*, 59(3), 817-858.

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint arXiv:1906.06583. https://arxiv.org/abs/1906.06583.

A. Zeileis (2004). Econometric computing with HC and HAC covariance matrix estimators.

confint.slm

Confidence intervals for the Model Parameters

Description

Computes confidence intervals for the model parameters.

Usage

```
## S3 method for class 'slm'
confint(object, parm = NULL, level = 0.95, ...)
```

Arguments

object a fitted model object of class slm.

parm a specification of which parameters are to be given confidence intervals, that is

a vector of numbers. If missing, all parameters are considered.

level the confidence level required.

... additional argument(s) for methods.

Value

This function returns the confidence intervals for the parameters of the model.

4 cov_AR

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint *arXiv*:1906.06583. https://arxiv.org/abs/1906.06583.

See Also

```
confint.lm.
```

Examples

```
\label{eq:data("shan")} $$ reg1 = slm(shan\$PM_Xuhui ~ . , data = shan, method_cov_st = "fitAR", model_selec = -1) $$ confint(reg1, level = 0.8) $$ $$ data("co2") $$ y = as.vector(co2) $$ x = as.vector(time(co2)) - 1958 $$ reg2 = slm(y ~ x + I(x^2) + I(x^3) + sin(2*pi*x) + cos(2*pi*x) + sin(4*pi*x) + cos(4*pi*x) + sin(6*pi*x) + cos(6*pi*x) + sin(8*pi*x) + cos(8*pi*x), $$ method_cov_st = "fitAR", model_selec = -1, plot = TRUE) $$ confint(reg2, level = 0.9) $$
```

cov_AR

Covariance estimation by AR fitting

Description

Fit an autoregressive model to the process and compute the theoretical autocovariances of the fitted AR process. By default, the order is chosen by using the AIC criterion (model_selec = -1).

Usage

```
cov_AR(epsilon, model_selec = -1, plot = FALSE)
```

Arguments

epsilon an univariate process.

model_selec the order of the method. If model_selec = -1, it is chosen automatically by

using the AIC criterion.

plot logical. By default, plot = FALSE. If plot = TRUE, then the ACF and the PACF

of the vector epsilon is plotted.

Value

The function returns the vector of the theoretical autocovariances of the AR process fitted on the process epsilon.

model_selec the order selected.

cov_st the vector of theoretical autocovariances of the fitted AR process.

cov_efromovich 5

References

P.J. Brockwell and R.A. Davis (1991). Time Series: Theory and Methods. *Springer Science & Business Media*.

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint arXiv:1906.06583. https://arxiv.org/abs/1906.06583.

Examples

```
x = arima.sim(list(ar=c(0.4,0.2)),1000)
cov_AR(x, model_selec = 2, plot = TRUE)
```

cov_efromovich

Spectral density estimation: Efromovich method

Description

This method estimates the spectral density and the autocovariances of the error process via a lagwindow estimator based on the rectangular kernel (see P.J. Brockwell and R.A. Davis (1991). Time Series: Theory and Methods. *Springer Science & Business Media*, page 330). The lag is computed according to Efromovich's algorithm (Efromovich (1998)).

Usage

```
cov_efromovich(epsilon, plot = FALSE)
```

Arguments

epsilon an univariate process.

plot logical. By default, plot = FALSE. If plot = TRUE, the ACF of the process

epsilon is plotted.

Value

The function returns the estimated autocovariances of the process, that is the Fourier coefficients of the spectral density estimates, and the order chosen by the algorithm.

model_selec the number of selected autocovariance terms.

cov_st the estimated autocovariances.

References

- P.J. Brockwell and R.A. Davis (1991). Time Series: Theory and Methods. *Springer Science & Business Media*.
- E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint *arXiv*:1906.06583. https://arxiv.org/abs/1906.06583.
- S. Efromovich (1998). Data-driven efficient estimation of the spectral density. *Journal of the American Statistical Association*, 93(442), 762-769.

6 cov_kernel

Examples

```
x = arima.sim(list(ar=c(0.4,0.2)),1000)
cov_efromovich(x)
```

cov_kernel

Kernel estimation: bootstrap method

Description

This method estimates the spectral density and the autocovariances of the error process via a lagwindow (or kernel) estimator (see P.J. Brockwell and R.A. Davis (1991). Time Series: Theory and Methods. *Springer Science & Business Media*, page 330). The weights are computed according to a kernel K and a bandwidth h (or a lag), to be chosen by the user. The lag can be computed automatically by using a bootstrap technique (as in Wu and Pourahmadi (2009)), via the Rboot function.

Usage

```
cov_kernel(epsilon, model_selec = -1,
  model_max = min(50,length(epsilon)/2), kernel_fonc = triangle,
  block_size = length(epsilon)/2, block_n = 100, plot = FALSE)
```

Arguments

an univariate process. epsilon model_selec the order of the method. If model_selec = -1, the method chooses the treshold automatically. If model_selec = k, then only k autocovariance terms are kept and smoothed by the kernel. model_max the maximal order. kernel_fonc define the kernel to use in the method. The user can give his own kernel function. size of the bootstrap blocks. block_size must be greater than model_max. block_size blocks number to use for the bootstrap. block_n plot logical. By default, plot = FALSE. If plot = TRUE, the risk curve is returned and the ACF of the process.

Value

The method returns the tapered autocovariance vector with model_selec autocovariance terms.

model_selec the number of selected autocovariance terms.

cov_st the estimated autocovariances.

cov_matrix_estimator 7

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint *arXiv*:1906.06583. https://arxiv.org/abs/1906.06583.

W.B. Wu, M. Pourahmadi (2009). Banding sample autocovariance matrices of stationary processes. *Statistica Sinica*, pp. 1755–1768.

Examples

```
x = arima.sim(list(ar=c(0.7)),1000)

cov_kernel(x, model_selec = -1, block_n = 10, plot = TRUE)
```

cov_matrix_estimator Covariance matrix estimator for slm object

Description

This function gives the estimation of the asymptotic covariance matrix of the normalized least squares estimator in the case of the linear regression model with strictly stationary errors.

Usage

```
cov_matrix_estimator(object)
```

Arguments

object

an object of class slm.

Details

The function computes the covariance matrix estimator of the normalized least squares estimator from the vector cov_st of a slm object. If the user has given the argument Cov_ST in the slm object, then it is used to compute the final covariance matrix. If the method used is the "hac" method, then the final covariance matrix is computed via the kernHAC function of the sandwich package, by using the Quadratic Spectral kernel and the bandwidth described in Andrews (1991). For the methods "efromovich", "kernel" and "select", the covariance matrix estimator may not be positive definite. Then we apply the "Positive definite projection" algorithm, which consists in replacing all eigenvalues lower or equal to zero with the smallest positive eigenvalue of the covariance matrix.

Value

This function returns the estimation of the asymptotic covariance matrix of the normalized least squares estimator.

8 cov_method

References

D. Andrews (1991). Heteroskedasticity and autocorrelation consistent covariant matrix estimation. *Econometrica*, 59(3), 817-858.

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

A. Zeileis (2004). Econometric computing with HC and HAC covariance matrix estimators.

See Also

The R package sandwich. kernHAC for HAC methods.

cov_method

Methods to estimate the autocovariances of a process

Description

This function gives the estimation of the autocovariances of the error process, with the method chosen by the user. Five methods are available: "fitAR", "spectralproj", "efromovich", "kernel" and "select".

Usage

```
cov_method(epsilon, method_cov_st = "fitAR", model_selec = -1,
  model_max = NULL, kernel_fonc = NULL, block_size = NULL,
  block_n = NULL, plot = FALSE)
```

Arguments

epsilon an univariate process. method_cov_st the method chosen by the user. model_selec the order of the method. If $model_selec = -1$, the method works automatically. model_max maximal dimension of the method. kernel_fonc to use if method_cov_st = kernel. Define the kernel to use in the method. The user can give his own kernel function. block_size size of the bootstrap blocks if method_cov_st = kernel. block_size must be greater than model_max. block_n blocks number to use for the bootstrap if method_cov_st = kernel. plot logical. By default, plot = FALSE.

Value

The function returns the autocovariances computed with the chosen method.

cov_select 9

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

Examples

```
x = arima.sim(list(ar=c(0.4,0.2)),1000)
cov_method(x, method_cov_st = "fitAR", model_selec = -1)
```

cov_select

Covariances Selection

Description

Allows the user to select the lags of the autocovariance terms of the process to be kept.

Usage

```
cov_select(epsilon, model_selec, plot = FALSE)
```

Arguments

epsilon an univariate process.

model_selec a vector with the positive lags of the selected autocovariance terms. The variance

(lag = 0) is automatically selected.

plot logical. By default, plot = FALSE. If plot = TRUE the ACF of the process is

plotted.

Details

In the framework of slm, this is a manual method for estimating the covariance matrix of the error process by only selecting some autocovariance terms from the residual autocovariances.

Value

This function returns the estimated autocovariance terms.

model_selec the vector with the positive lag of the selected autocovariance terms.

cov_st the vector of the selected autocovariances.

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

```
x = arima.sim(list(ar=c(0.2,0.1,0.25)),1000)

cov_select(x, c(1,3,5))
```

10 cov_spectralproj

cov_spectralproj	Data-driven spectra	ıl density estimation
cov_specti dipi oj	Daia arriver specifa	a acrisity estimation

Description

Computes a data-driven histogram estimator of the spectral density of a process and compute its Fourier coefficients, that is the associated autocovariances. For a dimension d, the estimator of the spectral density is an histogram on a regular basis of size d. Then we use a penalized criterion in order to choose the dimension which balance the bias and the variance, as proposed in Comte (2001). The penalty is of the form c * d/n, where c is the constant and n the sample size. The dimension and the constant of the penalty are chosen with the slope heuristic method, with the dimension jump algorithm (from package "capushe").

Usage

```
cov_spectralproj(epsilon, model_selec = -1,
  model_max = min(100,length(epsilon)/2), plot = FALSE)
```

Arguments

epsilon an univariate process.

model_selec the dimension of the method. If model_selec = -1, the method works automat-

ically and take a dimension between 1 and model_max.

model_max the maximal dimension. By default, it is equal to the minimum between 100 and

the length of the process divided by 2.

plot logical. By default, plot = FALSE. If plot = TRUE, plot the spectral density esti-

mator of the process.

Value

The function returns the estimated autocovariances of the process, that is the Fourier coefficients of the spectral density estimates, and the dimension chosen by the algorithm.

```
model_selec the dimension selected.

cov_st the estimated autocovariances.
```

References

J.P. Baudry, C. Maugis B. and Michel (2012). Slope heuristics: overview and implementation. *Statistics and Computing*, 22(2), 455–470.

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

F. Comte (2001). Adaptive estimation of the spectrum of a stationary Gaussian sequence. *Bernoulli*, 7(2), 267-298.

generative_model 11

See Also

```
The R package capushe.
```

Slope heuristic algorithm DDSE.

Dimension jump algorithm Djump.

Examples

```
x = arima.sim(list(ar=c(0.2), ma=c(0.3,0.05)), n=100)

cov\_spectralproj(x, model\_selec = -1)
```

generative_model

Some linear model

Description

This function returns a design for the regression linear model, without the intercept. The user can choose one of the two models: "mod1" or "mod2". The first model "mod1" contains just one column, equal to $i^2 + X_i$, i = 1, ..., n, where X is an AR(1) process with phi_1 = 0.5.

The second model "mod2" contains two columns, the first equal to $log(i) + sin(i) + X_i$ and the second equal to i, for i = 1, ..., n. The process X is again an AR(1) process with phi_1 = 0.5. More information about "mod2" is available in the paper of E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm.

Usage

```
generative_model(n, model = "mod1")
```

Arguments

n samples size.

model a list of character to choose the model.

Value

This function returns a data-frame which contains a simulated random design.

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

```
generative_model(500,"mod1")
```

12 generative_process

generative_process Some stationary processes

Description

This is a generative function. The user chooses one of the process: "iid", "AR1", "AR12", "MA12", "Nonmixing", "sysdyn", and it generates the chosen process. These processes are fully described in the paper of E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

Usage

```
generative_process(n, process = "AR1", phi = "numeric",
    theta = "numeric")
```

Arguments

n sample size.

process a list of character to choose the process.

phi a numeric vector with AR parameters if the process is "AR1" or "AR12".

theta a numeric vector with MA parameters if the process is "MA12".

Value

This function returns a vector of observations drawn according to the selected process.

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint arXiv:1906.06583. https://arxiv.org/abs/1906.06583.

```
generative_process(200, "Nonmixing")
```

plot.slm 13

plot.slm

Plot.slm

Description

Same function as the plot.lm function.

Usage

```
## S3 method for class 'slm' plot(x, ...)
```

Arguments

x slm object.

... other parameters to be passed through to plotting functions.

Value

This function returns the graphics of plot. lm(x).

Examples

```
data("shan")
reg = slm(shan$PM_Xuhui ~ . , data = shan, method_cov_st = "fitAR", model_selec = -1)
plot(reg)
```

predict.slm

Predict for slm object

Description

Predicted values based on slm object.

Usage

```
## S3 method for class 'slm'
predict(object, newdata = NULL, interval = "confidence",
  level = 0.95, ...)
```

14 predict.slm

Arguments

object an object of class slm.

newdata an optional data frame in which to look for variables with which to predict.

newdata must contain only variables and not the intercept. If omitted, the fitted

values are used.

interval type of interval calculation. It can be only interval = "confidence", the de-

fault value. It computes the confidence intervals for x'beta, where x' is a new

observation of the design.

level tolerance/confidence level.

... further arguments passed to or from other methods.

Details

This function produces predicted values, obtained by evaluating the regression function in the frame newdata (which defaults to model.frame(object)). If newdata is omitted the predictions are based on the data used for the fit.

Value

This function produces a vector of predictions or a matrix of predictions and bounds with column names fit, lwr, and upr if interval is set.

See Also

```
predict.lm.
```

```
data("shan")
reg1 = slm(shan$PM_Xuhui ~ . , data = shan, method_cov_st = "fitAR", model_selec = -1)
predict(reg1)

data("co2")
y = as.vector(co2)
x = as.vector(time(co2)) - 1958
reg2 = slm(y ~ x + I(x^2) + I(x^3) + sin(2*pi*x) + cos(2*pi*x) + sin(4*pi*x) +
cos(4*pi*x) + sin(6*pi*x) + cos(6*pi*x) + sin(8*pi*x) + cos(8*pi*x),
method_cov_st = "fitAR", model_selec = -1)
predict(reg2)
```

Rboot 15

Rboot	Risk estimation for a tapered covariance matrix estimator via boot- strap method

Description

This function computes an estimation of the risk for the tapered covariance matrix estimator of a process via a bootstrap method, for a specified treshold and a specified kernel.

Usage

```
Rboot(epsilon, treshold, block_size, block_n, model_max, kernel_fonc)
```

Arguments

epsilon	an univariate process.
treshold	number of estimated autocovariance terms that we consider for the estimation of the covariance matrix.
block_size	the size of the bootstrap blocks. block_size must be greater than model_max.
block_n	blocks number used for the bootstrap.
model_max	the maximal dimension, that is the maximal number of terms available to estimate the covariance matrix.
kernel_fonc	the kernel to use. The user can define his own kernel and put it in the argument.

Value

This function returns a list with:

risk for one treshold, the value of the estimated risk.

SE the standard-error due to the bootstrap.

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

16 shan

rectangle

Rectangular kernel

Description

Rectangular kernel

Usage

```
rectangle(x)
```

Arguments

Х

a vector of real numbers.

Value

This function computes the values of the rectangular kernel at points x.

Examples

```
x = seq(-2,2,length=1000)
y = rectangle(x)
plot(x,y)
```

shan

PM2.5 Data of Shanghai

Description

This dataset comes from a study about fine particle pollution in five Chinese cities. The data are available on the following website https://archive.ics.uci.edu/ml/datasets/PM2.5+Data+of+Five+Chinese+Cities#. The present dataset concerns the city of Shanghai. From the initial dataset, we have removed the lines that contain NA observations and we then extract the first 5000 observations. Then we consider only pollution variables and weather variables.

Usage

```
data("shan")
```

17 slm

Format

```
A data frame with 5000 rows and 10 variables:
```

```
PM_Xuhui PM2.5 concentration in the Xuhui district (ug/m3).
```

PM_Jingan PM2.5 concentration in the Jing'an district (ug/m3).

PM_US.Post PM2.5 concentration in the U.S diplomatic post (ug/m3).

DEWP Dew Point (CelsiusDegree).

TEMP Temperature (CelsiusDegree).

HUMI Humidity (%).

PRES Pressure (hPa).

Iws Cumulated wind speed (m/s).

precipitation hourly precipitation (mm).

Iprec Cumulated precipitation (mm).

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

X. Liang, S. Li, S. Zhang, H. Huang, S.X. Chen (2016). PM2.5 data reliability, consistency, and air quality assessment in five Chinese cities. Journal of Geophysical Research: Atmospheres, 121(17), 10-220.

slm

Fitting Stationary Linear Models

Description

slm is used to fit linear models when the error process is assumed to be strictly stationary.

Usage

```
slm(myformula, data = NULL, model = TRUE, x = FALSE, y = FALSE,
 qr = TRUE, method_cov_st = "fitAR", cov_st = NULL, Cov_ST = NULL,
 model_selec = -1, model_max = 50, kernel_fonc = NULL,
 block_size = NULL, block_n = NULL, plot = FALSE)
```

Arguments

myformula

an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.

data

an optional data frame, list or environment (or object coercible by as. data. frame to a data frame) containing the variables in the model. If not found in data, the variables are taken from environment (formula), typically the environment

from which slm is called.

18 slm

model, x, y, qr	logicals. If TRUE the corresponding components of the fit (the model frame, the model matrix, the response, the QR decomposition) are returned.
method_cov_st	the method chosen by the user to estimate the autocovariances of the error process. The user has the choice between the methods "fitAR", "spectralproj", "efromovich", "kernel", "select" or "hac". By default, the "fitAR" method is used.
cov_st	the estimated autocovariances of the error process. The user can give his own vector.
Cov_ST	an argument given by the user if he wants to use his own covariance matrix estimator.
model_selec	the order of the method. If $model_selec = -1$, the method works automatically.
model_max	maximal order of the method.
kernel_fonc	to use if method_cov_st = kernel. Define the kernel to use in the method. The user can give his own kernel function.
block_size	size of the bootstrap blocks if method_cov_st = kernel. block_size must be greater than model_max.
block_n	blocks number to use for the bootstrap if method_cov_st = kernel.
plot	logical. By default, plot = FALSE.

Details

The slm function is based on the architecture of the lm function. Models for slm are specified symbolically. A typical model has the form response ~ terms where response is the (numeric) response vector and terms is a series of terms which specifies a linear predictor for response. See the documentation of lm for more details.

Value

slm returns an object of class "slm". The function summary is used to obtain and print a summary of the results. The generic accessor functions coefficients, effects, fitted.values and residuals extract various useful features of the value returned by slm. An object of class "slm" is a list containing at least the following components:

method_cov_st	print the method chosen.
cov_st	the estimated autocovariances of the error process. NA if "hac" is used.
Cov_ST	if given by the user, the estimated covariance matrix of the error process. NA if "hac" is used.
model_selec	the order of the method.
norm_matrix	the normalization matrix of the least squares estimator.
design_qr	the matrix $(X^tX)^{-1}$.
coefficients	a named vector of the estimated coefficients.
residuals	the residuals, that is response minus fitted values.
fitted.values	the fitted values.
rank	the numeric rank of the fitted linear model.

slm-class 19

df.residual	the number of observations minus the number of variables.
call	the matched call.
terms	the terms object used.
xlevels	(only where relevant) a record of the levels of the factors used in fitting.
У	if requested, the response used.
Х	if requested, the model matrix used.
model	if requested (the default), the model frame used.

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

See Also

summary for summaries.

The generic functions coef, effects, residuals, fitted, vcov.

predict for prediction, including confidence intervals for x'beta, where x' is a new observation of the design.

confint for confidence intervals of parameters.

Examples

slm-class

slm class

Description

An S4 class to create an s1m object.

20 summary.slm

Slots

method_cov_st the method used to compute the autocovariance vector of the error process.

Cov_ST the estimated covariance matrix of the error process, computed from the method_cov_st method.

model_selec the order of the chosen method. If model_selec = -1, the method works automatically.

 $norm_matrix$ the normalization matrix of the design X.

```
design_qr the matrix (X^tX)^{-1}.
```

References

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv preprint arXiv:1906.06583*. https://arxiv.org/abs/1906.06583.

summary.slm

Summarizing Stationary Linear Model Fits

Description

Summary method for class "slm".

Usage

```
## S3 method for class 'slm'
summary(object, correlation = FALSE,
    symbolic.cor = FALSE, ...)
```

Arguments

object	an object of class "slm", usually, a result of a call to slm.
correlation	logical; if TRUE, the correlation matrix of the estimated parameters is returned and printed.
symbolic.cor	logical. If TRUE, print the correlations in a symbolic form (see symnum) rather than as numbers.

. . . further arguments passed to or from other methods.

summary.slm 21

Value

The function summary.slm computes and returns a list of summary statistics of the fitted linear model given in object, using the components (list elements) "call" and "terms" from its argument, plus:

the residuals, that is response minus fitted values. residuals coefficients a p * 4 matrix with columns for the estimated coefficient, its standard error, z-statistic and corresponding (two-sided) p-value. Aliased coefficients are omitted. aliased named logical vector showing if the original coefficients are aliased. sigma the square root of the estimated variance of the error process. df degrees of freedom, a 3-vector (p, n-p, p*), the first being the number of nonaliased coefficients, the last being the total number of coefficients. chi2statistic a 2-vector with the value of the chi2-statistic with its degree of freedom. R^2 , the 'fraction of variance explained by the model'. r.squared cov.unscaled the matrix $(X^t X)^{-1}$. correlation the correlation matrix corresponding to the above cov.unscaled, if correlation = TRUE is specified.

References

symbolic.cor

E. Caron, J. Dedecker and B. Michel (2019). Linear regression with stationary errors: the R package slm. *arXiv* preprint *arXiv*:1906.06583. https://arxiv.org/abs/1906.06583.

(only if correlation is true.) The value of the argument symbolic.cor.

See Also

The model fitting function slm, summary.

The function coef extracts the matrix of coefficients with standard errors, z-statistics and p-values.

```
data("shan")  
reg1 = slm(shan$PM_Xuhui ~ . , data = shan, method_cov_st = "fitAR", model_selec = -1)  
summary(reg1)  
data("co2")  
y = as.vector(co2)  
x = as.vector(time(co2)) - 1958  
reg2 = slm(y ~ x + I(x^2) + I(x^3) + sin(2*pi*x) + cos(2*pi*x) + sin(4*pi*x) + cos(4*pi*x) + sin(6*pi*x) + cos(6*pi*x) + sin(8*pi*x) + cos(8*pi*x),  
method_cov_st = "fitAR", model_selec = -1)  
summary(reg2)
```

22 triangle

trapeze

Trapeze kernel

Description

Trapeze kernel

Usage

```
trapeze(x, width = 0.8)
```

Arguments

x a vector of real numbers. width a number between 0 and 1.

Value

This function computes the values of the trapeze kernel at points x.

Examples

```
x = seq(-2,2,length=1000)
y = trapeze(x, width=0.5)
plot(x,y)
```

triangle

Kernel triangle

Description

Kernel triangle

Usage

```
triangle(x)
```

Arguments

Χ

a vector of real numbers.

Value

This function computes the values of the triangle kernel at points x.

vcov.slm 23

Examples

```
x = seq(-2,2,length=1000)
y = triangle(x)
plot(x,y)
```

vcov.slm

Calculate Variance-Covariance Matrix for a Fitted Model Object of class slm

Description

Returns the variance-covariance matrix of the (non-normalized) least squares estimators for an object of class slm.

Usage

```
## S3 method for class 'slm'
vcov(object, ...)
```

Arguments

object a fitted model object of class slm.

... additional arguments for method functions.

Value

The variance-covariance matrix of the (non-normalized) least squares estimators for an object of class slm.

See Also

The generic function vcov.

The function cov_matrix_estimator.

```
n = 500
eps = generative_process(n,"AR1",c(0.7))
X = as.matrix(generative_model(n,"mod2"))
Y = 3 + 2*X[,2] + eps
reg = slm(Y ~ X, method_cov_st = "fitAR", model_selec = -1)
vcov(reg)
```

Index

*Topic datasets shan, 16	rectangle, 16 residuals, <i>19</i>
as.data.frame, 17	sandwich, 8 shan, 16
<pre>capushe, 10, 11 class, 18 coef, 19, 21 confint, 19 confint.lm, 4 confint.slm, 3 cov_AR, 4 cov_efromovich, 5</pre>	slm, 10 slm, 17, 21 slm-class, 19 slm-package, 2 slm.class (slm-class), 19 summary, 19, 21 summary. slm, 20 symnum, 20
<pre>cov_kernel, 6 cov_matrix_estimator, 7, 23 cov_method, 8</pre>	terms, <i>19</i> trapeze, 22 triangle, 22
<pre>cov_select, 9 cov_spectralproj, 10</pre>	vcov, 19, 23 vcov.slm, 23
DDSE, <i>11</i> Djump, <i>11</i>	,
effects, 19	
fitted, 19 formula, 17	
<pre>generative_model, 11 generative_process, 12</pre>	
kernHAC, 7, 8	
lm, 18	
plot.lm, 13 plot.slm, 13 predict, 19 predict.lm, 14 predict.slm, 13	
Rboot, 6, 15	