

Package ‘segmented’

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Type Package

Title Regression Models with Break-Points / Change-Points Estimation

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Description

Given a regression model, segmented ‘updates’ it by adding one or more segmented (i.e., piece-wise linear) relationships.

Several variables with multiple breakpoints are allowed. The estimation method is discussed in Muggeo (2003, <[doi:10.1002/sim.1545](https://doi.org/10.1002/sim.1545)>)

and illustrated in Muggeo (2008, <https://www.r-project.org/doc/Rnews/Rnews_2008-1.pdf>). An approach for hypothesis testing is presented

in Muggeo (2016, <[doi:10.1080/00949655.2016.1149855](https://doi.org/10.1080/00949655.2016.1149855)>), and interval estimation for the breakpoint is discussed

in Muggeo (2017, <[doi:10.1111/anzs.12200](https://doi.org/10.1111/anzs.12200)>).

Depends MASS

License GPL

NeedsCompilation no

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R topics documented:

segmented-package	2
aapc	4
broken.line	5
confint.segmented	7
davies.test	9
down	11
draw.history	12
intercept	13

lines.segmented	14
plant	15
plot.segmented	16
points.segmented	19
predict.segmented	20
print.segmented	21
pscore.test	22
pwr.seg	24
seg.control	26
seg.lm.fit	29
segmented	31
selgmented	37
slope	39
stagnant	41
summary.segmented	42
vcov.segmented	44
Index	46

segmented-package	<i>Segmented relationships in regression models with breakpoints / changepoints estimation</i>
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Description

Estimation and Inference of Regression Models with piecewise linear relationships having a fixed number of break-points. The estimation method is described in Muggeo (2003) <doi: 10.1002/sim.1545>.

Details

Package: segmented
 Type: Package
 Version: 1.5-0
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 License: GPL

Package segmented is aimed to estimate linear and generalized linear models (and virtually any regression model) having one or more segmented relationships in the linear predictor. Estimates of the slopes and breakpoints are provided along with standard errors. The package includes testing/estimating functions and methods to print, summarize and plot the results.

The algorithm used by segmented is *not* grid-search. It is an iterative procedure (Muggeo, 2003) that needs starting values *only* for the breakpoint parameters and therefore it is quite efficient even with several breakpoints to be estimated. Moreover since version 0.2-9.0, segmented implements the bootstrap restarting (Wood, 2001) to make the algorithm less sensitive to starting values.

Since version 0.5-0.0 a default method `segmented.default` has been added. It may be employed to include segmented relationships in *general* regression models where specific methods do not exist. Examples include quantile, Cox, and lme regressions (where the random effects do not refer to the breakpoints!). See examples in [segmented.default](#).

Since version 1.0-0 the estimating algorithm has been slightly modified and it appears to be much stabler (in examples with noisy segmented relationships and flat log likelihoods) than previous versions.

Hypothesis testing (about the existence of the breakpoint) and confidence intervals are performed via appropriate methods and functions.

A tentative approach to deal with unknown number of breakpoints is also provided, see option `fix.npsi` in [seg.control](#). Also, as version 1.3-0, the [selgmented](#) function has been introduced to select the number of breakpoints via the BIC or sequential hypothesis testing.

Author(s)

Vito M.R. Muggeo <vito.muggeo@unipa.it>

References

- Muggeo, V.M.R. (2017) Interval estimation for the breakpoint in segmented regression: a smoothed score-based approach. *Australian & New Zealand Journal of Statistics* **59**, 311–322.
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- Muggeo, V.M.R., Adelfio, G. (2011) Efficient change point detection in genomic sequences of continuous measurements. *Bioinformatics* **27**, 161–166.
- Wood, S. N. (2001) Minimizing model fitting objectives that contain spurious local minima by bootstrap restarting. *Biometrics* **57**, 240–244.
- Muggeo, V.M.R. (2010) Comment on ‘Estimating average annual per cent change in trend analysis’ by Clegg et al., *Statistics in Medicine*; 28, 3670-3682. *Statistics in Medicine*, **29**, 1958–1960.

aapc

Average annual per cent change in segmented trend analysis

Description

Computes the average annual per cent change to summarize piecewise linear relationships in segmented regression models.

Usage

```
aapc(ogg, parm, exp.it = FALSE, conf.level = 0.95, wrong.se = TRUE,
     .vcov=NULL, .coef=NULL, ...)
```

Arguments

ogg	the fitted model returned by segmented.
parm	the <i>single</i> segmented variable of interest. It can be missing if the model includes a single segmented covariate. If missing and ogg includes several segmented variables, the first one is considered.
exp.it	logical. If TRUE, the per cent change is computed, namely $\exp(\hat{\mu}) - 1$ where $\mu = \sum_j \beta_j w_j$, see ‘Details’.
conf.level	the confidence level desired.
wrong.se	logical, if TRUE, the ‘wrong’ standard error (as discussed in Clegg et al. (2009)) ignoring uncertainty in the breakpoint estimate is returned as an attribute “wrong.se”.
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by <code>vcov(ogg, ...)</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef(ogg)</code> .
...	further arguments to be passed on to <code>vcov.segmented()</code> , such as <code>var.diff</code> or <code>is</code> .

Details

To summarize the fitted piecewise linear relationship, Clegg et al. (2009) proposed the ‘average annual per cent change’ (AAPC) computed as the sum of the slopes (β_j) weighted by corresponding covariate sub-interval width (w_j), namely $\mu = \sum_j \beta_j w_j$. Since the weights are the breakpoint differences, the standard error of the AAPC should account for uncertainty in the breakpoint estimate, as discussed in Muggeo (2010) and implemented by `aapc()`.

Value

`aapc` returns a numeric vector including point estimate, standard error and confidence interval for the AAPC relevant to variable specified in `parm`.

Note

`exp.it=TRUE` would be appropriate only if the response variable is the log of (any) counts.

Author(s)

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References

Clegg LX, Hankey BF, Tiwari R, Feuer EJ, Edwards BK (2009) Estimating average annual per cent change in trend analysis. *Statistics in Medicine*, **28**; 3670-3682

Muggeo, V.M.R. (2010) Comment on 'Estimating average annual per cent change in trend analysis' by Clegg et al., *Statistics in Medicine*; 28, 3670-3682. *Statistics in Medicine*, **29**, 1958–1960.

Examples

```
set.seed(12)
x<-1:20
y<-2-.5*x+.7*pmax(x-9,0)-.8*pmax(x-15,0)+rnorm(20)*.3
o<-lm(y~x)
os<-segmented(o, psi=c(5,12))
aapc(os)
```

broken.line

Fitted values for segmented relationships

Description

Given a segmented model (typically returned by a segmented method), `broken.line` computes the fitted values (and relevant standard errors) for each 'segmented' relationship.

Usage

```
broken.line(ogg, term = NULL, link = TRUE, interc=TRUE, se.fit=TRUE, isV=FALSE,
            .vcov=NULL, .coef=NULL, ...)
```

Arguments

<code>ogg</code>	A fitted object of class <code>segmented</code> (returned by any segmented method).
<code>term</code>	Three options. i) A named list (whose name should be one of the segmented covariates in the model <code>ogg</code>) including the covariate values for which segmented predictions should be computed; ii) a character meaning the name of any segmented covariate in the model, and predictions corresponding to the observed covariate values are returned; iii) It can be <code>NULL</code> if the model includes a single segmented covariate, and predictions corresponding to the observed covariate values are returned;

link	Should the predictions be computed on the scale of the link function? Default to TRUE.
interc	Should the model intercept be added? (provided it exists).
se.fit	If TRUE also standard errors for predictions are returned.
isV	A couple of logicals indicating if the segmented terms $(x-\psi)_+$ and $I(x > \psi)$ in the model matrix should be replaced by their smoothed counterparts. If a single logical is provided, it is applied to both terms.
.vcov	Optional. The <i>full</i> covariance matrix of estimates. If NULL (and se.fit=TRUE), the matrix is computed internally via <code>vcov.segmented()</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef(ogg)</code> .
...	Additional arguments to be passed on to <code>vcov.segmented()</code> when computing the standard errors for the predictions, namely <code>is</code> , <code>var.diff</code> , <code>p.df</code> . See summary.segmented and vcov.segmented .

Details

If `term=NULL` or `term` is a valid segmented covariate name, predictions for each segmented variable are the relevant fitted values from the model. If `term` is a (correctly named) list with numerical values, predictions corresponding to such specified values are computed. If `link=FALSE` and `ogg` inherits from the class "glm", predictions and possible standard errors are returned on the response scale. The standard errors come from the Delta method. Argument `link` is ignored whether `ogg` does not inherit from the class "glm".

Value

A list having one component if (if `se.fit=FALSE`), and two components (if `se.fit=TRUE`) list representing predictions and standard errors for the segmented covariate values.

Author(s)

Vito M. R. Muggeo

See Also

[segmented](#), [predict.segmented](#), [plot.segmented](#), [vcov.segmented](#)

Examples

```
set.seed(1234)
z<-runif(100)
y<-rpois(100,exp(2+1.8*pmax(z-.6,0)))
o<-glm(y~z,family=poisson)
o.seg<-segmented(o,seg.Z=~z,psi=.5)
## Not run: plot(z,y)
## Not run: points(z,broken.line(o.seg,link=FALSE)$fit,col=2) #ok, but use plot.segmented()!
```

confint.segmented *Confidence intervals for breakpoints*

Description

Computes confidence intervals for the breakpoints in a fitted ‘segmented’ model.

Usage

```
## S3 method for class 'segmented'
confint(object, parm, level=0.95, method=c("delta", "score", "gradient"),
        rev.sgn=FALSE, var.diff=FALSE, is=FALSE, digits=max(4, getOption("digits") - 1),
        .coef=NULL, .vcov=NULL, ...)
```

Arguments

object	a fitted segmented object.
parm	the segmented variable of interest. If missing the first segmented variable in object is considered.
level	the confidence level required, default to 0.95.
method	which confidence interval should be computed. One of "delta", "score", or "gradient". Can be abbreviated.
rev.sgn	vector of logicals. The length should be equal to the length of parm; recycled otherwise. when TRUE it is assumed that the current parm is ‘minus’ the actual segmented variable, therefore the sign is reversed before printing. This is useful when a null-constraint has been set on the last slope.
var.diff	logical. If method="delta", and there is a single segmented variable, var.diff=TRUE leads to standard errors based on sandwich-type formula of the covariance matrix. See Details in summary.segmented .
is	logical. If method="delta", is=TRUE means that the full covariance matrix is computed via vcov(..., is=TRUE)
digits	controls the number of digits to print when returning the output.
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by coef(object).
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by vcov(object).
...	additional parameters referring to Score-based confidence intervals, such as "h", "d.h", "bw", "msgWarn", and "n.values" specifying the number of points used to profile the Score (or Gradient) statistic.

Details

confint.segmented computes confidence limits for the breakpoints. Currently there are three options, see argument method. method="delta" uses the standard error coming from the Delta method for the ratio of two random variables. This value is an approximation (slightly) better than the one reported in the 'psi' component of the list returned by any segmented method. The resulting confidence intervals are based on the asymptotic Normal distribution of the breakpoint estimator which is reliable just for clear-cut kink relationships. See Details in [segmented](#). method="score" or method="gradient" compute the confidence interval via profiling the Score or the Gradient statistics smoothed out by the induced smoothing paradigm, as discussed in the reference below.

Value

A matrix including point estimate and confidence limits of the breakpoint(s) for the segmented variable possibly specified in parm.

Note

Currently method="score" or method="gradient" only works for segmented *linear* model. For segmented *generalized linear* model, currently only method="delta" is available.

Author(s)

Vito M.R. Muggeo

References

Muggeo, V.M.R. (2017) Interval estimation for the breakpoint in segmented regression: a smoothed score-based approach. *Australian & New Zealand Journal of Statistics* **59**, 311–322.

See Also

[segmented](#) and [lines.segmented](#) to plot the estimated breakpoints with corresponding confidence intervals.

Examples

```
set.seed(10)
x<-1:100
z<-runif(100)
y<-2+1.5*pmax(x-35,0)-1.5*pmax(x-70,0)+10*pmax(z-.5,0)+rnorm(100,0,2)
out.lm<-lm(y~x)
o<-segmented(out.lm,seg.Z=~x+z,psi=list(x=c(30,60),z=.4))
confint(o) #delta CI for the 1st variable
confint(o, "x", method="score") #also method="g"
```

davies.test	<i>Testing for a change in the slope</i>
-------------	--

Description

Given a generalized linear model, the Davies' test can be employed to test for a non-constant regression parameter in the linear predictor.

Usage

```
davies.test(obj, seg.Z, k = 10, alternative = c("two.sided", "less", "greater"),
            type=c("lrt","wald"), values=NULL, dispersion=NULL)
```

Arguments

obj	a fitted model typically returned by <code>glm</code> or <code>lm</code> . Even an object returned by <code>segmented</code> can be set (e.g. if interest lies in testing for an additional breakpoint).
seg.Z	a formula with no response variable, such as <code>seg.Z~x1</code> , indicating the (continuous) segmented variable being tested. Only a single variable may be tested and an error is printed when <code>seg.Z</code> includes two or more terms. <code>seg.Z</code> can be omitted if i) <code>obj</code> is a segmented fit with a single segmented covariate (and that variable is taken), or ii) if it is a "lm" or "glm" fit with a single covariate (and that variable is taken)
k	number of points where the test should be evaluated. See Details.
alternative	a character string specifying the alternative hypothesis (relevant to the slope difference parameter).
type	the test statistic to be used (only for GLM, default to <code>lrt</code>). Ignored if <code>obj</code> is a simple linear model.
values	optional. The evaluation points where the Davies approximation is computed. See Details for default values.
dispersion	the dispersion parameter for the family to be used to compute the test statistic. When <code>NULL</code> (the default), it is inferred from <code>obj</code> . Namely it is taken as 1 for the Binomial and Poisson families, and otherwise estimated by the residual Chi-squared statistic (calculated from cases with non-zero weights) divided by the residual degrees of freedom.

Details

`davies.test` tests for a non-zero difference-in-slope parameter of a segmented relationship. Namely, the null hypothesis is $H_0 : \beta = 0$, where β is the difference-in-slopes, i.e. the coefficient of the segmented function $\beta(x - \psi)_+$. The hypothesis of interest $\beta = 0$ means no breakpoint. Roughly speaking, the procedure computes `k` 'naive' (i.e. assuming fixed and known the breakpoint) test statistics for the difference-in-slope, seeks the 'best' value and corresponding naive p-value (according to the alternative hypothesis), and then corrects the selected (minimum) p-value by means

of the k values of the test statistic. If `obj` is a LM, the Davies (2002) test is implemented. This approach works even for small samples. If `obj` represents a GLM fit, relevant methods are described in Davies (1987), and the Wald or the Likelihood ratio test statistics can be used, see argument `type`. This is an asymptotic test. The k evaluation points are k equally spaced values between the second and the second-last values of the variable reported in `seg.Z`. k should not be small; I find no important difference for k larger than 10, so default is $k=10$.

Value

A list with class 'htest' containing the following components:

<code>method</code>	title (character)
<code>data.name</code>	the regression model and the segmented variable being tested
<code>statistic</code>	the point within the range of the covariate in <code>seg.Z</code> at which the maximum (or the minimum if <code>alternative="less"</code>) occurs
<code>parameter</code>	number of evaluation points
<code>p.value</code>	the adjusted p-value
<code>process</code>	a two-column matrix including the evaluation points and corresponding values of the test statistic

Warning

The Davies test is *not* aimed at obtaining the estimate of the breakpoint. The Davies test is based on k evaluation points, thus the value returned in the `statistic` component (and printed as "'best' at") is the best among the k points, and typically it will differ from the maximum likelihood estimate returned by `segmented`. Use `segmented` if you are interested in the point estimate.

To test for a breakpoint in *linear* models with small samples, it is suggested to use `davies.test()` with objects of class "lm". If `obj` is a "glm" object with gaussian family, `davies.test()` will use an approximate test resulting in smaller p-values when the sample is small. However if the sample size is large ($n > 300$), the exact Davies (2002) upper bound cannot be computed (as it relies on `gamma()` function) and the *approximate* upper bound of Davies (1987) is returned.

Note

Strictly speaking, the Davies test is not confined to the segmented regression; the procedure can be applied when a nuisance parameter vanishes under the null hypothesis. The test is slightly conservative, as the computed p-value is actually an upper bound.

Results should change slightly with respect to previous versions where the evaluation points were computed as k equally spaced values between the second and the second last observed values of the segmented variable.

Author(s)

Vito M.R. Muggeo

References

Davies, R.B. (1987) Hypothesis testing when a nuisance parameter is present only under the alternative. *Biometrika* **74**, 33–43.

Davies, R.B. (2002) Hypothesis testing when a nuisance parameter is present only under the alternative: linear model case. *Biometrika* **89**, 484–489.

See Also

See also [pscore.test](#) which is more powerful, especially when the signal-to-noise ratio is low.

Examples

```
## Not run:
set.seed(20)
z<-runif(100)
x<-rnorm(100,2)
y<-2+10*pmax(z-.5,0)+rnorm(100,0,3)

o<-lm(y~z+x)
davies.test(o,~z)
davies.test(o,~x)

o<-glm(y~z+x)
davies.test(o,~z) #it works but the p-value is too small..

## End(Not run)
```

down

Down syndrome in babies

Description

The down data frame has 30 rows and 3 columns. Variable cases means the number of babies with Down syndrome out of total number of births births for mothers with mean age age.

Usage

```
data(down)
```

Format

A data frame with 30 observations on the following 3 variables.

age the mothers' mean age.

births count of total births.

cases count of babies with Down syndrome.

Source

Davison, A.C. and Hinkley, D. V. (1997) *Bootstrap Methods and their Application*. Cambridge University Press.

References

Geyer, C. J. (1991) Constrained maximum likelihood exemplified by isotonic convex logistic regression. *Journal of the American Statistical Association* **86**, 717–724.

Examples

```
data(down)
```

```
draw.history
```

```
History for the breakpoint estimates
```

Description

Displays breakpoint iteration values for segmented fits.

Usage

```
draw.history(obj, term, ...)
```

Arguments

obj	a segmented fit returned by any "segmented" method.
term	a character to mean the 'segmented' variable whose breakpoint values throughout iterations have to be displayed.
...	graphic parameters to be passed to <code>matplot()</code> .

Details

For a given term in a segmented fit, `draw.history()` produces two plots. On the left panel it displays the different breakpoint values obtained during the estimating process, since the starting values up to the final ones, while on the right panel the objective values at different iterations. When bootstrap restarting is employed, `draw.history()` produces two plots, the values of objective function and the number of distinct solutions against the bootstrap replicates.

Value

None.

Author(s)

Vito M.R. Muggeo

Examples

```
data(stagnant)
os<-segmented(lm(y~x,data=stagnant),seg.Z=~x,psi=-.8)
draw.history(os) #diagnostics with boot restarting

os<-segmented(lm(y~x,data=stagnant),seg.Z=~x,psi=-.8, control=seg.control(n.boot=0))
draw.history(os) #diagnostics without boot restarting
```

intercept	<i>Intercept estimates from segmented relationships</i>
-----------	---

Description

Computes the intercepts of each ‘segmented’ relationship in the fitted model.

Usage

```
intercept(ogg, parm, rev.sgn = FALSE, var.diff=FALSE,
          .vcov=NULL, .coef=NULL, digits = max(4, getOption("digits") - 2),...)
```

Arguments

ogg	an object of class "segmented", returned by any segmented method.
parm	the segmented variable whose intercepts have to be computed. If missing all the segmented variables in the model are considered.
rev.sgn	vector of logicals. The length should be equal to the length of parm, but it is recycled otherwise. When TRUE it is assumed that the current parm is ‘minus’ the actual segmented variable, therefore the order is reversed before printing. This is useful when a null-constraint has been set on the last slope.
var.diff	Currently ignored as only point estimates are computed.
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by <code>vcov(ogg)</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef(ogg)</code> .
digits	controls number of digits in the returned output.
...	Further arguments to be passed on to <code>vcov.segmented</code> , such as <code>var.diff</code> and <code>is</code> . See Details in vcov.segmented .

Details

A broken-line relationship means that a regression equation exists in the intervals ‘ $\min(x)$ to ψ_1 ’, ‘ ψ_1 to ψ_2 ’, and so on. `intercept` computes point estimates of the intercepts of the different regression equations for each segmented relationship in the fitted model.

Value

intercept returns a list of one-column matrices. Each matrix represents a segmented relationship.

Author(s)

Vito M. R. Muggeo, <vito.muggeo@unipa.it>

See Also

See also [slope](#) to compute the slopes of the different regression equations for each segmented relationship in the fitted model.

Examples

```
## see ?slope
## Not run:
intercept(out.seg)

## End(Not run)
```

lines.segmented

Bars for interval estimate of the breakpoints

Description

Draws bars relevant to breakpoint estimates (point estimate and confidence limits) on the current device

Usage

```
## S3 method for class 'segmented'
lines(x, term, bottom = TRUE, shift=FALSE, conf.level = 0.95, k = 50,
      pch = 18, rev.sgn = FALSE, .vcov=NULL, .coef=NULL, ...)
```

Arguments

x	an object of class segmented.
term	the segmented variable of the breakpoints being drawn. It may be unspecified when there is a single segmented variable.
bottom	logical, indicating if the bars should be plotted at the bottom (TRUE) or at the top (FALSE).
shift	logical, indicating if the bars should be ‘shifted’ on the y-axis before plotting. Useful for multiple breakpoints with overlapped confidence intervals.
conf.level	the confidence level of the confidence intervals for the breakpoints.
k	a positive integer regulating the vertical position of the drawn bars. See Details.

pch	either an integer specifying a symbol or a single character to be used in plotting the point estimates of the breakpoints. See points .
rev.sgn	should the signs of the breakpoint estimates be changed before plotting? see Details .
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by <code>vcov(x)</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef(x)</code> .
...	further arguments passed to segments , for instance 'col' that can be a vector.

Details

`lines.segmented` simply draws on the current device the point estimates and relevant confidence limits of the estimated breakpoints from a "segmented" object. The y coordinate where the bars are drawn is computed as `usr[3]+h` if `bottom=TRUE` or `usr[4]-h` when `bottom=FALSE`, where $h=(usr[4]-usr[3])/abs(k)$ and `usr` are the extremes of the user coordinates of the plotting region. Therefore for larger values of `k` the bars are plotted on the edges. The argument `rev.sgn` allows to change the sign of the breakpoints before plotting. This may be useful when a null-right-slope constraint is set.

See Also

[plot.segmented](#) to plot the fitted segmented lines, and [points.segmented](#) to add the fitted join-points.

Examples

```
## See ?plot.segmented
```

plant	<i>Plan organ dataset</i>
-------	---------------------------

Description

The `plant` data frame has 103 rows and 3 columns.

Usage

```
data(plant)
```

Format

A data frame with 103 observations on the following 3 variables:

`y` measurements of the plant organ.

`time` times where measurements took place.

`group` three attributes of the plant organ, RKV, RKW, RWC.

Details

Three attributes of a plant organ measured over time where biological reasoning indicates likelihood of multiple breakpoints. The data are scaled to the maximum value for each attribute and all attributes are measured at each time.

Source

The data have been kindly provided by Dr Zongjian Yang at School of Land, Crop and Food Sciences, The University of Queensland, Brisbane, Australia.

Examples

```
## Not run:
data(plant)
attach(plant)

lattice::xyplot(y~time, groups=group, auto.key=list(space="right"))

## End(Not run)
```

plot.segmented

Plot method for segmented objects

Description

Takes a fitted segmented object returned by segmented() and plots (or adds) the fitted broken-line for the selected segmented term.

Usage

```
## S3 method for class 'segmented'
plot(x, term, add=FALSE, res=FALSE, conf.level=0, interc=TRUE,
     link=TRUE, res.col=1, rev.sgn=FALSE, const=0, shade=FALSE,
     rug=!add, dens.rug=FALSE, dens.col = grey(0.8), transf=I,
     isV=FALSE, is=FALSE, var.diff=FALSE, p.df="p", .vcov=NULL, .coef=NULL,
     prev.trend=FALSE, smoos=NULL, ...)
```

Arguments

x	a fitted segmented object.
term	the segmented variable having the piece-wise relationship to be plotted. If there is a single segmented variable in the fitted model x, term can be omitted.
add	when TRUE the fitted lines are added to the current device.
res	when TRUE the fitted lines are plotted along with corresponding partial residuals. See Details.

conf.level	If greater than zero, it means the confidence level at which the pointwise confidence intervals have to be plotted.
interc	If TRUE the computed segmented components include the model intercept (if it exists).
link	when TRUE (default), the fitted lines are plotted on the link scale, otherwise they are transformed on the response scale before plotting. Ignored for linear segmented fits.
res.col	when res=TRUE it means the color of the points representing the partial residuals.
rev.sgn	when TRUE it is assumed that current term is 'minus' the actual segmented variable, therefore the sign is reversed before plotting. This is useful when a null-constraint has been set on the last slope.
const	constant to add to each fitted segmented relationship (on the scale of the linear predictor) before plotting.
shade	if TRUE and conf.level>0 it produces shaded regions (in grey color) for the pointwise confidence intervals embracing the fitted segmented line.
rug	when TRUE the covariate values are displayed as a rug plot at the foot of the plot. Default is to !add.
dens.rug	when TRUE then smooth covariate distribution is plotted on the x-axis.
dens.col	if dens.rug=TRUE, it means the colour to be used to plot the density.
transf	A possible function to convert the fitted values before plotting. It is only effective if res=FALSE. If res=TRUE any transformation is ignored.
isV	logical value (to be passed to broken.line). Ignored if conf.level=0
is	logical value (to be passed to broken.line) indicating if the covariance matrix based on the induced smoothing should be used. Ignored if conf.level=0
var.diff	logical value to be passed to summary.segmented to compute the standard errors of fitted values (if conf.level>0).
p.df	degrees of freedom when var.diff=TRUE, see summary.segmented
.vcov	The <i>full</i> covariance matrix of estimates to be used when conf.level>0. If unspecified (i.e. NULL), the covariance matrix is computed internally by the function <code>vcov.segmented</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef()</code> .
prev.trend	logical. If TRUE dashed lines corresponding to the 'previous' trends (i.e. the trends if the breakpoints would not have occurred) are also drawn.
smoos	logical, indicating if the residuals (provided that res=TRUE) will be drawn using a <i>smoothed</i> scatterplot. If NULL (default) the smoothed scatterplot will be employed when the number of observation is larger than 10000.
...	other graphics parameters to pass to plotting commands: 'col', 'lwd' and 'lty' (that can be vectors and are recycled if necessary, see the example below) for the fitted piecewise lines; 'ylab', 'xlab', 'main', 'sub', 'cex.axis', 'cex.lab', 'xlim' and 'ylim' when a new plot is produced (i.e. when add=FALSE); 'pch' and 'cex' for the partial residuals (when res=TRUE, res.col is for the color); col.shade for the shaded regions (provided that shade=TRUE and conf.level>0).

Details

Produces (or adds to the current device) the fitted segmented relationship between the response and the selected term. If the fitted model includes just a single 'segmented' variable, term may be omitted.

The partial residuals are computed as 'fitted + residuals', where 'fitted' are the fitted values of the segmented relationship relevant to the covariate specified in term. Notice that for GLMs the residuals are the response residuals if `link=FALSE` and the working residuals if `link=TRUE`.

Value

None.

Note

For models with offset, partial residuals on the response scale are not defined. Thus `plot.segmented` does not work when `link=FALSE`, `res=TRUE`, and the fitted model includes an offset.

Author(s)

Vito M. R. Muggeo

See Also

[segmented](#) to fit the model, [lines.segmented](#) to add the estimated breakpoints on the current plot. [points.segmented](#) to add the joinpoints of the segmented relationship. [predict.segmented](#) to compute standard errors and confidence intervals for predictions from a "segmented" fit.

Examples

```
set.seed(1234)
z<-runif(100)
y<-rpois(100,exp(2+1.8*pmax(z-.6,0)))
o<-glm(y~z,family=poisson)
o.seg<-segmented(o) #single segmented covariate and one breakpoint: 'seg.Z' and 'psi' can be omitted
par(mfrow=c(1,2))
plot(o.seg, conf.level=0.95, shade=TRUE)
points(o.seg, link=TRUE, col=2)
## new plot
plot(z,y)
## add the fitted lines using different colors and styles..
plot(o.seg,add=TRUE,link=FALSE,lwd=2,col=2:3, lty=c(1,3))
lines(o.seg,col=2,pch=19,bottom=FALSE,lwd=2) #for the CI for the breakpoint
points(o.seg,col=4, link=FALSE)
## using the options 'is', 'isV', 'shade' and 'col.shade'.
par(mfrow=c(1,2))
plot(o.seg, conf.level=.9, is=TRUE, isV=TRUE, col=1, shade = TRUE, col.shade=2)
plot(o.seg, conf.level=.9, is=TRUE, isV=FALSE, col=2, shade = TRUE, res=TRUE, res.col=4, pch=3)
```

points.segmented *Points method for segmented objects*

Description

Takes a fitted segmented object returned by segmented() and adds on the current plot the joinpoints of the fitted broken-line relationships.

Usage

```
## S3 method for class 'segmented'
points(x, term, interc = TRUE, link = TRUE, rev.sgn=FALSE,
       transf=I, .vcov=NULL, .coef=NULL, ...)
```

Arguments

x	an object of class segmented.
term	the segmented variable of interest. It may be unspecified when there is a single segmented variable.
interc	If TRUE the computed joinpoints include the model intercept (if it exists).
link	when TRUE (default), the fitted joinpoints are plotted on the link scale
rev.sgn	when TRUE, the fitted joinpoints are plotted on the ‘minus’ scale of the current term variable. This is useful when a null-constraint has been set on the last slope.
transf	A possible function to convert the fitted values before plotting.
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by vcov().
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by coef(x).
...	other graphics parameters to pass on to points() function.

Details

We call ‘joinpoint’ the plane point having as coordinates the breakpoint (on the x scale) and the fitted value of the segmented relationship at that breakpoint (on the y scale). points.segmented() simply adds the fitted joinpoints on the current plot. This could be useful to emphasize the changes of the piecewise linear relationship.

See Also

[plot.segmented](#) to plot the fitted segmented lines.

Examples

```
## Not run:
#see examples in ?plot.segmented

## End(Not run)
```

predict.segmented	<i>Predict method for segmented model fits</i>
-------------------	--

Description

Returns predictions and optionally associated quantities (standard errors or confidence intervals) from a fitted segmented model object.

Usage

```
## S3 method for class 'segmented'
predict(object, newdata, .coef=NULL, ...)
```

Arguments

object	a fitted segmented model coming from segmented.lm or segmented.glm.
newdata	An optional data frame in which to look for variables with which to predict. If omitted, the fitted values are used.
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by coef().
...	further arguments passed to predict.lm or predict.glm. Usually these are se.fit, or interval or type.

Details

Basically predict.segmented builds the right design matrix accounting for breakpoint and passes it to predict.lm or predict.glm depending on the actual model fit object.

Value

predict.segmented produces a vector of predictions with possibly associated standard errors or confidence intervals. See predict.lm or predict.glm.

Note

If type="terms", predict.segmented returns predictions for each component of the segmented term. Namely if 'my.x' is the segmented variable, predictions for 'my.x', 'U1.my.x' and 'psi1.my.x' are returned. These are meaningless individually, however their sum provides the predictions for the segmented term.

Author(s)

Vito Muggeo

See Also[segmented](#), [plot.segmented](#), [broken.line](#), [predict.lm](#), [predict.glm](#)**Examples**

```
n=10
x=seq(-3,3,l=n)
set.seed(1515)
y <- (x<0)*x/2 + 1 + rnorm(x,sd=0.15)
segm <- segmented(lm(y ~ x), ~ x, psi=0.5)
predict(segm,se.fit = TRUE)$se.fit

#wrong (smaller) st.errors (assuming known the breakpoint)
olm<-lm(y~x+pmax(x-segm$psi[,2],0))
predict(olm,se.fit = TRUE)$se.fit
```

print.segmented

*Print method for the segmented class***Description**

Printing the most important features and coefficients (including the breakpoints) of a segmented model.

Usage

```
## S3 method for class 'segmented'
print(x, digits = max(3, getOption("digits") - 3), ...)

## S3 method for class 'segmented'
coef(object, include.psi=FALSE, ...)
```

Arguments

x	object of class segmented
digits	number of digits to be printed
object	object of class segmented
include.psi	logical. If TRUE, the breakpoints are returned along with the regression coefficients
...	arguments passed to other functions

Author(s)

Vito M.R. Muggeo

See Also

[summary.segmented](#), [print.summary.segmented](#)

pscore.test *Testing for existence of one breakpoint*

Description

Given a (generalized) linear model, the (pseudo) Score statistic tests for the existence of one breakpoint.

Usage

```
pscore.test(obj, seg.Z, k = 10, alternative = c("two.sided", "less", "greater"),
  values=NULL, dispersion=NULL, df.t=NULL, more.break=FALSE, n.break=1,
  only.term=FALSE, break.type=1)
```

Arguments

obj	a fitted model typically returned by <code>glm</code> or <code>lm</code> . Even an object returned by <code>segmented</code> can be set. Offset and weights are allowed.
seg.Z	a formula with no response variable, such as <code>seg.Z=~x1</code> , indicating the (continuous) segmented variable being tested. Only a single variable may be tested and an error is printed when <code>seg.Z</code> includes two or more terms. <code>seg.Z</code> can be omitted if i) <code>obj</code> is a segmented fit with a single segmented covariate (and that variable is taken), or ii) if it is a "lm" or "glm" fit with a single covariate (and that variable is taken).
k	optional. Number of points (equi-spaced from the min to max) used to compute the pseudo Score statistic. See Details.
alternative	a character string specifying the alternative hypothesis (relevant to the slope difference parameter).
values	optional. The evaluation points where the Score test is computed. See Details for default values.
dispersion	optional. the dispersion parameter for the family to be used to compute the test statistic. When NULL (the default), it is inferred from <code>obj</code> . Namely it is taken as 1 for the Binomial and Poisson families, and otherwise estimated by the residual Chi-squared statistic in the model <code>obj</code> (calculated from cases with non-zero weights divided by the residual degrees of freedom).
df.t	optional. The degrees-of-freedom used to compute the p-value. When NULL, the <code>df</code> extracted from <code>obj</code> are used.
more.break	optional, logical. If <code>obj</code> is a 'segmented' fit, <code>more.break=FALSE</code> tests for the actual breakpoint for the variable 'seg.Z', while <code>more.break=TRUE</code> tests for an <i>additional</i> breakpoint(s) for the variable 'seg.Z'. Ignored when <code>obj</code> is not a segmented fit.

n.break	optional. Number of breakpoints postulated under the alternative hypothesis.
only.term	logical. If TRUE, only the pseudo covariate(s) relevant to the testing for the breakpoint is returned, and no test is computed.
break.type	The kind of breakpoint being tested. 1 is for piecewise-linear relationships, 2 means piecewise-constant, i.e. a step-function, relationships.

Details

`pscore.test` tests for a non-zero difference-in-slope parameter of a segmented relationship. Namely, the null hypothesis is $H_0 : \beta = 0$, where β is the difference-in-slopes, i.e. the coefficient of the segmented function $\beta(x - \psi)_+$. The hypothesis of interest $\beta = 0$ means no breakpoint. Simulation studies have shown that such Score test is more powerful than the Davies test (see reference) when the alternative hypothesis is ‘one changepoint’. If there are two or more breakpoints (for instance, a sinusoidal-like relationships), `pscore.test` can have lower power, and `davies.test` can perform better.

The dispersion value, if unspecified, is taken from `obj`. If `obj` represents the fit under the null hypothesis (no changepoint), the dispersion parameter estimate will be usually larger, leading to a (potentially severe) loss of power.

The `k` evaluation points are `k` equally spaced values in the range of the segmented covariate. `k` should not be small. Specific values can be set via `values`, although I have found no important difference due to number and location of the evaluation points, thus default is `k=10` equally-spaced points. However, when the possible breakpoint is believed to lie into a specified narrower range, the user can specify `k` values in that range leading to higher power in detecting it, i.e. typically lower p-value.

If `obj` is a (segmented) *lm* object, the returned p-value comes from the t-distribution with appropriate degrees of freedom. Otherwise, namely if `obj` is a (segmented) *glm* object, the p-value is computed wrt the Normal distribution.

Value

A list with class ‘`hstest`’ containing the following components:

method	title (character)
data.name	the regression model and the segmented variable being tested
statistic	the empirical value of the statistic
parameter	number of evaluation points
p.value	the p-value
process	the alternative hypothesis set

Author(s)

Vito M.R. Muggeo

References

Muggeo, V.M.R. (2016) Testing with a nuisance parameter present only under the alternative: a score-based approach with application to segmented modelling. *J of Statistical Computation and Simulation*, **86**, 3059–3067.

See Also

See also [davies.test](#).

Examples

```
## Not run:
set.seed(20)
z<-runif(100)
x<-rnorm(100,2)
y<-2+10*pmax(z-.5,0)+rnorm(100,0,3)

o<-lm(y~z+x)

#testing for one changepoint
#use the simple null fit
pscore.test(o,~z) #compare with davies.test(o,~z)..

#use the segmented fit
os<-segmented(o, ~z)
pscore.test(os,~z) #smaller p-value, as it uses the dispersion under the alternative (from 'os')

#test for the 2nd breakpoint in the variable z
pscore.test(os,~z, more.break=TRUE)

## End(Not run)
```

pwr.seg

Power Analysis in segmented regression

Description

Given the appropriate input values, the function computes the power (sample size) corresponding to the specified sample size (power). If a segmented fit object is provided, the power is computed taking the parameter estimates as input values.

Usage

```
pwr.seg(oseg, pow, n, z = "1:n/n", psi, d, s, n.range = c(10,300),
        X = NULL, break.type=1, alpha = 0.01, round.n = TRUE,
        alternative = c("two.sided", "greater", "less"), msg = TRUE, ci.pow=0)
```

Arguments

oseg	The fitted segmented object. If provided, the power is computed at the model parameter estimates, and all the remaining arguments but alternative and alpha are ignored.
pow	The desired power level. If provided n has to be missing

n	The fixed sample size. If provided pow has to be missing
z	The covariate understood to have a segmented effect. Default is "1:n/n", i.e. equispaced values in (0,1). More generally a string indicating the quantile function having p and possible other numerical values as arguments. For instance "qunif(p,0,1)", "qnorm(p,2,5)", or "qexp(p)". "qunif(p,1,n)" can be also specified, but attention should be paid to guarantee psi within the covariate range. Finally, it could be also a numerical vector meaning the actual covariate, but pow has to be missing. Namely if the covariate is supplied (and n is known), only the relevant power can be estimated.
psi	The breakpoint value within the covariate range
d	The slope difference
s	The response standard deviation
n.range	When pow is provided and the relevant sample size estimate has to be returned, the function evaluates 50 sample sizes equally spaced in n.range. However the function can also compute, via spline interpolation, sample sizes outside the specified range.
X	The design matrix including additional linear variables in the regression equation. Default to NULL which means intercept and linear term for the segmented covariate.
break.type	Numerical, meaning the type of breakpoint. break.type=1 means piecewise linear (segmented), break.type=2 refers to piecewise constant.
alpha	The type-I error probability. Default to 0.01.
round.n	logical. If TRUE the (possible) returned sample size value is rounded.
alternative	a character string specifying the alternative hypothesis, must be one of "two.sided", "greater" or "less". Note, this refers to the sign of the slope difference.
msg	logical. If TRUE the output is returned along with a simple message, otherwise only the values are returned
ci.pow	Numerical. If oseg has been supplied, ci.pow replicates are drawn to build a 95% confidence interval for the power.

Details

The function exploits the sampling distribution of the pseudo Score statistic under the alternative hypothesis of one breakpoint.

Value

The computed power *or* sample size, with or without message (depending on msg)

Note

Currently the function assumes just 1 breakpoint in one covariate

Author(s)

Nicoletta D'Angelo and Vito Muggeo

References

D'Angelo N, Muggeo V.M.R. (2021) Power analysis in segmented regression, working paper <https://www.researchgate.net/publication/355885747>.

Muggeo, V.M.R. (2016) Testing with a nuisance parameter present only under the alternative: a score-based approach with application to segmented modelling. *J of Statistical Computation and Simulation*, **86**, 3059–3067.

See Also

[pscore.test](#)

Examples

```
## pwr.seg(pow=.7, psi=.5, d=1.5, s=.5) #returns the sample size
## pwr.seg(n=219, psi=.5, d=1.5, s=.5) #returns the power
## pwr.seg(n=20,z="qnorm(p, 2,5)", psi=3, d=.5, s=2) #the covariate is N(2,5)
## pwr.seg(n=20,z="qexp(p)", psi=.1, d=.5, s=.1) #the covariate is Exp(1)
```

seg.control

Auxiliary for controlling segmented model fitting

Description

Auxiliary function as user interface for 'segmented' fitting. Typically only used when calling any 'segmented' method (`segmented.lm`, `segmented.glm`, `segmented.Arima` or `segmented.default`).

Usage

```
seg.control(n.boot=10, display = FALSE, tol = 1e-05, it.max = 30, fix.npsi=TRUE,
  K = 10, quant = TRUE, maxit.glm = 25, h = 1, break.boot=5, size.boot=NULL,
  jt=FALSE, nonParam=TRUE, random=TRUE, seed=12345, fn.obj=NULL, digits=NULL,
  conv.psi=FALSE, alpha=.05, min.step=.0001,
  powers=c(1,1), last = TRUE, stop.if.error = NULL, gap=FALSE, fc=.95)
```

Arguments

n.boot	number of bootstrap samples used in the bootstrap restarting algorithm. If 0 the standard algorithm, i.e. without bootstrap restart, is used. Default to 10 that appears to be sufficient in most of problems. However when multiple breakpoints have to be estimated it is suggested to increase n.boot, e.g. n.boot=50.
display	logical indicating if the value of the objective function should be printed (along with current breakpoint estimates) at each iteration or at each bootstrap resample. If bootstrap restarting is employed, the values of objective and breakpoint estimates should not change at the last runs.
tol	positive convergence tolerance.

it.max	integer giving the maximal number of iterations.
fix.npsi	logical (it replaces previous argument stop.if.error) If TRUE (default) the <i>number</i> (and not location) of breakpoints is held fixed throughout iterations. Otherwise a sort of ‘automatic’ breakpoint selection is carried out, provided that several starting values are supplied for the breakpoints, see argument psi in segmented.lm or segmented.glm . The idea, relying on removing the ‘non-admissible’ breakpoint estimates at each iteration, is discussed in Muggeo and Adelfio (2011) and it is not compatible with the bootstrap restart algorithm. fix.npsi=FALSE, indeed, should be considered as a preliminary and tentative approach to deal with an unknown number of breakpoints. Alternatively, see selgmented .
K	the number of quantiles (or equally-spaced values) to supply as starting values for the breakpoints when the psi argument of segmented is set to NA. K is ignored when psi is different from NA.
quant	logical, indicating how the starting values should be selected. If FALSE equally-spaced values are used, otherwise the quantiles. Ignored when psi is different from NA.
maxit.glm	integer giving the maximum number of inner IWLS iterations (see details).
h	positive factor modifying the increments in breakpoint updates during the estimation process (see details).
break.boot	Integer, less than n.boot. If break.boot consecutive bootstrap samples lead to the same objective function, the algorithm stops without performing all n.boot ‘trials’. This can save computational time considerably.
size.boot	the size of the bootstrap samples. If NULL, it is taken equal to the actual sample size.
jt	logical. If TRUE the values of the segmented variable(s) are jittered before fitting the model to the bootstrap resamples.
nonParam	if TRUE nonparametric bootstrap (i.e. case-resampling) is used, otherwise residual-based. Currently working only for LM fits. It is not clear what residuals should be used for GLMs.
random	if TRUE, when the algorithm fails to obtain a solution, random values are employed to obtain candidate values.
seed	The seed to be passed on to set.seed() when n.boot>0. Set to NULL to use a random seed. Fixing the seed can be useful to replicate the results when the bootstrap restart algorithm is employed. The segmented fit includes seed representing the integer vector saved just before the bootstrap resampling. Re-use it if you want to replicate the bootstrap restarting algorithm with the <i>same</i> samples.
fn.obj	A <i>character string</i> to be used (optionally) only when segmented.default is used. It represents the function (with argument 'x') to be applied to the fit object to extract the objective function to be <i>minimized</i> . Thus for "lm" fits (although unnecessary) it should be fn.obj="sum(x\$residuals^2)", for "coxph" fits it should be fn.obj="-x\$loglik[2]". If NULL the ‘minus log likelihood’ extracted from the object, namely "-logLik(x)", is used. See segmented.default .
digits	optional. If specified it means the desired number of decimal points of the breakpoint to be used during the iterative algorithm.

conv.psi	optional. Should convergence of iterative procedure to be assessed on changes of breakpoint estimates or changes in the objective? Default to FALSE.
alpha	optional numerical value. The breakpoint is estimated within the quantiles alpha and 1-alpha of the relevant covariate.
min.step	optional. The minimum step size to break the iterative algorithm. Default to 0.0001.
powers	The powers of the pseudo covariates employed by the algorithm. These can be altered during the iterative process to stabilize the estimation procedure. Usually of no interest for the user. <i>This argument will be removed in next releases.</i>
last	logical indicating if output should include only the last fitted model. <i>This argument will be removed in next releases</i>
stop.if.error	same than fix.npsi. <i>This argument will be removed in next releases</i> , and replaced by fix.npsi. If provided, and different from NULL, it overwrites fix.npsi
gap	logical, if FALSE the gap coefficients are <i>always</i> constrained to zero at the convergence. <i>This argument will be removed in next releases.</i>
fc	A proportionality factor (≤ 1) to adjust the breakpoint estimates <i>if</i> these come close to the boundary or too close each other. For instance, if psi turns up close to the maximum, it will be changed to psi*fc or to psi/fc if close to the minimum. This is useful to get finite point estimate and standard errors for each slope parameter.

Details

Fitting a 'segmented' GLM model is attained via fitting iteratively standard GLMs. The number of (outer) iterations is governed by `it.max`, while the (maximum) number of (inner) iterations to fit the GLM at each fixed value of `psi` is fixed via `maxit.glm`. Usually three-four inner iterations may be sufficient.

When the starting value for the breakpoints is set to NA for any segmented variable specified in `seg.Z`, `K` values (quantiles or equally-spaced) are selected as starting values for the breakpoints. In this case, it may be useful to set also `fix.npsi=FALSE` to automate the procedure, see Muggeo and Adelfio (2011). The maximum number of iterations (`it.max`) should be also increased when the 'automatic' procedure is used.

If `last=TRUE`, the object resulting from `segmented.lm` (or `segmented.glm`) is a list of fitted GLM; the *i*-th model is the segmented model with the values of the breakpoints at the *i*-th iteration.

Since version 0.2-9.0 `segmented` implements the bootstrap restarting algorithm described in Wood (2001). The bootstrap restarting is expected to escape the local optima of the objective function when the segmented relationship is flat. Notice bootstrap restart runs `n.boot` iterations regardless of `tol` that only affects convergence within the inner loop.

Value

A list with the arguments as components.

Author(s)

Vito Muggeo

References

Muggeo, V.M.R., Adelfio, G. (2011) Efficient change point detection in genomic sequences of continuous measurements. *Bioinformatics* **27**, 161–166.

Wood, S. N. (2001) Minimizing model fitting objectives that contain spurious local minima by bootstrap restarting. *Biometrics* **57**, 240–244.

Examples

```
#decrease the maximum number inner iterations and display the
#evolution of the (outer) iterations
#seg.control(display = TRUE, maxit.glm=4)
```

seg.lm.fit

Fitter Functions for Segmented Linear Models

Description

seg.lm.fit is called by segmented.lm to fit segmented linear (gaussian) models. Likewise, seg.glm.fit is called by segmented.glm to fit generalized segmented linear models, and seg.def.fit is called by segmented.default to fit segmented relationships in general regression models (e.g., quantile regression and Cox regression). seg.lm.fit.boot, seg.glm.fit.boot, and seg.def.fit.boot are employed to perform bootstrap restart. These functions should usually not be used directly by the user.

Usage

```
seg.lm.fit(y, XREG, Z, PSI, w, offs, opz, return.all.sol=FALSE)
```

```
seg.lm.fit.boot(y, XREG, Z, PSI, w, offs, opz, n.boot=10,
  size.boot=NULL, jt=FALSE, nonParam=TRUE, random=FALSE, break.boot=n.boot)
```

```
seg.glm.fit(y, XREG, Z, PSI, w, offs, opz, return.all.sol=FALSE)
```

```
seg.glm.fit.boot(y, XREG, Z, PSI, w, offs, opz, n.boot=10,
  size.boot=NULL, jt=FALSE, nonParam=TRUE, random=FALSE, break.boot=n.boot)
```

```
seg.def.fit(obj, Z, PSI, mfExt, opz, return.all.sol=FALSE)
```

```
seg.def.fit.boot(obj, Z, PSI, mfExt, opz, n.boot=10, size.boot=NULL,
  jt=FALSE, nonParam=TRUE, random=FALSE, break.boot=n.boot)
```

```
seg.Ar.fit(obj, XREG, Z, PSI, opz, return.all.sol=FALSE)
```

```
seg.Ar.fit.boot(obj, XREG, Z, PSI, opz, n.boot=10, size.boot=NULL, jt=FALSE,
  nonParam=TRUE, random=FALSE, break.boot=n.boot)
```

```
seg.num.fit(y, XREG, Z, PSI, w, opz, return.all.sol=FALSE)
```

```
seg.num.fit.boot(y, XREG, Z, PSI, w, opz, n.boot=10, size.boot=NULL, jt=FALSE,
  nonParam=TRUE, random=FALSE, break.boot=n.boot)
```

Arguments

y	vector of observations of length n.
XREG	design matrix for standard linear terms.
Z	appropriate matrix including the segmented variables whose breakpoints have to be estimated.
PSI	appropriate matrix including the starting values of the breakpoints to be estimated.
w	possible weights vector.
offs	possible offset vector.
opz	a list including information useful for model fitting.
n.boot	the number of bootstrap samples employed in the bootstrap restart algorithm.
break.boot	Integer, less than n.boot. If break.boot consecutive bootstrap samples lead to the same objective function, the algorithm stops without performing all n.boot 'trials'. This can save computational time considerably.
size.boot	the size of the bootstrap resamples. If NULL (default), it is taken equal to the sample size. values smaller than the sample size are expected to increase perturbation in the bootstrap resamples.
jt	logical. If TRUE the values of the segmented variable(s) are jittered before fitting the model to the bootstrap resamples.
nonParam	if TRUE nonparametric bootstrap (i.e. case-resampling) is used, otherwise residual-based.
random	if TRUE, when the algorithm fails to obtain a solution, random values are used as candidate values.
return.all.sol	if TRUE, when the algorithm fails to obtain a solution, the values visited by the algorithm with corresponding deviances are returned.
obj	the starting regression model where the segmented relationships have to be added.
mfExt	the model frame.

Details

The functions call iteratively `lm.wfit` (or `glm.fit`) with proper design matrix depending on XREG, Z and PSI. `seg.lm.fit.boot` (and `seg.glm.fit.boot`) implements the bootstrap restarting idea discussed in Wood (2001).

Value

A list of fit information.

Note

These functions should usually not be used directly by the user.

Author(s)

Vito Muggeo

References

Wood, S. N. (2001) Minimizing model fitting objectives that contain spurious local minima by bootstrap restarting. *Biometrics* **57**, 240–244.

See Also

[segmented.lm](#), [segmented.glm](#)

Examples

```
##See ?segmented
```

segmented

Segmented relationships in regression models

Description

Fits regression models with segmented relationships between the response and one or more explanatory variables. Break-point estimates are provided.

Usage

```
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
  model = TRUE, ...)
```

```
## Default S3 method:
```

```
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
  model = TRUE, keep.class=FALSE, ...)
```

```
## S3 method for class 'lm'
```

```
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
  model = TRUE, keep.class=FALSE, ...)
```

```
## S3 method for class 'glm'
```

```
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
  model = TRUE, keep.class=FALSE, ...)
```

```
## S3 method for class 'Arima'
```

```
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
```

```

model = TRUE, keep.class=FALSE, ...)

## S3 method for class 'numeric'
segmented(obj, seg.Z, psi, npsi, fixed.psi=NULL, control = seg.control(),
          model = TRUE, keep.class=FALSE, adjX=FALSE, weights=NULL, ...)

```

Arguments

<code>obj</code>	standard 'linear' model of class "lm", "glm" or "Arima". Since version 0.5.0-0 any regression fit may be supplied (see 'Details'). <code>obj</code> could include any covariate, and if there is also the variable specified in <code>seg.Z</code> , then all the slopes of the fitted segmented relationship will be estimated. If <code>obj</code> misses the segmented variable, then the 1st (the leftmost) slope is assumed to be zero. Since version 1.5.0, <code>obj</code> can be a simple numeric or <code>ts</code> object, see examples below.
<code>seg.Z</code>	the segmented variables(s), i.e. the continuous covariate(s) understood to have a piecewise-linear relationship with response. It is a formula with no response variable, such as <code>seg.Z~x</code> or <code>seg.Z~x1+x2</code> . It can be missing when <code>obj</code> includes only one covariate which is taken as segmented variable. Currently, formulas involving functions, such as <code>seg.Z~log(x1)</code> , or selection operators, such as <code>seg.Z~d["x1"]</code> or <code>seg.Z~d\$x1</code> , are <i>not</i> allowed. Also, variable names formed by U or V only (with or without numbers) are not permitted.
<code>psi</code>	starting values for the breakpoints to be estimated. If there is a single segmented variable specified in <code>seg.Z</code> , <code>psi</code> is a numeric vector, and it can be missing when 1 breakpoint has to be estimated (and the median of the segmented variable is used as a starting value). If <code>seg.Z</code> includes several covariates, <code>psi</code> has to be specified as a <i>named</i> list of vectors whose names have to match the variables in the <code>seg.Z</code> argument. Each vector of such list includes starting values for the break-point(s) for the corresponding variable in <code>seg.Z</code> . A NA value means that 'K' quantiles (or equally spaced values) are used as starting values; K is fixed via the seg.control auxiliary function.
<code>npsi</code>	A named vector or list meaning the <i>number</i> (and not locations) of breakpoints to be estimated. The starting values will be internally computed via the quantiles or equally spaced values, as specified in argument <code>quant</code> in seg.control . <code>npsi</code> can be missing and <code>npsi=1</code> is assumed for all variables specified in <code>seg.Z</code> . If <code>psi</code> is provided, <code>npsi</code> is ignored.
<code>fixed.psi</code>	An optional named list meaning the breakpoints to be kept fixed during the estimation procedure. The names should be a subset of (or even the same) variables specified in <code>seg.Z</code> . If there is a single variable in <code>seg.Z</code> , a simple numeric vector can be specified. Note that, in addition to the values specified here, <code>segmented</code> will estimate additional breakpoints. To keep fixed all breakpoints (to be specified in <code>psi</code>) use <code>it.max=0</code> in seg.control
<code>control</code>	a list of parameters for controlling the fitting process. See the documentation for seg.control for details.
<code>model</code>	logical value indicating if the <code>model.frame</code> should be returned.
<code>keep.class</code>	logical value indicating if the final fit returned by <code>segmented.default</code> should keep the class 'segmented' (along with the class of the original fit <code>obj</code>). Ignored by the segmented methods.

...	optional arguments (to be ignored safely). Notice specific arguments relevant to the original call (via <code>lm</code> or <code>glm</code> for instance), such as <code>weights</code> or <code>offset</code> , have to be included in the starting model <code>obj</code>
<code>adjX</code>	if <code>obj</code> is a <code>ts</code> , the segmented variable (if not specified in <code>seg.Z</code>) is computed by taking information from the time series (e.g., years starting from 2000, say). If <code>adjX=TRUE</code> , the segmented variable is shifted such that its min equals zero. Default is using the unshifted values, but if there are several breakpoints to be estimated, it is strongly suggested to set <code>adjX=TRUE</code> .
<code>weights</code>	the weights if <code>obj</code> is a vector or a <code>ts</code> object, otherwise the weights should be specified in the starting fit <code>obj</code> .

Details

Given a linear regression model usually of class "lm" or "glm" (or even a simple numeric/ts vector), `segmented` tries to estimate a new regression model having broken-line relationships with the variables specified in `seg.Z`. A segmented (or broken-line) relationship is defined by the slope parameters and the break-points where the linear relation changes. The number of breakpoints of each segmented relationship is fixed via the `psi` argument, where initial values for the break-points must be specified. The model is estimated simultaneously yielding point estimates and relevant approximate standard errors of all the model parameters, including the break-points.

Since version 0.2-9.0 `segmented` implements the bootstrap restarting algorithm described in Wood (2001). The bootstrap restarting is expected to escape the local optima of the objective function when the segmented relationship is flat and the log likelihood can have multiple local optima.

Since version 0.5-0.0 the default method `segmented.default` has been added to estimate segmented relationships in general (besides "lm" and "glm" fits) regression models, such as Cox regression or quantile regression (for a single percentile). The objective function to be minimized is the (minus) value extracted by the `logLik` function or it may be passed on via the `fn.obj` argument in `seg.control`. See example below. While the default method is expected to work with any regression fit (where the usual `coef()`, `update()`, and `logLik()` returns appropriate results), it is not recommended for "lm" or "glm" fits (as `segmented.default` is slower than the specific methods `segmented.lm` and `segmented.glm`), although final results are the same. However the object returned by `segmented.default` is *not* of class "segmented", as currently the segmented methods are not guaranteed to work for 'generic' (i.e., besides "lm" and "glm") regression fits. The user could try each "segmented" method on the returned object by calling it explicitly (e.g. via `plot.segmented()` or `confint.segmented()` wherein the regression coefficients and relevant covariance matrix have to be specified, see `.coef` and `.vcov` in `plot.segmented()`, `confint.segmented()`, `slope()`).

Value

The returned object depends on the last component returned by `seg.control`. If `last=TRUE`, the default, `segmented` returns an object of class "segmented" which inherits from the class "lm" or "glm" depending on the class of `obj`. Otherwise a list is returned, where the last component is the fitted model at the final iteration, see [seg.control](#).

An object of class "segmented" is a list containing the components of the original object `obj` with additionally the followings:

<code>psi</code>	estimated break-points and relevant (approximate) standard errors
------------------	---

<code>it</code>	number of iterations employed
<code>epsilon</code>	difference in the objective function when the algorithm stops
<code>model</code>	the model frame
<code>psi.history</code>	a list or a vector including the breakpoint estimates at each step
<code>seed</code>	the integer vector containing the seed just before the bootstrap resampling. Returned only if bootstrap restart is employed
<code>..</code>	Other components are not of direct interest of the user

Warning

It is well-known that the log-likelihood function for the break-point may be not concave, especially for poor clear-cut kink-relationships. In these circumstances the initial guess for the break-point, i.e. the `psi` argument, must be provided with care. For instance visual inspection of a, possibly smoothed, scatter-plot is usually a good way to obtain some idea on breakpoint location. However bootstrap restarting, implemented since version 0.2-9.0, is relatively more robust to starting values specified in `psi`. Alternatively an automatic procedure may be implemented by specifying `psi=NA` and `fix.npsi=FALSE` in `seg.control`: experience suggests to increase the number of iterations via `it.max` in `seg.control()`. This automatic procedure, however, is expected to overestimate the number of breakpoints.

Note

1. The algorithm will start if the `it.max` argument returned by `seg.control` is greater than zero. If `it.max=0` `segmented` will estimate a new linear model with break-point(s) fixed at the values reported in `psi`.
2. In the returned fit object, 'U.' is put before the name of the segmented variable to mean the difference-in-slopes coefficient.
3. Methods specific to the class "segmented" are
 - `print.segmented`
 - `summary.segmented`
 - `print.summary.segmented`
 - `plot.segmented`
 - `lines.segmented`
 - `confint.segmented`
 - `vcov.segmented`
 - `predict.segmented`
 - `points.segmented`
 - `coef.segmented`

Others are inherited from the class "lm" or "glm" depending on the class of obj.

Author(s)

Vito M. R. Muggeo, <vito.muggeo@unipa.it>

References

Muggeo, V.M.R. (2003) Estimating regression models with unknown break-points. *Statistics in Medicine* **22**, 3055–3071.

Muggeo, V.M.R. (2008) Segmented: an R package to fit regression models with broken-line relationships. *R News* **8/1**, 20–25.

See Also

[lm](#), [glm](#)

Examples

```
set.seed(12)
xx<-1:100
zz<-runif(100)
yy<-2+1.5*pmax(xx-35,0)-1.5*pmax(xx-70,0)+15*pmax(zz-.5,0)+rnorm(100,0,2)
dati<-data.frame(x=xx,y=yy,z=zz)
out.lm<-lm(y~x,data=dati)

#the simplest example: the starting model includes just 1 covariate
#.. and 1 breakpoint has to be estimated for that
o<-segmented(out.lm) #1 breakpoint for x

#the single segmented variable is not in the starting model and 1 breakpoint for that:
#.. you need to specify the variable via seg.Z, but no starting value for psi
o<-segmented(out.lm, seg.Z=~z)
#note the leftmost slope is constrained to be zero (since out.lm does not include z)

#2 segmented variables, 1 breakpoint each (again no need to specify npsi or psi)
o<-segmented(out.lm,seg.Z=~z+x)

#1 segmented variable, 2 breakpoints: you have to specify starting values (vector) for psi:
o<-segmented(out.lm,seg.Z=~x,psi=c(30,60), control=seg.control(display=FALSE))

#or by specifying just the *number* of breakpoints
#o<-segmented(out.lm,seg.Z=~x, npsi=2, control=seg.control(display=FALSE))

slope(o) #the slopes of the segmented relationship

#2 segmented variables: starting values requested via a named list
out.lm<-lm(y~z,data=dati)
o1<-update(o,seg.Z=~x+z,psi=list(x=c(30,60),z=.3))
#or by specifying just the *number* of breakpoints
#o1<-update(o,seg.Z=~x+z, npsi=c(x=2,z=1))

#the default method leads to the same results (but it is slower)
```

```

#o1<-segmented.default(out.lm,seg.Z=~x+z,psi=list(x=c(30,60),z=.3))
#o1<-segmented.default(out.lm,seg.Z=~x+z,psi=list(x=c(30,60),z=.3),
#   control=seg.control(fn.obj="sum(x$residuals^2)"))

#automatic procedure to estimate breakpoints in the covariate x (starting from K quantiles)
# Hint: increases number of iterations. Notice: bootstrap restart is not allowed!
#o<-segmented.lm(out.lm,seg.Z=~x+z,psi=list(x=NA,z=.3),
#   control=seg.control(fix.npsi=FALSE, n.boot=0, tol=1e-7, it.max = 50, K=5, display=TRUE))

#assess the progress of the breakpoint estimates throughout the iterations
## Not run:
par(mfrow=c(1,2))
draw.history(o, "x")
draw.history(o, "z")

## End(Not run)
#try to increase the number of iterations and re-assess the
#convergence diagnostics

# A simple segmented model with continuous responses and no linear covariates
# No need to fit the starting lm model:
segmented(yy, npsi=2)
#if seg.Z is unspecified, the segmented variable is taken as 1:n/n

# An example using the Arima method:
## Not run:
n<-50
idt <-1:n #the time index

mu<-50-idt +1.5*pmax(idt-30,0)
set.seed(6969)
y<-mu+arima.sim(list(ar=.5),n)*3.5

o<-arima(y, c(1,0,0), xreg=idt)
os1<-segmented(o, ~idt, control=seg.control(display=TRUE))

#note using the .coef argument is mandatory!
slope(os1, .coef=os1$coef)
plot(y)
plot(os1, add=TRUE, .coef=os1$coef, col=2)

## End(Not run)

#####Three examples using the default method:

#==> 1. Cox regression with a segmented relationship
## Not run:
library(survival)

```

```

data(stanford2)

o<-coxph(Surv(time, status)~age, data=stanford2)
os<-segmented(o, ~age, psi=40) #estimate the breakpoint in the age effect
summary(os) #actually it means summary.coxph(os)
plot(os) #it does not work
plot.segmented(os) #call explicitly plot.segmented() to plot the fitted piecewise lines

# ==> 2. Linear mixed model via the nlme package
dati$g<-gl(10,10) #the cluster 'id' variable
library(nlme)
o<-lme(y~x+z, random=~1|g, data=dati)
os<-segmented.default(o, ~x+z, npsi=list(x=2, z=1))

#summarizing results (note the '.coef' argument)
slope(os, .coef=fixef(os))
plot.segmented(os, "x", .coef=fixef(os), conf.level=.95)
confint.segmented(os, "x", .coef=fixef(os))
dd<-data.frame(x=c(20,50),z=c(.2,.6), g=1:2)
predict.segmented(os, newdata=dd, .coef=fixef(os))

# ==> 3. segmented quantile regression via the quantreg package
library(quantreg)
data(Mammals)
y<-with(Mammals, log(speed))
x<-with(Mammals, log(weight))
o<-rq(y~x, tau=.9)
os<-segmented.default(o, ~x) #it does NOT work. It cannot compute the vcov matrix..

#Let's define the vcov.rq function.. (I don't know if it is the best option..)
vcov.rq<-function(x,...) {
  V<-summary(x,cov=TRUE,se="nid",...)$cov
  rownames(V)<-colnames(V)<-names(x$coef)
  V}

os<-segmented.default(o, ~x) #now it does work
plot.segmented(os, res=TRUE, col=2, conf.level=.95)

## End(Not run)

```

Description

This function selects the number of breakpoints of the segmented relationship according to the BIC criterion or sequential hypothesis testing.

Usage

```
segmented(olm, seg.Z, alpha = 0.05, type = c("score", "davies", "bic", "aic"),
  control = seg.control(), return.fit = TRUE, bonferroni = FALSE, Kmax = 2, msg=TRUE)
```

Arguments

olm	A starting lm or glm object
seg.Z	A one-side formula for the segmented variable. Only one term can be included, and it can be omitted if olm includes just one covariate.
alpha	The fixed type I error probability.
type	Which criterion should be used? Currently via sequential hypothesis testing score and davies, only Kmax=2 is allowed. Using the BIC allows to compare models with more than 2 breakpoints.
control	See seg.control .
return.fit	If TRUE, the fitted model (with the selected number of breakpoints according to type) is returned.
bonferroni	If TRUE, the Bonferroni correction is employed, i.e. alpha/Kmax is always taken as threshold value to reject or not. If FALSE, alpha is used in the second level of hypothesis testing.
Kmax	The maximum number of breakpoints being tested. If type='bic' any integer value can be specified, otherwise at most Kmax=2 breakpoints can be tested via the Score or Davies statistics.
msg	If FALSE the final fit is returned silently with the selected number of breakpoints, otherwise the message including information about the selection procedure (i.e. the BIC values) is printed.

Details

The function uses properly the functions `segmented`, `pscore.test` or `davies.test` to select the 'right' number of breakpoints.

Value

The returned object depends on argument `return.fit`. If FALSE, the returned object is a list with some information on the compared models (i.e. the BIC values), otherwise a classical segmented object with the component `selection.psi` including the aforementioned information. See [segmented](#) for details.

Note

This is a sperimental function. Please use that with caution.

Author(s)

Vito Muggeo

References

Muggeo V (2020) Selecting number of breakpoints in segmented regression: implementation in the R package segmented <https://www.researchgate.net/publication/343737604>

See Also

[segmented](#), [pscore.test](#), [davies.test](#)

Examples

```
## from ?segmented
## Not run:
set.seed(12)
xx<-1:100
zz<-runif(100)
yy<-2+1.5*pmax(xx-35,0)-1.5*pmax(xx-70,0)+15*pmax(zz-.5,0)+rnorm(100,0,2)
dati<-data.frame(x=xx,y=yy,z=zz)
out.lm<-lm(y~x,data=dati)

os<-selgmented(out.lm) ## selects number of breakpoints via the Score test

os <-selgmented(out.lm, Kmax=3, type="bic") #BIC-based selection

## End(Not run)
```

slope

Slope estimates from segmented relationships

Description

Computes the slopes of each ‘segmented’ relationship in the fitted model.

Usage

```
slope(ogg, parm, conf.level = 0.95, rev.sgn=FALSE,
      APC=FALSE, .vcov=NULL, .coef=NULL,
      use.t=NULL, ..., digits = max(4, getOption("digits") - 2))
```

Arguments

ogg	an object of class "segmented", returned by any segmented method.
parm	the segmented variable whose slopes have to be computed. If missing all the segmented variables are considered.
conf.level	the confidence level required.
rev.sgn	vector of logicals. The length should be equal to the length of parm, but it is recycled otherwise. When TRUE it is assumed that the current parm is ‘minus’ the actual segmented variable, therefore the sign is reversed before printing. This is useful when a null-constraint has been set on the last slope.

APC	logical. If APC=TRUE the ‘annual percent changes’, i.e. $100 \times (\exp(\beta) - 1)$, are computed for each interval (β is the slope). Only point estimates and confidence intervals are returned.
.vcov	The <i>full</i> covariance matrix of estimates. If unspecified (i.e. NULL), the covariance matrix is computed internally by <code>vcov(ogg)</code> .
.coef	The regression parameter estimates. If unspecified (i.e. NULL), it is computed internally by <code>coef(ogg)</code> .
use.t	Which quantiles should be used to compute the confidence intervals? If NULL (default) the <i>t</i> distribution is used only for objects obtained by <code>segmented.lm</code> .
...	Further arguments to be passed on to <code>vcov.segmented</code> , such as <code>var.diff</code> and <code>is</code> . See Details in vcov.segmented and summary.segmented .
digits	controls number of digits in the returned output.

Details

To fit broken-line relationships, `segmented` uses a parameterization whose coefficients are not the slopes. Therefore given an object "segmented", `slope` computes point estimates, standard errors, t-values and confidence intervals of the slopes of each segmented relationship in the fitted model.

Value

`slope` returns a list of matrices. Each matrix represents a segmented relationship and its number of rows equal to the number of segments, while five columns summarize the results.

Note

The returned summary is based on limiting Gaussian distribution for the model parameters involved in the computations. Sometimes, even with large sample sizes such approximations are questionable (e.g., with small difference-in-slope parameters) and the results returned by `slope` might be unreliable. Therefore is responsibility of the user to gauge the applicability of such asymptotic approximations. Anyway, the t values may be not assumed for testing purposes and they should be used just as guidelines to assess the estimate uncertainty.

Author(s)

Vito M. R. Muggeo, <vito.muggeo@unipa.it>

References

Muggeo, V.M.R. (2003) Estimating regression models with unknown break-points. *Statistics in Medicine* **22**, 3055–3071.

See Also

See also [davies.test](#) and [pscore.test](#) to test for a nonzero difference-in-slope parameter.

Examples

```

set.seed(16)
x<-1:100
y<-2+1.5*pmax(x-35,0)-1.5*pmax(x-70,0)+rnorm(100,0,3)
out<-glm(y~1)
out.seg<-segmented(out,seg.Z=~x,psi=list(x=c(20,80)))
## the slopes of the three segments...
slope(out.seg)
rm(x,y,out,out.seg)
#
## an heteroscedastic example..
set.seed(123)
n<-100
x<-1:n/n
y<- -x+1.5*pmax(x-.5,0)+rnorm(n,0,1)*ifelse(x<=.5,.4,.1)
o<-lm(y~x)
oseg<-segmented(o,seg.Z=~x,psi=.6)
slope(oseg)
slope(oseg,var.diff=TRUE) #better CI

```

stagnant

Stagnant band height data

Description

The stagnant data frame has 28 rows and 2 columns.

Usage

```
data(stagnant)
```

Format

A data frame with 28 observations on the following 2 variables.

x log of flow rate in g/cm sec.

y log of band height in cm

Details

Bacon and Watts report that such data were obtained by R.A. Cook during his investigation of the behaviour of stagnant surface layer height in a controlled flow of water.

Source

Bacon D.W., Watts D.G. (1971) Estimating the transition between two intersecting straight lines. *Biometrika* **58**: 525 – 534.

Originally from the PhD thesis by R.A. Cook

Examples

```
data(stagnant)
## plot(stagnant)
```

summary.segmented	<i>Summarizing model fits for segmented regression</i>
-------------------	--

Description

summary method for class segmented.

Usage

```
## S3 method for class 'segmented'
summary(object, short = FALSE, var.diff = FALSE, p.df="p", .vcov=NULL, ...)

## S3 method for class 'summary.segmented'
print(x, short=x$short, var.diff=x$var.diff,
      digits = max(3, getOption("digits") - 3),
      signif.stars = getOption("show.signif.stars"),...)
```

Arguments

object	Object of class "segmented".
short	logical indicating if the 'short' summary should be printed.
var.diff	logical indicating if different error variances should be computed in each interval of the segmented variable, see Details. If .vcov is provided, var.diff is set to FALSE.
p.df	A character as a function of 'p' (number of parameters) and 'K' (number of groups or segments) affecting computations of the group-specific variance (and the standard errors) if var.diff=TRUE, see Details.
.vcov	Optional. The full covariance matrix for the parameter estimates. If provided, standard errors are computed (and displayed) according to this matrix.
x	a summary.segmented object produced by summary.segmented().
digits	controls number of digits printed in output.
signif.stars	logical, should stars be printed on summary tables of coefficients?
...	further arguments.

Details

If `short=TRUE` only coefficients of the segmented relationships are printed. If `var.diff=TRUE` and there is only one segmented variable, different error variances are computed in the intervals defined by the estimated breakpoints of the segmented variable. For the j th interval with n_j observations, the error variance is estimated via $RSS_j/(n_j - p)$, where RSS_j is the residual sum of squares in interval j , and p is the number of model parameters. This number to be subtracted from n_j can be changed via argument `p.df`. For instance `p.df="0"` uses $RSS_j/(n_j)$, and `p.df="p/K"` leads to $RSS_j/(n_j - p/K)$, where K is the number of groups (segments), and p/K can be interpreted as the average number of model parameter in that group.

Note `var.diff=TRUE` only affects the estimates covariance matrix. It does *not* affect the parameter estimates, neither the log likelihood and relevant measures, such as AIC or BIC. In other words, `var.diff=TRUE` just provides 'alternative' standard errors, probably appropriate when the error variances are different before/after the estimated breakpoints. Also p -values are computed using the t-distribution with 'naive' degrees of freedom (as reported in `object$df.residual`).

If `var.diff=TRUE` the variance-covariance matrix of the estimates is computed via the sandwich formula,

$$(X^T X)^{-1} X^T V X (X^T X)^{-1}$$

where V is the diagonal matrix including the different group-specific error variance estimates. Standard errors are the square root of the main diagonal of this matrix.

Value

A list (similar to one returned by `segmented.lm` or `segmented.glm`) with additional components:

<code>psi</code>	estimated break-points and relevant (approximate) standard errors
<code>Ttable</code>	estimates and standard errors of the model parameters. This is similar to the matrix <code>coefficients</code> returned by <code>summary.lm</code> or <code>summary.glm</code> , but without the rows corresponding to the breakpoints. Even the p -values relevant to the difference-in-slope parameters have been replaced by NA, since they are meaningless in this case, see davies.test .
<code>gap</code>	estimated coefficients, standard errors and t-values for the 'gap' variables
<code>cov.var.diff</code>	if <code>var.diff=TRUE</code> , the covariance matrix accounting for heteroscedastic errors.
<code>sigma.new</code>	if <code>var.diff=TRUE</code> , the square root of the estimated error variances in each interval.
<code>df.new</code>	if <code>var.diff=TRUE</code> , the residual degrees of freedom in each interval.

Author(s)

Vito M.R. Muggeo

See Also

[print.segmented](#), [davies.test](#)

Examples

```
##continues example from segmented()
# summary(segmented.model,short=TRUE)

## an heteroscedastic example..
# set.seed(123)
# n<-100
# x<-1:n/n
# y<- -x+1.5*pmax(x-.5,0)+rnorm(n,0,1)*ifelse(x<=.5,.4,.1)
# o<-lm(y~x)
# oseg<-segmented(o,seg.Z=~x,psi=.6)
# summary(oseg,var.diff=TRUE)$sigma.new
```

vcov.segmented	<i>Variance-Covariance Matrix for a Fitted Segmented Model</i>
----------------	--

Description

Returns the variance-covariance matrix of the parameters (including breakpoints) of a fitted segmented model object.

Usage

```
## S3 method for class 'segmented'
vcov(object, var.diff = FALSE, is = FALSE, ...)
```

Arguments

object	a fitted model object of class "segmented", returned by any segmented method.
var.diff	logical. If var.diff=TRUE and there is a single segmented variable, the covariance matrix is computed using a sandwich-type formula. See Details in summary.segmented .
is	logical. If TRUE, the <i>asymptotic</i> covariance matrix based on the idea of induced smoothing is returned. If is=TRUE, var.diff=FALSE is set. is=TRUE only works with segmented (g)lm fits.
...	additional arguments.

Details

The returned covariance matrix is based on an approximation of the nonlinear segmented term. Therefore covariances corresponding to breakpoints are reliable only in large samples and/or clear cut segmented relationships. If is=TRUE, the returned covariance matrix depends on the design matrix having the term $I(x > \psi)$ replaced by its smooth counterpart.

Value

The full matrix of the estimated covariances between the parameter estimates, including the breakpoints.

Note

`var.diff=TRUE` works when there is a single segmented variable.

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See Also

[summary.segmented](#)

Examples

```
##continues example from summary.segmented()
# vcov(oseg)
# vcov(oseg, var.diff=TRUE)
# vcov(oseg, is=TRUE)
```

Index

- * **datasets**
 - down, 11
 - plant, 15
 - stagnant, 41
- * **hplot**
 - plot.segmented, 16
- * **htest**
 - davies.test, 9
 - pscore.test, 22
 - slope, 39
- * **models**
 - predict.segmented, 20
 - print.segmented, 21
- * **nonlinear**
 - broken.line, 5
 - confint.segmented, 7
 - draw.history, 12
 - lines.segmented, 14
 - plot.segmented, 16
 - points.segmented, 19
 - seg.lm.fit, 29
 - segmented, 31
 - segmented-package, 2
- * **regression**
 - aapc, 4
 - broken.line, 5
 - confint.segmented, 7
 - draw.history, 12
 - intercept, 13
 - lines.segmented, 14
 - plot.segmented, 16
 - points.segmented, 19
 - predict.segmented, 20
 - seg.control, 26
 - seg.lm.fit, 29
 - segmented, 31
 - segmented-package, 2
 - slope, 39
 - summary.segmented, 42
 - vcov.segmented, 44
- aapc, 4
- broken.line, 5, 17, 21
- coef.segmented, 34
- coef.segmented (print.segmented), 21
- confint.segmented, 7, 34
- davies.test, 9, 23, 24, 39, 40, 43
- down, 11
- draw.history, 12
- glm, 35
- intercept, 13
- lines.segmented, 8, 14, 18, 34
- lm, 35
- plant, 15
- plot.segmented, 6, 15, 16, 19, 21, 34
- points, 15
- points.segmented, 15, 18, 19, 34
- predict.glm, 21
- predict.lm, 21
- predict.segmented, 6, 18, 20, 34
- print.segmented, 21, 34, 43
- print.summary.segmented, 22, 34
- print.summary.segmented (summary.segmented), 42
- pscore.test, 11, 22, 26, 39, 40
- pwr.seg, 24
- seg.Ar.fit (seg.lm.fit), 29
- seg.control, 3, 26, 32–34, 38
- seg.def.fit (seg.lm.fit), 29
- seg.glm.fit (seg.lm.fit), 29
- seg.lm.fit, 29
- seg.num.fit (seg.lm.fit), 29

segmented, [6](#), [8](#), [10](#), [18](#), [21](#), [31](#), [38](#), [39](#)
segmented-package, [2](#)
segmented.default, [3](#), [27](#)
segmented.glm, [27](#), [31](#)
segmented.lm, [27](#), [31](#)
segments, [15](#)
selgmented, [3](#), [27](#), [37](#)
slope, [14](#), [39](#)
stagnant, [41](#)
summary.segmented, [6](#), [7](#), [17](#), [22](#), [34](#), [40](#), [42](#),
[44](#), [45](#)

vcov.segmented, [6](#), [13](#), [34](#), [40](#), [44](#)