

Package ‘SiZer’

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Title SiZer: Significant Zero Crossings

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Depends R (>= 2.4.0), splines, boot

Suggests

Description Calculates and plots the SiZer map for scatterplot data.

A SiZer map is a way of examining when the p -th derivative of a scatterplot-smoother is significantly negative, possibly zero or significantly positive across a range of smoothing bandwidths.

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Arkansas

Time Series of Macroinvertebrates Abundance in the Arkansas River.

Description

A time series of 16 years (5 replicates per year) of mayfly (Ephemeroptera:Heptageniidae) abundance in the fall at the monitoring station AR1.

Usage

```
data(Arkansas)
```

Format

A data frame with 90 observations on the following 2 variables.

year a numeric vector

sqrt.mayflies a numeric vector of abundance values

Source

Sonderegger, D.L., Wang, H., Clements, W.H., and Noon, B.R. 2009. Using SiZer to detect thresholds in ecological data. *Frontiers in Ecology and the Environment* 7:190-195.

Examples

```
data(Arkansas)
plot(Arkansas)
```

bent.cable

Fits a bent-cable model to the given data

Description

Fits a bent-cable model to the given data by exhaustively searching the 2-dimensional parameter space to find the maximum likelihood estimators for α and γ .

Usage

```
bent.cable(x, y, grid.size=100)
```

Arguments

x The independent variable

y The dependent variable

grid.size How many α and γ values to examine. The total number of parameter combinations examined is grid.size squared.

Details

Fit the model which is essentially a piecewise linear model with a quadratic curve of length 2γ connecting the two linear pieces.

The reason for searching the space exhaustively is because the bent-cable model often has a likelihood surface with a very flat ridge instead of definite peak. While the exhaustive search is slow, at least it is possible to examine the contour plot of the likelihood surface.

Value

A list of 7 elements:

log.likelihood	A matrix of log-likelihood values.
SSE	A matrix of sum-of-square-error values.
alphas	A vector of alpha values examined.
gammas	A vector of gamma values examined.
alpha	The MLE estimate of alpha.
gamma	The MLE estimate of gamma.
model	The lm fit after <i>alpha</i> and <i>gamma</i> are known.

Author(s)

Derek Sonderegger

References

Chiu, G. S., R. Lockhart, and R. Routledge. 2006. Bent-cable regression theory and applications. *Journal of the American Statistical Association* 101:542-553.

Toms, J. D., and M. L. Lesperance. 2003. Piecewise regression: a tool for identifying ecological thresholds. *Ecology* 84:2034-2041.

See Also

[piecewise.linear](#)

Examples

```
data(Arkansas)
x <- Arkansas$year
y <- Arkansas$sqrt.mayflies

# For a more accurate estimate, increase grid.size
model <- bent.cable(x,y, grid.size=20)
plot(x,y)
x.grid <- seq(min(x), max(x), length=200)
lines(x.grid, predict(model, x.grid), col='red')
```

 locally.weighted.polynomial

Locally-Weighted Polynomial Regression Smoother

Description

Smooths the given bivariate data using kernel regression.

Usage

```
locally.weighted.polynomial(x, y, h = NA, x.grid = NA,
  degree = 1, kernel.type = "Normal")
```

Arguments

x	Vector of data for the independent variable
y	Vector of data for the dependent variable
h	The bandwidth for the kernel
x.grid	What x-values should the value of the smoother be calculated at.
degree	The degree of the polynomial to be fit at each x-value. The default is to fit a linear regression, ie degree=1.
kernel.type	What kernel to use. Valid choices are 'Normal', 'Epanechnikov', 'biweight', and 'triweight'

Details

The confidence intervals are created using the row-wise method of Hannig and Marron (2006).

Notice that the derivative to be estimated must be less than or equal to the degree of the polynomial initially fit to the data.

If the bandwidth is not given, the Sheather-Jones bandwidth selection method is used.

Value

Returns a `LocallyWeightedPolynomial` object that has the following elements:

data	A structure of the data used to generate the smoothing curve
h	The bandwidth used to generate the smoothing curve.
x.grid	The grid of x-values that we have estimated function value and derivative(s) for.
degrees.freedom	The effective sample size at each grid point
Beta	A matrix of estimated beta values. The number of rows is degrees+1, while the number of columns is the same as the length of x.grid. Notice that

$$\hat{f}(x_i) = \beta[1, i]$$

$$\hat{f}'(x_i) = \beta[2, i] * 1!$$

$$\hat{f}''(x_i) = \beta[3, i] * 2!$$

and so on...

Beta.var Matrix of estimated variances for Beta. Same structure as Beta.

Author(s)

Derek Sonderegger

References

Chaudhuri, P., and J. S. Marron. 1999. SiZer for exploration of structures in curves. *Journal of the American Statistical Association* 94 807-823.

Hannig, J., and J. S. Marron. 2006. Advanced distribution theory for SiZer. *Journal of the American Statistical Association* 101 484-499.

Sonderegger, D.L., Wang, H., Clements, W.H., and Noon, B.R. 2009. Using SiZer to detect thresholds in ecological data. *Frontiers in Ecology and the Environment* 7:190-195

See Also

[SiZer](#), [plot.LocallyWeightedPolynomial](#), [spm](#) in package 'SemiPar', [loess](#), [smooth.spline](#), [interpSpline](#)

Examples

```
data(Arkansas)
x <- Arkansas$year
y <- Arkansas$sqrt.mayflies

layout(cbind(1,2,3))
model <- locally.weighted.polynomial(x,y)
plot(model, main='Smoothed Function', xlab='Year', ylab='Sqrt.Mayflies')

model2 <- locally.weighted.polynomial(x,y,h=.5)
plot(model2, main='Smoothed Function', xlab='Year', ylab='Sqrt.Mayflies')

model3 <- locally.weighted.polynomial(x,y, degree=1)
plot(model3, derv=1, main='First Derivative', xlab='Year', ylab='1st Derivative')
```

piecewise.linear

Fit a piecewise linear model

Description

Fit a degree 1 spline with 1 knot point where the location of the knot point is unknown.

Usage

```
piecewise.linear(x, y, middle = 1, CI = FALSE,
bootstrap.samples = 1000, sig.level = 0.05)
```

Arguments

<code>x</code>	Vector of data for the x-axis.
<code>y</code>	Vector of data for the y-axis
<code>middle</code>	A scalar in $[0, 1]$. This represents the range that the change-point can occur in. 0 means the change-point must occur at the middle of the range of x-values. 1 means that the change-point can occur anywhere along the range of the x-values.
<code>CI</code>	Whether or not a bootstrap confidence interval should be calculated.
<code>bootstrap.samples</code>	The number of bootstrap samples to take.
<code>sig.level</code>	What significance level to use for the confidence intervals.

Details

The bootstrap samples are taken by resampling the raw data points. Often a more appropriate bootstrap sample would be to calculate the residuals and then add a randomly selected residual to each y-value.

Value

A list of 5 elements is returned:

<code>change.point</code>	The estimate of α .
<code>model</code>	The resulting <code>lm</code> object once α is known.
<code>x</code>	The x-values used.
<code>y</code>	The y-values used.
<code>CI</code>	Whether or not the confidence interval was calculated.
<code>intervals</code>	If the CIs were calculated, this is a matrix of the upper and lower intervals.
<code>...</code>	

Author(s)

Derek Sonderegger

References

Chiu, G. S., R. Lockhart, and R. Routledge. 2006. Bent-cable regression theory and applications. *Journal of the American Statistical Association* 101:542-553.

Toms, J. D., and M. L. Lesperance. 2003. Piecewise regression: a tool for identifying ecological thresholds. *Ecology* 84:2034-2041.

See Also

~~objects to See Also as [help](#), ~~~

Examples

```
data(Arkansas)
x <- Arkansas$year
y <- Arkansas$sqrt.mayflies

model <- piecewise.linear(x,y, CI=FALSE)
plot(model)
print(model)
predict(model, 2001)
```

```
plot.LocallyWeightedPolynomial
Plot a LocallyWeightedPolynomial object
```

Description

Creates a plot of an object created by `locally.weighted.polynomial`.

Usage

```
## S3 method for class 'LocallyWeightedPolynomial'
plot(x, deriv = 0, CI.method = 2,
     alpha = 0.05, use.ess = TRUE, draw.points = TRUE, ...)
```

Arguments

<code>x</code>	LocallyWeightedPolynomial object
<code>deriv</code>	Derivative to be plotted. Default is 0 - which plots the smoothed function.
<code>CI.method</code>	What method should be used to calculate the confidence interval about the estimated line. The methods are from Hannig and Marron (2006), where 1 is the point-wise estimate, and 2 is the row-wise estimate.
<code>alpha</code>	The CI has a $1-\alpha/2$ level of significance.
<code>use.ess</code>	ESS stands for the estimated sample size. If at any point along the x-axis, the ESS is too small, then we will not plot unless <code>use.ess=FALSE</code> .
<code>draw.points</code>	Should the data points be included in the graph?
<code>...</code>	Additional arguments to be passed to the graphing functions.

Author(s)

Derek Sonderegger

References

Hannig, J., and J. S. Marron. 2006. Advanced distribution theory for SiZer. *Journal of the American Statistical Association* 101:484-499.

Sonderregger, D.L., Wang, H., Clements, W.H., and Noon, B.R. 2009. Using SiZer to detect thresholds in ecological data. *Frontiers in Ecology and the Environment* 7:190-195.

See Also

[locally.weighted.polynomial](#)

Examples

```
data('Arkansas')
x <- Arkansas$year
y <- Arkansas$sqrt.mayflies

model <- locally.weighted.polynomial(x,y)
plot(model)

model <- locally.weighted.polynomial(x,y,degree=2)
plot(model, derv=1)
plot(model, derv=2)
```

plot.SiZer

Plot a SiZer map

Description

Plot a SiZer object that was created using SiZer().

Usage

```
## S3 method for class 'SiZer'.
## S3 method for class 'SiZer'
plot(x, ylab = expression(log[10](h)),
colorlist = c("red", "purple", "blue", "grey"), ...)
```

Arguments

x	An object created using SiZer()
ylab	What the y-axis should be labeled.
colorlist	What colors should be used. This is a vector that corresponds to 'decreasing', 'possibly zero', 'increasing', 'insufficient data'
...	Any other parameters to be passed to the function image.

Details

The white lines in the SiZer map give a graphical representation of the bandwidth. The horizontal distance between the lines is $2h$.

Value

None

Author(s)

Derek Sonderegger

References

Chaudhuri, P., and J. S. Marron. 1999. SiZer for exploration of structures in curves. *Journal of the American Statistical Association* 94:807-823.

Hannig, J., and J. S. Marron. 2006. Advanced distribution theory for SiZer. *Journal of the American Statistical Association* 101:484-499.

Sonderegger, D.L., Wang, H., Clements, W.H., and Noon, B.R. 2009. Using SiZer to detect thresholds in ecological data. *Frontiers in Ecology and the Environment* 7:190-195.

See Also

[SiZer](#), [locally.weighted.polynomial](#)

Examples

```
# data('Arkansas')
# x <- Arkansas$year
# y <- Arkansas$sqrt.mayflies

# Calculate the SiZer map for the first derivative
# SiZer.1 <- SiZer(x, y, h=c(.5,10), degree=1, derv=1)
# plot(SiZer.1)

# Calculate the SiZer map for the second derivative
# SiZer.2 <- SiZer(x, y, h=c(.5,10), degree=2, derv=2);
# plot(SiZer.2)

# By setting the grid.length larger, we get a more detailed SiZer
# map but it takes longer to compute.
#
# SiZer.3 <- SiZer(x, y, h=c(.5,10), grid.length=100, degree=1, derv=1)
# plot(SiZer.3)
```

SiZer

*Calculate SiZer Map***Description**

Calculates the SiZer map from a given set of X and Y variables.

Usage

```
SiZer(x, y, h=NA, x.grid=NA, degree=NA, derv=1, grid.length=41)
```

Arguments

x	data vector for the independent axis
y	data vector for the dependent axis
h	An integer representing how many bandwidths should be considered, or vector of length 2 representing the upper and lower limits h should take, or a vector of length greater than two indicating which bandwidths to examine.
x.grid	An integer representing how many bins to use along the x-axis, or a vector of length 2 representing the upper and lower limits the x-axis should take, or a vector of length greater than two indicating which x-values the derivative should be evaluated at.
grid.length	The default length of the h.grid or x.grid if the length of either is not given.
derv	The order of derivative for which to make the SiZer map.
degree	The degree of the local weighted polynomial used to smooth the data. This must be greater than or equal to derv.

Details

SiZer stands for the Significant Zero crossings of the derivative. There are two dominate approaches in smoothing bivariate data: locally weighted regression or penalized splines. Both approaches require the use of a 'bandwidth' parameter that controls how much smoothing should be done. Unfortunately there is no uniformly best bandwidth selection procedure. SiZer (Chaudhuri and Marron, 1999) is a procedure that looks across a range of bandwidths and classifies the p-th derivative of the smoother into one of three states: significantly increasing (blue), possibly zero (purple), or significantly negative (red).

Value

Returns an SiZer object which has the following components:

x.grid	Vector of x-values at which the derivative was evaluated.
h.grid	Vector of bandwidth values for which a smoothing function was calculated.
slopes	Matrix of what category a particular x-value and bandwidth falls into (Increasing=1, Possibly Zero=0, Decreasing=-1, Not Enough Data=2).

Author(s)

Derek Sonderegger

References

Chaudhuri, P., and J. S. Marron. 1999. SiZer for exploration of structures in curves. *Journal of the American Statistical Association* 94:807-823.

Hannig, J., and J. S. Marron. 2006. Advanced distribution theory for SiZer. *Journal of the American Statistical Association* 101:484-499.

Sonderegger, D.L., Wang, H., Clements, W.H., and Noon, B.R. 2009. Using SiZer to detect thresholds in ecological data. *Frontiers in Ecology and the Environment* 7:190-195.

See Also

[plot.SiZer](#), [locally.weighted.polynomial](#)

Examples

```
data('Arkansas')
x <- Arkansas$year
y <- Arkansas$sqrt.mayflies

plot(x,y)

# Calculate the SiZer map for the first derivative
SiZer.1 <- SiZer(x, y, h=c(.5,10), degree=1, derv=1)
plot(SiZer.1)

# Calculate the SiZer map for the second derivative
SiZer.2 <- SiZer(x, y, h=c(.5,10), degree=2, derv=2);
plot(SiZer.2)

# By setting the grid.length larger, we get a more detailed SiZer
# map but it takes longer to compute.
#
# SiZer.3 <- SiZer(x, y, h=c(.5,10), grid.length=100, degree=1, derv=1)
# plot(SiZer.3)
```

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