

Package ‘robustgam’

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Author Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee

Maintainer Raymond K. W. Wong <raymondkww.dev@gmail.com>

Description This package provides robust estimation for generalized additive models. It implements a fast and stable algorithm in Wong, Yao and Lee (2013). The implementation also contains three automatic selection methods for smoothing parameter. They are designed to be robust to outliers. For more details, see Wong, Yao and Lee (2013).

License GPL (>= 2)

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robustgam-package *Robust Estimation for Generalized Additive Models*

Description

This package provides robust estimation for generalized additive models. It implements a fast and stable algorithm in Wong, Yao and Lee (2013). The implementation also contains three automatic selection methods for smoothing parameter. They are designed to be robust to outliers. For more details, see Wong, Yao and Lee (2013).

Details

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Author(s)

Raymond K. W. Wong Maintainer: Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

Examples

```
# see example in function robustgam
```

cplot.robustgam *Plot of Component Smooth Functions from robustgam fit*

Description

This function plots the component smooth functions.

Usage

```
cplot.robustgam(fit, ranges, len=100)
```

Arguments

fit the robustgam fit
 ranges matrix with each column containing the range of predictor in each plot
 len grid size for the plot; it is used for all component smooth functions

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#), [pred.robustgam](#)

Examples

```
# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

# robust GAM fit
robustfit <- robustgam(x, y, family=true.family, p=3, c=1.6, sp=0.000143514, show.msg=FALSE,
  smooth.basis='tp')
## the smoothing parameter is selected by the RBIC, the command is:
# robustfit.gic <- robustgam.GIC.optim(x, y, family=true.family, p=3, c=1.6, show.msg=FALSE,
```

```

# count.lim=400, smooth.basis='tp', lsp.initial=log(1e-2) ,lsp.min=-15, lsp.max=10,
# gic.constant=log(n), method="L-BFGS-B"); robustfit <- robustfit.gic$optim.fit

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))

# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
      lty=c(1,1,1))

# plot component smooth function
cplot.robustgam(robustfit, ranges=range(x))

```

<code>pred.robustgam</code>	<i>Prediction method for robustgam</i>
-----------------------------	----------------------------------------

Description

A prediction function for output of [robustgam](#).

Usage

```
pred.robustgam(fit, data, type="response")
```

Arguments

<code>fit</code>	fit object of robustgam
<code>data</code>	a data.frame object. Call the variable X if there is only one covariate. Otherwise, call the first variable X1, the second one X2, and so on.
<code>type</code>	type of output

Value

<code>predict.comp</code>	a matrix containing the individual additive components for each covariates.
<code>predict.values</code>	the type of output required by <code>type</code> . For example, if <code>type="response"</code> , the predicted values of the data is outputted.

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam](#)

Examples

```
# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

# robust GAM fit
robustfit <- robustgam(x, y, family=true.family, p=3, c=1.6, sp=0.000143514, show.msg=FALSE,
  smooth.basis='tp')

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))
```

```
# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
      lty=c(1,1,1))
```

robustgam

*Pseudo Data Algorithm for Robust Estimation of GAMs***Description**

This function implements a fast and stable algorithm developed in *Wong, Yao and Lee (2013)* for robust estimation of generalized additive models. Currently, this implementation only covers binomial and poisson distributions. This function does not choose smoothing parameter by itself and the user has to specify it manually. For implementation with automatic smoothing parameter selection, see [robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#). For prediction, see [pred.robustgam](#).

Usage

```
robustgam(X, y, family, p=3, K=30, c=1.345, sp=-1, show.msg=FALSE, count.lim=200,
          w.count.lim=50, smooth.basis="tp", wx=FALSE)
```

Arguments

X	a vector or a matrix (each covariate form a column) of covariates
y	a vector of responses
family	A family object specifying the distribution and the link function. See glm and family .
p	order of the basis. It depends on the option of smooth.basis.
K	number of knots of the basis; dependent on the option of smooth.basis.
c	tunning parameter for Huber function; a smaller value of c corresponds to a more robust fit. It is recommended to set as 1.2 and 1.6 for binomial and poisson distribution respectively.
sp	a vector of smoothing parameter. If only one value is specified, it will be used for all smoothing parameters.
show.msg	If show.msg=T, progress is displayed.
count.lim	maximum number of iterations of the whole algorithm
w.count.lim	maximum number of updates on the weight. It corresponds to zeta in <i>Wong, Yao and Lee (2013)</i>
smooth.basis	the specification of basis. Four choices are available: "tp" = thin plate regression spline, "cr" = cubic regression spline, "ps" = P-splines, "tr" = truncated power spline. For more details, see smooth.construct .
wx	If wx=T, robust weight on the covariates are applied. For details, see Real Data Example in <i>Wong, Yao and Lee (2013)</i>

Value

fitted.values	fitted values
initial.fitted	the starting values of the algorithm
beta	estimated coefficients (corresponding to the basis)
B	the basis: fitted linear estimator = $B \backslash * \beta$
sD	for internal use
basis	the smooth construct object. For more details, see smooth.construct
converge	If converge=T, the algorithm converged.
w	for internal use
family	the family object
wx	Indicate whether robust weight on covariates is applied
beta.fit	for internal use

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#), [pred.robustgam](#)

Examples

```
# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
```

```

y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

# robust GAM fit
robustfit <- robustgam(x, y, family=true.family, p=3, c=1.6, sp=0.000143514, show.msg=FALSE,
  smooth.basis='tp')
## the smoothing parameter is selected by the RBIC, the command is:
# robustfit.gic <- robustgam.GIC.optim(x, y, family=true.family, p=3, c=1.6, show.msg=FALSE,
#   count.lim=400, smooth.basis='tp', lsp.initial=log(1e-2) ,lsp.min=-15, lsp.max=10,
#   #   gic.constant=log(n), method="L-BFGS-B"); robustfit <- robustfit.gic$optim.fit

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))

# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
  lty=c(1,1,1))

```

robustgam.CV

Smoothing parameter selection by robust cross validation

Description

This function combines the [robustgam](#) with automatic smoothing parameter selection. The smoothing parameter is selected through robust cross validation criterion described in *Wong, Yao and Lee (2013)*. The criterion is designed to be robust to outliers. This function uses grid search to find the smoothing parameter that minimizes the criterion.

Usage

```

robustgam.CV(X, y, family, p=3, K=30, c=1.345, show.msg=FALSE, count.lim=200,
  w.count.lim=50, smooth.basis="tp", wx=FALSE, sp.min=1e-7, sp.max=1e-3,
  len=50, show.msg.2=TRUE, ngroup=length(y), seed=12345)

```


Arguments

X	a vector or a matrix (each covariate form a column) of covariates
y	a vector of responses
family	A family object specifying the distribution and the link function. See glm and family .
p	order of the basis. It depends on the option of smooth.basis.
K	number of knots of the basis; dependent on the option of smooth.basis.
c	tunning parameter for Huber function; a smaller value of c corresponds to a more robust fit. It is recommended to set as 1.2 and 1.6 for binomial and poisson distribution respectively.
show.msg	If show.msg=T, progress of robustgam is displayed.
count.lim	maximum number of iterations of the whole algorithm
w.count.lim	maximum number of updates on the weight. It corresponds to zeta in <i>Wong, Yao and Lee (2013)</i>
smooth.basis	the specification of basis. Four choices are available: "tp" = thin plate regression spline, "cr" = cubic regression spline, "ps" = P-splines, "tr" = truncated power spline. For more details, see smooth.construct .
wx	If wx=T, robust weight on the covariates are applied. For details, see Real Data Example in <i>Wong, Yao and Lee (2013)</i>
sp.min	A vector of minimum values of the searching range for smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
sp.max	A vector of maximum values of the searching range for smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
len	A vector of grid sizes. If only one value is specified, it will be used for all smoothing parameters.
show.msg.2	If show.msg.2=T, progress of the grid search is displayed.
ngroup	number of group used in the cross validation. If ngroup=length(y), full cross validation is implemented. If ngroup=2, two-fold cross validation is implemented.
seed	The seed for random generator used in generating partitions.

Value

fitted.values	fitted values (of the optimum fit)
initial.fitted	the starting values of the algorithm (of the optimum fit)
beta	estimated coefficients (corresponding to the basis) (of the optimum fit)
optim.index	the index of the optimum fit
optim.index2	the index of the optimum fit in another representation: optim.ndex2=arrayInd(optim.index, len)
optim.criterion	the optimum value of robust cross validation criterion

optim.sp	the optimum value of the smoothing parameter
criteria	the values of criteria for all fits during grid search
sp	the grid of smoothing parameter
optim.fit	the robustgam fit object of the optimum fit. It is handy for applying the prediction method.

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#), [pred.robustgam](#)

Examples

```
# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

## Not run:

# robust GAM fit
robustfit.gic <- robustgam.CV(x, y, family=true.family, p=3, c=1.6, show.msg=FALSE,
```

```

count.lim=400, smooth.basis='tp',ngroup=5); robustfit <- robustfit.gic$optim.fit

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))

# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
      lty=c(1,1,1))

## End(Not run)

```

robustgam.GIC

Smoothing parameter selection by GIC (grid search)

Description

This function combines the `robustgam` with automatic smoothing parameter selection. The smoothing parameter is selected through generalized information criterion (GIC) described in *Wong, Yao and Lee (2013)*. The GIC is designed to be robust to outliers. There are two criteria for user to choose from: Robust AIC and Robust BIC. The robust BIC usually gives smoother surface than robust AIC. This function uses grid search to find the smoothing parameter that minimizes the criterion.

Usage

```

robustgam.GIC(X, y, family, p=3, K=30, c=1.345, show.msg=FALSE, count.lim=200,
              w.count.lim=50,smooth.basis="tp", wx=FALSE, sp.min=1e-7, sp.max=1e-3,
              len=50, show.msg.2=TRUE, gic.constant=log(length(y)))

```

Arguments

X	a vector or a matrix (each covariate form a column) of covariates
y	a vector of responses
family	A family object specifying the distribution and the link function. See <code>glm</code> and <code>family</code> .
p	order of the basis. It depends on the option of <code>smooth.basis</code> .
K	number of knots of the basis; dependent on the option of <code>smooth.basis</code> .

<code>c</code>	tunning parameter for Huber function; a smaller value of <code>c</code> corresponds to a more robust fit. It is recommended to set as 1.2 and 1.6 for binomial and poisson distribution respectively.
<code>show.msg</code>	If <code>show.msg=T</code> , progress of robustgam is displayed.
<code>count.lim</code>	maximum number of iterations of the whole algorithm
<code>w.count.lim</code>	maximum number of updates on the weight. It corresponds to <code>zeta</code> in <i>Wong, Yao and Lee (2013)</i>
<code>smooth.basis</code>	the specification of basis. Four choices are available: "tp" = thin plate regression spline, "cr" = cubic regression spline, "ps" = P-splines, "tr" = truncated power spline. For more details, see smooth.construct .
<code>wx</code>	If <code>wx=T</code> , robust weight on the covariates are applied. For details, see Real Data Example in <i>Wong, Yao and Lee (2013)</i>
<code>sp.min</code>	A vector of minimum values of the searching range for smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
<code>sp.max</code>	A vector of maximum values of the searching range for smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
<code>len</code>	A vector of grid sizes. If only one value is specified, it will be used for all smoothing parameters.
<code>show.msg.2</code>	If <code>show.msg.2=T</code> , progress of the grid search is displayed.
<code>gic.constant</code>	If <code>gic.constant=log(length(y))</code> , robust BIC is used. If <code>gic.constant=2</code> , robust AIC is used.

Value

<code>fitted.values</code>	fitted values (of the optimum fit)
<code>initial.fitted</code>	the starting values of the algorithm (of the optimum fit)
<code>beta</code>	estimated coefficients (corresponding to the basis) (of the optimum fit)
<code>optim.index</code>	the index of the optimum fit
<code>optim.index2</code>	the index of the optimum fit in another representation: <code>optim.index2=arrayInd(optim.index, len)</code>
<code>optim.gic</code>	the optimum value of robust AIC or robust BIC
<code>optim.sp</code>	the optimum value of the smoothing parameter
<code>fit.list</code>	a list object containing all fits during the grid search
<code>gic</code>	the values of GIC for all fits during grid search
<code>gic.comp1</code>	for internal use
<code>gic.comp2</code>	for internal use
<code>sp</code>	the grid of smoothing parameter
<code>gic.constant</code>	the <code>gic.constant</code> specified in the input
<code>optim.fit</code>	the robustgam fit object of the optimum fit. It is handy for applying the prediction method.

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#), [pred.robustgam](#)

Examples

```
# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

## Not run:

# robust GAM fit
robustfit.gic <- robustgam.GIC(x, y, family=true.family, p=3, c=1.6, show.msg=FALSE,
  count.lim=400, smooth.basis='tp'); robustfit <- robustfit.gic$optim.fit

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
```

```

nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))

# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
      lty=c(1,1,1))

## End(Not run)

```

robustgam.GIC.optim *Smoothing parameter selection by GIC (by optim)*

Description

This function is the same as [robustgam.GIC](#), except that the R internal optimization function [optim](#) is used to find the smoothing parameter that minimizes the RAIC or RBIC criterion.

Usage

```

robustgam.GIC.optim(X, y, family, p=3, K=30, c=1.345, show.msg=FALSE, count.lim=200,
  w.count.lim=50, smooth.basis="tp", wx=FALSE, lsp.initial=log(1e-4),
  lsp.min=-20, lsp.max=10, gic.constant=log(length(y)),
  method="L-BFGS-B", optim.control=list(trace=1))

```

Arguments

X	a vector or a matrix (each covariate form a column) of covariates
y	a vector of responses
family	A family object specifying the distribution and the link function. See glm and family .
p	order of the basis. It depends on the option of smooth.basis.
K	number of knots of the basis; dependent on the option of smooth.basis.
c	tunning parameter for Huber function; a smaller value of c corresponds to a more robust fit. It is recommended to set as 1.2 and 1.6 for binomial and poisson distribution respectively.
show.msg	If show.msg=T, progress of robustgam is displayed.
count.lim	maximum number of iterations of the whole algorithm
w.count.lim	maximum number of updates on the weight. It corresponds to zeta in <i>Wong, Yao and Lee (2013)</i>
smooth.basis	the specification of basis. Four choices are available: "tp" = thin plate regression spline, "cr" = cubic regression spline, "ps" = P-splines, "tr" = truncated power spline. For more details, see smooth.construct .

wx	If wx=T, robust weight on the covariates are applied. For details, see Real Data Example in <i>Wong, Yao and Lee (2013)</i>
lsp.initial	A vector of initial values of the <i>log</i> of smoothing parameters used to start the optimization algorithm.
lsp.min	a vector of minimum values of the searching range for the <i>log</i> of smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
lsp.max	a vector of maximum values of the searching range for the <i>log</i> of smoothing parameters. If only one value is specified, it will be used for all smoothing parameters.
gic.constant	If gic.constant=log(length(y)), robust BIC is used. If gic.constant=2, robust AIC is used.
method	method of optimization. For more details, see optim
optim.control	setting for optim . For more details, see optim

Value

fitted.values	fitted values (of the optimum fit)
beta	estimated coefficients (corresponding to the basis) (of the optimum fit)
beta.fit	for internal use
gic	the optimum value of robust AIC or robust BIC
sp	the optimum value of the smoothing parameter
gic.optim	the output of optim
w	for internal use
gic.constant	the gic.constant specified in the input
optim.fit	the robustgam fit object of the optimum fit. It is handy for applying the prediction method.

Author(s)

Raymond K. W. Wong <raymondkww.dev@gmail.com>

References

Raymond K. W. Wong, Fang Yao and Thomas C. M. Lee (2013) *Robust Estimation for Generalized Additive Models*. *Journal of Graphical and Computational Statistics*, to appear.

See Also

[robustgam.GIC](#), [robustgam.GIC.optim](#), [robustgam.CV](#), [pred.robustgam](#)

Examples

```

# load library
library(robustgam)

# test function
test.fun <- function(x, ...) {
  return(2*sin(2*pi*(1-x)^2))
}

# some setting
set.seed(1234)
true.family <- poisson()
out.prop <- 0.05
n <- 100

# generating dataset for poisson case
x <- runif(n)
x <- x[order(x)]
true.eta <- test.fun(x)
true.mu <- true.family$linkinv(test.fun(x))
y <- rpois(n, true.mu) # for poisson case

# create outlier for poisson case
out.n <- trunc(n*out.prop)
out.list <- sample(1:n, out.n, replace=FALSE)
y[out.list] <- round(y[out.list]*runif(out.n,min=3,max=5)^(sample(c(-1,1),out.n,TRUE)))

## Not run:

# robust GAM fit
robustfit.gic <- robustgam.GIC.optim(x, y, family=true.family, p=3, c=1.6, show.msg=FALSE,
  count.lim=400, smooth.basis='tp', lsp.initial=log(1e-2), lsp.min=-15, lsp.max=10,
  gic.constant=log(n), method="L-BFGS-B"); robustfit <- robustfit.gic$optim.fit

# ordinary GAM fit
nonrobustfit <- gam(y~s(x, bs="tp", m=3),family=true.family) # m = p for 'tp'

# prediction
x.new <- seq(range(x)[1], range(x)[2], len=1000)
robustfit.new <- pred.robustgam(robustfit, data.frame(X=x.new))$predict.values
nonrobustfit.new <- as.vector(predict.gam(nonrobustfit,data.frame(x=x.new),type="response"))

# plot
plot(x, y)
lines(x.new, true.family$linkinv(test.fun(x.new)), col="blue")
lines(x.new, robustfit.new, col="red")
lines(x.new, nonrobustfit.new, col="green")
legend(0.6, 23, c("true mu", "robust fit", "nonrobust fit"), col=c("blue","red","green"),
  lty=c(1,1,1))

```


End(Not run)

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