Package ‘assist’

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Description

The acid data frame has 112 rows and 4 columns of data derived based on the Eastern Lakes Survey of 1984 implemented by the Environmental Protection Agency of the USA.

Usage

data(acid)

Format

The data frame contains the following columns:

- ph a numeric vector of surface pH values.
- t1 a numeric vector of calcium concentrations in log10 milligrams per liter.
- x1, x2 numeric vectors of the lakes' geographic locations.

Details

112 lakes are extracted in the southern Blue Ridge mountains area. The surface pH values were recorded together with the calcium concentration and geographic locations.

Source

References

### alogit

**Calculate the Inverse Logit Transformation**

**Description**
Perform an inverse logit calculation

**Usage**
alogit(x)

**Arguments**
- **x**: a numeric value

**Value**
Returned is $e^x/(1 + e^x)$.

**Author(s)**
Chunlei ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

### anova.ssr

**Testing a Non-parametric Function Fitted via Smoothing Splines**

**Description**
For smoothing spline models with a single smoothing parameter, test the hypothesis that the unknown function lies in the null space using the local most powerful (LMP) test and a GCV or GML test.

**Usage**
```
## S3 method for class 'ssr'
anova(object, simu.size=100, ...)
```

**Arguments**
- **object**: an object of class "ssr" fitted with a single smoothing parameter.
- **simu.size**: an optional integer giving the number of simulations to calculate p-values based on simulation. Default is 100.
- **...**: other available arguments, currently unused.
Details

For Gaussian data with one smoothing parameter, test the hypothesis that the function is in the null space $H_0$, i.e. the parametric part of the fitted model is sufficient. Available are the LMP and GCV or GML methods. However, the p-values cannot be calculated analytically since the null distributions for these testing statistics under $H_0$ are unknown. Monte Carlo simulation is used to approximate the p-values for the LMP, and GCV (if spar="v") or GML (if spar="m") methods. Due to computation burden, the smoothing parameters are fixed at their estimate in the current calculation.

When spar="m", an approximate p-value based on a mixture of two Chi-square distributions is also provided for the GML test, which tends to be conservative (Pinherio and Bates, 2002).


Value

a list including test values.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

ssr, print.anova.ssr

Examples

data(acid)

# fit a partial thin-plate spline
temp <- ssr(ph~t1+x1+x2, rk=tp(t1), data=acid, spar="m")
anova(temp, 500)
Arosa

Monthly Mean Ozone Thickness in Arosa of Switzerland

Description

The Arosa data frame has 518 rows and 3 columns of data for monthly mean ozone thickness.

Usage

data(Arosa)

Format

The data frame contains the following columns:

- year: a vector of integers from 1 to 46 indicating the years when the measures were taken from 1926.
- month: a vector of integers from 1 to 12 representing the months in a year.
- thick: a numeric vector of mean ozone thickness (Dobson units).

Details

Monthly mean ozone thickness in Arosa, Switzerland was recorded from 1926-1971.

Source


bdiag

Construct a Block Diagonal Matrix

Description

Return a block diagonal matrix formed from the input list of matrices

Usage

bdiag(x)

Arguments

- x: a list of matrices

Value

Returned is a matrix of the form diag(x1, ..., xn) where n is the length of the list.
**bond**  
*Treasury and GE bonds*

**Description**

The bond data frame has 1234 rows and 5 columns of data derived from 144 General Electronic Company bonds and 78 Treasury bonds.

**Usage**

data(bond)

**Format**

The data frame contains the following columns:

name: a vector of index for individual bond
price: a numeric vector of current price
time: a numeric vector of future time points at which the payments are made
payment: a numeric vector of future payments
type: a vector of character strings identifying the groups, "govt" or "ge", which the individual bonds belong to.

**Source**

Bloomberg

**references**


---

**canadaTemp**  
*Monthly Mean Temperatures*

**Description**

The canadaTemp data frame has 420 rows and 3 columns of data for monthly mean temperatures in Canada

**Usage**

data(canadaTemp)
**Format**

The data frame contains the following columns:
- `temp` a numeric vector of mean temperatures at some stations in Canada.
- `month` a vector of integers from 1 to 12 representing the months in a year.
- `station` a vector of integers from 1 to 35 indicating the stations where the temperatures were recorded.

**Source**

The data set was downloaded from [http://www.psych.mcgill.ca/faculty/ramsay/datasets.html](http://www.psych.mcgill.ca/faculty/ramsay/datasets.html).

**References**


---

**Description**

The `chickenpox` data frame has 498 rows and 3 columns of data recording the number of Chickenpox occurrences in New York City.

**Usage**

data(chickenpox)

**Format**

The data frame contains the following columns:
- `count` the number of monthly reported Chickenpox cases.
- `month` a vector of integers from 1 to 12 representing the month for the reported cases.
- `year` a numeric vector representing the year when the cases were reported.

**Details**

This data frame contains monthly number of reported cases of chickenpox in New York City from 1931 to the first six months of 1972.

**Source**

chol.new

A Modified Cholesky Decomposition

Description
Returned a matrix forming Cholesky Decomposition

Usage
chol.new(Q)

Arguments
Q  a symmetric matrix, maybe non-positive.

Details
This is used internally as an extension of chol that works on a positive matrix.

Value
A matrix M such that $XX' = Q$.

See Also
chol

climate
Winter Average Temperatures

Description
The data frame climate, obtained from the Carbon Dioxide Information and Analysis Center, has 690 rows and 5 columns of data representing station winter temperature measurements.

Usage
data(climate)

Format
The data frame contains the following columns:
temp a numeric vector of temperatures in celsius.
lat, long numeric vectors identifying the latitudes and longitudes of the stations in.
lat.degree, long.degree numeric vectors identifying the latitudes and longitudes of the stations in degree.
Details
The station winter average temperatures were the averages of the December, January and February monthly average temperatures obtained from the Jones/Wigley data files obtainable from the CDIAC at Oak Ridge National Laboratory in the files ndp020r1/jonesnh.data.Z and ndp020r1/jonessh.dat.Z in the pbu directory at 128.219.24.36.

Source

dcrdr

Interface to Fortran Subroutine dcrdr

Description
Calculate some matrix operations needed to construct Bayesian confidence intervals

Usage

dcrdr(rkpk.obj, r)

Arguments

rkpk.obj an object returned from calling dsidr
r a matrix to evaluate reproducing kernels on grid points

Value
See the document for the corresponding Fortran subroutine.

deviance.ssr

Model Deviance

Description
Extract deviance from a fitted ssr object

Usage

## S3 method for class 'ssr'
deviance(object, residuals=FALSE, ...)

Arguments

object  a fitted ssr object.
residuals  a logical value. If 'TRUE', deviance residuals are returned. If 'FALSE', the sum of deviance residuals squares is returned. Default is FALSE.

...  other arguments, currently unused.

Details

This is a method for the function deviance for objects inheriting from class ssr.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

ssr

dmudr  Interface of dmudr subroutine in RKPACK

Description

To calculate a spline estimate with multiple smoothing parameters

Usage

dmudr(y, q, s, weight = NULL, vmu = "v", theta = NULL, varht = NULL, tol = 0, init = 0, prec = 1e-06, maxit = 30)

Arguments

y  a numerical vector representing the response.
q  a list, or an array, of square matrices of the same order as the length of y, which are the reproducing kernels evaluated at the design points.
s  the design matrix of the null space $H_0$ of size (length-of-y, dim($H_0$)), with elements equal to the bases of $H_0$ evaluated at design points.
weight  a weight matrix for penalized weighted least-square: $(y - f)'W(y - f) + n\lambda J(f)$. Default is NULL for iid random errors.
vmu  a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. "u~", only used for non-Gaussian family, specifies UBR with estimated variance. Default is "v".
theta  If 'init=1', theta includes intial values for smoothing parameters. Default is NULL.
varht

needed only when vmu="u", which gives the fixed variance in calculation of the UBR function. Default is NULL.

tol

the tolerance for truncation in the tridiagonalization. Default is 0.0.

init

an integer of 0 or 1 indicating if initial values are provided for theta. If init=1, initial values are provided using theta. Default is 0.

prec

precision requested for the minimum score value, where precision is the weaker of the absolute and relative precisions. Default is \(1e^{-06}\).

maxit

maximum number of iterations allowed. Default is 30.

Value

info

an integer that provides error message. info=-1 indicates dimension error, info=-2 indicates \(F_2^T Q_2^* F_2! > = 0\), info=-3 indicates tuning parameters are out of scope, info=-4 indicates fails to converge within maxite steps, info=-5 indicates fails to find a reasonable descent direction, info>0 indicates the matrix \(S\) is rank deficient with \(info = rank(S) + 1\).

fit

fitted values.

c

estimates of \(c\).

d

estimates of \(d\).

resi

vector of residuals.

varht

estimate of variance.

theta

estimates of parameters \(log10(\theta)\).

nlaht

the estimate of \(log10(nobs * \lambda)\).

score

the minimum GCV/GML/UBR score at the estimated smoothing parameters.

df

equivalent degree of freedom.

nobs

length(y), number of observations.

nnull

dim(\(H_0\)), number of bases.

nq

length(rk), number of reproducing kernels.

s,q,y

changed from the inputs.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also
dsidr, gdsidr, gdmudr, ssr
**Description**

The dog data frame has 252 rows and 4 columns of data considered by Grizzle and Alen (1969).

**Usage**

data(dog)

**Format**

The data frame contains the following columns:

- `y`: a numeric vector of measurements of coronary sinus potassium concentrations.
- `group`: a vector of group index for the four groups of dogs.
- `dog`: a vector of integers identifying dogs.
- `time`: a numeric vector of time points measurements were made.

**Details**

The data are coronary sinus potassium concentrations measured on each of 36 dogs. These 36 dogs were divided into 4 treatment groups, and the measurements on each dog were taken every two minutes from 1 to 13 minutes after occlusion.

**Source**

Arguments

\(y\) a numerical vector representing the response.

\(q\) a square matrix of the same order as the length of \(y\), with elements equal to the reproducing kernel evaluated at the design points.

\(s\) the design matrix of the null space \(H_0\) of size \((\text{length}(y), \text{dim}(H_0))\), with elements equal to the bases of \(H_0\) evaluated at design points. Default is NULL, representing an empty NULL space.

\(\text{weight}\) A weight matrix for penalized weighted least-square: \((y - f)'W(y - f) + n\lambda J(f)\). Default is NULL for iid random errors.

\(\text{vmu}\) a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. "u~", only used for non-Gaussian family, specifies UBR with estimated variance. Default is "v".

\(\text{varht}\) needed only when \(\text{vmu} = "u"\), which gives the fixed variance in calculation of the UBR function. Default is NULL.

\(\text{limnla}\) a vector of length 2, specifying a search range for the n times smoothing parameter on \(\log_{10}\) scale. Default is \((-10, 3)\).

\(\text{job}\) an integer representing the optimization method used to find the smoothing parameter. The options are \(\text{job} = -1\); golden-section search on \((\text{limnla}(1), \text{limnla}(2))\); \(\text{job} = 0\); golden-section search with interval specified automatically; \(\text{job} > 0\); regular grid search on \([\text{limnla}(1), \text{limnla}(2)]\) with \#(grids) = \(\text{job} + 1\). Default is -1.

\(\text{tol}\) tolerance for truncation used in ‘dsidr’. Default is 0.0, which sets to square of machine precision.

Value

\(\text{info}\) an integer that provides error message. \(\text{info}=0\) indicates normal termination, \(\text{info}=-1\) indicates dimension error, \(\text{info}=-2\) indicates \(F_2^TQF_2! >= 0\), \(\text{info}=-3\) indicates \(\text{vmu}\) is out of scope, and \(\text{info}>0\) indicates the matrix \(S\) is rank deficient with \(\text{info} = \text{rank}(S)+1\).

\(\text{fit}\) fitted values.

\(c\) estimates of c.

\(d\) estimates of d.

\(\text{resi}\) vector of residuals.

\(\text{varht}\) estimate of variance.

\(\text{nlaht}\) the estimate of \(\log_{10}(\text{nobs}*\text{lambda})\).

\(\text{limnla}\) searching range for \(\text{nlaht}\).

\(\text{score}\) the minimum GCV/GML/UBR score at the estimated smoothing parameter. When \(\text{job}>0\), it gives a vector of GCV/GML/UBR functions evaluated at regular grid points.

\(\text{df}\) equivalent degree of freedom.

\(\text{nobs}\) length(\(y\)), number of observations.
null \ dim(H_0), number of bases.
\quad s, q, aux, jpv \quad QR \ decomposition \ of \ S=FR, \ as \ from \ Linpack \ 'dqrdc'.
\quad q \quad \text{first dim}(H_0) \ columns \ gives \ F^T Q F_1, \ and \ its \ bottom-right \ corner \ gives \ tridiagonalization \ of \ F_2^T Q F_2.

\textbf{Author(s)}

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

\textbf{References}


\textbf{See Also}

dmudr, gdsidr, gdmudr, ssr

dsms

\textbf{Description}

Calculate a matrix operation needed to construct Bayesian confidence intervals

\textbf{Usage}

dsms(rkpk.obj)

\textbf{Arguments}

rkpk.obj \quad an \ object \ returned \ from \ calling \ dsidr

\textbf{Value}

a matrix. See the corresponding Fortran subroutine.
**gdmdr**

*Interface of dbmdr, dbimdr, dgmdr, dpmdr in GRKPACK.*

**Description**

To calculate a spline estimate with multiple smoothing parameters for non-Gaussian data

**Usage**

```
gdmudr(y, q, s, family, vmu = "v", varht = NULL, 
   init = 0, theta = NULL, tol1 = 0, tol2 = 0, prec1 = 1e-06, 
   maxit1 = 30, prec2 = 1e-06, maxit2 = 30)
```

**Arguments**

- **y**: a numerical vector representing the response, or a matrix of two columns for binomial data with the first column as the largest possible counts and the second column as the counts actually observed.
- **q**: a list, or an array, of square matrices of the same order as the length of y, which are the reproducing kernels evaluated at the design points.
- **s**: the design matrix of the null space $H_0$ of size $(\text{length-of-y}, \text{dim}(H_0))$, with elements equal to the bases of $H_0$ evaluated at design points.
- **family**: a string specifying the family of distribution. Families supported are "binary", "binomial", "poisson" and "gamma" for Bernoulli, binomial, poisson, and gamma distributions respectively. Canonical links are used except for Gamma family where log link is used.
- **vmu**: a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. "u~", only used for non-Gaussian family, specifies UBR with estimated variance. Default is "v".
- **varht**: needed only when vmu="u", which gives the fixed variance in calculation of the UBR function. Default is 1.0.
- **init**: an integer of 0 or 1 indicating if initial values are provided for theta. If init=1, initial values are provided using theta. Default is 0.
- **theta**: If 'init=1', theta includes intial values for smoothing parameters. Default is NULL.
- **tol1**: the tolerance for elements of w’s. Default is 0.0 which sets to square of machine precision.
- **tol2**: tolerance for truncation used in ‘dsidr’. Default is 0.0 which sets to square of machine precision.
- **prec1**: precision requested for the minimum score value, where precision is the weaker of the absolute and relative precisions. Default is $1e$-06.
- **maxit1**: maximum number of iterations allowed for DMUDR subroutine. Default is 30.
- **prec2**: precision requested for stopping the iteration. Default is $1e$ – 06.
- **maxit2**: maximum number of iterations allowed for the iteration in GRKPACK. Default is 30.
gdmudr

Value

info
an integer that provides error message. info=-1 indicates dimension error, info=-2 indicates $F_2^T Q_{\theta}^T F_2 \geq 0$, info=-3 indicates tuning parameters are out of scope, info=-4 indicates dmu dr fails to converge within maxit1 steps, info=-5 indicates dmu dr fails to find a reasonable descent direction, info=-6 indicates GRKPACK fails to converge within maxit2 steps, info=-7 indicates there are some $w$'s equals to zero, info>0 indicates the matrix $S$ is rank deficient with info = rank($S$) + 1.

fit
estimate of the function at design points.

c
estimates of $c$.

d
estimates of $d$.

resi
vector of working residuals.

varht
estimate of dispersion parameter.

theta
estimates of parameters $\log(\theta)$.

nlaht
the estimate of $\log(nobs \times \lambda)$.

score
the minimum GCV/GML/UBR score at the estimated smoothing parameters.

df
equivalent degree of freedom.

nobs
length-of-$y$, number of observations.

nnull
$dim(H_0)$, number of bases.

nq
length(rk), number of reproducing kernels.

s,q,y,init,maxit2
changed from the inputs.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

dsdr, dmu dr, gdsdr, ssr
Description

To calculate a spline estimate with single smoothing parameter for non-Gaussian data.

Usage

gdsidr(y, q, s, family, vmu="v", varht=NULL, limnla=c(-10, 3),
maxit=30, job=-1, tol1=0, tol2=0, prec=1e-06)

Arguments

- **y**: a numerical vector representing the response, or a matrix of two columns for binomial data with the first column as the largest possible counts and the second column as the counts actually observed.
- **q**: a square matrix of the same order as the length of y, with elements equal to the reproducing kernel evaluated at the design points.
- **s**: the design matrix of the null space $H_0$ of size (length-of-y, dim($H_0$)), with elements equal to the bases of $H_0$ evaluated at design points.
- **family**: a string specifying the family of distribution. Families supported are "binary", "binomial", "poisson" and "gamma" for Bernoulli, binomial, poisson, and gamma distributions respectively. Canonical links are used except for Gamma family where a log link is used.
- **vmu**: a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. "u~", only used for non-Gaussian family, specifies UBR with estimated variance. Default is "v".
- **varht**: needed only when vmu="u", which gives the fixed variance in calculation of the UBR function. Default is 1.0.
- **limnla**: a vector of length 2, specifying a search range for the n times smoothing parameter on log10 scale. Default is (-10, 3).
- **maxit**: maximum number of iterations allowed for the iteration in GRKPACK.
- **job**: an integer representing the optimization method used to find the smoothing parameter. The options are job=-1: golden-section search on (limnla(1), limnla(2)); job=0: golden-section search with interval specified automatically; job >0: regular grid search on [limnla(1), limnla(2)] with \#(grids) = job + 1. Default is -1.
- **tol1**: the tolerance for elements of w’s. Default is 0.0 which sets to square of machine precision.
- **tol2**: tolerance for truncation used in ‘dsidr’. Default is 0.0 which sets to square of machine precision.
- **prec**: precision requested for stopping the iteration. Default is 1e − 06.
Value

info
  an integer that provides error message. info=0 indicates normal termination, info=-1 indicates dimension error, info=-2 indicates $F^T Q F_2 \geq 0$, info=-3 indicates vmu is out of scope, info=-4 indicates the algorithm fails to converge at the muiter steps, info=-5 indicates there are some w's equals to zero, and info>0 indicates the matrix S is rank deficient with info=rank(S)+1.

fit
  estimate of the function at design points.

c
  estimates of c.

d
  estimates of d.

resi
  vector of working residuals.

varht
  estimate of dispersion parameter.

nlaht
  the estimate of $\log_{10}(nobs \cdot \lambda)$.

limla
  searching range for nlaht.

score
  the minimum GCV/GML/UBR score at the estimated smoothing parameter. When job>0, it gives a vector of GCV/GML/UBR functions evaluated at regular grid points.

df
  equivalent degree of freedom.

nobs
  length-of-y, number of observations.

nnull
  $\dim(H_0)$, number of bases.

s, qraux, jpvt
  QR decomposition of S=FR, as from Linpack ‘dqrdc’.

q
  first $\dim(H_0)$ columns gives $F^T Q F_1$, and its bottom-right corner gives tridiagonalization of $F^T Q F_2$.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

dsidr, dmudr, gdmudr, ssr
hat.ssr

Extract the Hat Matrix from a ssr Object

Description

Calculate the hat matrix for a ssr object.

Usage

hat.ssr(ssr.obj)

Arguments

ssr.obj a fitted ssr object.

Details

The hat matrix may be used for diagnosis. Note that the full name hat.ssr should be used since the function hat already exist.

Value

returned is the hat (influence, smoother) matrix.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

ssr

Examples

```r
## Not run: library(MASS)
## Not run: fit1 <- ssr(accel~times, data=mcycle, scale=T, rk=cubic(times))
## Not run: h <- hat.ssr(fit1)
```
### Description

The `horm.cort` data frame has 425 rows and 4 columns of data representing measurement of cortisol on 36 individuals.

### Usage

```r
data(horm.cort)
```

### Format

The data frame contains the following columns:

- **ID**: a vector of index indicating individuals on whom measures were made.
- **time**: a numeric vector of time points of every 2 hours in 24 hours. The time is scaled into [0, 1].
- **type**: a vector of character strings identifying the groups, "normal", "depressed", or "cushing", which the individuals belong to.
- **conc**: cortisol concentration measurements in $\log_{10}$ scale.

### Details

Blood samples were collected every 2 hours for 24 hours from three group of healthy normal volunteers and volunteers with depression and suching syndrome. They were analyzed for parameters that measure hormones of the hypothalamic-pituitary axis. Human circadian rhythm is one of the research objective. In this data set, only measurements of cortisol concentration were included.

### Source

This data set was extracted from a stress study conducted in the medical center of the University of Michigan.

### References


**ident**

*Scaling a Vector*

**Description**

Perform standardization of vector relative to another.

**Usage**

`ident(x, y = x)`

**Arguments**

- **x**: a numeric vector, matrix or data frame
- **y**: an optional numeric vector, matrix or data frame. Default is `x`.

**Details**

Scale `y` based on `x` component by component. For example, if both are a matrix, $y_{[,i]} = (y_{[,i]} - \min(x_{[,i]}))/(\max(x_{[,i]}) - \min(x_{[,i]}))$.

**Value**

a scaled `y`.

---

**inc**

*Fit a Monotone Curve Using a Cubic Spline*

**Description**

Return a spline fit of an increasing curve.

**Usage**

`inc(y, x, spar = "v", limmla = c(-6, 0), grid = x, prec = 1e-06, maxit = 50, verbose = F)`

**Arguments**

- **y**: a vector, used as the response data
- **x**: a vector, used as the covariate. Assume an increasing relationship of `y` on `x`
- **spar**: a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. Default is "v" for GCV
**intervals.nnr**

- **limnla**: a vector of length one or two, specifying a search range for \(\log_{10}(n\lambda)\), where \(\lambda\) is the smoothing parameter and \(n\) is the sample size. If it is a single value, the smoothing parameter will be fixed at this value.

- **grid**: a vector of \(x\) used to assess convergence. Default is \(x\).

- **prec**: a numeric value used to assess convergence. Default is \(1e-6\).

- **maxit**: an integer representing the maximum iterations. Default is 50.

- **verbose**: an optional logical value. If 'TRUE', detailed iteration results are displayed. Default is "FALSE".

**Details**

This function is to fit an increasing function to the data. The monotone function is expressed as the integral of an unknown function that a cubic spline is used to estimate.

**Value**

a split fit together with the convergence information

**Author(s)**

Yuedong Wang &lt;yuedong@pstat.ucsb.edu&gt; and Chunlei Ke &lt;chunlei\_ke@yahoo.com&gt;

**See Also**

`ssr`

---

**Description**

Approximate posterior standard deviations are calculated for the spline estimate of nonparametric functions from a `nnr` object, based on which approximate Bayesian confidence intervals may be constructed.

**Usage**

```r
## S3 method for class 'nnr'
intervals(object, level=0.95, newdata=NULL, terms, pstd=TRUE, ...)
```
Arguments

object an object inheriting from class nnr, representing a nonlinear nonparametric regression model fit.

newdata a data frame on which the fitted spline estimates are to be evaluated. Only those predictors, referred in func of nnr fitting, have to be present. The variable names of the data frame should correspond to the function(s)’ arguments appearing in the option func= of nnr. Default is NULL, where predictions are made at the same values used to fit the object.

terms an optional named list of vectors or matrices containing 0’s and 1’s collecting one or several combinations of the components of spline estimates in the fitted snr object. The length and names of the list shall match those of the unknown functions appearing in the ‘snr’ fit object. For the case of a single function, a vector of 0’s and 1’s can also be accepted. A value “1” at a particular position means that the component at that position is collected. Default is a vector of 1’s, representing the overall fits of all unknown functions.

pstd an optional logic value. If TRUE (the default), the posterior standard deviations are calculated. Orelse, only the predictions are calculated. Computation required for posterior standard deviations could be intensive.

level a numeric value set as 0.95.

... other arguments, currently unused.

Details

The standard deviation returned is based on approximate Bayesian confidence intervals as formulated in Ke and Wang (2002).

Value

an object of class bCI is returned, which is a list of length 2. Its first element is a matrix which contains predictions for combinations specified by terms, and second element is a matrix which contains corresponding posterior standard deviations.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

nnr, plot.bCI
Examples

```r
## fit a generalized varying coefficient models
data(Arosa)
Arosa$csmonth <- (Arosa$month-0.5)/12
Arosa$csyear <- (Arosa$year-1)/45
ozone.fit <- nnr(thick~f1(csyear)+exp(f2(csyear))*f3(csmonth),
  func=list(f1(x)=list(-I(x-.5),cubic(x)), f2(x)=list(-I(x-.5)-1,cubic(x)),
  f3(x)=list(-sin(2*pi*x)+cos(2*pi*x)-1,lspline(x,type="sine"))),
data=Arosa[!Arosa$year%%2==1,], spar="m", start=list(f1=mean(thick),f2=0,f3=mean(csmonth)),
  control=list(backfit=1))
x <- seq(0,1,len=50)
u <- seq(0,1,len=50)

## calculate Bayesian confidence limits for all components of all functions
p.ozone.fit <- intervals(ozone.fit, newdata=list(csyear=x,csmonth=u),
  terms=list(f1=matrix(c(1,1,1,1,0,0,0),nrow=3,byrow=TRUE),
  f2=matrix(c(1,1,0,0,1),nrow=3,byrow=TRUE),
  f3=matrix(c(1,1,1,1,0,0,0),nrow=3,byrow=TRUE)))
```

Description

Provide a way to calculate approximate posterior standard deviations and fitted values at any specified values for any combinations of elements of the spline estimate of nonparametric functions from a slm object, based on which approximate Bayesian confidence intervals may be constructed.

Usage

```
## S3 method for class 'slm'
intervals(object, level=0.95, newdata=NULL, terms, pstd=TRUE, ...)
```

Arguments

- `object`: an object inheriting from class "slm", representing a semi-parametric nonlinear regression model fit.
- `level`: set as 0.95, unused currently
- `newdata`: an optional data frame on which the fitted spline estimate is to be evaluated.
- `terms`: an optional vector of 0's and 1's collecting a combination of components, or a matrix of 0's and 1's collecting several combinations of components, in a fitted ssr object. All components include bases on the right side of ~ in the formula and reproducing kernels in the rk list. Note that the first component is usually
intervals.slm

a constant function if it is not specifically excluded in the formula. A value "1" at a particular position means that the component at that position is collected. Default is a vector of 1’s, representing the overall fit.

pstd

an optional logic value. If TRUE (the default), the posterior standard deviations are calculated. Orelse, only the predictions are calculated. Computation required for posterior standard deviations could be intensive.

... other arguments, currently unused.

Details

The standard deviation returned is based on approximate Bayesian confidence intervals as formulated in Wang (1998).

Value

an object of class bCI is returned, which is a list of length 2. Its first element is a matrix which contains predictions for combinations specified by terms, and second element is a matrix which contains corresponding posterior standard deviations.

Author(s)

Chunlei Ke <chunlei.ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

slm, plot.bCI, predict.ssr

Examples

data(dog)
  # fit a SLM model with random effects for dogs
dog.fit<-slm(y~group*time, rk=list(cubic(time), shrink(group),
  rk.prod(kron(time-0.5),shrink(group)),rk.prod(cubic(time),
  shrink(group))), random=list(dog=-1), data=dog)

  intervals(dog.fit)
Description

Provide a way to calculate approximate posterior standard deviations and fitted values at any specified values for any combinations of elements of the spline estimate of nonparametric functions from a snm object, based on which approximate Bayesian confidence intervals may be constructed.

Usage

```r
## S3 method for class 'snm'
intervals(object, level = 0.95, newdata = NULL, terms, pstd = TRUE, ...)
```

Arguments

- `object`: an object inheriting from class snm, representing a semi-parametric nonlinear mixed effects model fit.
- `newdata`: a data frame on which the fitted spline estimates are to be evaluated. Only those predictors, referred in `func` of `snm` fitting, have to be present. The variable names of the data frame should correspond to the function(s)' arguments appearing in the option `func` of snm. Default is NULL, where predictions are made at the same values used to fit the object.
- `terms`: an optional vector of 0's and 1's collecting a combination of components, or a matrix of 0's and 1's collecting several combinations of components of spline estimates in a fitted snm object. Note that in the cases of multiple functions, the order of all components is collection of base functions for all functions followed by RK's. A value "1" at a particular position means that the component at that position is collected. Default is a vector of 1's, representing the overall fit.
- `pstd`: an optional logic value. If TRUE (the default), approximate posterior standard deviations are calculated. Orelse, only the predictions are calculated. Computation required for posterior standard deviations could be intensive.
- `level`: a numeric value set as 0.95.
- `...`: other arguments, currently unused.

Details

The standard deviation returned is based on approximate Bayesian confidence intervals as formulated in Ke and Wang (2001).

Value

An object of class bCI is returned, which is a list of length 2. Its first element is a matrix which contains predictions for combinations specified by "terms", and second element is a matrix which contains corresponding posterior standard deviations.
Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References


See Also

snm, plot.bCI, predict.ssr

Examples

data(horm.cort)

## extract normal subjects
cort.nor<- horm.cort[horm.cort$type=="normal",]

## fit a self-modelling model with random effects
cort.fit<- snm(conc~b1+exp(b2)*f(time-alogit(b3)),
func=f(u)-list(periodic(u)), fixed=list(b1~1),
random=pDiag(b1+b2+b3~1), data=cort.nor,
groups= ~ID,start=mean(cort.nor$conc))

## note the variable name of newdata
intervals(cort.fit, newdata=data.frame(u=seq(0,1,len=50)))

intervals.snr

Calculate Predictions and Approximate Posterior Standard Deviations for Spline Estimates From a snr Object

Description

Approximate posterior standard deviations are calculated for the spline estimate of nonparametric functions from a snr object, based on which approximate Bayesian confidence intervals may be constructed.

Usage

## S3 method for class 'snr'
intervals(object, level=0.95,newdata=NULL, terms=list(), pstd=TRUE, ...)

intervals.snr

Arguments

object an object inheriting from class 'snr', representing a semi-parametric nonlinear regression model fit.

level set as 0.95, unused currently

newdata a data frame on which the fitted spline estimates are to be evaluated. Only those predictors, referred in 'func' of 'snr' fitting, have to be present. The variable names of the data frame should correspond to the function(s)' arguments appearing in the option func= of snr. Default is NULL, where predictions are made at the same values used to fit the object.

terms an optional named list of vectors or matrices containing 0's and 1's collecting one or several combinations of the components of spline estimates in the fitted snr object. The length and names of the list shall match those of the unknown functions appearing in the 'snr' fit object. For the case of a single function, a vector of 0's and 1's can also be accepted. A value "1" at a particular position means that the component at that position is collected. Default is a vector of 1's, representing the overall fits of all unknown functions.

pstd an optional logic value. If TRUE (the default), the posterior standard deviations are calculated. Orelse, only the predictions are calculated. Computation required for posterior standard deviations could be intensive.

... other arguments, currently unused.

Details

The standard deviation returned is based on approximate Bayesian confidence intervals as formulated in Ke (2000).

Value

a named list of objects of class "bCI" is returned, each component of which is a list of length 2. Within each component, the first element is a matrix which contains predictions for combinations specified by "terms", and the second element is a matrix which contains corresponding posterior standard deviations.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

snr, plot.bCI, predict.ssr
Examples

```r
## Not run:
data(CO2)
options(contrasts=rep("contr.treatment", 2))

## get start values
c02.fit1 <- nlme(uptake~exp(a1)*(1-exp(-exp(a2)*(conc-a3))),
                  fixed=list(a1+a2~Type*Treatment,a3~1),
                  random=a1~1, groups=Plant,
                  start=c(log(30),0,0,log(0.01),0,0,0,50),
                  data=CO2)

M <- model.matrix(~Type*Treatment, data=CO2)[,-1]

## fit a SNR model
c02.fit2 <- snr(uptake~exp(a1)*f(exp(a2)*(conc-a3)),
                func=f(u)-list(-I(1-exp(-u))-1,lspline(u, type="exp")),
                params=list(a1~M-1, a3~1, a2~Type*Treatment),
                start=list(params=c02.fit1$coefficients[c(2:4,9,5:8)]), data=CO2)
p.c02.fit2<- intervals(c02.fit2, newdata=data.frame(u=seq(0,10,len=50)))
```

## End(Not run)

---

**kron**  
*Calculate reproducing kernels for one-dimensional space*

**Description**

Return a matrix evaluating reproducing kernels for the one-dimensional space usually spanned by a vector.

**Usage**

```r
kron(x,y=x)
```

**Arguments**

- `x`  
a vector or a list of numerical values which spans the one-dimensional space.

- `y`  
a vector or a list of numerical values. Default is `x`.

**Value**

a matrix with the numbers of row and column equal to the length of `x` and `y` respectively. The `[i, j]` element is the reproducing kernel evaluated at the `i`th element of `x` and `j`th element of `y`. 
Author(s)
Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
kronecker, ssr

Examples
\begin{verbatim}
x<-runif(10)
kron(x)
\end{verbatim}

Description
Return a matrix evaluating reproducing kernels for some L-splines at observed points.

Usage
\begin{verbatim}
lspline(x, y=x, type="exp", ...)
\end{verbatim}

Arguments
\begin{itemize}
  \item \textbf{x} a numeric vector on which reproducing kernels are evaluated.
  \item \textbf{y} an optional vector, specifying the second argument of reproducing kernels. Default is \text{x}.
  \item \textbf{type} a string indicating the type of L-splines. Available options are "exp", "logit","sine", "sine1", and "linSinCos". Default is "exp".
  \item ... other arguments needed.
\end{itemize}

Details
Denote $L$ as the differential operator, $H_0$ as the null (kernel) space. The available kernels correspond to the following $L$:
\begin{itemize}
  \item exp: $L = rD + D^2$, $H_0 = \text{span}\{1, \exp(-rx)\}$, $r > 0$, default to be 1;
  \item logit: $L = D - 1/(1 + e^t)$, $H_0 = \text{span}\{e^t/(1 + e^t)\}$;
  \item sine0: $L = D^2 + (2\pi)^2$, $H_0 = \text{span}\{\sin(2\pi x), \cos(2\pi x)\}$;
  \item sine1: $L = D(D^2 + (2\pi)^2)$, $H_0 = \text{span}\{1, \sin(2\pi x), \cos(2\pi x)\}$;
  \item linSinCos: $L = D^4 + D^2$, $H_0 = \text{span}\{1, x, \sin(x), \cos(x)\}$.
\end{itemize}
Value

A matrix with the numbers of row and column equal to the lengths of x and y respectively. The [i, j]
element is the reproducing kernel evaluated at (x[i], y[j]).

Author(s)

Chunlei Ke <chunlei_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

ssr

Examples

```r
x <- seq(0, 1, len=20)
lspline(x, type="exp", r=1.5)
```

Description

Fit a nonlinear nonparametric regression models with spline smoothing based on extended Gauss-
Newton/Newton-Raphson and backfitting.

Usage

```r
nnr(formula, func, spar="v", data=sys.parent(),
    start=list(), verbose=FALSE, control=list())
```

Arguments

- `formula`: a model formula, with the response on the left of a ~ operator and on the right
  an expression representing the mean function with a nonparametric function
  appearing with a symbol, e.g. f.
- `func`: a required formula specifying the spline components necessary to estimate the
  non-parametric function. On the left of a ~ operator is the unknown function
  symbol as well as its arguments, while the right side is a list of two components,
  an optional nb and a required rk. nb and rk are similar to `formula` and `rk` in
  ssr. A missing nb denotes an empty null space.
spar  a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. Default is "v" for GCV.

data  an optional data frame.

start  a list of vectors or expressions which input initial values for the unknown functions. If expressions, the argument(s) inside should be the same as in func. The length of start should be the same as the number of unknown functions. If named, the names of the list should match those in "func". If not named, the order of the list is taken as that appearing in "func".

verbose  an optional logical numerical value. If TRUE, information on the evolution of the iterative algorithm is printed. Default is FALSE.

control  an optional list of control values to be used. See nnr.control for details.

Details
A nonlinear nonparametric model is fitted using the algorithms developed in Ke and Wang (2002).

Value
an object of class nmr is returned, containing fitted values, fitted function values as well as other information used to assess the estimate.

Author(s)
Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References

See Also
nnr.control, ssr, print.nnr, summary.nnr, intervals.nnr

Examples
x <- 1:100/100
y <- exp(sin(2*pi*x))+0.3*runif(x)
fit <- nmr(y~exp(f(x)), func=list(f(u=0, cubic(u))), start=list(0))

## fit a generalized varying coefficient models
data(Arosa)
Arosa$csmonth <= (Arosa$month-0.5)/12
Arosa$cseyear <= (Arosa$year-1)/45
ozone.vc.fit <- nmr(thick~f1(cseyear)+exp(f2(cseyear))*f3(csmonth),
  func=list(f1(x)=list(-1(x-.5), cubic(x)), f2(x)=list(-1(x-.5)-1,cubic(x)),
  f3(x)=list(-sin(2*pi*x)+cos(2*pi*x)-1,lspline(x,type="sine"))),
  data=Arosa[Arosa$year==1,], spar="m", start=list(f1=mean(thick),f2=0,f3=sin(csmonth)),
  control=list(backfit=11))
nnr.control

Set Control Parameters for nnr

Description

Control parameters supplied in the function call replace the defaults to be used in calling nnr.

Usage

```r
nnr.control(job = -1, tol = 0, max.iter = 50, init = 0, limnla = c(-10, 0), varht = NULL, theta = NULL, prec = 1e-06, maxit = 30,
method = "NR", increment = 1e-04, backfit = 5, converg = "coef",
toler = 0.001)
```

Arguments

- **job**: an integer representing the optimization method used to find the smoothing parameter. The options are job=-1: golden-section search on \((\text{limnla}(1), \text{limnla}(2))\); job=0: golden-section search with interval specified automatically; job >0: regular grid search on \([\text{limnla}(1), \text{limnla}(2)]\) with \(\#(\text{grids}) = \text{job} + 1\). Default is -1.
- **tol**: tolerance for truncation used in ‘dsidr’. Default is 0.0, which sets to square of machine precision.
- **max.iter**: maximum number of iterations allowed for the Gauss-Newton/Newton-Raphson iteration.
- **init**: an integer of 0 or 1 indicating if initial values are provided for theta. If init=1, initial values are provided using theta. Default is 0.
- **limnla**: a vector of length 2, specifying a search range for the n times smoothing parameter on log10 scale. Default is (-10, 0).
- **varht**: needed only when vmu="u", which gives the fixed variance in calculation of the UBR function. Default is NULL.
- **theta**: If ‘init=1’, theta includes initial values for smoothing parameters. Default is NULL.
- **prec**: precision requested for the minimum score value, where precision is the weaker of the absolute and relative precisions. Default is 1e-06.
- **maxit**: maximum number of iterations allowed. Default is 30.
- **method**: a character string specifying a method for iterations, "GN" for Gauss-Newton and "NR" for Newton-Raphson. Default is "GN".
- **increment**: specifies a small value as increment to calculate derivatives. Default is 1e-04.
- **backfit**: an integer representing the number of backfitting iterations for multiple functions. Default is 5.
paramecium

| converg | an optional character, with possible values "coef" and "ortho", specifying the convergence criterion to be used. "coef" uses the change of estimate of parameters and functions to assess convergence, and "ortho" uses a criterion similar to the relative offset used in nls. Default is "coef".
| toler | tolerance for convergence of the algorithm. Default is 0.001.

Value

terminated is a list includes all re-seted control parameters.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

nrr, dsidr, dmufr

Examples

## use Newton-Raphson
nrr.control(method="NR")

<table>
<thead>
<tr>
<th>paramecium</th>
<th>Growth of paramecium caudatum population</th>
</tr>
</thead>
</table>

Description

The 'paramecium' data frame has 25 rows and 2 columns of data from an experiment that grow paramecium caudatum

Usage

data(paramecium)

Format

The data frame contains the following columns:

day a numeric vector of days since the start of the experiment

density a numeric vector of mean number of individuals in 0.5 ml of medium of four different cultures started simultaneously

Source


References

### periodic

**Calculate Reproducing Kernels for Periodic Polynomial Splines with Period 1**

---

**Description**

Return a matrix evaluating reproducing kernels for periodic polynomial splines at observed points.

**Usage**

```r
periodic(s, t=s, order=2)
```

**Arguments**

- `s` a numeric vector.
- `t` an optional vector. Default is the same as `s`.
- `order` an optional integer specifying the order of the polynomial spline. Default is 2 for the periodic cubic spline.

**Details**

The general formula of the reproducing kernel is sum of an infinite series, which is approximated by taking the first 50 terms. For the case of order=2, the close form is available and used.

**Value**

a matrix with the numbers of row and column equal to the lengths of `s` and `t` respectively. The `[i, j]` element is the reproducing kernel evaluated at `(s[i], t[j])`.

**References**


**See Also**

- `cubic`
- `lspline`

**Examples**

```r
x <- seq(0, 1, len=100)
periodic(x, order=3)
```
plot.bCI

Bayesian Confidence Interval Plot of a Smoothing Spline Fit

Description

Create trellis plots of a nonparametric function fit together with its (approximate) 95% Bayesian confidence intervals from a ssr/slm/snr/snm object.

Usage

```r
## S3 method for class 'bCI'
plot(x, x.val=NULL, type.name=NULL, ...)
```

Arguments

- `x`: an object of class "bCI" containing point evaluation of the unknown function and/or corresponding posterior standard deviations.
- `x.val`: an optional vector representing values of argument based on which the function is to evaluate.
- `type.name`: an optional character vector specifying the names of fits.
- `...`: options suitable for `xyplot`.

Details

This function is to visualize a spline fit by use of trellis graphic facility with Bayesian confidence intervals superposed. Multi-panel plots, based on `xyplot`, are suitable for SS ANOVA decomposition of a spline estimate.

Author(s)

Chunlei Ke <chunlei_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

- `predict.ssr`
- `intervals.slm`
- `intervals.snr`
- `intervals.snm`

Examples

```r
x <- seq(0, 1, len=100)
y <- 2*sin(2*pi*x)+rnorm(x)*0.5

fit <- ssr(y~x, cubic(x))
p.fit <- predict(fit)
## Not run: plot(p.fit)
## Not run: plot(p.fit,type.name="fit")
```
plot.ssr

Generate Diagnostic Plots for a ssr Object

Description

Creates a set of plots suitable for assessing a fitted smoothing spline model of class ssr.

Usage

```r
## S3 method for class 'ssr'
plot(x, ask=FALSE, ...)
```

Arguments

- `x` a `ssr` object.
- `ask` if TRUE, plot.ssr operates in interactive mode.
- `...` Other options used for plot, currently inactive.

Details

This function is a method for the generic function plot for class ssr. It can be invoked by calling plot for an object of the appropriate class, or directly by calling plot.ssr regardless of the class of the object.

An appropriate x-y plot is produced to display diagnostic plots. These can be one or all of the following choices:

- Estimate of function with CIs
- Residuals against Fitted values
- Response against Fitted values
- Normal QQplot of Residuals

The first plot of estimate of function with CIs is only useful for univariate smoothing spline fits.

When `ask=TRUE`, rather than produce each plot sequentially, plot.ssr displays a menu listing all the plots that can be produced. If the menu is not desired but a pause between plots is still wanted one must set `par(ask=TRUE)` before invoking this command with argument `ask=FALSE`.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

plot, ssr, predict.ssrm
Examples

## Not run: library(MASS)
## Not run: fit1 <- ssr(accel~times, data=mcycle, scale=TRUE, rk=cubic(times))
## Not run: plot(fit1, ask=TRUE)

### Polynomial

#### Calculate Reproducing Kernels for Polynomial Splines on [0, 1]

**Description**

Return a matrix evaluating reproducing kernels for polynomial splines at observed points.

**Usage**

- `linear(s, t=s)`
- `cubic(s, t=s)`
- `quintic(s, t=s)`
- `septic(s, t=s)`

**Arguments**

- `s`: a vector of values in [0, 1], at which the kernels are evaluated.
- `t`: an optional vector in [0, 1]. Default is the same as `s`.

**Details**

The reproducing kernels implemented in these functions are based on Bernoulli functions with domain [0, 1].

**Value**

A matrix with the numbers of row and column equal to the lengths of `s` and `t` respectively. The [i, j] element is the reproducing kernel of linear, cubic, quintic, or septic spline evaluated at (s[i], t[j]).

**Author(s)**

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

**References**


**See Also**

- `ssr`
- `linear2`
- `cubic2`
- `quintic2`
- `septic2`
### Description

Return a matrix evaluating reproducing kernels for polynomial splines at observed points.

### Usage

```r
linear2(s, t = s)
cubic2(s, t = s)
quintic2(s, t = s)
septic2(s, t = s)
```

### Arguments

- `s`: a vector of non-negative values, at which the kernels are evaluated.
- `t`: an optional non-negative vector. Default is the same as `s`.

### Details

The reproducing kernels implemented in these functions are based on Green functions. The domain is \([0, T]\), where \(T\) is a given positive number.

### Value

A matrix with the numbers of row and column equal to the length of `s` and `t` respectively. The \([i, j]\) element is the reproducing kernel of linear, cubic, quintic, or septic spline evaluated at \((s[i], t[j])\).

### Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

### References


### See Also

- `ssr`, `linear`, `cubic`, `quintic`, `septic`

### Examples

```r
x <- seq(0, 5, len = 10)
linear2(x)
```
predict.slm

Predict Method for Semiparametric Linear Mixed Effects Model Fits

Description

Predicted Values on different levels of random effects with the spline fit as part of fixed effects

Usage

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

Arguments

- `object`: an object inheriting from class `slm`, representing a semi-parametric linear mixed effects model fit.
- `newdata`: a data frame containing the values at which predictions are required. Only those predictors, referred to in the right side of the formula in the object, need to be present by name in newdata. Default is NULL, where predictions are made at the same values used to compute the object.
- `...`: other arguments, but currently unused.

Value

Returned is a data.frame with columns given by the predictions at different levels and the grouping factors. Note that the smooth part of the spline fit is regarded as fixed.

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References


See Also

`slm`

Examples

```r
data(dog)
dog.fit<-slm(y~group*time, rk=list(cubic(time), shrink1(group),
            rk.prod(kron(time-0.5), shrink1(group)), rk.prod(cubic(time),
            shrink1(group))), random=list(dog=-1), data=dog)
predict(dog.fit)
```
**predict.snm**

Predictions from a Semiparametric Nonlinear Mixed Effects Model Fit

**Description**

The predictions are obtained on a semiparametric nonlinear mixed effects model object by replacing the unknown functions and the unknown parameters with their estimates. Of note, only a population level of predictions is available.

**Usage**

```r
## S3 method for class 'snm'
predict(object, newdata, ...)
```

**Arguments**

- `object` a fitted `snm` object.
- `newdata` a data frame containing the values at which predictions are required. Default are data used to fit the object.
- `...` other arguments, but currently unused.

**Details**

This function is a method for the generic function `predict` for class `snm`.

**Value**

a vector of prediction values, obtained by evaluating the model in the frame `newdata`

**Author(s)**

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

**References**


**See Also**

`snm`, `predict`
predict.snr

Predict Method from a Semiparametric Nonlinear Regression Model Fit

Description

The predictions on a semiparametric nonlinear regression model object are obtained by substituting the unknown functions together with unknown parameters with their estimates and evaluating the regression functional based on provided or default covariate values.

Usage

```r
## S3 method for class 'snr'
predict(object, newdata, ...)  
```

Arguments

- `object`: a fitted `snr` object.
- `newdata`: a data frame containing the values at which predictions are required. Default are NULL, where data used to produce the fit are to be taken.
- `...`: other arguments, but currently unused.

Details

This function is a method for the generic function predict for class `snr`.

Value

A vector of prediction values, obtained by evaluating the model in the frame `newdata`.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

`snr`
predict.ssr

*Calculate Predictions and Posterior Standard Deviations for a ssr Object*

**Description**

Provide a way to calculate predictions at any specified values for any combinations of elements in the fitted model. Posterior standard deviations may be used to construct Bayesian confidence intervals.

**Usage**

```r
## S3 method for class 'ssr'
predict(object, newdata=NULL, terms, pstd=TRUE, ...)
```

**Arguments**

- **object**: a fitted ssr object.
- **newdata**: an optional data frame containing the values at which predictions are required. Default is NULL, where predictions are made at the same values used to compute the object. Note that if scale=T, the newdata is on the original scale before transformation.
- **terms**: an optional vector of 0’s and 1’s collecting a combination of components, or a matrix of 0’s and 1’s collecting several combinations of components, in a fitted ssr object. All components include bases on the right side of ~ in the formula and reproducing kernels in the rk list. Note that the first component is usually a constant function if it is not specifically excluded in the formula. A value "1" at a particular position means that the component at that position is collected. Default is a vector of 1’s, representing the overall fit.
- **pstd**: an optional logic value. If TRUE (the default), the posterior standard deviations are calculated. Otherwise, only the predictions are calculated. Computation required for posterior standard deviations could be intensive.
- **...**: other arguments, but currently unused.

**Details**

This function is a method for the generic function predict for class ssr. It can be used to construct Bayesian confidence intervals for any combinations of components in the fitted model.

**Value**

an object of class bCI is returned, which is a list of length 2. Its first element is a matrix which contains predictions for combinations specified by terms, and second element is a matrix which contains corresponding posterior standard deviations.
print.anova.ssr

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References


See Also

ssr, plot.bCI

Examples

data(acid)

# tp.pseudo calculates the pseudo kernel
acid.fit<- ssr( ph ~ t1 + x1 + x2, rk = list(tp.pseudo(t1),
   tp.pseudo(list(x1, x2))), spar = "m", data=acid)

# extract the main effect of t1
grid <- seq(min(acid$t1), max(acid$t1), length=100)
p <- predict(acid.fit, data.frame(t1=grid,x1=0,x2=0),
   terms=c(0,1,0,0,1,0))

# extract the main effect of (x1,x2)
grid <- expand.grid(x1=seq(min(acid$x1), max(acid$x1), length=20),
   x2=seq(min(acid$x2), max(acid$x2), length=20))
p <- predict(acid.fit, data.frame(t1=0,x1=grid,x1,x2=grid,x2),
   terms=c(0,0,1,1,0,1), pstd=FALSE)

print.anova.ssr

Print an anova.ssr Object

Description

Calculate and output p-values for tests available.

Usage

## S3 method for class 'anova.ssr'
print(x, ...)

Arguments

x an object inheriting from class anova.ssr, generally obtained by applying the
   anova.ssr method to an ssr object.

... other available arguments, currently unused.
Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

anova.ssr, ssr

print.nnr

Print Values

Description

Print the arguments of a 'nnr' object.

Usage

```r
## S3 method for class 'nnr'
print(x, ...)
```

Arguments

- `x` a nnr object
- `...` unused argument

Details

This is a method for the function print for objects inheriting from class nnr.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

nnr
**print.slm**  
*Print Values*

---

### Description

Print the arguments of a slm object.

### Usage

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

### Arguments

- `x`: a `slm` object
- `...`: unused argument

### Details

This is a method for the function `print` for objects inheriting from class `slm`.

### Author(s)

Chunlei Ke `<chunlei__ke@pstat.ucsb.edu>` and Yuedong Wang `<yuedong@pstat.ucsb.edu>`

### See Also

- `slm`

---

**print.snm**  
*Print Values*

---

### Description

Print the arguments of a 'snm' object.

### Usage

```r
## S3 method for class 'snm'
print(x, ...)  
```

### Arguments

- `x`: a `snm` object
- `...`: unused argument
Details
This is a method for the function `print` for objects inheriting from class 'snr'.

Author(s)
Chunlei Ke <chunlei@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
slm, print

---

**print.snr**

```r
# S3 method for class 'snr'
print(x, ...)
```

Arguments

- `x` a snr object
- `...` unused argument

Details
This is a method for the function `print` for objects inheriting from class snr.

Author(s)
Chunlei Ke <chunlei@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
snr
print.ssr

Description

Print the arguments of a ssr object.

Usage

```r
## S3 method for class 'ssr'
print(x, ...)
```

Arguments

- `x`  
  a ssr object
- `...`  
  unused argument

Details

This is a method for the function print for objects inheriting from class ssr.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

- `ssr`

print.summary.nnr

Description

Print the arguments of a summary.nnr object

Usage

```r
## S3 method for class 'summary.nnr'
print(x, ...)
```

Arguments

- `x`  
  an object of class summary.nnr
- `...`  
  unused argument
Details
This is a method for the function `print` for objects inheriting from class `summary.slm`.

Author(s)
Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
nnr, summary.nnr

Print Values

Description
Print the arguments of a `summary.slm` object

Usage
```r
## S3 method for class 'summary.slm'
print(x, ...)
```

Arguments
- `x` an object of class `summary.slm`
- `...` unused argument

Details
This is a method for the function `print` for objects inheriting from class `summary.slm`.

Author(s)
Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
`slm`, `summary.slm`
Description

Print the arguments of a summary.snm object

Usage

```r
## S3 method for class 'summary.snm'
print(x, ...)
```

Arguments

- `x` an object of class summary.snm
- `...` unused argument

Details

This is a method for the function `print` for objects inheriting from class summary.snm.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

- `snm`, `summary.snm`

---

Description

Print the arguments of a summary.snr object

Usage

```r
## S3 method for class 'summary.snr'
print(x, ...)
```

Arguments

- `x` an object of class summary.snr
- `...` unused argument
Details

This is a method for the function `print` for objects inheriting from class `summary.snr`.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

`snr`, `summary.snr`

---

### Description

Print the arguments of a `summary.ssr` object

### Usage

```r
## S3 method for class 'summary.ssr'
print(x, ...)
```

### Arguments

- `x`: an object of class `summary.ssr`
- `...`: unused argument.

### Details

This is a method for the function `print` for objects inheriting from class `summary.ssr`.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

`ssr`, `summary.ssr`
rk.prod

Calculate product of reproducing kernels

Description

Return a matrix as product of reproducing kernels

Usage

rk.prod(x, ...)

Arguments

x                  a matrix evaluating a reproducing kernel, or a vector.
...               optional lists of matrices evaluating reproducing kernels or vectors. All matrices
                 must have the same dimensions. All vectors must have the same length. The
                 length of each vector must equal to the column and row numbers of each matrix.

Details

The product of reproducing kernels is again a reproducing kernel. In SS ANOVA, product of reproducing kernels is often used to model interaction spline terms.

Value

a matrix as the product of reproducing kernels. If one argument is a vector, a kron kernel is constructed first.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

kron, ssr

Examples

x1<- 1:10/10
x2<- runif(10)
rk.prod(cubic(x1), periodic(x2))
The 'seizure' data frame has 60,000 rows and 3 columns of data from an IEEG time series

Usage

data(seizure)

Details

The baseline segment contains 5-minute IEEG signal extracted at least four hours before the seizure’s onset. The preseizure segment contains 5-minute IEEG signal right before a seizure’s clinical onset. The sampling rate of the IEEG signal is 200 observations per second. Therefore there are 60,000 time points in each segment.

Format

The data frame contains the following columns:

- t a numeric vector of the observation number
- base a numeric vector of the baseline segment
- preseizure a numeric vector of the segment right before a seizure

Source


references

Shrinkage

Calculate reproducing kernels for Stein shrinkage estimate

Description
Return a matrix evaluating reproducing kernels for the discrete shrinkage towards zero or the mean estimate

Usage
shrink0(x, y=x)
shrink1(x, y=x)

Arguments
x a vector of numerical values or factor indicating different levels.
y a vector of numerical values or factor indicating different levels. Default is x.

Value
a matrix with the numbers of row and column equal to the length of x and y respectively. The \([i,j]\) element is the reproducing kernel evaluated at the ith element of x and jth element of y.
shrink0 shrinks towards zero, and shrink1 shrinks towards the mean.

Author(s)
Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
shrink0, ssr

Examples
x<-rep(1:10,2)
shrink1(x)
sine4p

*Calculate Reproducing Kernels for Periodic L-Splines with Period 1/2*

**Description**

Return a matrix evaluating reproducing kernels for periodic L-splines at observed points.

**Usage**

`sine4p(s, t=s)`

**Arguments**

- `s`: a numeric vector.
- `t`: an optional vector. Default is the same as `s`.

**Details**

The general formula of the reproducing kernel is provided in Gu (2001). The close form is not available, so an approximate based on the first 50 terms of the series is used.

**Value**

A matrix with the numbers of row and column equal to the lengths of `s` and `t` respectively. The `[i, j]` element is the reproducing kernel evaluated at `(s[i], t[j])`.

**References**


**See Also**

- `cubic`
- `lspline`

**Examples**

```r
x <- seq(0, 1, len=100)
sine4p(x)
```
Fit a Semi-parametric Linear Mixed Effects Model

Description

Returns an object of class `slm` that represents a semi-parametric linear mixed effects model fit.

Usage

```r
slm(formula, rk, data=sys.parent(), random, weights=NULL,
correlation=NULL, control=list(apvar=FALSE))
```

Arguments

- `formula`: a formula object, with the response on the left of a `∼` operator, and the bases of the null space `H_0` of the non-parametric function and other terms, separated by `+` operators, on the right.
- `rk`: a list of expressions that specify the reproducing kernels of the spline function(s), `R_1, ..., R_p` for spaces `H_1, ..., H_p`. See the help file of `ssr` for more details.
- `data`: An optional data frame containing the variables appearing in `formula`, `random`, `rk`, `correlation`, `weights`. By default, the variables are taken from the environment from which `slm` is called.
- `random`: A named list of formulae, lists of formulae, or pdMat objects, which defines nested random effects structures. See help file of `lme` for more details.
- `weights`: An optional varFun object or one-sided formula describing the within-group heteroscedasticity structure. See the help file of `lme` for more details.
- `correlation`: An optional corStruct object specifying the within-group correlation structure. See `lme` for more details.
- `control`: an optional list of any applicable control parameters from `lme`.

Details

This generic function fits a semi-parametric linear mixed effects model (or non-parametric mixed effects models) as described in Wang (1998), but allowing for general random and correlation structures. Because the connection to a linear mixed effects model is adopted, only GML is available to choose smoothing parameters.

Value

An object of class `slm` is returned. Generic functions such as print, summary, predict and intervals have methods to show the results of the fit.

Note: output from earlier versions of `slm` shows incorrect smoothing spline parameters for SSANOVA, which is corrected in this version.
**Author(s)**

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

**References**


**See Also**

ssr, predict.slm, intervals.slm, print.slm, summary.slm

**Examples**

```r
## SS ANOVA is used to model "time" and "group"
## with random intercept for "dog".
data(dog)
dog.fit<- slm(y~group*time, rk=list(cubic(time), shrink(group)),
            rk=prod(kron(time-0.5), shrink(group)),
            shrink(group)),
            random=list(dog=-1), data=dog)
```

---

**snm**

*Fit a Semi-parametric Nonlinear Mixed-effects Model*

**Description**

This generic function fits a semi-parametric nonlinear mixed-effects model in the formulation described in Ke and Wang (2001). Current version only allows linear dependence on non-parametric functions.

**Usage**

```r
snm(formula, func, data=sys.parent(), fixed, random=fixed,
    groups, start, spar="v", verbose=FALSE, method="REML", control=NULL,
    correlation=NULL, weights=NULL)
```

**Arguments**

- `formula` a formula object, with the response on the left of a ~ operator, and an expression of variables, parameters and non-parametric functions on the right.
- `func` a list of spline formulae each specifying the spline components necessary to estimate each non-parametric function. On the left of a ~ operator of each component is the unknown function symbol(s) as well as its arguments, while the right side is a list of two components nb, an optional one-side formula for representing the null space's bases, and a required rk structure. nb and rk are similar to `formula` and `rk` in `ssr`. A missing nb denotes an empty null space.
fixed  a two-sided formula specifying models for the fixed effects. The syntax of fixed
in nlme is adopted.

start  a numeric vector, the same length as the number of fixed effects, supplying
starting values for the fixed effects.

spar  a character string specifying a method for choosing the smoothing parameter.
"v", "m" and "u" represent GCV, GML and UBR respectively. Default is "v" for
GCV.

data  An optional data frame containing the variables appearing in formula, random,
rk, correlation, weights. By default, the variables are taken from the envi-
ronment from which snm is called.

random  an optional random effects structure specifying models for the random effects.
The same syntax of random in nlme is assumed.

groups  an optional one-sided formula of the form ~g1 (single level) or ~g1/.../gQ (mul-
tiple levels of nesting), specifying the partitions of the data over which the ran-
dom effects vary. g1,....gQ must evaluate to factors in data. See nlme for
details.

verbose  an optional logical numerical value. If TRUE, information on the evolution of the
iterative algorithm is printed. Default is FALSE.

method  a character string. If 'REML' the model is fit by maximizing the restricted log-
likelihood. If 'ML' the log-likelihood is maximized. Default is 'REML.

control  a list of parameters to control the performance of the algorithm.

correlation  an optional corStruct object describing the within-group correlation structure.
See the documentation of corClasses for a description of the available corStruct
classes. Default is NULL, corresponding to no within-in group correlations.

weights  an optional varFunc object or one-sided formula describing the within-group
heteroscedasticity structure. If given as a formula, it is used as the argument to
varFixed, corresponding to fixed variance weights. See the documentation on
varClasses for a description of the available varFunc classes. Defaults to NULL,
corresponding to homoscesdatic within-group errors.

Value  an object of class snm is returned, representing a semi-parametric nonlinear mixed effects model fit.
Generic functions such as print, summary, predict and intervals have methods to show the results of
the fit.

Author(s)  Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References  Ke, C. and Wang, Y. (2001). Semi-parametric Nonlinear Mixed Effects Models and Their Applica-
tions. JASA 96:1272-1298.
See Also

predict.snm, intervals.snm, snm.control, print.snm, summary.snm

Examples

```r
## Not run:
data(CO2)

options(contrasts=rep("contr.treatment", 2))
co2.fit1 <- nlme(uptake=exp(a1)*(1-exp(-exp(a2)*(conc-a3))),
               fixed=list(a1+a2-Type*Treatment, a3=1),
               random=a1~1, groups=-Plant,
               start=c(log(30),0,0,0,log(0.01),0,0,0,50),
               data=CO2)

M <- model.matrix(~Type*Treatment, data=CO2)[,-1]
co2.fit2 <- snm(uptake=exp(a1)+f(exp(a2)*(conc-a3)),
               func=f(u)-list(-I(1-exp(-u))-1, lpspline(u, type="exp")),
               fixed=list(a1-M-1,a3=1,a2-Type*Treatment),
               random=list(a1~1), group=-Plant, verbose=TRUE,
               start=co2.fit1$coefficient[2:4,9,5:8], data=CO2)

## End(Not run)
```

### snm.control

Set Control Parameters for snm

**Description**

Control parameters supplied in the function call replace the defaults to be used in calling snm.

**Usage**

```r
snm.control(rkpk.control, nlme.control, prec.out=0.0005,
             maxit.out=30, converg="COEF", incdelta)
```

**Arguments**

- `rkpk.control`: an optional list of control parameters for dsidr or dmudr to estimate the unknown functions.
- `nlme.control`: a list of control parameters for the nonlinear regression step, the same as nlme-Control. Default is `list(returnObject = T, maxIter = 5)`.
- `prec.out`: tolerance for convergence criterion. Default is 0.0005.
- `maxit.out`: maximum number of iterations for the algorithm. Default is 30.
- `converg`: an optional character, with possible values "COEF" and "PRSS", specifying the convergence criterion to be used. "COEF" uses the change of estimate of parameters and functions to assess convergence, and "PRSS" uses penalized residual sums of squares. Default is "COEF".
- `incdelta`: specifies a small value as increment to calculate derivatives. Default is 0.001.
Value

Returned is a list includes all re-seted control parameters.

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

snm, dsidr, dmudr

Examples

```r
## set maximum iteration to be 50
snm.control(maxit.out=50)
```

---

**snr**

*Fit A Semi-parametric Nonlinear Regression Model*

Description

This generic function fits a Semi-parametric Nonlinear Regression Model as formulated in Ke (2000).

Usage

```r
snr(formula, func, params, data = sys.parent(), start,
    spar = "v", verbose = FALSE, control = list(), correlation = NULL,
    weights = NULL)
```

Arguments

- **formula**: a model formula, with the response on the left of a ~ operator and on the right an expression representing the mean function with at least one unknown function appearing with a symbol, e.g. f. If "data" is present, all names except the nonparametric function(s) used in the formula should be defined as parameters or variables in the data frame.

- **func**: a list of spline formulae each specifying the spline components necessary to estimate each non-parametric function. On the left of a ~ operator of each component is the unknown function symbol(s) as well as its arguments, while the right side is a list of two components nb, an optional one-side formula for representing the null space's bases, and a required rk structure. nb and rk are similar to formula and rk in ssr. A missing nb denotes an empty null space.

- **params**: a two-sided formula specifying models for the parameters. The syntax of params in gnlss is adopted. See gnlss for details.
data

an optional data frame containing the variables named in model, params, correlation and weights. By default the variables are taken from the environment from which snr is called.

start

a numeric list with two components: "params=", a vector of the size of the length of the unknown parameters, providing initial values for the parameters, and "f=" a list of vectors or expressions which input initial values for the unknown functions. If the unknown functions appear linear in the model, the initial values then are not necessary.

spar

a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. Default is "v" for GCV.

verbose

an optional logical numerical value. If TRUE, information on the evolution of the iterative algorithm is printed. Default is TRUE.

control

an optional list of control parameters. See snr.control for details.

correlation

an optional corStruct as in glmm. Default is NULL, corresponding to uncorrelation.

weights

an optional varFunc structure as in glmm. Default is NULL, representing equal variances.

Details

A semi-parametric regression model is generalization of self-modeling regression, nonlinear regression and smoothing spline models, including as special cases (nonlinear) partial spline models, varying coefficients models, PP regression and some other popular models. 'snr' is implemented with an alternate iterative procedures with smoothing splines to estimate the unknown functions and general nonlinear regression to estimate parameters.

Value

An object of class snr is returned, representing a semi-parametric nonlinear regression fit. Generic functions such as print, summary, intervals and predict have methods to show the results of the fit.

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

References


See Also

intervals.snr, predict.snr, snr.control
Examples

```r
## Not run:
data(CO2)
options(contrasts=rep("contr.treatment", 2))
co2.fit1 <- nlme(uptake~exp(a1)*(1-exp(-exp(a2))*((conc-a3))),
  fixed=list(a1+a2~Type*Treatment, a3=1),
  random=a1~1, groups="Plant",
  start=c(log(30),0,0,0,log(0.01),0,0,0,50),
data(CO2)

M <- model.matrix(~Type*Treatment, data=CO2)[,-1]

## fit a SNR model
co2.fit2 <- snr(uptake~exp(a1)*f(exp(a2)*((conc-a3)),
  func=f(u)~list(-1(1-exp(-u))^-1, lspline(u, type="exp")),
  params=list(a1=M, a3=1, a2~Type*Treatment),
  start=list(params=co2.fit1$coef$fixed[c(2:4,9,5:8)]), data=CO2)

## End(Not run)
```

**snr.control**  
*Set Control Parameters for snr*

**Description**

Control parameters supplied in the function call replace the defaults to be used in calling `snr`.

**Usage**

```r
snr.control(rkpk.control = list(job = -1, tol = 0, init = 0, limnla = c(-10, 0), varht = NULL, theta = NULL, prec = 1e-06, maxit = 30),
nls.control = list(returnObject = TRUE, maxIter = 5), incdelta = 0.001,
prec.out = 0.001, maxit.out = 30, converg = "COEF", method = "GN",
backfit = 5)
```

**Arguments**

- **rkpk.control** a optional list of control parameters for dsidr or dmudr to estimate the unknown functions. Default is "list(job = -1, tol = 0, init = 0, limnla = c(-10, 0), varht = NULL, theta = NULL, prec = 1e-06, maxit = 30)".
- **nls.control** a list of control parameters for the nonlinear regression step, the same as `gnlsControl`. Default is "list(returnObject = TRUE, maxIter = 5)
- **incDelta** the incremental value to be used to calculate derivatives for the unknown functions. Default is 0.001
- **prec.out** tolerance for convergence criterion. Default is 0.0001.
- **maxit.out** maximum number of iterations for the algorithm. Default is 30.
converg an optional character, with possible values COEF and PRSS, specifying the convergence criterion to be used. COEF uses the change of estimate of parameters and functions to assess convergence, and PRSS uses penalized residual sums of squares. Default is COEF.

method an optional string of value either GN for Gauss-Newton or NR for Newton-Raphson iteration methods to estimate the unknown functions. Default is GN.

backfit an integer to set the number of backfitting iterations inside the loop. Default is 5.

Value returned is a list includes all re-seted control parameters.

Author(s) Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>.

See Also snr, dsidr, dmudr

Examples

```r
## use Newton-Raphson iteration and only a single backfitting
snr.control(method="NR", backfit=1)
```

---

### sphere

**Calculate Pseudo Reproducing Kernels for Spherical Splines**

Description Return a matrix evaluating reproducing kernels for splines on a sphere.

Usage

```r
sphere(x, y=x, order=2)
```

Arguments

- `x` a matrix of two columns or a list of two components, representing observed latitude and longitude respectively.
- `y` a matrix of two columns or a list of two components, representing latitude and longitude respectively. Default is the same as `x`.
- `order` an optional integer specifying the order of the spherical spline. Available are 2, 3, 4, 5 and 6, with a default 2.
Details

The kernel for spherical splines is a series inconvenient to compute. This pseudo kernel is based on a topological equivalence as described in Wahba (1981), for which cases the closed form can be derived.

Value

A matrix with the numbers of row and column equal to the lengths of x and y respectively. The [i, j] element is the reproducing kernel evaluated at (x[i], y[j]) (or ((x[1][i], x[2][i]), (y[1][j], y[2][j])) for lists).

Author(s)

Chunlei Ke <chunlei\_ke@pstat.ucsb.edu> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References


See Also

periodic

Examples

```r
x <- seq(0, 2*pi, len=10)
y <- seq(-pi/2, pi/2, len=10)
s.ker <- sphere(cbind(x, y), order=3)
```

---

ssr

Fit a General Smoothing Spline Regression Model

Description

Returns an object of class ssr which is a general/generalized/correlated smoothing spline fit.

Usage

```r
ssr(formula, rk, data = sys.parent(), subset, weights = NULL, correlation = NULL, family = "gaussian", scale = FALSE, spar = "v", varht = NULL, limmla = c(-10, 3), control = list())
```
Arguments

formula a formula object, with the response on the left of a ~ operator, and the bases of the null space $H_0$, separated by + operators, on the right. Thus it specifies the parametric part of the model that contains functions which are not penalized.

rk a list of expressions specifying reproducing kernels $R_1, \ldots, R_p$ for $H_1, \ldots, H_p$. For $p = 1$, rk may be specified with given functions. Supported functions are: "linear", "cubic", "quintic", and "septic" for linear, cubic, quintic and septic polynomial splines with "linear2", "cubic2", "quintic2", and "septic2" for another construction; "periodic" for periodic splines; "shrink0" and "shrink1" for Stein’s shrink-toward-zero and shrink-toward-mean estimates; "tp" for thin-plate-splines; "lspline" for L-splines. For details on these kernels, see their help files. Users may also write their own functions.

data a data frame containing the variables occurring in the formula and the rk. If this option is not specified, the variables should be on the search list. Missing values are not allowed.

subset an optional expression indicating which subset of the rows of the data should be used in the fit. This can be a logical vector (which is replicated to have length equal to the number of observations), a numeric vector indicating which observation numbers are to be included, or a character vector of the row names to be included. All observations are included by default.

weights a vector or a matrix specifying known weights for weighted smoothing, or a varFunc structure specifying a variance function structure. Its length, if a vector, or its number of columns and rows, if a matrix, must be equal to the length of responses. See documentations of nlme for available varFunc structures. The default is that all weights are equal.

correlation a corStruct object describing the correlation structure for random errors. See documentations of corClasses for available correlation structures. Default is NULL for no correlation.

family an optional string specifying the distribution family. Families supported are "binary", "binomial", "poisson", "gamma" and "gaussian" for Bernoulli, binomial, poisson, gamma and Gaussian distributions respectively. Default is "gaussian".

scale an optional logical value. If TRUE, all covariates appearing in "rk" will be scaled into interval [0, 1]. This transformation will affect predict.ssr. Default is FALSE.

spar a character string specifying a method for choosing the smoothing parameter. "v", "m" and "u" represent GCV, GML and UBR respectively. "u~", only used for non-Gaussian families, specifies UBR with an estimated variance. Default is "v".

varht needed only when 'u' is chosen for 'method', which gives the fixed variance in calculation of the UBR function. Default is NULL for 'family="gaussian"' and 1 for all other families.

limnl a vector of length one or two, specifying a search range for log10(n*lambda), where lambda is the smoothing parameter and n is the sample size. If it is a single value, the smoothing parameter will be fixed at this value. This option is only applicable to spline smoothing with a single smoothing parameter.
control a list of iteration and algorithmic constants. See ssr.control for details and default values.

Details
We adopt notations in Wahba (1990) for the general spline and smoothing spline ANOVA models. Specifically, the functional relationship between the predictor and independent variable is unknown and is assumed to be in a reproducing kernel Hilbert space $H$. $H$ is decomposed into $H_0$ and $H_1 + \ldots + H_p$, where the null space $H_0$ is a finite dimensional space spanned by bases specified at the right side of in formula, and $H_1, \ldots, H_p$ are reproducing kernel Hilbert spaces with reproducing kernel specified in the list rk.

The function is estimated from weighted penalized least square. ssr can be used to fit the general spline and smoothing spline ANOVA models (Wahba, 1990), generalized spline models (Wang, 1997) and correlated spline models (Wang, 1998). ssr can also fit partial spline model with additional parametric terms specified in the formula (Wahba, 1990).

ssr could be slow and memory intensive, especially for large sample size and/or when $p$ is large. For fitting a cubic spline with CV or GCV estimate of the smoothing parameter, the S-Plus function smooth.spline is more efficient.

Components can be extracted using extractor functions predict, deviance, residuals, and summary. The output can be modified using update.

Value
an object of class ssr is returned. See ssr.object for details.

Note: output from earlier versions of ssr shows incorrect smoothing spline parameters for SSANOVA, which is corrected in this version.

Author(s)
Yuedong Wang <yuedong@pstat.ucsb.edu> and Chunlei Ke <chunlei\_ke@yahoo.com>

References

See Also
deviance.ssr, hat.ssr, plot.ssr, ssr.control, predict.ssr, print.ssr, ssr.object, summary.ssr, smooth.spline.
Examples

```r
# fitting a cubic spline
# Not run: fit1 <- ssr(accel~times, data=mcycle, scale=T, rk=cubic(times))
# Not run: summary(fit1)

# using GML to choose the smoothing parameter
# Not run: fit2 <- update(fit1, spar="m")

data(acid)
# fit an additive thin plate spline
acid.fit <- ssr(ph ~ t + x1 + x2, rk=list(tp(t), tp(list(x1, x2))),
data = acid, spar = "m", scale = FALSE)
acid.fit
```

**ssr.control**  
*Set Control Parameters for ‘ssr’*

**Description**

The values supplied in the function call replace the defaults and a list with all possible arguments is returned. The returned list is used as the ‘control’ argument to the ‘ssr’ function.

**Usage**

```r
ssr.control(job=-1, tol=0.0, init=0.0, theta, prec=1e-06, maxit=30, tol.g=0.0, prec.g=1e-06, maxit.g=30)
```

**Arguments**

- **job**: an integer representing the optimization method used to find the smoothing parameter. The options are job=-1: golden-section search on (limnla(1), limnla(2)); job=0: golden-section search with interval specified automatically; job >0: regular grid search on [limnla(1), limnla(2)] with \#(grids) = job + 1. Default is -1. This is only applicable to smoothing spline model with a single smoothing parameter.

- **tol**: tolerance for truncation used in ‘dsidr’ or ‘dmudr’. Default is 0.0 which sets to square of machine precision.

- **init**: init=0 means no initial values are provided for smoothing parameters theta; init=1 means initial values are provided for the theta. Default is 0. This option is only applicable to smoothing spline models with multiple smoothing parameters.

- **theta**: If init=1, theta includes intial values for smoothing parameters. Default is NULL. This is only applicable to smoothing spline models with multiple smoothing parameters.
prec precision requested for the minimum score value in ‘dmudr’, where precision is the weaker of the absolute and relative precisions. Default is 1e-06. This is only applicable to smoothing spline models with multiple smoothing parameters.

maxit maximum number of iterations allowed in ‘dmudr’. Default is 30. This is only applicable to smoothing spline model with multiple smoothing parameters.

tol.g the tolerance for elements of w’s in GRKPK. Default is 0.0 which means using the machine precision. This is only applicable to generalized spline smoothing.

prec.g precision for stopping the iteration in GRKPK. Default is 1e-06. This is only applicable to generalized spline smoothing.

maxit.g maximum number of iterations allowed for the iteration in GRKPACK. Default is 30. This is only applicable to generalized spline smoothing.

Value

a list with components for each of the possible arguments.

See Also

ssr

Examples

# use regular grid search method with 100 grid points
ssr.control(job=99)

ssr.object A fitted ssr Object

Description

An object returned by the ssr function, inheriting from class ssr, and representing a fitted smoothing spline model. Objects of this class have methods for the generic functions predict, print and summary.

Value

The following components must be included in a legitimate ssr object:

call a list containing an image of the ssr call that produced the object
coef estimated coefficients for the spline estimate
lambda a vector representing the estimate smoothing parameters
fitted fitted values of the unknown mean function
family the distribution family used
cor.est estimated parameters, if any, in corMatrix
var.est estimated parameters, if any, in varFunc
s  design matrix extracted from formula
q  a list of matrices representing reproducing kernels evaluated at design points.
residuals  working residuals from the fit.
df  equivalent degrees of freedom. It is calculated as the trace of the hat matrix.
weight  a matrix representing the covariance matrix. It is NULL for iid data.
rkpk.obj  an object representing fits from dsidr/dmudr/gdsidr/gdmudr. See help files for
dsidr/dmudr/gdsidr/gdmudr for more details.
scale  a logical value, specifying if scaling is used.

Author(s)
Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
ssr, predict.ssr, summary.ssr, plot.ssr, dsidr, dmudr, gdsidr, gdmudr

star  

Magnitude of the Mira Variable R Hydrae

Description

The star data frame has 1086 rows and 2 columns of data from the Mira Variable R Hydrae

Usage
data(star)

Details

This dataset contains magnitude (brightness) of the Mira variable R Hydrae during 1900-1950.

Format

The data frame contains the following columns:
time a numeric vector of the observation time in days
magnitude a numeric vector of brightness of the Mira variable R Hydrae

Source


references

Stratford

Daily maximum temperatures in Stratford

Description

The Stratford data frame has 73 rows and 2 columns of data containing daily maximum temperatures in Stratford every five days in 1990.

Usage

data(Stratford)

Details

Daily maximum temperatures from the station in Stratford, Texas, in the year 1990 were extracted. The year was divided into 73 five-day periods and measurements on the third day in each period were selected as observations.

Format

The data frame contains the following columns:

- x a numeric vector representing time in a year scaled into [0,1]
- y a numeric vector of the observed maximum temperature in Fahrenheit

Source

This is part of a climate dataset downloaded from the Carbon Dioxide Information Analysis Center at http://cdiac.ornl.gov/ftp/ndp070.

summary.nnr

Object Summaries

Description

Summarize a nnr object

Usage

## S3 method for class 'nnr'
summary(object, ...)

Arguments

- object a fitted nnr object.
- ... unused argument
Details

This is a method for the function summary for objects inheriting from class nnr. See summary for the general behavior of this function.

Author(s)

Chunlei Ke <chunlei_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

nnr, print.nnr

summary.slm

Object Summaries

Description

Summarize a slm object

Usage

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

Arguments

object       a fitted slm object.
...          unused argument

Details

This is a method for the function summary for objects inheriting from class slm.

Author(s)

Chunlei Ke <chunlei_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

slm, print.slm
summary.snm

Object Summaries

Description
Summarize a snm object

Usage
```
## S3 method for class 'snm'
snmm::summary(object, ...)
```

Arguments
- `object` : a fitted 'snm' object.
- `...` : unused argument

Details
This is a method for the function summary for objects inheriting from class snm.

Author(s)
Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also
- snm, print.snm

summary.snr

Object Summaries

Description
Summarize a snr object

Usage
```
## S3 method for class 'snr'
snmm::summary(object, ...)
```

Arguments
- `object` : a fitted snr object.
- `...` : unused argument
Details

This is a method for the function `summary` for objects inheriting from class `snr`. See `summary` for the general behavior of this function.

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

`snr`, `print.snr`

---

`summary.ssr`  
*Summarize a ssr object*

Description

Provides a synopsis of a ssr object and perform tests.

Usage

```r
## S3 method for class 'ssr'
summary(object, ...)```

Arguments

- `object`  
a fitted ssr object.
- `...`  
unused option.

Details

This is a method for the function `summary` for objects inheriting from class `ssr`.

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

See Also

`ssr`, `print.ssr`
Calculate Reproducing Kernels for Thin Plate Splines

Description

Return a matrix evaluating reproducing kernels for thin plate splines at observed points.

Usage

\[
\text{tp.pseudo}(s, u=s, \text{order}=2) \\
\text{tp}(s, u=s, \text{order}=2) \\
\text{tp.linear}(s, u=s)
\]

Arguments

- **s**: a list or matrix of observations. One component, if a list, and one column, if a matrix, contains observations on one variable. If a list, all components must be of the same length.
- **u**: a list or matrix of observations. If a list, all components must be of the same length. The number of components of the list, or the number of column of the matrix must be the same as that for s. Default is s.
- **order**: an optional integer specifying the order of the thin plate spline. Default is 2. Let d be the dimension of s (and u). Then order must satisfy \(2 \times \text{order} - d > 0\).

Details

The pseudo kernel, which is conditional definite positive instead of definite positive, is easy to calculate, while the true reproducing kernel is complicated. Pseudo Kernels are enough to compute spline estimates, but to calculate Bayesian confidence intervals, the true kernel is required. For the special case of d=2 and order=2, the function \text{tp.linear} computes evaluations of the reproducing kernel of the space spanned by linear basis.

Value

A matrix with the numbers of row and column equal to the common length of components or the number of row of s and t respectively. The [i, j] element is the pseudo, true, or linear reproducing kernel evaluated at the i_th element of s and j_th element of u.

Author(s)

Chunlei Ke <chunlei\_ke@yahoo.com> and Yuedong Wang <yuedong@pstat.ucsb.edu>

References

The data frame `txtemp`, obtained from the Carbon Dioxide Information and Analysis Center at Oak Ridge National Laboratory, has 17280 rows and 6 columns of data representing monthly temperature records for stations in Texas.

Usage

data(txtemp)

Format

The data frame contains the following columns:

- `stacode`: a numeric vector of the unique station code formed by combining the two-digit state number [state numbers range from 1 to 48] and the four-digit station number (values range from 0008 to 9933);
- `lat`, `long`: numeric vectors identifying the latitudes and longitudes of the stations in decimal degree.
- `year`: a numeric vector comprising the year for the records.
- `month`: a numeric vector of values 1 to 12, representing the month for the data.
- `mmtemp`: a numeric vector of monthly average temperature in Fahrenheit scale.

Details

The data set was extracted from a large national historical climate data, containing data for 48 stations in Texas from 1961 to 1990. Monthly temperature records as well as the latitude and longitude for each station were available.

Of note, the missing values were coded as -99.99.

Source

Data are downloadable from [http://cdiac.ornl.gov/ftp/ndp019/](http://cdiac.ornl.gov/ftp/ndp019/)
ultrasound

---

Ultrasound imaging of the tongue shape

Description

The 'ultrasound' data frame has 1,215 rows and 4 columns of data from an ultrasound experiment.

Usage

data(ultrasound)

Details

A Russian speaker produced the consonant sequence, /gd/, in three different linguistic environments: '2words', 'cluster' and 'Schwa', with three replications for each environment. 15 points from each of 9 slices of tongue curves separated by 30 ms (milliseconds) are extracted. Therefore, in total there are $15 \times 9 \times 3 \times 3 = 1,215$ observations.

Format

The data frame contains the following columns:
- height: a numeric vector of tongue height in mm
- length: a numeric vector of tongue length in mm
- time: a numeric vector of time in ms
- env: a factor with three levels: 1, 2 and 3 for environment '2words', 'cluster' and 'Schwa' respectively.

Source

Phonetics-Phonology Lab of New York University.

References

USAtemp

Average Winter temperature in the United States

Description

The USAtemp data frame has 1214 rows and 3 columns of data containing average Winter temperatures in 1981 from 1205 stations in USA.

Usage

data(USAtemp)

Format

The data frame contains the following columns:

- temp a numeric vector of average temperatures (Fahrenheit)
- lat a numeric vector of the latitude of a station
- long a numeric vector of the longitude of a station

details

The average Winter temperatures are calculated as the averages of temperatures in December, January and February. The geological locations of 1214 stations are given in terms of longitude and latitude.

wesdr

Wisconsin Epidemiological Study of Diabetic Retinopathy

Description

The wesdr data frame has 669 rows and 5 columns of data from an ongoing epidemiological study of a cohort of patients receiving their medical care in an 11-country area in southern Wisconsin.

Usage

data(wesdr)

Details

The progression of diabetic retinopathy was assessed together with a number of medical, demographic, ocular and other covariates and the retinopathy scores.
Format

This data frame contains the following columns:
num a numeric vector giving IDs for individuals.
dur a numeric vector of duration of at baseline in year.
gly a numeric vector of glycosylated hemoglobin, a measuer of hyperglycemia.
bmi a numeric vector of body mass index, weight in $kg/(height \text{in meter})^2$.
prg a vector of 0 or 1’s representing disease progression for each individual.

Source

Klein, R., Klein, B. E. K., Moss, S. E., Davis, M. D. and Demets, D. L. (1989b). The Wisconsin epidemiologic study of diabetic retinopathy. X. Four year incidence and progression of diabetic retinopathy when age at diagnosis is less than 30 years. Arch. Ophthalmal. 107, 244-249.

Description

Extend xyplot to superpose one or more symbols to each panel.

Usage

xyplot2(formula, data, type = "l", ...)

Arguments

formula a two-sided formula as accepted in xyplot
data a list of data frames. Each component shall be able to evaluate the variables appearing in formula
type a vector of characters to indicate what type of plots are to draw. Default is line.
... any options as accepted in xyplot

Value

On each panel, several plot types, the length of data, are superposed.
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