

# Package ‘CommonSplines’

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**Title** Regression Spline and Smoothing Spline

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**Authors** Xingchen LIU <e0225109@u.nus.edu>, Yuchen SHI <yuchenshinus@gmail.com>, Xiaozhou Yang <yang\_xiaozhou@icloud.com>

**Description** This is an R package that covers commonly seen regression spline and smoothing spline. For regression spline, commonly seen basis functions are provided such as truncated power basis, natural spline basis and B-spline basis. For smoothing spline, penalties on second order derivative are provided, i.e., cubic smoothing spline.

**Depends** R (>= 3.3.2)

**License** Apache License 2.0

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 6.0.1

**Suggests** knitr,  
rmarkdown

**VignetteBuilder** knitr

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bs_predict	<i>Regression using B-spline basis</i>
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### Description

This function provides nonparametric regressions using B-splines. The B-splines are generated by the function `bs_train`. The return value of `bs_train` is required as an argument of `bs_predict`

### Usage

```
bs_predict(x_test, order = NULL, knots = NULL, beta = NULL,
          basis = NULL)
```

### Arguments

<code>x_test</code>	The input values at which evaluations are required.
<code>basis</code>	The return value of function <code>bs_train</code> .

### Value

The evaluated output at `x_test`.

### Examples

```
x<-seq(0, 1, 0.001)
y <- x^3 * 3 - x^2 * 2 + x + exp(1)+rnorm(length(x),0,0.1)
plot(x,y)
innerknots <- seq(0.1, 0.9, 0.01)
order<-4
basis<-bs_train(x,y,order,innerknots)

x_test<-seq(0, 1, 0.01)
fit<-bs_predict(x_test,basis)
plot(x_test,fit)
lines(x_test,x_test^3 * 3 - x_test^2 * 2 + x_test + exp(1),col="red")
```

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bs_train	<i>Generating B-spline basis</i>
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### Description

This function generates B-spline basis. The B-splines are defined following the recursive formulas due to de Boor. Only univariate input can be used.

### Usage

```
bs_train(x, y, order, real_knots)
```

**Arguments**

x	The input vector of training dataset.
y	The output vector of training dataset.
order	The order of B-spline functions. The default is order=4 for cubic B-splines.
knots	The innerknots and boundary knots that define the spline. knots should contain knots on the boundary.

**Value**

A list with the following components:

beta	The coefficients of nonparametric regression.
basis	The B-spline basis matrix of dimension $c(\text{length}(x), \text{df})$ . $\text{df} = \text{length}(\text{innerknots}) + \text{order}$ .
knots	The knots used to construct the B-splines, including innerknots, boundary knots and phantom knots
order	The order of basis functions. $\text{order} = \text{degree} + 1$

**Examples**

```
x<-seq(0, 1, 0.001)
y <- x^3 * 3 - x^2 * 2 + x + exp(1)+rnorm(length(x),0,0.1)
plot(x,y)
knots <- seq(0, 1, 0.1)
order<-4

basis<-bs_train(x,y,order,knots)
plot(x,rep(0,length(x)),type="l",ylim=c(0,1))
for (i in 1: (length(knots)+order)){
  lines(x,basis$basismatrix[,i])
}
```

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cal_loo_cv_error	<i>Calculate leave-one-out CV error</i>
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**Description**

Calculate leave-one-out CV error

**Usage**

```
cal_loo_cv_error(y, f_hat, S)
```

**Arguments**

y	response variable values
f_hat	fitted response variable values
S	smoother matrix

**Value**

leave-one-out cross-validation error

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css_predict	<i>Prediction using smoothing spline with squared 2nd derivative penalty</i>
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### Description

This function takes the coefficients trained by `CubicSmoothingSpline.Train` and evaluate the output at `x_test`

### Usage

```
css_predict(basis, x_test)
```

### Arguments

<code>basis</code>	The return value of function <code>CubicSmoothingSpline.Train</code> .
<code>x_test</code>	The input values at which evaluations are required.

### Value

The evaluated output at `x_test`.

### Examples

```
x<-seq(0, 1, 0.0015)
y <- x^3 * 3 - x^2 * 2 + x + exp(1)+rnorm(length(x),0,0.1)
plot(x,y)
lambda<-0.001
basis<-css_train(x,y,lambda)

x_test<-seq(0, 1, 0.1)
fit<-css_predict(basis,x_test)

plot(x_test,fit)
lines(x_test,x_test^3 * 3 - x_test^2 * 2 + x_test + exp(1),col="red")
```

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css_train	<i>Train a smoothing spline with squared 2nd derivative penalty using natural cubic spline</i>
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### Description

This function trains a smoothing spline with squared 2nd derivative penalty. It has an explicit,finite-dimensional, unique minimizer which is a natural cubic spline. This function can be used for small or moderate number of knots. When the number of data  $N \leq 50$ , all knots are included. When  $N > 50$ , 50 knots are uniformly chosen from the training dataset.

### Usage

```
css_train(x, y, lambda)
```

**Arguments**

x	The input vector of training dataset.
y	The output vector of training dataset.
lambda	A fixed smoothing parameter.

**Value**

A list with the following components:

beta	The coefficients of natural splines.
S	The smoother matrix.
knots	The knots used to construct the B-splines, including innerknots, boundary knots and phantom knots

**Examples**

```
x<-seq(0, 1, 0.001)
y <- x^3 * 3 - x^2 * 2 + x + exp(1)+rnorm(length(x),0,0.1)
plot(x,y)
lambda<-0.001

basis<-css_train(x,y,lambda)
cat("the knots chosen are: ",basis$knots)
```

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ncs_eval_basis	<i>Evaluate basis functions as each x and return the evaluated basis matrix N</i>
----------------	---

---

**Description**

Evaluate basis functions as each x and return the evaluated basis matrix N

**Usage**

```
ncs_eval_basis(x, knots)
```

**Arguments**

x	Predictor variable vector.
knots	Knots location in terms of quantiles of x_train, optional, default will be evenly spaced quantiles based on number of knots.
nknots	Number of knots used in training.

**Value**

Basis matrix evaluated at each x value.

---

ncs_predict	<i>Prediction based on trained regression model</i>
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**Description**

Prediction based on trained regression model

**Usage**

```
ncs_predict(x_test, beta, knots)
```

**Arguments**

x_test	The input values at which evaluations are required.
knots	Knots location in terms of quantiles of x_train, optional, default will be evenly spaced quantiles based on number of knots.
betas	Least square fit parameters obtained from training.

**Value**

y_pred	A vector of dimension length(x), the prediction vector evaluated at x_test values.
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ncs_train	<i>Generate an evaluated basis matrix for natural cubic splines</i>
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**Description**

Generate an evaluated basis matrix for natural cubic splines

**Usage**

```
ncs_train(x_train, y_train, df = NULL, knots = NULL)
```

**Arguments**

x_train	The input vector of training dataset.
y_train	The output vector of training dataset.
df	Degrees of freedom. One can supply df rather than knots; ncs() then chooses (df + 1) knots at uniform quantiles of x. The default, df = 4, sets 5 knots with 3 inner knots at uniform quantiles of x.
knots	Breakpoints that define the spline, in terms of quantiles of x. The default is five knots at uniform quantiles c(0, .25, .5, .75, 1). Typical values are the mean or median for one knot, quantiles for more knots.

**Value**

A list of following components:

nknots	Number of knots.
knots	A vector of knot locations.
N	Basis matrix evaluated at each x value.
betas	Least square fit parameters.

**Examples**

```
x_train <- seq(1, 10, 0.1)
y_train <- cos(x_train)^3 * 3 - sin(x_train)^2 * 2 + x_train + exp(1)+rnorm(length(x_train),0,1)
plot(x_train,y_train)
x_test <- seq(1, 10, 0.1)
df <- 10
train_result <- ncs_train(x_train, y_train, df)
print(train_result$betas)
print(train_result$N[1:5,1:5])
```

np\_reg

*Nonparametric Regression using spline based methods***Description**

This function provides regression using natural cubic splines with truncated power basis functions. Only univariate input can be used.

**Usage**

```
np_reg(x_train, y_train, x_test, func = "bs", order = 3, df = NULL,
       knots = NULL, lambda = 0.001)
```

**Arguments**

x_train	The input vector of training dataset.
y_train	The output vector of training dataset.
x_test	The input values at which evaluations are required.
df	Degrees of freedom. One can supply df rather than knots; (df + 1) knots are chosen at uniform quantiles of x. The default, df = 4, sets 5 knots with 3 inner knots at uniform quantiles of x.
knots	Breakpoints that define the spline. The default is five knots at uniform quantiles c(0, .25, .5, .75, 1). Typical values are the mean or median for one knot, quantiles for more knots.

**Value**

y\_pred A vector of dimension length(x), the prediction vector evaluated at x\_test values.

**Examples**

```

x_train <- seq(1, 10, 0.1)
y_train <- cos(x_train)^3 * 3 - sin(x_train)^2 * 2 + x_train + exp(1)+rnorm(length(x_train),0,1)
plot(x_train,y_train)
title('Comparison of Different Degrees of Freedom')
x_test <- seq(1, 10, 0.1)
lines(x_test,cos(x_train)^3 * 3 - sin(x_train)^2 * 2 + x_train + exp(1),col="red")

df <- 2
y_pred <- np_reg(x_train, y_train, x_test,func="ncs", df=df)
lines(x_test,y_pred, col='blue')
df <- 4
y_pred <- np_reg(x_train, y_train, x_test,func="ncs", df=df)
lines(x_test,y_pred, col='green')
df <- 10
y_pred <- np_reg(x_train, y_train, x_test,func="ncs", df=df)
lines(x_test,y_pred, col='black')
legends <- c("Actual", "Prediction: 2 df", "Prediction: 4 df", "Prediction: 10 df")
legend('topleft', legend=legends, col=c('red', 'blue', 'green', 'black'), lty=1, cex=0.8)

```

---

pbs\_basis

*Evaluate basis functions as each x and return the evaluated basis matrix N*

---

**Description**

Evaluate basis functions as each x and return the evaluated basis matrix N

**Usage**

```
pbs_basis(x, order, knots)
```

**Arguments**

x	Predictor variable vector.
knots	Knots location in terms of quantiles of x_train, optional, default will be evenly spaced quantiles based on number of knots.
nknots	Number of knots used in training.

**Value**

Basis matrix evaluated at each x value.



pbs\_train

*Regression using Power Basis spline***Description**

This function provides regressions using Power Basis splines. The basis are defined as  $1, x, x^2, \dots, x^m, (x-k_1)^{+(m-1)}, (x-k_2)^{+(m-1)}, \dots, (x-k_n)^{+(m-1)}$  where  $m$  is the order,  $k_1, k_2$  and  $k_n$  are  $n$  knots, '+' denotes the positive part.

**Usage**

```
pbs_train(x, y, order, knots)
```

**Arguments**

<code>x</code>	The input vector of training dataset.
<code>y</code>	The output vector of training dataset.
<code>order</code>	The order that defines the spline.
<code>knots</code>	The internal knots that define the spline.
<code>x_test</code>	The input values at which evaluations are required.

**Details**

Only univariate input can be used.

**Value**

A list with the following components:

<code>beta</code>	The coefficients of nonparametric regression.
<code>basis</code>	The spline basis matrix of dimension $c(\text{length}(x), \text{length}(\text{knots}) + \text{order})$
<code>f</code>	The evaluated output at <code>x_test</code> .

**Examples**

```
n <- 100
t <- seq(0, 2*pi, length.out = 100)
a <- 3
b <- 2
c.unif <- runif(n)
amp <- 2
set.seed(1)
y1 <- a*sin(b*t) + c.unif*amp # uniform error
knots <- c(min(t), 2*pi*c(1/4, 2/4, 3/4), max(t))
order <- 4
basis <- pbs_train(t, y1, order, knots)
fit <- pbs_predict(t, basis=basis)
y.hat <- fit
plot(t, y1, t="l")
lines(t, y.hat, col=2)
```

---

place_knots	<i>Find evenly spaced knots by quantile</i>
-------------	---

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**Description**

Knots found include boundary knots at 0th and 100th quantile.

**Usage**

```
place_knots(nknots, x)
```

**Arguments**

nknots	Number of knots to be located.
x	Data vector on which knots are placed.

**Value**

A named vector with knot quantiles and values.

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sel_smoothing_para	<i>Select smoothing parameter based on leave-one-out CV error</i>
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---

**Description**

Select smoothing parameter based on leave-one-out CV error

**Usage**

```
sel_smoothing_para(x, y, cv_lambda)
```

**Arguments**

x	predictor variable
y	response variable
cv_lambda	vector of candidate lambda values

**Value**

lambda value that minimizes leave-one-out CV error

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