

# Package ‘rgcvpack’

August 28, 2007

**Version** 0.1-2

**Date** 2007/08/28

**Title** R Interface for GCVPACK Fortran Package

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**Description** Thin plate spline fitting and prediction

**Depends** R (>= 2.1.0)

**License** GPL version 2 or later

**URL** <http://www.stat.wisc.edu/~xie>

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fitTps	<i>Fitting Thin Plate Smoothing Spline</i>
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## Description

Fit thin plate splines of any order with user specified knots

## Usage

```
fitTps(x, y, m = 2, knots = NULL, scale.type = "range", method = "v",  
       lambda = NULL, cost = 1, nstep.cv = 80, verbose = FALSE, tau = 0)
```

**Arguments**

x	the design data points
y	the observation vector
m	the order of the spline
knots	the placement the thin plate spline basis
scale.type	"range" (default), the x and knots will be rescaled with respect to x; "none", nothing is done on x and knots
method	"v", GCV is used for choosing lambda; "d", user specified lambda
lambda	only used when method="d"
cost	the fudge factor for inflating the model degrees of freedom, default to be 1
nstep.cv	the number of initial steps for GCV grid search
verbose	whether some computational details should be outputed
tau	the truncation ratio used in SVD when knots is specified by the user, some possible values are 1, 10, 100, ...

**Details**

The minimization problem for this function is

$$\sum_{i=1}^n (y_i - f(x_i))^2 + \lambda * J_m(f),$$

where  $J_m(\cdot)$  is the m-the order thin plate spline penalty functional.

If scale.type="range", each column of x is rescaled to [0 1] in the following way  $x' = (x - \min(x))/\text{range}(x)$ , and the knots is rescaled w.r.t.  $\min(x)$  and  $\text{range}(x)$  in the same way.

When the cost argument is used, the GCV score is computed as

$$\text{GCV}(\lambda) = \frac{n * \text{RSS}(\lambda)}{(n - \text{cost} * \text{tr}(A))^2}.$$

**Value**

A Tps object of the following components

x	same as input
y	same as input
m	same as input
knots	same as input
scale.type	same as input
method	same as input
lambda	same as input
cost	same as input
nstep.cv	same as input
tau	same as input
df	model degrees of freedom
gcv	gcv score of the model adjusted for the fudge factor

xs	scaled design points
ks	scaled knots design
c	coefficient c
d	coefficient d
yhat	predicted values at the data points
svals	singular values of the matrix decomposition
gcv.grid	gcv grid table, number of rows=nstep.cv
call	the call to this function

### Note

This function uses GCVPACK fortran code with some addition and modification by the author.

### Author(s)

Xianhong Xie

### References

D. Bates, M. Lindstrom, G. Wahba, B. Yandell (1987), GCVPACK – routines for generalized cross-validation. Commun. Statist.-Simula., 16(1), 263-297.

### See Also

[predict.Tps](#)

### Examples

```
#define the test function
f <- function(x, y) { .75*exp(-((9*x-2)^2 + (9*y-2)^2)/4) +
  .75*exp(-((9*x+1)^2/49 + (9*y+1)^2/10)) +
  .50*exp(-((9*x-7)^2 + (9*y-3)^2)/4) -
  .20*exp(-((9*x-4)^2 + (9*y-7)^2)) }

#generate a data set with the test function
set.seed(200)
N <- 13; xr <- (2*(1:N) - 1)/(2*N); yr <- xr
zr <- outer(xr, yr, f); zrmax <- max(abs(zr))

noise <- rnorm(N^2, 0, 0.07*zrmax)
zr <- zr + noise #this is the noisy data we will use

#convert the data into column form
xc <- rep(xr, N)
yc <- rep(yr, rep(N,N))
zc <- as.vector(zr)

#fit the thin plate spline with all the data points as knots
tpsfit1 <- fitTps(cbind(xc,yc), zc, m=2, scale.type="none")
persp(xr, yr, matrix(predict(tpsfit1),N,N), theta=130, phi=20,
  expand=0.45, xlab="x1", ylab="x2", zlab="y", xlim=c(0,1),
  ylim=c(0,1),zlim=range(zc), ticktype="detailed", scale=FALSE,
  main="GCV Smooth I")
```

```
#fit the thin plate spline with subset of data points as knots
grid.list <- list(xc=seq(2/13,11/13,len=10),
                  yc=seq(2/13,11/13,len=10))
knots.grid <- expand.grid(grid.list)

tpsfit2 <- fitTps(cbind(xc,yc), zc, m=2, knots=knots.grid)
persp(xr, yr, matrix(predict(tpsfit2),N,N), theta=130, phi=20,
      expand=0.45, xlab="x1", ylab="x2", zlab="y", xlim=c(0,1),
      ylim=c(0,1),zlim=range(zc), ticktype="detailed", scale=FALSE,
      main="GCV Smooth II")
```

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predict.Tps

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*Predicting Thin Plate Smoothing Spline*


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## Description

Predict the thin plate spline fitting at given new data points.

## Usage

```
predict.Tps(object, newdata = NULL, ...)
```

## Arguments

object	a Tps object returned by fitTps
newdata	the new data to be predicted at
...	currently not used

## Value

A vector with the length = the number of rows in newdata.

## Note

This function uses GCVPACK fortran code with some addition and modification by the author.

## Author(s)

Xianhong Xie

## See Also

[fitTps](#)

**Examples**

```

#the same test function as in fitTps
f <- function(x, y) { .75*exp(-((9*x-2)^2 + (9*y-2)^2)/4) +
  .75*exp(-((9*x+1)^2/49 + (9*y+1)^2/10)) +
  .50*exp(-((9*x-7)^2 + (9*y-3)^2)/4) -
  .20*exp(-((9*x-4)^2 + (9*y-7)^2)) }

#generate a data set with the test function
set.seed(200)
N <- 13; xr <- (2*(1:N) - 1)/(2*N); yr <- xr
zr <- outer(xr, yr, f); zrmax <- max(abs(zr))

noise <- rnorm(N^2, 0, 0.07*zrmax)
zr <- zr + noise #this is the noisy data we will use

#convert the data into column form
xc <- rep(xr, N)
yc <- rep(yr, rep(N,N))
zc <- as.vector(zr)

#fit the thin plate spline with all the data points as knots
tpsfit1 <- fitTps(cbind(xc,yc), zc, m=2, scale.type="none")

#predict the thin plate spline on a finer grid (50x50)
xf <- seq(1/26, 25/26, length=50); yf <- xf
zf <- predict(tpsfit1, expand.grid(xc=xf,yc=yf))

#plot the predicted result
persp(xf, yf, matrix(zf,50,50), theta=130, phi=20, expand=0.45,
  xlab="x1", ylab="x2", zlab="y", xlim=c(0,1), ylim=c(0,1),
  zlim=range(zc), ticktype="detailed", scale=FALSE,
  main="GCV Smoothing")

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