



# DierckxSpline: An R Package For Minimal Knot Splines

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July 29, 2007



# Agenda

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- **Splines in R**
- **FITPACK Routines**
- **Univariate Splines**
  - Smoothing splines
  - Least square splines
  - Free knot splines
- **The DierckxSpline package for R**
- **Examples**
- **Software Status and Extensions**

# Splines in R

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- **Many algorithms have been improved since Dierckx**
  - Better free knot selection algorithms
  - Applications for functional data analysis
  
- **Purpose of the package is to make available Dierckx FITPACK functions**
  - Univariate splines
  - Free knot splines
  - Bivariate splines
  
- **R lacks a comprehensive spline package**
  - `spline`
  - `smooth.spline`
  - Several packages
    - `splines` – Spline package for B-splines
    - `fda` – Functional Data Analysis
    - `ssr` – Spline Smoothing Regression
  - No splines package for free knots or constrained splines

# Dierckx FITPACK

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- The FITPACK library is available in Fortran from NETLIB

<http://www.netlib.org/dierckx>

- Includes

- Code to accompany *Curve and Surface Fitting with Splines*

Dierckx, P. (1993). *Curve and Surface Fitting with Splines*. Oxford Science Publications, New York.

- Examples and data
- Currently R package interfaces with approximately half of the provided functions

- Not to be confused with commercial FITPACK library <http://www.netlib.org/fitpack>

# Smoothing Splines

## Given

- Data:  $(x_r, y_r), r = 1, \dots, m$
- Constraints:  $a \leq x_r \leq x_{r+1} \leq b$
- Weights:  $w_r$

## Goal

- Determine spline  $s(x)$  on  $[a, b]$
- Degree:  $k$
- Knots:  $a = \lambda_0, \lambda_1, \dots, \lambda_g, \lambda_{g+1} = b$

## Unconstrained minimization

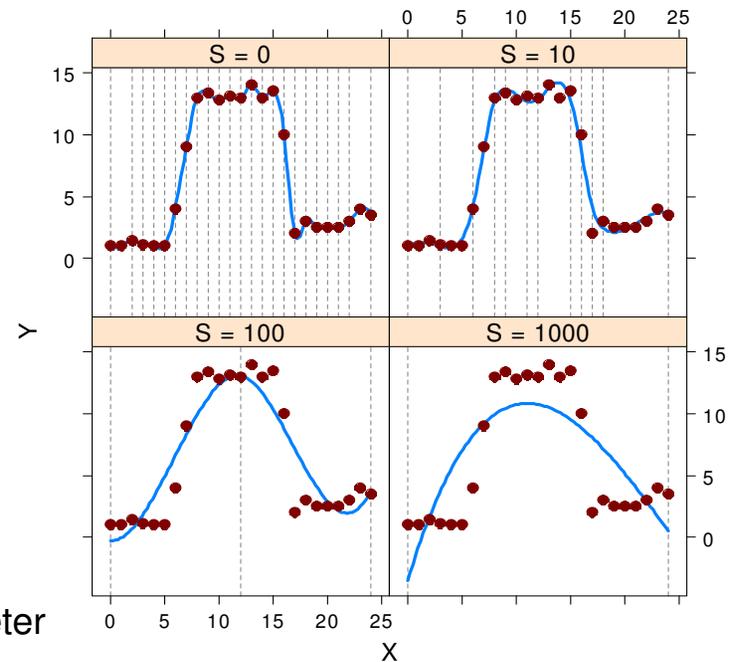
$$\text{Minimize } \tilde{\eta} := \sum_{i=1}^g (s^{(k)}(\lambda_i +) - s^{(k)}(\lambda_i -))^2$$

$$\text{Subject to } \delta := \sum_{i=1}^m (w_r (y_r - s(x_r)))^2 < S$$

- where  $S$  is some user-specified smoothing parameter
- Increase  $S \rightarrow$  increase smoothing

```

ss <- list()
s <- c(0, 10, 100, 1000)
for(i in seq(s)) {
  ss[[i]] <- curfit(x, y,
    s = s[i], method = "ss")
}
    
```



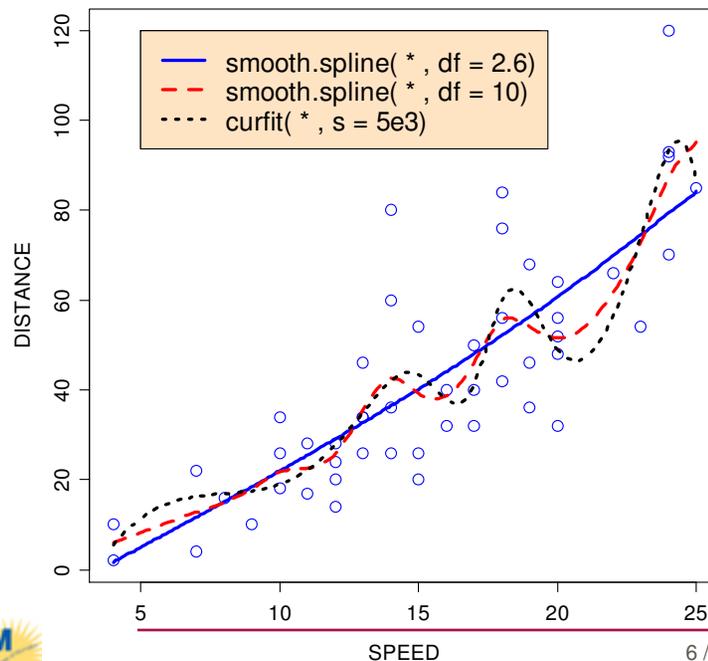
Vertical lines are knot placements

# Comparison To `smooth.spline`

- R has `smooth.spline` which is a competing function for smoothing splines

```
## example from ?smooth.spline
## This example has duplicate points, so avoid cv = TRUE
cars.spl.0 <- smooth.spline(cars$speed, cars$dist)
cars.spl.1 <- smooth.spline(cars$speed, cars$dist, df = 10)
cars.spl.2 <- curfit(cars$speed, cars$dist, s = 5e3)
```

data(cars) & smoothing splines



`smooth.spline` uses cross validation or equivalent degrees of freedom to determine the amount of smoothing

`curfit` constrains the model deviance

# Least Squares Splines With Fixed Knots

## ■ Fixed knots

- $a = \lambda_0, \lambda_1, \dots, \lambda_g, \lambda_{g+1} = b$

## ■ Minimize

$$\delta = \sum_{r=1}^m (w_r (y_r - s(x_r)))^2$$

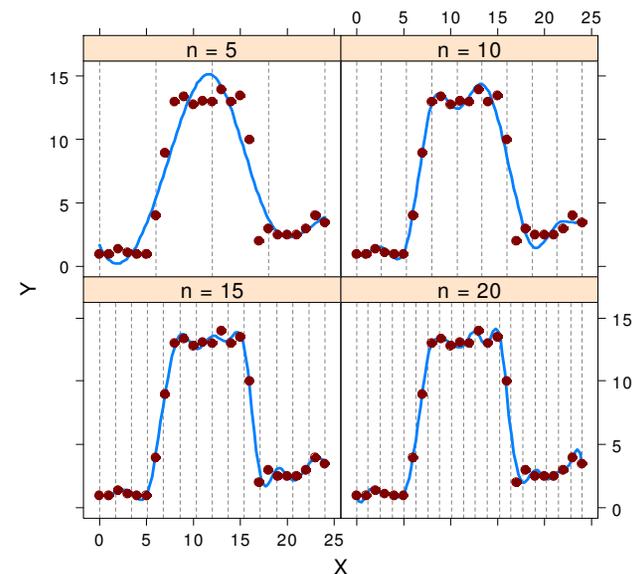
$$= \sum_{r=1}^m \left( w_r y_r - \sum_{i=-k}^g c_i w_r N_{i,k+1}(x_r) \right)^2,$$

where  $N_{i,k+1}$  are B-splines of degree  $k$  and  $c_i$  are the B-spline coefficients of  $s(x)$

## ■ Knots are user-determined

- There is no known R equivalent
- R function `spline` places a knot at each observation

```
n <- c(5, 10, 15, 25)
ls <- list()
for(i in seq(n)) {
  kn <- seq(0, 24, len = n[i])
  ls[[i]] <- curfit(x, y,
    method = "ls", knots = kn)
}
```



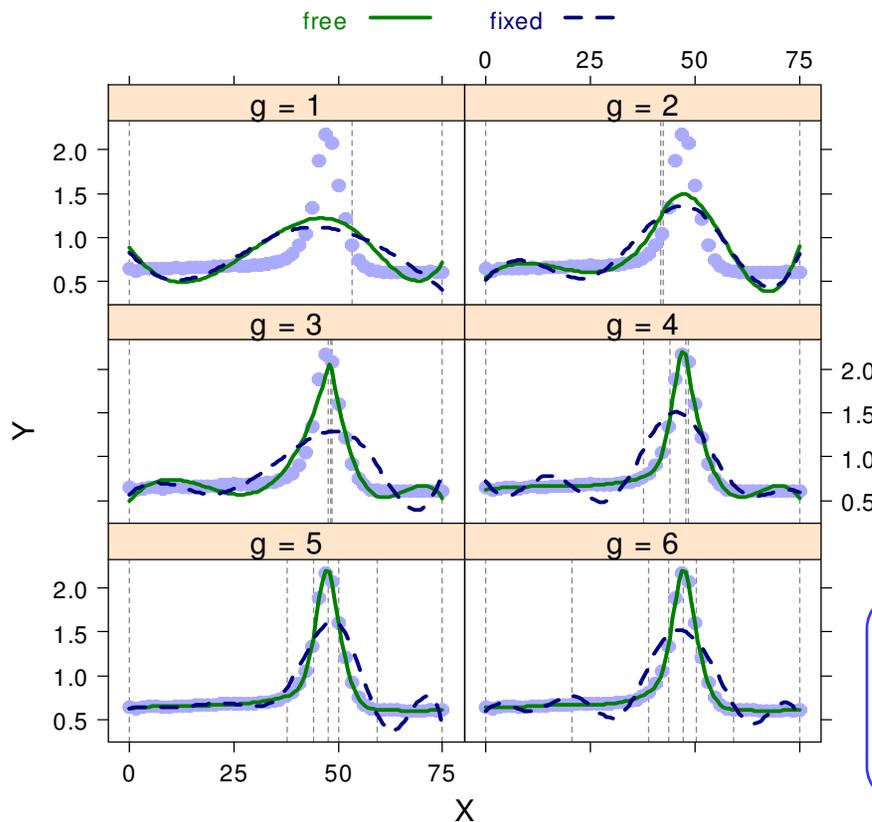
Vertical lines are knot placements

# Least Squares Splines with Variable Knots

## Titanium data (de Boors and Rice, 1968)

- Dierckx (1993) for optimizing number of knots
- We use `optim` to minimize the residual sums of squares

```
data(titanium)
r <- curfit.free.knot(titanium$x2,
titanium$y, g = 10, eps = 5e-4)
```



| g | sigma    | T     |
|---|----------|-------|
| 1 | 8.29E-02 | 5.95  |
| 2 | 4.81E-02 | 5.41  |
| 3 | 1.10E-02 | 5.01  |
| 4 | 1.56E-03 | 3.58  |
| 5 | 1.88E-04 | -0.84 |
| 6 | 1.33E-04 | -2.21 |

Optimal solution uses 5 interior knots

Plots include both fixed equally-spaced knots (dashed) and free knots (solid)

| knots (5) |
|-----------|
| 37.58     |
| 43.96     |
| 47.42     |
| 50.15     |
| 59.35     |

# Selecting An Appropriate Number Of Knots

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## ■ Algorithm described by Dierckx (1993)

- Supply starting value of  $\lambda_1^0 = (a + b)/2$  for the first knot, where  $a = \min(x)$  and  $b = \max(x)$
- Determine  $\lambda$  by minimizing a penalized RSS with user-defined  $\varepsilon$  and  $g = \text{length}(\lambda)$

$$\xi(\lambda) = \text{RSS}(\lambda) + \varepsilon \frac{(b-a)\text{RSS}(\lambda^0)}{(g+1)^2} \sum_{j=0}^g (\lambda_{j+1} - \lambda_j)^{-1}$$

- For  $j = 0, 1, 2, \dots, g$ , determine the region between knots with the largest RSS

$$\text{RSS}_j = \frac{1}{m - m_j} \sum_{i=q_j+1}^{q_j+m_j} (w_i (y_i - s_g(x_i)))^2,$$

where

$$\lambda_j \leq x_{q_j+1} < x_{q_j+2} < \dots < x_{q_j+m_j} \leq \lambda_{j+1}$$

- Add a new knot at the midpoint of  $\lambda_j$  and  $\lambda_{j+1}$  where  $\text{RSS}_j$  is maximized

## ■ Stopping criteria

$$T_g = \frac{\sqrt{m-1} \sum_{i=2}^m r_i r_{i-1}}{\sum_{i=1}^m r_i^2}$$

- Number of optimal knots  $g$  is determined by the first  $T_g < 0$

# The DierckxSpline Package

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## ■ Package Functions

- Includes interfaces for computing univariate splines
- FORTRAN for modeling bivariate

## ■ Examples and data included

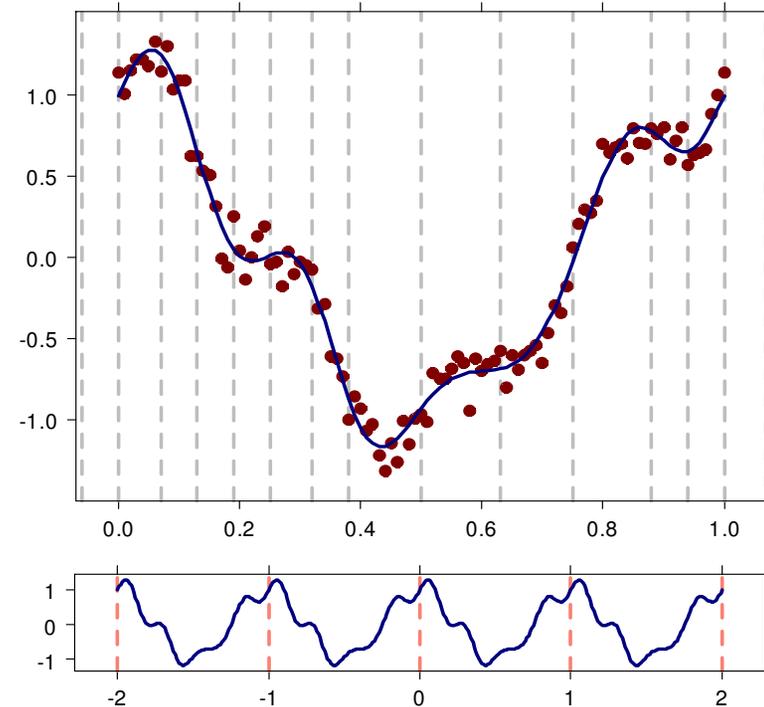
- Data
  - De Boor and Rice (1968) **titanium**
  - Dierckx (1980) volumetric **moisture** content
  - Soudan and Dierckx (1979) **knee** flexion-extension during walking
  - Additional data extracted from FITPACK
- **demo (DierckxSpline)**
  - Includes examples with data discussed
- **vignette (DierckxSpline)**
  - Provides more details on the spline fitting and algorithms
  - Includes relevant sections from DierckxSpline (1993)

# Example #1 – Smoothing With Periodic Splines

## ■ Quintic periodic smoothing spline

- Penalty: 90
- Periodic:  $s(a) == s(b)$

```
## periodic
set.seed(42)
n <- 100
r <- 1:n
x <- 0.01 * (r - 1)
e <- rnorm(n, 0, 0.1)
s2 <- var(e)
w <- rep(1/s2, n + 1)
y <- cos(2*pi*x) + 0.25*sin(8*pi*x) + e
x <- c(x, 1)
y <- c(y, y[1])
kn <- seq(0.01, 0.99, length = 12)
f1 <- percur(x, y, w, method = "ss",
             s = 90, k = 5)
```



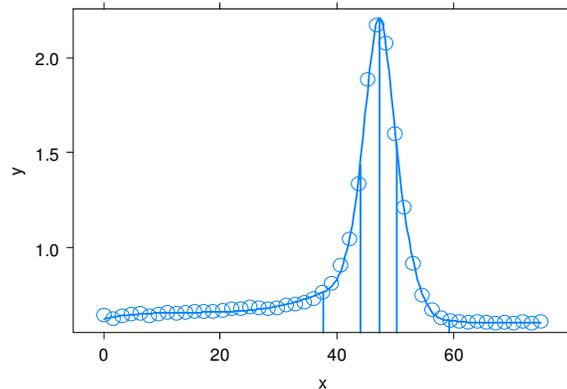
# Example #2 – Differentiation With Free Knot Splines

## ■ Obtain analytical spline derivatives with `deriv` function

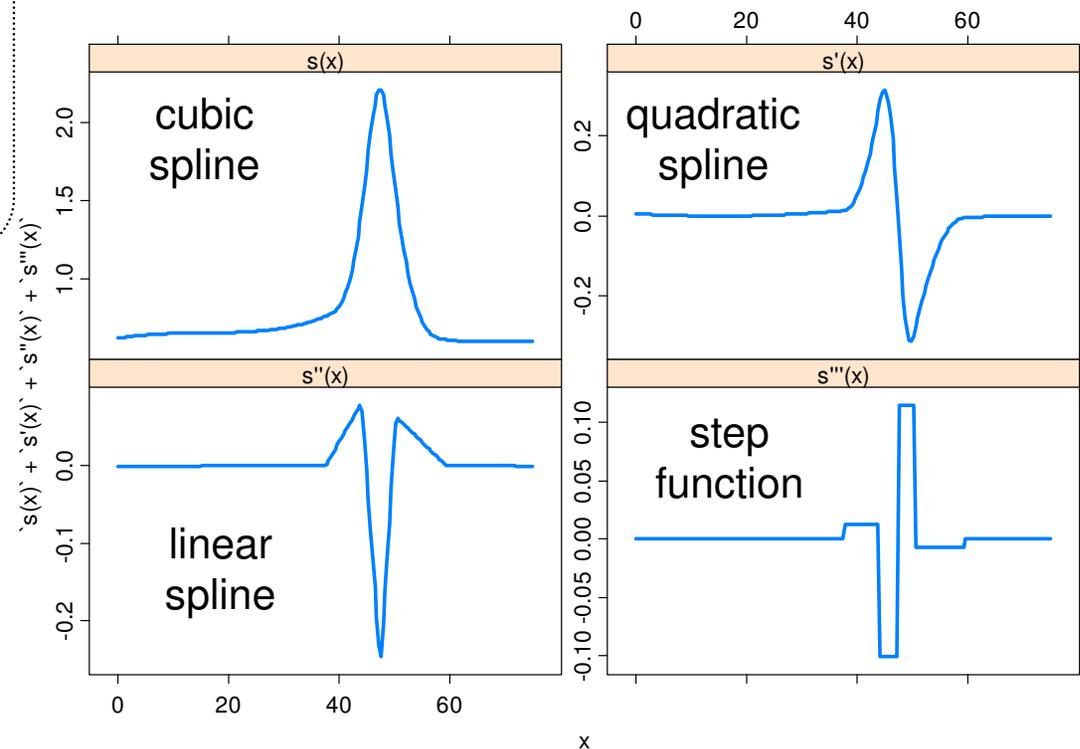
- Derivatives available from 0 (fitted spline value) to  $k$  (spline order)

```
data(titanium)
r <- curfit.free.knot(titanium$x2,
  titanium$y, g = 10, eps = 5e-4)
xyplot(r, show.knots = TRUE)

dr <- sapply(0:3, deriv,
  expr = r, at = titanium$x2)
```



Spline Derivatives For Titanium Data



# Example #3 – Splines With Convexity Constraints

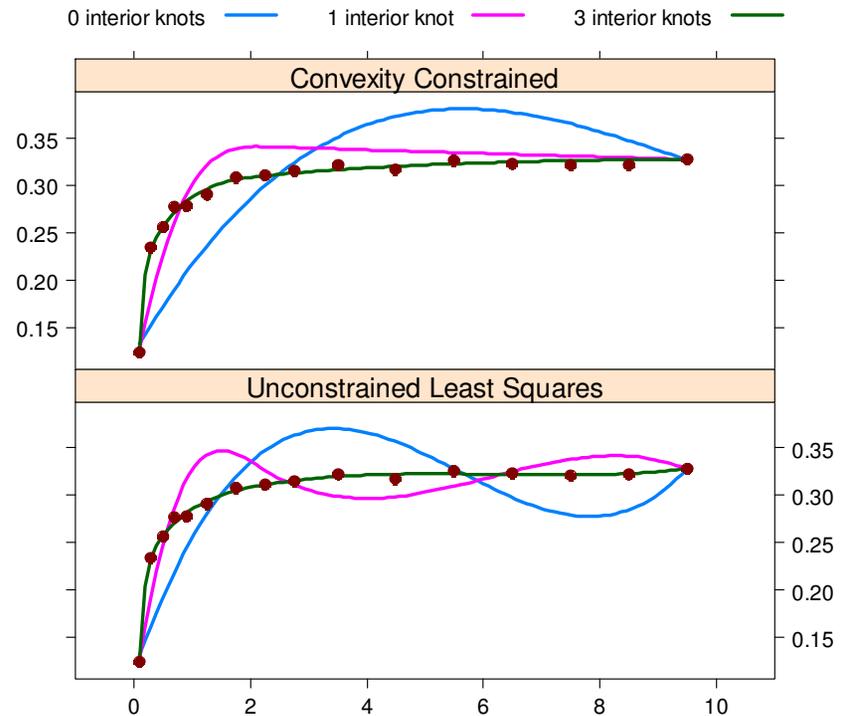
## ■ Volumetric moisture content data (Dierckx 1980)

- Force convex constraints for all data points

```
## convexity constraints
data(moisture)

f1 <- with(moisture,
  concon(x, y, w, v, s = 0.2))
f2 <- update(f1, s = 0.04)
f3 <- update(f1, s = 0.0002)

g1 <- with(moisture,
  curfit(x, y, w, method = "ls",
    knots = knots(f1)))
g2 <- update(g1, knots = knots(f2))
g3 <- update(g1, knots = knots(f3))
```



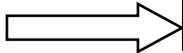
# Example #4 – Profile Likelihood

- Profiling the likelihood provides confidence intervals on knot placement

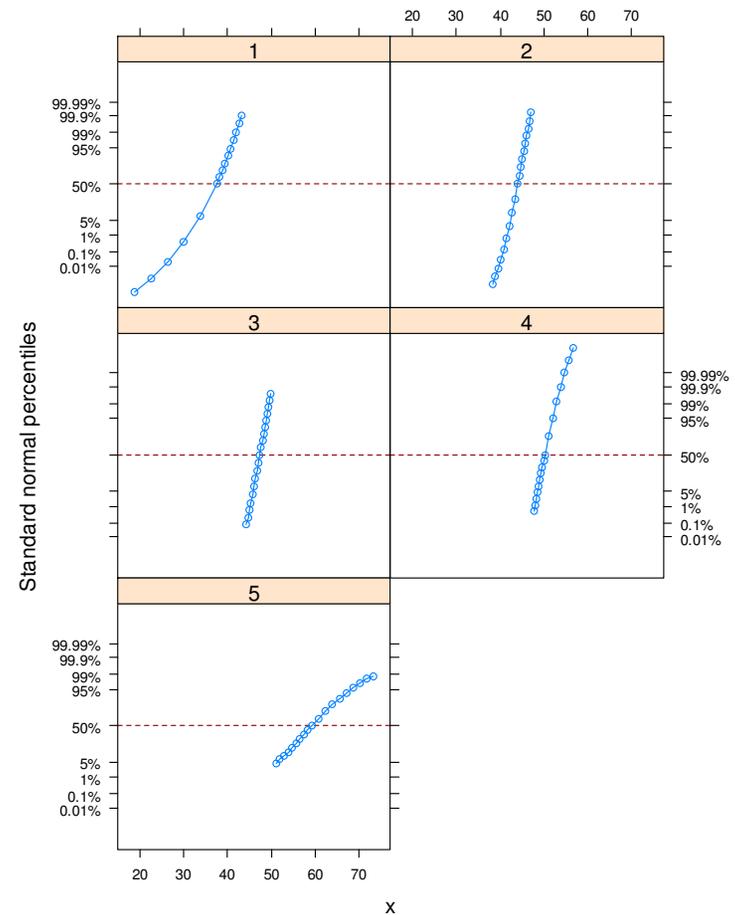
```
data(titanium)
r <- curfit.free.knot(titanium$x2,
  titanium$y, g = 10, eps = 5e-4)

pro <- confint(profile(r))
xyplot(pro)
```

|           | knots | 2.50% | 97.50% |
|-----------|-------|-------|--------|
| <b>1</b>  | 37.58 | 32.35 | 41.42  |
| <b>2</b>  | 43.98 | 42.00 | 45.80  |
| <b>3</b>  | 47.37 | 45.54 | 49.17  |
| <b>4</b>  | 50.19 | 48.26 | 52.36  |
| <b>*5</b> | 59.23 | 50.19 | 70.38  |



Lower bound for knot 5 is not achievable



# Software Status And Extensions

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- **Available for download from CRAN after JSM 2007**
  - Contact the author for bug reports and coding help
  
- **Create interfaces for remaining FITPACK routines**
  
- **Enhanced plotting for 3d splines with `lattice`**