Tools for General Parametric Splines

'spida' package in R developed for the Summer Program In Data Analysis (SPIDA) at York University

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Flexible tools to explore general parametric splines:

- Splines are conceptually simple but generally awkward to use in statistical analyses
- Goal: make splines as easy to use in regression as categorical variables
- "General parametric spline:"
 - o Polynomials of different degrees in each interval
 - Different degrees of smoothness at each knot including possible discontinuity
- Natural parametrization with interpretable coefficients: can use regular expression to test groups of coefficients
- Can estimate and test features of spline: slope, curvature, salti.
- Can interact with other variables (numerical or categorical), other splines, etc.
- Limitation: need to centre and scale variable

Installation

1) Install R http://cran.r-project.org/

Only once: Install 'car' and 'spida'. After starting R:

- 2) Each time you run R: load spida
 - > library(spida)

¹ The '>' sign is the prompt in R. Type only what comes after the '>'.

Main Functions:

gsp Creates general parametric splines

wald Uses regular expressions or hypothesis matrices to estimate coefficients and test hypotheses

Lmat Generates hypothesis matrices

Operates on spline functions to generate hypothesis matrices for spline derivatives: slope, curvature, salti at knots

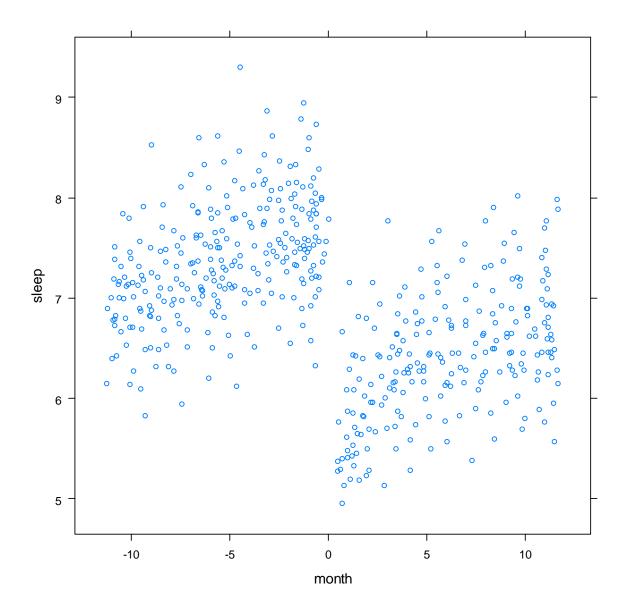
smsp Non-parametric splines using mixed models

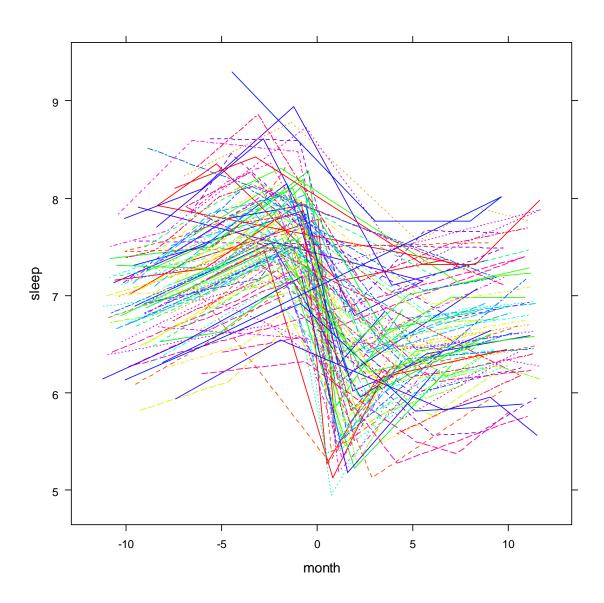
Example:

Simulated sleep data before and after birth. Inspired by Kunkel, Reitav, Monette (2010)

100 women each observed 5 times:

> xyplot(sleep ~ month, sim.data)





Defining a spline function:

Create your own spline function

degree parameter:

1,2,3,	linear, quadratic, cubic, etc.	
0	flat	
-1	equal to 0	

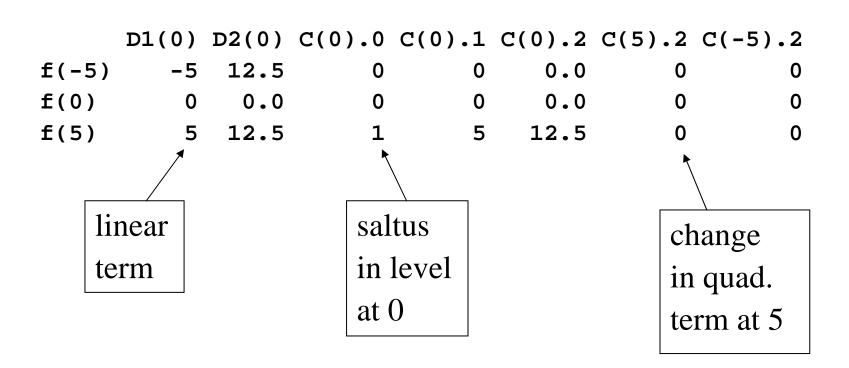
smooth parameter:

1	'smooth': same slope on	(
	both sides of knot	
2	same curvature on both sides	
0	continuous (no gap)	
-1	discontinuous	

What sp does:

Generates a portion of a model matrix:

$$> sp(c(-5,0,5))$$



Using your **sp** in a linear multilevel mixed model:

```
> library( nlme )
> fit <- lme( sleep ~ sp(month), sim.data,</pre>
              random = ~ 1 | id )
> summary( fit )
Linear mixed-effects model fit by REML
Data: sim.data
      AIC BIC logLik
  384.8794 426.8642 -182.4397
Random effects:
Formula: ~1 | id
        (Intercept) Residual
StdDev: 0.4922676 0.2439248
```

Fixed effects: sleep ~ sp(month)

```
Value Std.Error DF
                                         t-value p-value
                 7.770252 0.07562772 393 102.74344
                                                  0.0000
(Intercept)
sp(month)D1(0)
                 0.007316 0.04231428 393 0.17289 0.8628
sp(month)D2(0) -0.027195 0.01204080 393 -2.25853
                                                  0.0245
sp(month)C(0).0 -2.284334 0.10143225 393 -22.52078 0.0000
sp(month)C(0).1 0.295341 0.06867966 393 4.30027
                                                  0.0000
sp(month)C(0).2 -0.015195 0.01850084 393 -0.82130 0.4120
sp(month)C(5).2 0.027771 0.02041699 393 1.36018
                                                 0.1746
sp(month)C(-5).2 -0.048239 0.01943028 393
                                        -2.48267
                                                  0.0135
```

Standardized Within-Group Residuals:

Min Q1 Med Q3 Max -6.955982485 -0.326223723 -0.009829296 0.320549996 7.518591530

Number of Observations: 500

Number of Groups: 100

> round(model.matrix(fit) [1:5,],4) (Intercept) sp(month)D1(0) sp(month)D2(0) sp(month)C(0).0 -9.8637 48.6464 1 1 2 1 0 -6.9539 24.1783 3 1 -2.1141 2.2347 0 4 1 0.0534 0.0014 1 5 0.9739 0.4742 sp(month)C(0).1 sp(month)C(0).2 sp(month)C(5).2 sp(month)C(-5).21 0.0000 0.0000 -11.8278 0.0000 2 0.0000 0 -1.90883 0.0000 0.0000 0 0.0000 4 0.0534 0.0014 0 0.0000 5 0.9739 0.4742 0.0000

Natural spline

Linear in the extreme intervals

You can create almost every kind of spline used in regression

Fixed effects: sleep ~ spn(month) Value Std.Error t-value p-value \mathbf{DF} (Intercept) 7.770115 0.07572201 393 102.61369 0.0000 0.16829 0.8664 spn(month)D1(0) 0.007163 0.04256467 393 spn(month)D2(0) -0.027267 0.01214689 393 -2.24475 0.0253 spn(month)C(0).0 -2.278971 0.10190952 393 -22.36269 0.0000 spn(month)C(0).1 0.290233 0.06946431 393 4.17816 0.0000 spn(month)C(0).2 -0.013372 0.01882470 393 -0.71035 0.4779 spn(month)C(5).2 0.024200 0.02135186 393 1.13338 0.2577

Note: Although the estimable coefficients have the same interpretation for both splines, the natural spline matrix is necessarily different. There are no estimable coefficients attached to the boundary knots.

spn(month)C(-5).2 -0.048604 0.01975742 393

-2.46003

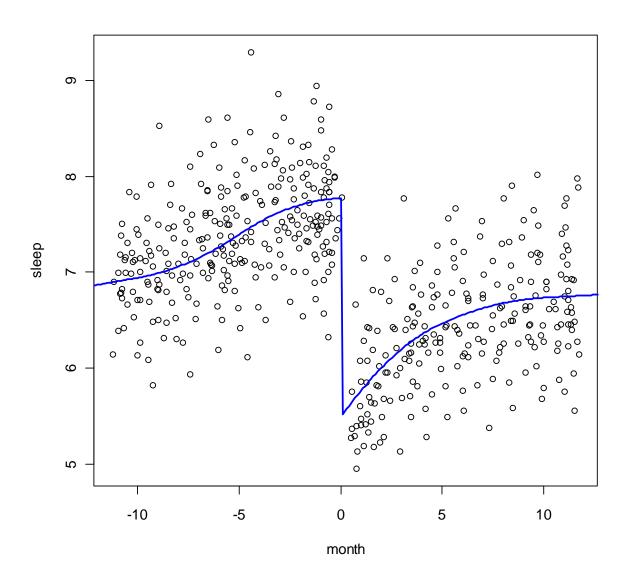
0.0143

Viewing the fitted model:

Create a data frame of values of functional predictors – here just months:

² I used expand.grid although it is really needed only for multiple functional predictors

> plot(sleep ~ month, sim.data)
> lines(sleep ~ month, pred, col = 'blue', lwd = 2)



Estimation and testing:

```
Overall test for effect of time:
                                            'month' is a "regular
> wald( fitn, 'month')
                                            expression" matching
                                            names of coefficients
      numDF denDF F.value p.value
              393 366.2914 <.00001
month
Coefficients
                     Estimate Std.Error
                                                t-value p-value
                                         \mathbf{DF}
                               0.042565 393
                                               0.168293 0.86644
  spn(month)D1(0)
                     0.007163
  spn(month)D2(0)
                    -0.027267
                               0.012147 393
                                              -2.244749 0.02534
  spn(month)C(0).0
                    -2.278971
                               0.101910 393
                                             -22.362687 <.00001
                   0.290233 0.069464 393
  spn(month)C(0).1
                                               4.178156 0.00004
  spn(month)C(0).2 -0.013372 0.018825 393
                                              -0.710346 0.47791
                   0.024200
  spn(month)C(5).2
                               0.021352 393
                                               1.133384 0.25774
  spn(month)C(-5).2 -0.048604
                               0.019757 393
                                              -2.460034 0.01432
```

Can we get rid of knots and just use a quadratic?

```
Coefficients Estimate Std.Error DF t-value p-value spn(month)C(0).0 -2.278971 0.101910 393 -22.362687 <.00001 spn(month)C(0).1 0.290233 0.069464 393 4.178156 0.00004 spn(month)C(0).2 -0.013372 0.018825 393 -0.710346 0.47791 spn(month)C(5).2 0.024200 0.021352 393 1.133384 0.25774 spn(month)C(-5).2 -0.048604 0.019757 393 -2.460034 0.01432
```

Can we get rid of non-zero knots?

Using a hypothesis matrix:

```
> Lm < - rbind("at 5 mos" = c(0,0,0,0,0,0,1,0),
           "at -5 mos" = c(0,0,0,0,0,0,0,1))
> Lm
         [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
at 5 mos
            0 0 0
                              0
at -5 mos 0 0 0 0 0
                                        0 1
> wald( fitn, Lm)
 numDF denDF F.value p.value
 2 393 3.646731 0.02696
          Estimate Std.Error DF t-value p-value
 at 5 mos 0.024200 0.021352 393 1.133384 0.25774
 at -5 mos -0.048604 0.019757 393 -2.460034 0.01432
```

Generating Standard Errors

Wald tests generate estimated standard errors for coefficients, we need to capture them for plotting

Example: plotting the estimated response

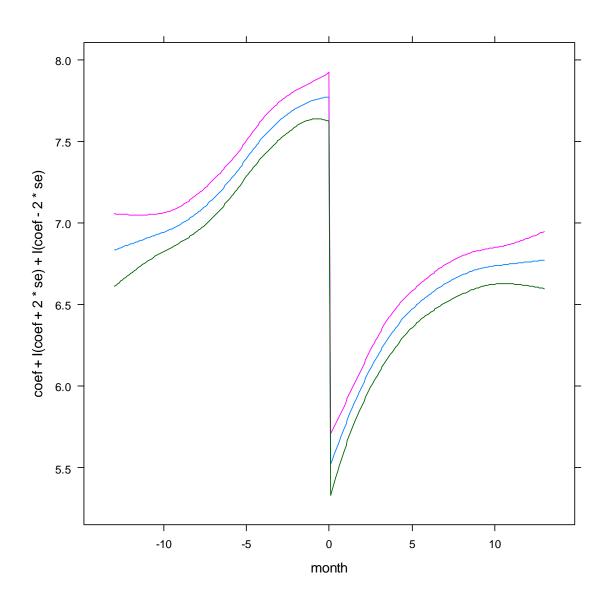
L matrix as model matrix evaluated over a range of predictors

```
> Lest <- with( pred, cbind( 1, spn(month)))</pre>
> dim(Lest)
[1] 261
> head(Lest)
          D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
f(-13)
        1 -13.0
                   80
                                                    -27.5
                                  0
                                         0
f(-12.9) 1 -12.9
                                                    -27.0
                   79
                                         0
                                                    -26.5
f(-12.8) 1 -12.8
                   78
                                  0
                                         0
f(-12.7) 1 -12.7
                                                    -26.0
                   77
                           0
                                  0
                                         0
f(-12.6) 1 -12.6
                   76
                                                    -25.5
                                  0
                                         0
f(-12.5) 1 -12.5
                                  0
                                         0
                                                    -25.0
                   75
                           0
```

```
> head( pred )
 month sleep
1 -13.0 6.832263
2 -12.9 6.835944
3 -12.8 6.839625
4 -12.7 6.843306
5 -12.6 6.846987
6 -12.5 6.850669
> ww <- as.data.frame(wald(fitn, Lest))</pre>
           # capture estimates and SEs
> ww <- cbind( ww, pred)</pre>
           # combine back with pred
> head( ww )
                   se month sleep
             coef
f(-13) 6.832263 0.1111460 -13.0 6.832263
f(-12.9) 6.835944 0.1089242 -12.9 6.835944
f(-12.8) 6.839625 0.1067182 -12.8 6.839625
f(-12.7) 6.843306 0.1045293 -12.7 6.843306
f(-12.6) 6.846987 0.1023583 -12.6 6.846987
f(-12.5) 6.850669 0.1002066 -12.5 6.850669
```

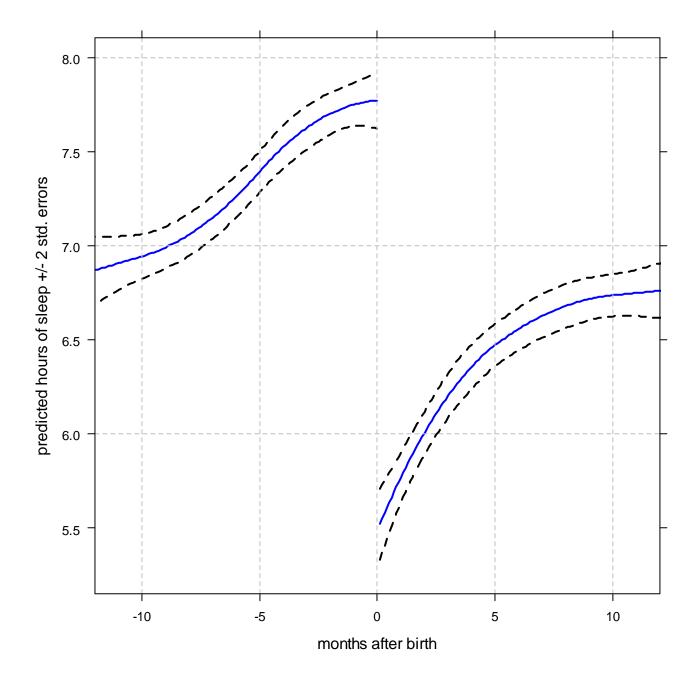
Easy plotting:

```
> xyplot(coef + I(coef + 2*se) + I(coef - 2*se) ~
month, ww, type = 'l')
```



Fancier plotting:

```
> td( col = c('blue','black','black'),
           lty = c(1,2,2), lwd = 2)
> xyplot( coef + I(coef + 2*se) + I(coef - 2*se) ~
              month,
       rbind(ww[1:131,],NA,ww[132:261,]),
       type = '1',
       panel = function(x,y,...) {
           panel.xyplot( x, y, ...)
           panel.abline( v = c(-10, -5, 0, 5, 10).
               col = 'gray', lty = 2)
           panel.abline( h = seq(0,10,1),
               col = 'gray', lty = 2)
       },
       xlim = c(-12,12),
       ylab =
         'predicted hours of sleep +/- 2 std. errors',
       xlab = "months after birth")
```



Inference concerning a spline

All fixed effects coefficients: > wald(fitn) numDF denDF F.value p.value 393 2688.578 < .00001 8 Coefficients Estimate Std.Error DF t-value p-value (Intercept) 7.770115 0.075722 393 102.613693 <.00001 spn(month)D1(0) 0.007163 0.042565 393 0.168293 0.86644 spn(month)D2(0) -0.027267 0.012147 393 -2.244749 0.02534 spn(month)C(0).0 -2.278971 0.101910 393 -22.362687 <.00001 spn(month)C(0).1 0.290233 0.069464 393 4.178156 0.00004 spn(month)C(0).2 -0.013372 0.018825 393 -0.710346 0.47791 spn(month)C(5).2 0.024200 0.021352 393 1.133384 0.25774 spn(month)C(-5).2 -0.048604 0.019757 393 -2.460034 0.01432 Test saltus at birth (month = 0): > wald(fitn , "C\\(0.*0\$") numDF denDF F.value p.value C\\(0.*0\$ 1 393 500.0898 <.00001

Better labelled:

Operating on splines: sc function

```
sc( spn, x, D, type)
```

generates a hypothesis matrix to estimate the Dth power coefficient of the spline **spn** at the points given by **x**. '**type**' only has an effect when the value of **x** corresponds to a knot. In that case, **type** determines whether the evaluation is to the left, to the right or across the knot.

sc parameters:			
	0	value of spline	
D	1	first derivative	
	2,	quadratic, terms	
type	0	limit from the left of knot	
	1	limit from the right of knot	
	2	saltus: limit from the right	
		minus limit from the left	

Example:

Estimate value, slope and quadratic component at -1 months:

```
> sc(spn, c(-1,-1,-1), D = c(1,2,3))
       D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
D1(-1)
           1
                -1
                               0
                                                      0
D2(-1)
           0
                 1
                               0
                                                      0
D3(-1)
           0
                 0
                        0
                               0
                                                      0
> wald( fitn, cbind(c(1,0,0,0),
        sc(spn, c(-1,-1,-1,-1), D = c(0,1,2,3)))
+
  numDF denDF F.value p.value
          393 6937.729 <.00001
1
         Estimate Std.Error DF t-value p-value
 g(-1) 7.749318 0.057332 393 135.165234 <.00001
 D1(-1) 0.034430 0.030741 393 1.120008 0.26339
 D2(-1) -0.027267 0.012147 393 -2.244749 0.02534
 D3(-1) 0.000000 0.000000 Inf
                                    NaN
                                           NaN
```

```
Using the sc function to test a saltus (type = 2):
     > Lm <- cbind(0, sc(spn, 0, D = 0, type = 2))
     > Lm
                  D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
     g(0+)-g(0-) 0
                  0 0 1
                                               0
                                                      0
     > wald( fitn, Lm )
       numDF denDF F.value p.value
           1 393 500.0898 <.00001
                   Estimate Std.Error DF t-value p-value
       g(0+)-g(0-) -2.278971 0.10191 393 -22.36269 <.00001
     > head( model.matrix( fitn))
       (Intercept) spn(month)D1(0) spn(month)D2(0) spn(month)C(0).0
     1
                   -9.86370968 48.646384300
     2
                   -6.95388563 24.178262681
               1 -2.11407625 2.234659188
     3
               1 0.05344575 0.001428224
     4
                                                       1
```

0.97390029 0.474240891

-7.66649560 29.387577401

5

1

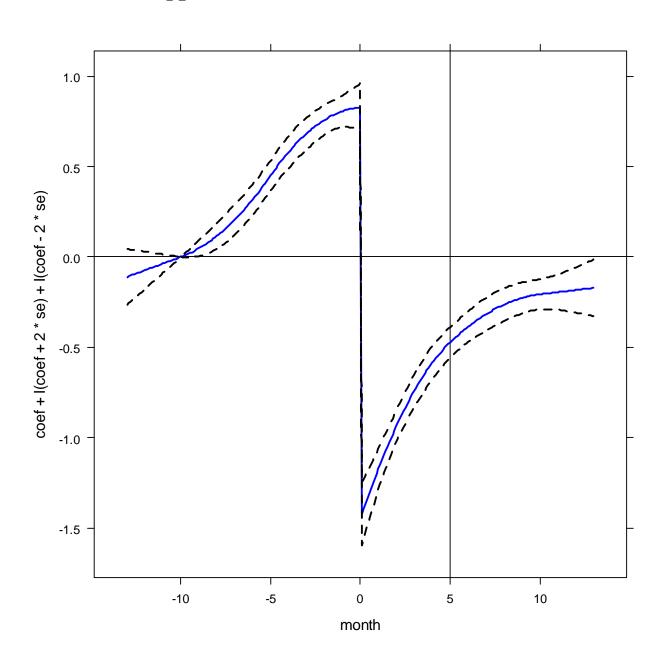
1

1

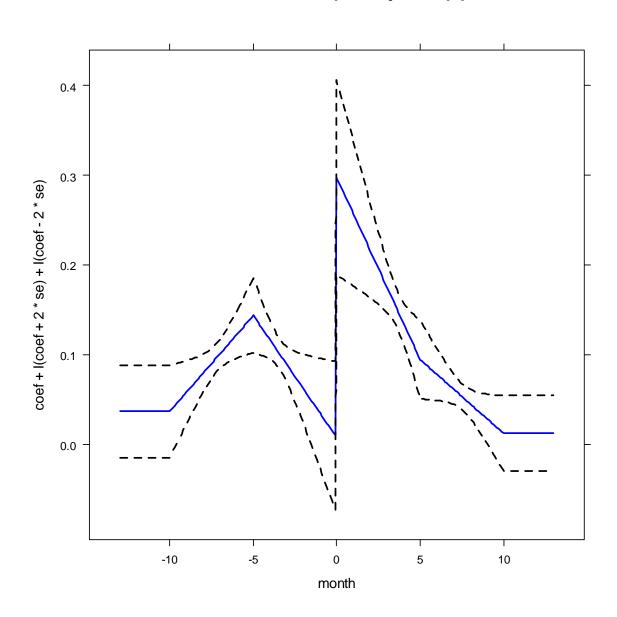
0

```
spn(month)C(0).1 spn(month)C(0).2 spn(month)C(5).2 spn(month)C(-5).2
       0.0000000
                      0.00000000
                                     0.000000e+00
1
                                                         -11.827836
2
       0.0000000
                      0.00000000
                                     0.000000e+00
                                                          -1,908835
3
       0.0000000
                      0.00000000
                                     0.000000e+00
                                                          0.000000
4
       0.05344575
                      0.001428224
                                    -1.441111e-18
                                                          0.000000
       0.97390029
                      0.474240891
                                      1.717766e-16
                                                          0.000000
5
       0.0000000
                      0.00000000
                                     0.000000e+00
6
                                                          -3.555099
> L.level <- cbind( 1, spn( c(-10, 10)))</pre>
> L.level
         D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
f(-10) 1
            -10
                   50
                            0
                                   0
                                           0
                                                0.0
                                                       -12.5
f(10) 1
            10
                   50
                            1
                                  10
                                                         0.0
                                          50
                                               12.5
> wald( fitn, L.level )
  numDF denDF F.value p.value
            393 8216.843 <.00001
1
       2
        Estimate Std.Error DF t-value p-value Lower 0.95 Upper 0.95
 f(-10) 6.942695 0.058820 393 118.0330 <.00001
                                               6.827054
                                                          7.058337
                                               6.625887
 f(10) 6.735666 0.055838 393 120.6283 <.00001
                                                          6.845445
```

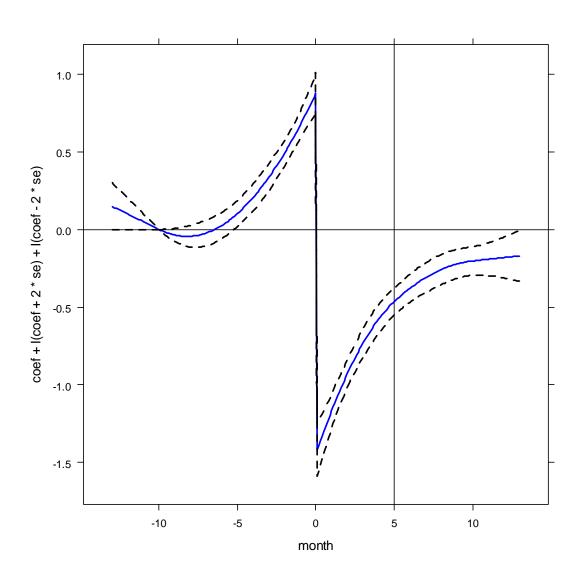
```
> wald( fitn, L.level[2,] - L.level[1,] )
  numDF denDF F.value p.value
1
      1
          393 24.4368 <.00001
      Estimate Std.Error DF t-value p-value Lower 0.95 Upper 0.95
 Larg -0.20703 0.04188 393 -4.943359 <.00001 -0.289367 -0.124692
> ?sc
> Lslope <- with( pred, cbind( 0 , sc( spn, month, D = 1)))</pre>
> ws <- as.data.frame(wald(fitn,Lslope))</pre>
> head(ws)
                 coef
                                se
D1(-13) 0.03681075 0.02576518
D1(-12.9) 0.03681075 0.02576518
D1(-12.8) 0.03681075 0.02576518
D1(-12.7) 0.03681075 0.02576518
D1(-12.6) 0.03681075 0.02576518
D1(-12.5) 0.03681075 0.02576518
> ws <- cbind(ws, pred)</pre>
```



```
> head( Lest )
          D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
f(-13) 1 -13.0 80
                                                 -27.5
                         0
                                0
                                      0
f(-12.9) 1 -12.9
                                                 -27.0
                  79
                         0
                                0
                                      0
f(-12.8) 1 -12.8 78
                                                 -26.5
                         0
                                0
                                      0
                                             0 -26.0
f(-12.7) 1 -12.7 77
                         0
                                0
                                      0
f(-12.6) 1 -12.6 76
                                                -25.5
                         0
                                0
                                      0
                                             0
f(-12.5) 1 -12.5 75
                                             0 -25.0
                         0
                                0
                                      0
> L.m10 <- cbind( 1, spn(-10))</pre>
> L.m10
        D1(0) D2(0) C(0).0 C(0).1 C(0).2 C(5).2 C(-5).2
f(-10) 1 -10 50 0 0 0 -12.5
> Ldiff.m10 <- Lest - L.m10[rep(1,nrow(Lpred)),]</pre>
> wd <- as.data.frame( wald( fitn, Ldiff.m10))</pre>
> wd <- cbind(wd, pred)</pre>
> wdr <- as.data.frame( wald( fitnr, Ldiff.m10))</pre>
> wdr <- cbind(wdr, pred)</pre>
```



```
> xyplot( coef + I(coef + 2*se)+I(coef - 2*se) ~
    month, wdr, type = 'l',
    abline = list(h=0, v=5))
```



Caution:

Since gsp uses raw parametrization (in contrasts with b-splines, for example), you should center and rescale the predictor. A range within -10 to 10 for a spline with no power higher than cubic should be adequate.