

# Package: spfda (via r-universe)

June 27, 2024

**Type** Package

**Title** Function-on-Scalar Regression with Group-Bridge Penalty

**Version** 0.9.1.9000

**License** MIT + file LICENSE

**Description** Implements a group-bridge penalized function-on-scalar regression model proposed by Wang et al. (2020)  [<arXiv:2006.10163>](https://arxiv.org/abs/2006.10163), to simultaneously estimate functional coefficient and recover the local sparsity.

**URL** <https://github.com/dipterix/spfda>, <http://dipterix.org/spfda/>

**BugReports** <https://github.com/dipterix/spfda/issues>

**Imports** stats, splines, graphics, mathjaxr

**Encoding** UTF-8

**RoxygenNote** 7.1.2

**RdMacros** mathjaxr

**Language** en-US

**Suggests** grpreg, refund

**Repository** <https://dipterix.r-universe.dev>

**RemoteUrl** <https://github.com/dipterix/spfda>

**RemoteRef** HEAD

**RemoteSha** 56649eee17dc2364372009e3131955b36cc0f06f

## Contents

fosr_vs . . . . .	2
spfda . . . . .	2
spfda_simulate . . . . .	4
spfda_weight . . . . .	5

<b>Index</b>	<b>6</b>
--------------	----------

---

 fosr\_vs

*Ported function from 'refund' package*


---

### Description

A modified version of [fosr.vs](#), but with groups parameter to allow grouping time points rather than the whole coefficient when the underlying functions are locally supported.

### Usage

```

fosr_vs(
  formula,
  data,
  nbasis = 10,
  method = c("ls", "grLasso", "grMCP", "grSCAD"),
  epsilon = 1e-05,
  max.iter_num = 100,
  groups = NULL
)

```

### Arguments

formula, data, nbasis, method, epsilon, max.iter\_num

see [fosr.vs](#)

groups integer vector with length of number of time-points of how time-points should be grouped; default is NULL, indicating there is no local sparsity.

---

 spfda

*Sparse Function-on-scalar Regression with Group Bridge Penalty*


---

### Description

Function-on-scalar regression model, denote  $n$  as total number of observations,  $p$  the number of coefficients,  $K$  as the number of B-splines,  $T$  as total time points.

### Usage

```

spfda(
  Y,
  X,
  lambda,
  time = seq(0, 1, length.out = ncol(Y)),
  nsp = "auto",
  ord = 4,
  alpha = 0.5,
)

```

```

W = NULL,
init = NULL,
max_iter = 50,
inner_iter = 50,
CI = FALSE,
...
)

```

### Arguments

Y	Numeric $n \times T$ matrix, response function.
X	Numeric $n \times p$ matrix, design matrix
lambda	Regularization parameter $\gamma$
time	Time domain, numerical length of $T$
nsp	Integer or 'auto', number of B-splines $K$ ; default is 'auto'
ord	B-spline order, default is 4; must be $\geq 3$
alpha	Bridge parameter $\alpha$ , default is 0.5
W	A $T \times T$ weight matrix or NULL (identity matrix); default is NULL
init	Initial $\gamma$ ; default is NULL
max_iter	Number of outer iterations
inner_iter	Number of <i>ADMM</i> iterations (inner steps)
CI	Logical, whether to calculate theoretical confidence intervals
...	Ignored

### Details

This function implements "Functional Group Bridge for Simultaneous Regression and Support Estimation" (<https://arxiv.org/abs/2006.10163>). The model estimates functional coefficients  $\beta(t)$  under model

$$y(t) = X\beta(t) + \epsilon(t)$$

with B-spline basis expansion

$$\beta(t) = \gamma B(t) + R(t),$$

where  $R(t)$  is B-spline approximation error. The objective function

$$\|(Y - X\gamma B)W\|_2^2 + \sum_{j,m} \|\gamma_j^T \mathbf{1}(B^t > 0)\|_1^\alpha.$$

The input response variable is a matrix. If  $y_i(t)$  are observed at different time points, please interpolate (e.g. [kernel](#)) before feeding in.

**Value**

A `spfda.model` object (environment) with following elements:

**B** B-spline basis functions used

**error** Root Mean Square Error ('RMSE')

**CI** Whether confidence intervals are calculated

**gamma** B-spline coefficient  $\gamma_{p \times K}$

**generate\_splines** Function to generate B-splines given time points

**K** Number of B-spline basis functions

**knots** B-spline knots used to fit the model

**predict** Function to predict responses  $\beta(t)$  given new  $X$  and/or time points

**raw** A list of raw variables

**Examples**

```
dat <- spfda_simulate()
x <- dat$X
y <- dat$Y

fit <- spfda(y, x, lambda = 5, CI = TRUE)

BIC(fit)

plot(fit, col = c("orange", "dodgerblue3", "darkgreen"),
     main = "Fitted with 95% CI", aty = c(0, 0.5, 1), atx = c(0,0.2,0.8,1))
matpoints(fit$time, t(dat$env$beta), type = 'l', col = 'black', lty = 2)
legend('topleft', c("Fitted", "Underlying"), lty = c(1,2))

print(fit)
coefficients(fit)
```

---

spfda\_simulate

*Generates toy example data*

---

**Description**

Synthesized functional signals with heterogeneous error. The underlying three coefficients correspond to 'dense', 'global sparse', and 'local sparse' functions. See <https://arxiv.org/abs/2006.10163> for detailed configurations.

**Usage**

```
spfda_simulate(n = 1000, n_timepoints = 100, err = 1, scale = c(1, 1, 1))
```

**Arguments**

n	Total number of observations
n_timepoints	Total number of time points
err	Error magnitude
scale	the scale of coefficients length of 1 or 3.

**Value**

A list of data generated: X is scalar predictor, Y is functional response.

---

spfda_weight	<i>Calculates weight matrices</i>
--------------	-----------------------------------

---

**Description**

Calculates weight matrices

**Usage**

```
spfda_weight(X, Y, bandwidth, part)
```

**Arguments**

X	design matrix
Y	response matrix
bandwidth	numeric band-width
part	list of time point boundaries

**Value**

the weight matrix

# Index

`fosr.vs`, [2](#)

`fosr_vs`, [2](#)

`kernel`, [3](#)

`spfda`, [2](#)

`spfda_simulate`, [4](#)

`spfda_weight`, [5](#)