

# Package ‘psme’

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**Title** Penalized Splines Mixed-Effects Models

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**Depends** R (>= 4.0.0)

**Imports** stats, Matrix, methods, mgcv, lme4

## Description

Fit penalized splines mixed-effects models (a special case of additive models) for large longitudinal datasets. The package includes a `psme()` function that (1) relies on package 'mgcv' for constructing population and subject smooth functions as penalized splines, (2) transforms the constructed additive model to a linear mixed-effects model, (3) exploits package 'lme4' for model estimation and (4) backtransforms the estimated linear mixed-effects model to the additive model for interpretation and visualization. See Pedersen et al. (2019) <[doi:10.7717/peerj.6876](https://doi.org/10.7717/peerj.6876)> and Bates et al. (2015) <[doi:10.18637/jss.v067.i01](https://doi.org/10.18637/jss.v067.i01)> for an introduction. Unlike the `gamm()` function in 'mgcv', the `psme()` function is fast and memory-efficient, able to handle datasets with millions of observations.

**License** GPL-3

**NeedsCompilation** no

**URL** <https://github.com/ZheyuanLi/psme>

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EvalSmooth

*Evaluate a fitted smooth term in a fitted model*

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**Description**

A helper function to evaluate the fitted population and subject smooth functions in a fitted penalized splines mixed-effects model.

**Usage**

```
EvalSmooth(mgcv.smooth, new.x)
```

**Arguments**

<code>mgcv.smooth</code>	A single smooth term taken from the smooth list of the fitted penalized splines mixed-effects model.
<code>new.x</code>	A numeric vector of new covariate values at which to evaluate the smooth term.

**Value**

These are two common cases of population smooth functions. For a single smooth function like  $s(x)$ , the function returns a vector of predicted values. For a factor ‘by’ smooth term like  $s(x, \text{by} = f)$ , the function returns a matrix of predicted values, where each column corresponds to a factor level.

For subject smooth functions specified as a factor-smooth interaction term  $s(x, f, \text{bs} = "fs")$ , the function returns a matrix of predicted values, where each column corresponds to an individual subject.

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**See Also**

[psme](#)

**Examples**

```
## see examples for the psme() function
```

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GetExtrema

*Identify all the local extrema of a curve with a grid search*

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### Description

Approximately identify all the local extrema of a 1-dimensional curve  $y = f(x)$ , by finding all the local minima and all the local maxima in a sequence of  $(x, y)$  values, discretely observed or sampled from the curve.

### Usage

```
GetExtrema(x, y)
```

### Arguments

**x** A numeric vector of x-coordinates for a curve.  
**y** A numeric vector of y-coordinates for a curve.

### Value

A list with four elements:

**min.x** The x-coordinates of the local minima.  
**min.y** The y-coordinates of the local minima.  
**max.x** The x-coordinates of the local maxima.  
**max.y** The y-coordinates of the local maxima.

### Author(s)

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### See Also

[EvalSmooth](#)

### Examples

```
x <- seq(0, 2 * pi, length.out = 100)
y <- sin(x)
ex <- GetExtrema(x, y)
plot(x, y, type = "l")
points(ex$min.x, ex$min.y, pch = 19, col = 2)
points(ex$max.x, ex$max.y, pch = 19, col = 3)
```

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 psme

*Fit penalized splines mixed-effects models*


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## Description

Fit penalized splines mixed-effects models by leveraging the known equivalence between penalized splines and mixed-effects models. `psme` reparameterizes each penalized spline term specified in a *mgcv*-style model formula as linear mixed-effects model terms, and fits such equivalent model using *lme4* as the computational engine.

## Usage

```
psme(mgcv.form, data, knots = NULL)
```

## Arguments

<code>mgcv.form</code>	a <i>mgcv</i> -style model formula that specify a penalized splines mixed-effects model in the fashion of penalized splines additive models.
<code>data</code>	a data frame with all variables in the formula, and any rows with NA must be manually removed.
<code>knots</code>	an optional named list providing knots.

## Details

A penalized splines mixed-effects model is a special case of additive models, most suitable for modelling longitudinal data. Such model typically contains some population smooth functions and a group of subject smooth functions, both nonlinear in a variable (for example, time). The function relies on *mgcv* for constructing smooth functions as penalized splines, and users should specify these terms via a *mgcv*-style model formula. Examples of population smooth functions include a single smooth function like  $s(x)$  and a factor ‘by’ smooth term like  $s(x, \text{by} = f)$ , where the smooth function in  $x$  changes with the levels of a factor variable  $f$  (for example, gender). Subject smooth functions are set up as a factor-smooth interaction term like  $s(x, f, \text{bs} = "fs")$ . The model may also involve smooth functions of other covariate variables constructed using `s`, `te` and `ti`, as well as any parametric terms that are legitimate in a `lm` formula. In addition, users can customize the class and basis dimension of each penalized spline, as they can when using *mgcv*.

To fit a penalized splines mixed-effects model, the function transforms each penalized spline into a combination of fixed and random effects, producing an equivalent linear mixed-effects model. This model is then estimated using *lme4*'s low-level functions. Finally, the function backtransforms the estimated fixed-effects and predicted random-effects to the basis coefficients of the original penalized splines, so that the users can extract and plot the estimated smooth functions. In particular, the transformation from a penalized splines mixed-effects model to a linear mixed-effects model is performed with great caution to preserve the sparsity of the design matrices. As a result, the function is able to handle large longitudinal datasets with millions of observations.

**Value**

a list containing:

<code>pform</code>	The parametric component of the model formula.
<code>pcoef</code>	The estimated coefficients of the parametric component of the model.
<code>smooth</code>	A list of constructed and estimated smooth functions corresponding to the smooth terms in the model formula.
<code>lme4.fit</code>	The raw output of <i>lme4</i> 's low-level functions.

**Author(s)**

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**See Also**

[gam](#), [gamm](#)

**Examples**

```
library(psme)

## simulate a toy dataset of 50 subjects, each with a random quadratic trajectory
n.subjects <- 50
y <- x <- vector("list", n.subjects)
## a subject has 5 to 10 random observations
n <- sample(5:10, size = n.subjects, replace = TRUE)
a <- rnorm(n.subjects, mean = 1, sd = 0.2)
b <- rnorm(n.subjects, mean = 0, sd = 0.3)
c <- rnorm(n.subjects, mean = 0, sd = 0.4)
for (i in 1:n.subjects) {
  x_i <- sort(runif(n[i], -1, 1))
  y_i <- a[i] * x_i ^ 2 + b[i] * x_i + c[i]
  x[[i]] <- x_i; y[[i]] <- y_i
}
x <- unlist(x)
y <- unlist(y)
## add noise to the true trajectories
y <- y + rnorm(length(y), sd = 0.1)
## compose a data frame in long format
id <- rep.int(1:n.subjects, n)
dat <- data.frame(x = x, y = y, id = as.factor(id))

## fit a penalized splines mixed-effects model
mgcv.form <- y ~ s(x, bs = 'ps', k = 8) +
  s(x, id, bs = 'fs', xt = 'ps', k = 8, m = c(2, 1))
fit <- psme(mgcv.form, data = dat)

## evaluate population smooth function and subject smooth functions
xp <- seq.int(-1, 1, length.out = 51)
pop.smooth <- EvalSmooth(fit$smooth[[1]], new.x = xp)
sub.smooth <- EvalSmooth(fit$smooth[[2]], new.x = xp)
```

```
smooth <- pop.smooth + sub.smooth + fit$pcoef[["(Intercept)"]]  
matplot(xp, smooth, type = "l", lty = 1, xlab = "x", ylab = "fitted trajectories")  
points(dat, pch = 20, col = 8)
```

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