

# Package ‘tensorMARM’

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**Type** Package

**Title** tensorMARM

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**Description** Multivariate additive regression for multi-view data via tensor estimation. The B-splines are used to approximate component functions. The spline coefficients are rearranged into multiple third-order tensors or even a fourth-order tensor with low-rankness under some specific conditions. Dimension reduction can be achieved by Tucker decomposition and group sparse penalty, for example, LASSO, MCP or SCAD.

**License** GPL (>= 2)

**Imports** splines, Rcpp (>= 0.11.15), RcppEigen (>= 0.3.2.3.0)

**LinkingTo** Rcpp, RcppEigen

**RoxygenNote** 6.0.1

**NeedsCompilation** yes

**Repository** github

**URL** <https://github.com/xliusufe/tensorMARM>

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## R topics documented:

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tensorMARM-package	<i>Multivariate additive regression for multi-view data via tensor estimation</i>
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### Description

For a multivariate additive model (MARM), the B-spline base functions are applied to approximate the additive components. We treat the coefficients as multiple third-order tensors or even a fourth-order tensor for some specific structure (called structural MARM). With the tensor low-rankness, the Tucker decomposition and group sparse penalty (lasso, mcp or scad) reduce the number of parameters. A blockwise coordinate descent algorithm is used to estimate the core tensors and factor matrices, and feature selection.

### Details

This package includes four main functions and four generating functions. `marm3` and `marm3.dr` yield the estimator of MARM. The difference is the former requires the fixed ranks and the latter can search the optimal ranks and regularization parameter simultaneously by BIC or CV method. `marm3.sim.fbs` and `marm3.sim.fsin` generate data of scenario I and II respectively. Scenario I assumes that the true functions are exactly residing in the space of B-spline basis functions. Scenario II assumes that the true functions are some linear combination of  $\sin(2\pi x)$  and  $\cos(\pi x)$ . Similarly, `marm4` and `marm4.dr` yield the estimator of structural MARM. `marm4.sim.fbs` and `marm4.sim.fsin` are two generating functions of scenario I and II, which have the same assumptions as MARM.

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### References

Multivariate additive regression for multi-view data via tensor estimation.

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marm3	<i>Fit MARM with sparsity assumption and fixed ranks.</i>
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### Description

Fit a multivariate additive model for multi-view data (MARM) using B-splines with given ranks ( $r_{1g}, r_{2g}, r_{3g}$ ). Multiple third-order coefficient tensors can be estimated by this function. The group sparse penalty such as LASSO, MCP or SCAD and the coordinate descent algorithm are used to yield a sparsity estimator. The BIC or cross-validation method are used to search the optimal regularization parameter.

## Usage

```
marm3 <- function(Y,X,group=NULL,K=6,r1=NULL,r2=NULL,r3=NULL,
  method="BIC",ncv=10,penalty="LASSO",lambda=NULL,D0=NULL,
  intercept=TRUE,degr=3,nlam=20,lam_min=0.01,
  eps=1e-4,max_step=20, eps1=1e-4,max_step1=20,
  gamma=2,dfmax=NULL,alpha=1)
```

## Arguments

Y	A $n \times q$ numeric matrix of responses.
X	A $n \times p$ numeric design matrix for the model, where $p = \sum_g p_g$ .
group	A $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
K	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic spline.
r1	The first dimension of single value matrix of the tensor. Default is 2.
r2	The second dimension of single value matrix of the tensor. Default is 2.
r3	The third dimension of single value matrix of the tensor. Default is 2.
method	The method to be applied to select regularization parameters. Either BIC (default), or CV.
ncv	The number of cross-validation folds. Default is 10. If method is not CV, ncv is useless.
penalty	The penalty to be applied to the model. Either LASSO (the default), MCP or SCAD.
lambda	A user-specified sequence of lambda values. By default, a sequence of values of length nlam is computed, equally spaced on the log scale.
D0	A user-specified list of initialized values, including ng sub-lists where ng is the number of groups. For each sub-list, it has four initialized matrix $S_{\{3\}}$ (called S), A, B and C. By default, a list of initialization satisfying fixed ranks is computed by random.
intercept	A logical value indicating whether the intercept is fitted. Default is TRUE or set to zero by FALSE.
degr	The number of knots of B-spline base function. Default is 3.
nlam	The number of lambda values. Default is 20.
lam_min	The smallest value for lambda, as a fraction of lambda.max. Default is 0.01.
eps	Convergence threshold. The algorithm iterates until the relative change in any coefficient is less than eps. Default is 1e-4.
max_step	Maximum number of iterations. Default is 20.
eps1	Convergence threshold. The Coordinate descent method algorithm iterates until the relative change in any coefficient is less than eps1. Default is 1e-4.
max_step1	The maximum iterates number of coordinate descent method. Default is 20.
gamma	The tuning parameter of the MCP/SCAD penalty.
dfmax	Upper bound for the number of nonzero coefficients. Default is no upper bound. However, for large data sets, computational burden may be heavy for models with a large number of nonzero coefficients.

alpha                      Tuning parameter for the Mnet estimator which controls the relative contributions from the LASSO, MCP or SCAD penalty and the ridge, or L2 penalty. alpha=1 is equivalent to LASSO, MCP or SCAD penalty, while alpha=0 would be equivalent to ridge regression. However, alpha=0 is not supported; alpha may be arbitrarily small, but not exactly 0.

### Details

This function gives pq functional coefficients' estimators of marm. Multiple third-order tensors with multiple ranks  $(r_{1g}, r_{2g}, r_{3g})$  need to be estimated. We fix these ranks and use an alternative updating algorithm to update its core tensor and factor matrices based on Tucker decomposition. Group LASSO, SCAD or MCP penalty is applied on the row of each factor matrix  $A^g$  to achieve variable selection.

### Value

D	Estimator of coefficients $D_{(3)} = (D_{(3)}^1, \dots, D_{(3)}^{ng})$ where $ng$ is the number of groups.
mu	Estimator of intercept $\mu$ .
S.opt	A length- $ng$ list including estimator of the core tensor $S_{(3)}$ of each coefficient tensor.
A.opt	A length- $ng$ list including estimator of the factor matrix $A$ of each coefficient tensor.
B.opt	A length- $ng$ list including estimator of the factor matrix $B$ of each coefficient tensor.
C.opt	A length- $ng$ list including estimator of the factor matrix $C$ of each coefficient tensor.
lambda.seq	The sequence of regularization parameter values in the path.
lambda_opt	The value of lambda with the minimum BIC or CV value.
rss	Residual sum of squares (RSS).
df	Degrees of freedom.
activeX	The active set of $X$ . A length- $p$ vector.
opts	Other related parameters used in algorithm. Some of them are set by default.
opts	Other related parameters used in algorithm (especially parameters in penalty). Some of them are set by default.

### References

Multivariate additive regression for multi-view data via tensor estimation.

### See Also

marm3\_dr

### Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
```

```

s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm3.sim.fbs(n,q,p,s,group,isfixedR=1)
fit <- with(mydata, marm3(Y,X,group,K,r10,r20,r30,D0=D0,nlam=5))

```

marm3.dr

*Fit MARM with sparsity assumption and unknown ranks.*

## Description

Fit a multivariate additive model for multi-view data (MARM) using B-splines with unknown ranks ( $r_{1g}, r_{2g}, r_{3g}$ ). Multiple third-order coefficient tensors can be estimated by this function. The group sparse penalty such as LASSO, MCP or SCAD and the coordinate descent algorithm are used to yield a sparsity estimator. The BIC or cross-validation method are used to search the optimal regularization parameter, multiple ranks and the number of B-spline basis functions simultaneously.

## Usage

```

marm3.dr <- function(Y,X,group=NULL,K_index=NULL,r1_index=NULL,r2_index=NULL,
  r3_index=NULL,method="BIC",ncv=10,penalty="LASSO",
  lambda=NULL,D0=NULL,intercept=TRUE,nlam=50,degr=3,
  lam_min=0.01,eps=1e-4,max_step=20,eps1=1e-4,
  max_step1=20,gamma=2,dfmax=NULL,alpha=1)

```

## Arguments

Y	A $n \times q$ numeric matrix of responses.
X	A $n \times p$ numeric design matrix for the model, where $p = \sum_g p_g$
group	A $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
K_index	The user-specified sequence of K. Default is a length-1 vector 6.
r1_index	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ .
r2_index	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ .
r3_index	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ .
method	The method to be applied to select the number of B-spline basis functions, regularization parameters and multiple ranks simultaneously. Either BIC (default), or CV.
ncv	The number of cross-validation folds. Default is 10. If method is not CV, ncv is useless.
penalty	The penalty to be applied to the model. Either LASSO (the default), MCP or SCAD.
lambda	A user-specified sequence of lambda values. By default, a sequence of values of length nlam is computed, equally spaced on the log scale.

D0	A user-specified list of initialized values, including ng sub-lists where ng is the number of groups. For each sub-list, it has four initialized matrix $S_{(3)}$ (called S), A, B and C. By default, a list of initialization satisfying fixed ranks is computed by random.
intercept	A logical value indicating whether the intercept is fitted. Default is TRUE or set to zero by FALSE.
degr	The number of knots of B-spline base function. Default is 3.
nlam	The number of lambda values. Default is 50.
lam_min	The smallest value for lambda, as a fraction of lambda.max. Default is 0.01.
eps	Convergence threshold. The algorithm iterates until the relative change in any coefficient is less than eps. Default is 1e-4.
max_step	Maximum number of iterations. Default is 20.
eps1	Convergence threshold. The Coordinate descent method algorithm iterates until the relative change in any coefficient is less than eps1. Default is 1e-4.
max_step1	The maximum iterates number of coordinate descent method. Default is 20.
gamma	The tuning parameter of the MCP/SCAD penalty.
dfmax	Upper bound for the number of nonzero coefficients. Default is no upper bound. However, for large data sets, computational burden may be heavy for models with a large number of nonzero coefficients.
alpha	Tuning parameter for the Mnet estimator which controls the relative contributions from the LASSO, MCP or SCAD penalty and the ridge, or L2 penalty. alpha=1 is equivalent to LASSO, MCP or SCAD penalty, while alpha=0 would be equivalent to ridge regression. However, alpha=0 is not supported; alpha may be arbitrarily small, but not exactly 0.

## Details

This function gives pq functional coefficients' estimators of marm. Multiple third-order tensors with unknown ranks  $(r_1, r_2, r_3)$  need to be estimated. The BIC or CV can be applied to select the optimal regularization parameters, multiple ranks and the number of B-spline basis functions simultaneously. An alternative updating algorithm can be used to update its core tensor and factor matrices based on Tucker decomposition. Group LASSO, SCAD or MCP penalty is applied on the row of each factor matrix  $A^g$  to achieve variable selection. Generally, the number of B-spline basis functions we need is fixed by 6, i.e., cubic splines are used to approximate the component functions.

## Value

D	Estimator of coefficients $D_{(3)} = (D_{(3)}^1, \dots, D_{(3)}^{ng})$ .
mu	Estimator of intercept $\mu$ .
S.opt	A length-ng list including estimator of the core tensor $S_{(3)}$ of each coefficient tensor.
A.opt	A length-ng list including estimator of the factor matrix $A$ of each coefficient tensor.
B.opt	A length-ng list including estimator of the factor matrix $B$ of each coefficient tensor.
C.opt	A length-ng list including estimator of the factor matrix $C$ of each coefficient tensor.

rk_opt	The optimal ranks and the number of B-spline basis functions that selected by BIC, or CV. It is a vector with length 4, which are selected $r_1$ , $r_2$ , $r_3$ , and $K$ .
lambda.seq	The sequence of regularization parameter values in the path.
lambda_opt	The value of lambda with the minimum BIC or CV value.
rss	Residual sum of squares (RSS).
df	Degrees of freedom.
activeX	The active set of $X$ . A length- $p$ vector.
opts	Other related parameters used in algorithm. Some of them are set by default.
opts	Other related parameters used in algorithm (especially parameters in peanlty). Some of them are set by default.

## References

Multivariate additive regression for multi-view data via tensor estimation.

## See Also

marm3

## Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm3.sim.fbs(n,q,p,s,group)
fit <- with(mydata, marm3.dr(Y,X,group,K,r1_index,r2_index,r3_index,D0=D0,nlam=5))
```

---

marm3.sim.fbs

*Generate scenario I data from MARM model.*

---

## Description

Generate scenario I data for MARM model.

## Usage

```
marm3.sim.fbs <- function(n,q,p,s,group=NULL,r10=2,r20=2,r30=2,isfixedR=0,
  D3=NULL,K=6,degr=3,sigma2=NULL,seed_id=NULL,
  r1_index=NULL,r2_index=NULL,r3_index=NULL,D0=NULL)
```

**Arguments**

<code>n</code>	Sample size.
<code>q</code>	The number of responses, $q \geq 1$ .
<code>p</code>	The number of covariates, $p \geq 1$ .
<code>s</code>	The true covariates associated with responses, $s \geq 1$ .
<code>group</code>	A length- $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
<code>r10</code>	The first dimension of the tensor. Default is 2.
<code>r20</code>	The second dimension of the tensor. Default is 2.
<code>r30</code>	The third dimension of the tensor. Default is 2.
<code>isfixedR</code>	A logical value indicating whether ranks are fixed.
<code>D3</code>	The mode of unfolding $D_{(3)}$ . By default, D3 is generated by random.
<code>K</code>	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic splines.
<code>degr</code>	The number of knots of B-spline base function. Default is 3.
<code>sigma2</code>	err variance. Default is 0.1.
<code>seed_id</code>	A positive integer, the seed for generating the random numbers. Default is 1000.
<code>r1_index</code>	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of single value matrix of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ . if $isfixedR = 1$ , $r1\_index$ is useless.
<code>r2_index</code>	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of single value matrix of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ . if $isfixedR = 1$ , $r2\_index$ is useless.
<code>r3_index</code>	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of single value matrix of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ . if $isfixedR = 1$ , $r3\_index$ is useless.
<code>D0</code>	A user-specified list of initialized values, including $ng$ sub-lists where $ng$ is the number of groups. For each sub-list, it has four initialized matrices $S_{(3)}$ (called S), A, B and C. By default, a list of initialization satisfying fixed ranks is computed by random.

**Details**

This function can generate scenario I data of marm model. `isfixedR` is required to yield a different initialization D0. In scenario I, the true functions are exactly residing in the space of B-spline basis functions.

**Value**

<code>Y</code>	Response, a $n \times q$ -matrix.
<code>X</code>	Design matrix, a $n \times p$ -matrix.
<code>f0</code>	True functions, a $n \times p$ -matrix.
<code>group</code>	The grouping index of predictors, a length- $p$ vector.
<code>D0</code>	The initialized values.
<code>...</code>	Other options for algorithm.



## References

Multivariate additive regression for multi-view data via tensor estimation.

## See Also

marm3.sim.fsin

## Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm3.sim.fbs(n,q,p,s,group)
```

---

marm3.sim.fsin

*Generate scenario II data from MARM model.*

---

## Description

Generate scenario II data for a multivariate additive model for multi-view data.

## Usage

```
marm3.sim.fsin <- function(n,q,p,s,group=NULL,r10=2,r20=2,r30=2,isfixedR=0,
                           D2=NULL,K=6,degr=3,sigma2=NULL,seed_id=NULL,
                           r1_index=NULL,r2_index=NULL,r3_index=NULL,D0=NULL)
```

## Arguments

n	Sample size.
q	The number of responses, $q \geq 1$ .
p	The number of covariates, $p \geq 1$ .
s	The true covariates associated with responses, $s \geq 1$ .
group	A length- $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
r10	The first dimension of the tensor. Default is 2.
r20	The second dimension of the tensor. Default is 2.
r30	The third dimension of the tensor. Default is 2.
isfixedR	A logical value indicating whether ranks are fixed.
D2	The mode of unfolding $D_{(2)}$ . By default, D2 is generated by random.
K	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic spline.

degr	The number of knots of B-spline base function. Default is 3.
sigma2	err variance. Default is 0.1.
seed_id	A positive integer, the seed for generating the random numbers. Default is 1000.
r1_index	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of single value matrix of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ . if $isfixedR = 1$ , $r1\_index$ is useless.
r2_index	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of single value matrix of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ . if $isfixedR = 1$ , $r2\_index$ is useless.
r3_index	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of single value matrix of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ . if $isfixedR = 1$ , $r3\_index$ is useless.
D0	A user-specified list of initialized values, including ng sub-lists where ng is the number of groups. For each sub-list, it has four initialized matrices $S_{(3)}$ (called S), A, B and C. By default, a list of initialization satisfying fixed ranks is computed by random.

### Details

This function can generate scenario II data of marm model. `isfixedR` is required to yield a different initialization D0. In scenario II, the true functions are the linear combination of  $\sin(2\pi x)$  and  $\cos(\pi x)$ .

### Value

Y	Response, a $n \times q$ -matrix.
X	Design matrix, a $n \times p$ -matrix.
f0	True functions, a $n \times p$ -matrix.
group	The grouping index of predictors, a length- $p$ vector.
D0	The initialized values.
...	Other options for algorithm.

### References

Multivariate additive regression for multi-view data via tensor estimation.

### See Also

marm3.sim.fbs

### Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng, each=p/ng)
mydata <- marm3.sim.fsin(n, q, p, s, group)
```

marm4

*Fit structural MARM with sparsity assumption and fixed ranks.***Description**

Fit a structural intergrative multi-view multivariate additive model (structural marm) using B-splines with given ranks  $(r_1, r_2, r_3, r_4)$ . A fourth-order coefficient tensor can be estimated by this function. The group sparse penalty such as LASSO, MCP or SCAD and the coordinate descent algorithm are used to yield a sparsity estimator. The BIC or cross-validation method are used to search the optimal regularization parameter.

**Usage**

```
marm4 <- function(Y,X,group=NULL,K=6,r1=NULL,r2=NULL,r3=NULL,r4=NULL,
  method="BIC",ncv=10,penalty="LASSO",lambda=NULL,D0=NULL,
  intercept=TRUE,nlam=20,degr=3,lam_min=0.01,
  eps=1e-4,max_step=10,eps1=1e-4,max_step1=10,
  gamma=2,dfmax=NULL,alpha=1)
```

**Arguments**

Y	A $n \times q$ numeric matrix of responses.
X	A $n \times p$ numeric design matrix for the model, where $p = \sum_g p_g$
group	A $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
K	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic splines.
r1	The first dimension of single value matrix of the tensor. Default is 2.
r2	The second dimension of single value matrix of the tensor. Default is 2.
r3	The third dimension of single value matrix of the tensor. Default is 2.
r4	The fourth dimension of single value matrix of the tensor. Default is 2.
method	The method to be applied to select regularization parameters. Either BIC (default), or CV.
ncv	The number of cross-validation folds. Default is 10. If method is not CV, ncv is useless.
penalty	The penalty to be applied to the model. Either LASSO (the default), MCP or SCAD.
lambda	A user-specified sequence of lambda values. By default, a sequence of values of length nlam is computed, equally spaced on the log scale.
D0	A user-specified list of initialized values, including five initialized matrix $S_{(4)}$ (called S), A, B, C and D. By default, a list of initialization satisfying fixed ranks is computed by random.
intercept	A logical value indicating whether the intercept is fitted. Default is TRUE or set to zero by FALSE.
degr	The number of knots of B-spline base function. Default is 3.

n <sub>lam</sub>	The number of lambda values. Default is 20.
lam <sub>min</sub>	The smallest value for lambda, as a fraction of lambda.max. Default is 0.01.
eps	Convergence threshold. The algorithm iterates until the relative change in any coefficient is less than eps. Default is 1e-4.
max_step	Maximum number of iterations. Default is 20.
eps1	Convergence threshold. The Coordinate descent method algorithm iterates until the relative change in any coefficient is less than eps1. Default is 1e-4.
max_step1	The maximum iterates number of coordinate descent method. Default is 20.
gamma	The tuning parameter of the MCP/SCAD penalty.
dfmax	Upper bound for the number of nonzero coefficients. Default is no upper bound. However, for large data sets, computational burden may be heavy for models with a large number of nonzero coefficients.
alpha	Tuning parameter for the Mnet estimator which controls the relative contributions from the LASSO, MCP/SCAD penalty and the ridge, or L2 penalty. alpha=1 is equivalent to LASSO, MCP/SCAD penalty, while alpha=0 would be equivalent to ridge regression. However, alpha=0 is not supported; alpha may be arbitrarily small, but not exactly 0.

### Details

This function gives pq functional coefficients' estimators of marm. A fourth-order tensor with multiple ranks  $(r_1, r_2, r_3, r_4)$  need to be estimated. We fix these ranks and use an alternative updating algorithm to update its core tensor and factor matrices based on Tucker decomposition. Group LASSO, SCAD or MCP penalty is applied on the row of the factor matrix  $A$  to achieve variable selection.

### Value

D	Estimator of $D_{(4)}$ .
mu	Estimator of intercept $\mu$ .
S.opt	A length- $ng$ list including estimator of the core tensor $S_{(3)}$ of each coefficient tensor.
A.opt	A length- $ng$ list including estimator of the factor matrix $A$ of each coefficient tensor.
B.opt	A length- $ng$ list including estimator of the factor matrix $B$ of each coefficient tensor.
C.opt	A length- $ng$ list including estimator of the factor matrix $C$ of each coefficient tensor.
D.opt	A length- $ng$ list including estimator of the factor matrix $D$ of each coefficient tensor.
lambda.seq	The sequence of regularization parameter values in the path.
lambda_opt	The value of lambda with the minimum BIC or CV value.
rss	Residual sum of squares (RSS).
df	Degrees of freedom.
activeX	The active set of $X$ . A length- $p$ vector.
opts	Other related parameters used in algorithm. Some of them are set by default.
opts	Other related parameters used in algorithm (especially parameters in penalty). Some of them are set by default.

## References

Multivariate additive regression for multi-view data via tensor estimation.

## See Also

marm4\_dr

## Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm4.sim.fbs(n,q,p,s,group,isfixedR=1)
fit <- with(mydata, marm4(Y,X,group,K,r10,r20,r30,r40,D0=D0,nlam=5))
```

---

marm4.dr

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*Fit structural MARM with sparsity assumption and unknown ranks.*


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## Description

Fit a structural multivariate additive model for multi-view data (structural MARM) using B-splines with unknown ranks  $(r_1, r_2, r_3, r_4)$ . A fourth-order coefficient tensor can be estimated by this function. The group sparse penalty such as LASSO, MCP or SCAD and the coordinate descent algorithm are used to yield a sparsity estimator. The BIC or cross-validation method is used to search the optimal regularization parameter, multiple ranks and the number of B-spline basis functions simultaneously.

## Usage

```
marm4.dr <- function(Y,X,group,K_index=NULL,r1_index=NULL,r2_index=NULL,
  r3_index=NULL,r4_index=NULL,method="BIC",ncv=10,
  penalty="LASSO",lambda=NULL,D0=NULL,
  intercept=TRUE,nlam=20,degr=3,lam_min=0.01,
  eps=1e-4,max_step=10,eps1=1e-4,max_step1=10,
  gamma=2,dfmax=NULL,alpha=1)
```

## Arguments

Y	A $n \times q$ numeric matrix of responses.
X	A $n \times p$ numeric design matrix for the model, where $p = \sum_g p_g$
group	A $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = rep(1, p)$ .
K_index	The user-specified sequence of K. Default is a length-1 vector 6.
r1_index	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of single value matrix of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ .

r2_index	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of single value matrix of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ .
r3_index	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of single value matrix of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, ng)$ .
r4_index	A user-specified sequence of $r_4$ values, where $r_4$ is the third dimension of single value matrix of the tensor. Default is $r4\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ .
method	The method to be applied to search the number of B-spline basis functions, regularization parameters and multiple ranks simultaneously. Either BIC (default), or CV.
ncv	The number of cross-validation folds. Default is 10. If method is not CV, ncv is useless.
penalty	The penalty to be applied to the model. Either LASSO (the default), MCP or SCAD.
lambda	A user-specified sequence of lambda values. By default, a sequence of values of length nlam is computed, equally spaced on the log scale.
D0	A user-specified list of initialized values, including five initialized matrix $S_{(4)}$ (called S), A, B, C and D. By default, a list of initialization satisfying fixed ranks is computed by random.
intercept	A logical value indicating whether the intercept is fitted. Default is TRUE or set to zero by FALSE.
degr	The number of knots of B-spline base function. Default is 3.
nlam	The number of lambda values. Default is 20.
lam_min	The smallest value for lambda, as a fraction of lambda.max. Default is 0.01.
eps	Convergence threshold. The algorithm iterates until the relative change in any coefficient is less than eps. Default is 1e-4.
max_step	Maximum number of iterations. Default is 20.
eps1	Convergence threshold. The Coordinate descent method algorithm iterates until the relative change in any coefficient is less than eps1. Default is 1e-4.
max_step1	The maximum iterates number of coordinate descent method. Default is 20.
gamma	The tuning parameter of the MCP or SCAD penalty (see details).
dfmax	Upper bound for the number of nonzero coefficients. Default is no upper bound. However, for large data sets, computational burden may be heavy for models with a large number of nonzero coefficients.
alpha	Tuning parameter for the Mnet estimator which controls the relative contributions from the LASSO, MCP or SCAD penalty and the ridge, or L2 penalty. alpha=1 is equivalent to LASSO, MCP or SCAD penalty, while alpha=0 would be equivalent to ridge regression. However, alpha=0 is not supported; alpha may be arbitrarily small, but not exactly 0.

## Details

This function gives pq functional coefficients' estimators of structural marm. A fourth-order tensor with unknown ranks ( $r_1, r_2, r_3, r_4$ ) need to be estimated. The BIC or CV can be applied to select the optimal regularization parameter, multiple ranks and the number of B-spline basis functions simultaneously. An alternative updating algorithm can be used to update its core tensor and factor matrices based on Tucker decomposition. Group LASSO, SCAD or MCP penalty is applied on the row of the factor matrix  $A$  to achieve variable selection.

**Value**

D	Estimator of $D_{(4)}$ .
mu	Estimator of intercept $\mu$ .
S.opt	A length- $ng$ list including estimator of the core tensor $S_{(3)}$ of each coefficient tensor.
A.opt	A length- $ng$ list including estimator of the factor matrix $A$ of each coefficient tensor.
B.opt	A length- $ng$ list including estimator of the factor matrix $B$ of each coefficient tensor.
C.opt	A length- $ng$ list including estimator of the factor matrix $C$ of each coefficient tensor.
D.opt	A length- $ng$ list including estimator of the factor matrix $D$ of each coefficient tensor.
rk_opt	The optimal ranks and the number of B-spline basis functions that selected by BIC, or CV. It is a vector with length 5, which are selected $r_1, r_2, r_3, r_4$ and $K$ .
lambda.seq	The sequence of regularization parameter values in the path.
lambda_opt	The value of lambda with the minimum BIC or CV value.
rss	Residual sum of squares (RSS).
df	Degrees of freedom.
activeX	The active set of $X$ . A length- $p$ vector.
opts	Other related parameters used in algorithm. Some of them are set by default.
opts	Other related parameters used in algorithm (especially parameters in peanlty). Some of them are set by default.

**References**

Multivariate additive regression for multi-view data via tensor estimation.

**See Also**

marm4

**Examples**

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm4.sim.fbs(n,q,p,s,group)
fit <- with(mydata, marm4.dr(Y,X,group,K,r1_index,r2_index,r3_index,r4_index,D0=D0,nlam=5))
```

marm4.sim.fbs

*Generate scenario I data from structural MARM model.***Description**

Generate scenario I data for a structural multivariate additive model for multi-view data.

**Usage**

```
marm4.sim.fbs <- function(n,q,p,s,group=NULL,r10=2,r20=2,r30=2,r40=2,
  isfixedR=0,D44=NULL,K=6,degr=3,sigma2=NULL,
  seed_id=NULL,r1_index=NULL,r2_index=NULL,
  r3_index=NULL,r4_index=NULL,D0=NULL)
```

**Arguments**

n	Sample size.
q	The number of responses, $q \geq 1$ .
p	The number of covariates, $p \geq 1$ .
s	The true covariates associated with responses, $s \geq 1$ .
group	A length- $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = c(rep(1, as.integer(p/2)), rep(2, p - as.integer(p/2)))$ .
r10	The first dimension of the tensor. Default is 2.
r20	The second dimension of the tensor. Default is 2.
r30	The third dimension of the tensor. Default is 2.
r40	The third dimension of the tensor. Default is 2.
isfixedR	A logical value indicating whether ranks are fixed.
D44	The mode of unfolding $D_{(4)}$ . By default, D44 is generated by random.
K	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic splines.
degr	The number of knots of B-spline base function. Default is 3.
sigma2	err variance. Default is 0.1.
seed_id	A positive integer, the seed for generating the random numbers. Default is 1000.
r1_index	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of single value matrix of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ . if $isfixedR = 1$ , $r1\_index$ is useless.
r2_index	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of single value matrix of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ . if $isfixedR = 1$ , $r2\_index$ is useless.
r3_index	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of single value matrix of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, ng)$ . if $isfixedR = 1$ , $r3\_index$ is useless.



r4_index	A user-specified sequence of $r_4$ values, where $r_4$ is the third dimension of single value matrix of the tensor. Default is $r4\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ . if $isfixedR = 1$ , $r4\_index$ is useless.
D0	A user-specified list of initialized values, including five initialized matrices $S_{(4)}$ (called S), A, B, C and D. By default, a list of initialization satisfying fixed ranks is computed by random.

### Details

This function can generate scenario I data of structural marm model. `isfixedR` are required to yield a different initialization `D0`. In scenario I, the true functions are exactly residing in the space of B-spline basis functions.

### Value

Y	Response, a $n \times q$ -matrix.
X	Design matrix, a $n \times p$ -matrix.
f0	True functions, a $n \times p$ -matrix.
group	The grouping index of predictors, a length- $p$ vector.
D0	The initialized values.
...	Other options for algorithm.

### References

Multivariate additive regression for multi-view data via tensor estimation.

### See Also

`marm4.sim.fsin`

### Examples

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng, each=p/ng)
mydata <- marm4.sim.fbs(n,q,p,s,group)
```

---

`marm4.sim.fsin`

*Generate scenario II data from structural MARM model.*

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### Description

Generate scenario II data for a structural multivariate additive model for multi-view data.

## Usage

```
marm4.sim.fsin <- function(n,q,p,s,group=NULL,r10=2,r20=2,r30=2,r40=2,
  isfixedR=0,D42=NULL,K=6,degr=3,sigma2=NULL,
  seed_id=NULL,r1_index=NULL,r2_index=NULL,
  r3_index=NULL,r4_index=NULL,D0=NULL)
```

## Arguments

n	Sample size.
q	The number of responses, $q \geq 1$ .
p	The number of covariates, $p \geq 1$ .
s	The true covariates associating to responses, $s \geq 1$ .
group	A length- $p$ vector of the grouping index of predictors, e.g., $group = c(1, 1, 1, 2, 2, 2)$ means there are 6 predictors in the model, and the first three predictors are in the same group and the last three predictors are in another one. By default, we set $group = c(rep(1, as.integer(p/2)), rep(2, p - as.integer(p/2)))$ .
r10	The first dimension of the tensor. Default is 2.
r20	The second dimension of the tensor. Default is 2.
r30	The third dimension of the tensor. Default is 2.
r40	The third dimension of the tensor. Default is 2.
isfixedR	A logical value indicating whether ranks are fixed.
D42	The mode of unfolding $D_{(2)}$ . By default, D2 is generated by random.
K	The number of B-spline basis functions, that is the plus of both degrees of basis functions and the number of knots. Default is 6, which means cubic splines.
degr	The number of knots of B-spline base function. Default is 3.
sigma2	err variance. Default is 0.1.
seed_id	A positive integer, the seed for generating the random numbers. Default is 1000.
r1_index	A user-specified sequence of $r_1$ values, where $r_1$ is the first dimension of single value matrix of the tensor. Default is $r1\_index = 1, \dots, \min(\lceil \log(n) \rceil, p)$ . if $isfixedR = 1$ , $r1\_index$ is useless.
r2_index	A user-specified sequence of $r_2$ values, where $r_2$ is the second dimension of single value matrix of the tensor. Default is $r2\_index = 1, \dots, \max\{K\_index\}$ . if $isfixedR = 1$ , $r2\_index$ is useless.
r3_index	A user-specified sequence of $r_3$ values, where $r_3$ is the third dimension of single value matrix of the tensor. Default is $r3\_index = 1, \dots, \min(\lceil \log(n) \rceil, ng)$ . if $isfixedR = 1$ , $r3\_index$ is useless.
r4_index	A user-specified sequence of $r_4$ values, where $r_4$ is the third dimension of single value matrix of the tensor. Default is $r4\_index = 1, \dots, \min(\lceil \log(n) \rceil, q)$ . if $isfixedR = 1$ , $r4\_index$ is useless.
D0	A user-specified list of initialized values, including five initialized matrices $S_{(4)}$ (called S), A, B, C and D. By default, a list of initialization satisfying fixed ranks is computed by random.

## Details

This function can generate scenario II data of structural marm model. `isfixedR` is required to yield a different initialization `D0`. In scenario II, the true functions are the linear combination of  $\sin(2\pi x)$  and  $\cos(\pi x)$ .

**Value**

Y	Response, a $n \times q$ -matrix.
X	Design matrix, a $n \times p$ -matrix.
f0	True functions, a $n \times p$ -matrix.
group	The grouping index of predictors, a length- $p$ vector.
D0	The initialized values.
...	Other options for algorithm.

**References**

Multivariate additive regression for multi-view data via tensor estimation.

**See Also**

marm4.sim.fbs

**Examples**

```
library(tensorMARM)
n <- 200
q <- 5
p <- 100
s <- 3
ng <- 4
group <- rep(1:ng,each=p/ng)
mydata <- marm4.sim.fsin(n,q,p,s,group)
```

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