

# Package ‘gfmR’

March 14, 2018

**Type** Package

**Title** Implements Group Fused Multinomial Regression

**Version** 1.1-0

**Date** 2018-03-13

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**Depends** R (>= 3.1.0), parallel, nnet, faraway

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

**Description** Software to implement methodology to perform automatic response category combinations in multinomial logistic regression. There are functions for both cross validation and AIC for model selection. The method provides regression coefficient estimates that may be useful for better understanding the true probability distribution of multinomial logistic regression when category probabilities are similar. These methods are not recommended for a large number of predictor variables.

**License** GPL-2

**LazyLoad** yes

**NeedsCompilation** yes

**URL** <https://www.r-project.org>

**Repository** CRAN

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## R topics documented:

GFMR.cv . . . . .	2
GroupFusedMulti . . . . .	3
predict.gfmR . . . . .	5
print.gfmR . . . . .	7
print.gfmR.cv . . . . .	8

<b>Index</b>	<b>10</b>
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GFMR.cv

*Tuning parameter selection using validation likelihood for GFMR.*


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### Description

This routine implements K fold cross validation for group fused multinomial regression.

### Usage

```
GFMR.cv(Y,X,lamb,sampID,H,n.cores=1,rho=10^-8,...)
```

### Arguments

Y	A matrix of response category counts where the columns represent the categories and rows represent the observations. Currently supported for n=1.
X	A matrix of predictor variables. The columns represent predictors and rows represent observations.
lamb	tuning parameter for fusion penalty
sampID	An identified or the sampleID for the cross validation routine. Should take values 1:k and be user supplied.
H	An indicator matrix representing the edge set of the penalty set. The matrix is square and symmetric with dimension number of response categories, and if two categories are in the penalty set a 1 should be in the row column combination.
n.cores	The number of cores for the mclapply function
rho	Step Size parameter of ADMM
...	Other arguments for Group fused multinomial regression

### Details

This routine implements the validation likelihood approach proposed by Price et. al. to obtain the tuning parameter for Group Fused Multinomial Regression which automatically combines response categories in multinomial regression. Show to do well with regard to predicting the true category probabilities.

### Value

A list is returned with elements

v1	The validation likelihood for all tuning parameters
v1.sd	The standard deviation of the validation likelihood
lambda	tuning parameter used
v1.mat	Validation likelihood for each

**Author(s)**

Brad Price, <brad.price@mail.wvu.edu>.

**References**

Price, B.S, Geyer, C.J. and Rothman, A.J. "Automatic Response Category Combination in Multinomial Logistic Regression." <https://arxiv.org/abs/1705.03594>.

**Examples**

```
## Not run: data(nes96)
attach(nes96)
Response=matrix(0,944,7)
for(i in 1:944){
  if(PID[i]=="strRep"){Response[i,1]=1}
  if(PID[i]=="weakRep"){Response[i,2]=1}
  if(PID[i]=="indRep"){Response[i,3]=1}
  if(PID[i]=="indind"){Response[i,4]=1}
  if(PID[i]=="indDem"){Response[i,5]=1}
  if(PID[i]=="weakDem"){Response[i,6]=1}
  if(PID[i]=="strDem"){Response[i,7]=1}
}

Hmat=matrix(1,dim(Response)[2],dim(Response)[2])
diag(Hmat)=0
ModMat<-lm(popul~age,x=TRUE)$x

X=cbind(ModMat[,1],apply(ModMat[,-1],2,scale))

set.seed(1010)
n=dim(Response)[1]
sampID=rep(5,n)
samps=sample(1:n)
mine=floor(n/5)
for(j in 1:4){
  sampID[samps[((j-1)*mine+1):(j*mine)]] = j
}

o1<-GFMR.cv(Response,X,lamb = 2^seq(4.2,4.3,.1),H=Hmat2,sampID = sampID,n.cores =5)

which(o1$v1==max(o1$v1))

## End(Not run)
```

**Description**

This routine fits the group fused multinomial logistic regression model, which uses fusion shrinkage to automatically combine response categories.

**Usage**

```
GroupFusedMulti(Y,X,lambda,H,tol1=10^-7,tol2=10^-7,TD=2,rho=10^-8,tau1=10^-9,iter=1e3)
```

**Arguments**

Y	A matrix of response category counts where the columns represent the categories and rows represent the observations. Currently supported for n=1.
X	A matrix of predictor variables. The columns represent predictors and rows represent observations.
lambda	tuning parameter for fusion penalty
H	An indicator matrix representing the edge set of the penalty set. The matrix is square and symmetric with dimension number of response categories, and if two categories are in the penalty set a 1 should be in the row column combination.
tol1	Convergence tolerance for ADMM
tol2	Convergence tolerance of ADMM
TD	Step size Adjustment for iterative step size
rho	ADMM step-size parameter, iterative implementation
tau1	The threshold parameter to 0 for final estimates
iter	Maximum number of iterations of the algorithm

**Details**

Implements the ADMM algorithm for the group fused multinomial regression estimates proposed by Price et. al.

**Value**

The function returns a list:

Coeff	Final coefficient estimates for the gfmr model
Approx	Final iterates of beta before threshold
Z	Final iterate of Z in the ADMM algorithm
lambda	Tuning Parameter
Converge	Indicator of algorithm convergence
NGroups	Number of groups in final estimates
Groups	The response category groups in the final estimates

**Author(s)**

Brad Price, <brad.price@mail.wvu.edu>.

## References

Price, B.S, Geyer, C.J. and Rothman, A.J. "Automatic Response Category Combination in Multinomial Logistic Regression." <https://arxiv.org/abs/1705.03594>.

## Examples

```
## Not run: data(nes96)
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Response=matrix(0,944,7)
for(i in 1:944){
  if(PID[i]=="strRep"){Response[i,1]=1}
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  if(PID[i]=="indDem"){Response[i,5]=1}
  if(PID[i]=="weakDem"){Response[i,6]=1}
  if(PID[i]=="strDem"){Response[i,7]=1}
}

Hmat=matrix(1,dim(Response)[2],dim(Response)[2])
diag(Hmat)=0
ModMat<-lm(popul~age, x=TRUE)$x

X=cbind(ModMat[,1], apply(ModMat[, -1], 2, scale))
mod<-GroupFusedMulti(Response,X, lambda=2^4.3, H=Hmat2, rho=10^2, iter=50, tol1=10^-4, tol2=10^-4)
predict(mod, X[1,])

## End(Not run)
```

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predict.gfmR

*Predict method for group fused multinomial logistic regression*

---

## Description

This routine fits the group fused multinomial logistic regression model, which uses fusion shrinkage to automatically combine response categories.

## Usage

```
## S3 method for class 'gfmR'
predict(object,newdata,type="probs",...)
```

## Arguments

object	A gfmr object which specifically is the output from the GroupFusedMulti function.
newdata	value to be predicted in model matrix form

type	Type of prediction, "probs" returns probabilities, while response returns log of the ratio of probabilities for baseline category.
...	Other arguments

**Details**

Prediction function for GFMR

**Value**

A vector or a matrix corresponding to type return.

**Author(s)**

Brad Price, <brad.price@mail.wvu.edu>.

**References**

Price, B.S, Geyer, C.J. and Rothman, A.J. "Automatic Response Category Combination in Multi-nomial Logistic Regression." <https://arxiv.org/abs/1705.03594>.

**Examples**

```
## Not run: data(nes96)
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Response=matrix(0,944,7)
for(i in 1:944){
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  if(PID[i]=="weakDem"){Response[i,6]=1}
  if(PID[i]=="strDem"){Response[i,7]=1}
}

Hmat=matrix(1,dim(Response)[2],dim(Response)[2])
diag(Hmat)=0
ModMat<-lm(popul~age, x=TRUE)$x

X=cbind(ModMat[,1],apply(ModMat[,-1],2,scale))
mod<-GroupFusedMulti(Response,X,lambda=2^4.3,H=Hmat2,rho=10^2,iter=50,tol1=10^-4,tol2=10^-4)
predict(mod,X[1,])

## End(Not run)
```

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print.gfmR	<i>print method for group fused multinomial logistic regression</i>
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## Description

This routine fits the group fused multinomial logistic regression model, which uses fusion shrinkage to automatically combine response categories.

## Usage

```
## S3 method for class 'gfmR'  
print(x,...)
```

## Arguments

x	A gfmr object which specifically is the output from the GroupFusedMulti function.
...	Other arguments

## Details

Prediction function for GFMR

## Value

A vector or a matrix corresponding to type return.

## Author(s)

Brad Price, <brad.price@mail.wvu.edu>.

## References

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  if(PID[i]=="indDem"){Response[i,5]=1}  
  if(PID[i]=="weakDem"){Response[i,6]=1}
```

```
  if(PID[i]=="strDem"){Response[i,7]=1}
}

Hmat=matrix(1,dim(Response)[2],dim(Response)[2])
diag(Hmat)=0
ModMat<-lm(popul~age,x=TRUE)$x

X=cbind(ModMat[,1],apply(ModMat[,-1],2,scale))
mod<-GroupFusedMulti(Response,X,lambda=2^4.3,H=Hmat2,rho=10^2,iter=50,to1=10^-4,to2=10^-4)
mod

## End(Not run)
```

---

print.gfmR.cv

*print method for group fused multinomial logistic regression validation likelihood tuning parameter selection*

---

## Description

This routine fits the group fused multinomial logistic regression model, which uses fusion shrinkage to automatically combine response categories. This specifically focuses on tuning parameter selection with validation likelihood.

## Usage

```
## S3 method for class 'gfmR.cv'
print(x,...)
```

## Arguments

x	A gfmR.cv object which specifically is the output from the GroupFusedMulti function.
...	Other arguments

## Details

print method for gfmR.cv objects.

## Value

A readable printout of cross validation

## Author(s)

Brad Price, <brad.price@mail.wvu.edu>.

## References

Price, B.S, Geyer, C.J. and Rothman, A.J. "Automatic Response Category Combination in Multinomial Logistic Regression." <https://arxiv.org/abs/1705.03594>.

## Examples

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## Not run: data(nes96)
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Hmat=matrix(1,dim(Response)[2],dim(Response)[2])
diag(Hmat)=0
ModMat<-lm(popul~age,x=TRUE)$x

X=cbind(ModMat[,1],apply(ModMat[,-1],2,scale))

set.seed(1010)
n=dim(Response)[1]
sampID=rep(5,n)
samps=sample(1:n)
mine=floor(n/5)
for(j in 1:4){
  sampID[samps[((j-1)*mine+1):(j*mine)]] = j
}

o1<-GFMR.cv(Response,X,lamb = 2^seq(4.2,4.3,.1),H=Hmat2,sampID = sampID,n.cores =5)
o1

## End(Not run)
```

# Index

## \*Topic **regression**

GFMR.cv, [2](#)

GroupFusedMulti, [3](#)

predict.gfmR, [5](#)

print.gfmR, [7](#)

print.gfmR.cv, [8](#)

GFMR.cv, [2](#)

GroupFusedMulti, [3](#)

predict.gfmR, [5](#)

print.gfmR, [7](#)

print.gfmR.cv, [8](#)