

# Package ‘gglasso’

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**Title** Group Lasso Penalized Learning Using a Unified BMD Algorithm

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**Description** A unified algorithm, blockwise-majorization-descent (BMD), for efficiently computing the solution paths of the group-lasso penalized least squares, logistic regression, Huberized SVM and squared SVM. The package is an implementation of Yang, Y. and Zou, H. (2015) DOI: <doi:10.1007/s11222-014-9498-5>.

**License** GPL-2

**Imports** methods

**URL** <https://github.com/emeryyi/gglasso>

**BugReports** <https://github.com/emeryyi/gglasso/issues>

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bardet	<i>Simplified gene expression data from Scheetz et al. (2006)</i>
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### Description

Gene expression data (20 genes for 120 samples) from the microarray experiments of mammalian eye tissue samples of Scheetz et al. (2006).

### Usage

bardet

### Format

An object of class `list` of length 2.

### Details

This data set contains 120 samples with 100 predictors (expanded from 20 genes using 5 basis B-splines, as described in Yang, Y. and Zou, H. (2015)).

### Value

A list with the following elements:

- x a [120 x 100] matrix (expanded from a [120 x 20] matrix) giving the expression levels of 20 filtered genes for the 120 samples. Each row corresponds to a subject, each 5 consecutive columns to a grouped gene.
- y a numeric vector of length 120 giving expression level of gene TRIM32, which causes Bardet-Biedl syndrome.

## References

Scheetz, T., Kim, K., Swiderski, R., Philp, A., Braun, T., Knudtson, K., Dorrance, A., DiBona, G., Huang, J., Casavant, T. et al. (2006), "Regulation of gene expression in the mammalian eye and its relevance to eye disease", *Proceedings of the National Academy of Sciences* **103**(39), 14429-14434.

Huang, J., S. Ma, and C.-H. Zhang (2008). "Adaptive Lasso for sparse high-dimensional regression models". *Statistica Sinica* 18, 1603-1618.

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.

BugReport: <https://github.com/emeryyi/glasso>

## Examples

```
# load gglasso library
library(gglasso)

# load data set
data(bardet)

# how many samples and how many predictors ?
dim(bardet$x)

# response y
bardet$y
```

---

coef.cv.glasso	<i>get coefficients or make coefficient predictions from a "cv.glasso" object.</i>
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---

## Description

This function gets coefficients or makes coefficient predictions from a cross-validated gglasso model, using the stored "gglasso.fit" object, and the optimal value chosen for lambda.

## Usage

```
## S3 method for class 'cv.glasso'
coef(object, s = c("lambda.1se", "lambda.min"), ...)
```

## Arguments

object	fitted <code>cv.glasso</code> object.
s	value(s) of the penalty parameter lambda at which predictions are required. Default is the value <code>s="lambda.1se"</code> stored on the CV object, it is the largest value of lambda such that error is within 1 standard error of the minimum. Alternatively <code>s="lambda.min"</code> can be used, it is the optimal value of lambda that gives minimum cross validation error <code>cvm</code> . If s is numeric, it is taken as the value(s) of lambda to be used.
...	not used. Other arguments to <code>predict</code> .

## Details

This function makes it easier to use the results of cross-validation to get coefficients or make coefficient predictions.

## Value

The coefficients at the requested values for lambda.

## Author(s)

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

## References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.

BugReport: <https://github.com/emeryyi/glasso>

Friedman, J., Hastie, T., and Tibshirani, R. (2010), "Regularization paths for generalized linear models via coordinate descent," *Journal of Statistical Software*, 33, 1.

<http://www.jstatsoft.org/v33/i01/>

## See Also

`cv.glasso`, and `predict.cv.glasso` methods.

## Examples

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)
```

```
# 5-fold cross validation using group lasso
# penalized logistic regression
cv <- cv.gglasso(x=colon$x, y=colon$y, group=group, loss="logit",
pred.loss="misclass", lambda.factor=0.05, nfolds=5)

# the coefficients at lambda = lambda.1se
pre = coef(cv$gglasso.fit, s = cv$lambda.1se)
```

---

coef.gglasso	<i>get coefficients or make coefficient predictions from an "gglasso" object.</i>
--------------	---

---

## Description

Computes the coefficients at the requested values for lambda from a fitted `gglasso` object.

## Usage

```
## S3 method for class 'gglasso'
coef(object, s = NULL, ...)
```

## Arguments

object	fitted <code>gglasso</code> model object.
s	value(s) of the penalty parameter lambda at which predictions are required. Default is the entire sequence used to create the model.
...	not used. Other arguments to predict.

## Details

s is the new vector at which predictions are requested. If s is not in the lambda sequence used for fitting the model, the coef function will use linear interpolation to make predictions. The new values are interpolated using a fraction of coefficients from both left and right lambda indices.

## Value

The coefficients at the requested values for lambda.

## Author(s)

Yi Yang and Hui Zou  
 Maintainer: Yi Yang <yi.yang6@mcgill.ca>

## References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
 BugReport: <https://github.com/emeryyi/gglasso>

**See Also**

[predict.gglasso](#) method

**Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)

# fit group lasso
m1 <- gglasso(x=colon$x,y=colon$y,group=group,loss="logit")

# the coefficients at lambda = 0.01 and 0.02
coef(m1,s=c(0.01,0.02))
```

---

colon

*Simplified gene expression data from Alon et al. (1999)*

---

**Description**

Gene expression data (20 genes for 62 samples) from the microarray experiments of colon tissue samples of Alon et al. (1999).

**Usage**

```
colon
```

**Format**

An object of class `list` of length 2.

**Details**

This data set contains 62 samples with 100 predictors (expanded from 20 genes using 5 basis B-splines, as described in Yang, Y. and Zou, H. (2015)): 40 tumor tissues, coded 1 and 22 normal tissues, coded -1.

**Value**

A list with the following elements:

- x a [62 x 100] matrix (expanded from a [62 x 20] matrix) giving the expression levels of 20 genes for the 62 colon tissue samples. Each row corresponds to a patient, each 5 consecutive columns to a grouped gene.
- y a numeric vector of length 62 giving the type of tissue sample (tumor or normal).

**Source**

The data are described in Alon et al. (1999) and can be freely downloaded from <http://microarray.princeton.edu/oncology/affydata/index.html>.

**References**

Alon, U. and Barkai, N. and Notterman, D.A. and Gish, K. and Ybarra, S. and Mack, D. and Levine, A.J. (1999). "Broad patterns of gene expression revealed by clustering analysis of tumor and normal colon tissues probed by oligonucleotide arrays", *Proc. Natl. Acad. Sci. USA*, **96**(12), 6745–6750.

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
BugReport: <https://github.com/emeryyi/glasso>

**Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# how many samples and how many predictors ?
dim(colon$x)

# how many samples of class -1 and 1 respectively ?
sum(colon$y==-1)
sum(colon$y==1)
```

---

cv.glasso

*Cross-validation for gglasso*

---

**Description**

Does k-fold cross-validation for gglasso, produces a plot, and returns a value for lambda. This function is modified based on the cv function from the glmnet package.

**Usage**

```

cv.gglasso(
  x,
  y,
  group,
  lambda = NULL,
  pred.loss = c("misclass", "loss", "L1", "L2"),
  nfolds = 5,
  foldid,
  delta,
  ...
)

```

**Arguments**

x	matrix of predictors, of dimension $n \times p$ ; each row is an observation vector.
y	response variable. This argument should be quantitative for regression (least squares), and a two-level factor for classification (logistic model, huberized SVM, squared SVM).
group	a vector of consecutive integers describing the grouping of the coefficients (see example below).
lambda	optional user-supplied lambda sequence; default is NULL, and <code>gglasso</code> chooses its own sequence.
pred.loss	loss to use for cross-validation error. Valid options are: <ul style="list-style-type: none"> <li>• "loss" for classification, margin based loss function.</li> <li>• "misclass" for classification, it gives misclassification error.</li> <li>• "L1" for regression, mean square error used by least squares regression loss="ls", it measure the deviation from the fitted mean to the response.</li> <li>• "L2" for regression, mean absolute error used by least squares regression loss="ls", it measure the deviation from the fitted mean to the response.</li> </ul> Default is "loss".
nfolds	number of folds - default is 5. Although nfolds can be as large as the sample size (leave-one-out CV), it is not recommended for large datasets. Smallest value allowable is nfolds=3.
foldid	an optional vector of values between 1 and nfold identifying what fold each observation is in. If supplied, nfold can be missing.
delta	parameter $\delta$ only used in huberized SVM for computing log-likelihood on validation set, only available with pred.loss = "loss", loss = "hsvm".
...	other arguments that can be passed to <code>gglasso</code> .

**Details**

The function runs `gglasso` `nfolds+1` times; the first to get the lambda sequence, and then the remainder to compute the fit with each of the folds omitted. The average error and standard deviation over the folds are computed.



**Value**

an object of class `cv.glasso` is returned, which is a list with the ingredients of the cross-validation fit.

<code>lambda</code>	the values of <code>lambda</code> used in the fits.
<code>cvm</code>	the mean cross-validated error - a vector of length <code>length(lambda)</code> .
<code>cvsd</code>	estimate of standard error of <code>cvm</code> .
<code>cvupper</code>	upper curve = <code>cvm+cvsd</code> .
<code>cvlower</code>	lower curve = <code>cvm-cvsd</code> .
<code>name</code>	a text string indicating type of measure (for plotting purposes).
<code>gglasso.fit</code>	a fitted <code>gglasso</code> object for the full data.
<code>lambda.min</code>	The optimal value of <code>lambda</code> that gives minimum cross validation error <code>cvm</code> .
<code>lambda.1se</code>	The largest value of <code>lambda</code> such that error is within 1 standard error of the minimum.

**Author(s)**

Yi Yang and Hui Zou  
 Maintainer: Yi Yang <yi.yang6@mcgill.ca>

**References**

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
 BugReport: <https://github.com/emeryyi/gglasso>

**See Also**

[gglasso](#), [plot.cv.gglasso](#), [predict.cv.gglasso](#), and [coef.cv.gglasso](#) methods.

**Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(bardet)

# define group index
group <- rep(1:20,each=5)

# 5-fold cross validation using group lasso
# penalized logistic regression
cv <- cv.gglasso(x=bardet$x, y=bardet$y, group=group, loss="ls",
  pred.loss="L2", lambda.factor=0.05, nfolds=5)
```

---

gglasso	<i>Fits the regularization paths for group-lasso penalized learning problems</i>
---------	--

---

## Description

Fits regularization paths for group-lasso penalized learning problems at a sequence of regularization parameters lambda.

## Usage

```
gglasso(
  x,
  y,
  group = NULL,
  loss = c("ls", "logit", "sqsvm", "hsvm", "wls"),
  nlambda = 100,
  lambda.factor = ifelse(nobs < nvars, 0.05, 0.001),
  lambda = NULL,
  pf = sqrt(bs),
  weight = NULL,
  dfmax = as.integer(max(group)) + 1,
  pmax = min(dfmax * 1.2, as.integer(max(group))),
  eps = 1e-08,
  maxit = 3e+08,
  delta,
  intercept = TRUE
)
```

## Arguments

x	matrix of predictors, of dimension $n \times p$ ; each row is an observation vector.
y	response variable. This argument should be quantitative for regression (least squares), and a two-level factor for classification (logistic model, huberized SVM, squared SVM).
group	a vector of consecutive integers describing the grouping of the coefficients (see example below).
loss	a character string specifying the loss function to use, valid options are: <ul style="list-style-type: none"> <li>• "ls" least squares loss (regression),</li> <li>• "logit" logistic loss (classification).</li> <li>• "hsvm" Huberized squared hinge loss (classification),</li> <li>• "sqsvm" Squared hinge loss (classification),</li> </ul> Default is "ls".
nlambda	the number of lambda values - default is 100.

<code>lambda.factor</code>	the factor for getting the minimal lambda in lambda sequence, where $\min(\text{lambda}) = \text{lambda.factor} * \max(\text{lambda})$ . $\max(\text{lambda})$ is the smallest value of lambda for which all coefficients are zero. The default depends on the relationship between $n$ (the number of rows in the matrix of predictors) and $p$ (the number of predictors). If $n \geq p$ , the default is $0.001$ , close to zero. If $n < p$ , the default is $0.05$ . A very small value of <code>lambda.factor</code> will lead to a saturated fit. It takes no effect if there is user-defined lambda sequence.
<code>lambda</code>	a user supplied lambda sequence. Typically, by leaving this option unspecified users can have the program compute its own lambda sequence based on <code>nlambda</code> and <code>lambda.factor</code> . Supplying a value of <code>lambda</code> overrides this. It is better to supply a decreasing sequence of lambda values than a single (small) value, if not, the program will sort user-defined lambda sequence in decreasing order automatically.
<code>pf</code>	penalty factor, a vector in length of <code>bn</code> ( <code>bn</code> is the total number of groups). Separate penalty weights can be applied to each group of $\beta$ s to allow differential shrinkage. Can be 0 for some groups, which implies no shrinkage, and results in that group always being included in the model. Default value for each entry is the square-root of the corresponding size of each group.
<code>weight</code>	a $n \times n$ observation weight matrix in the where $n$ is the number of observations. Only used if <code>loss='wls'</code> is specified. Note that cross-validation is NOT IMPLEMENTED for <code>loss='wls'</code> .
<code>dfmax</code>	limit the maximum number of groups in the model. Useful for very large <code>bs</code> (group size), if a partial path is desired. Default is <code>bs+1</code> .
<code>pmax</code>	limit the maximum number of groups ever to be nonzero. For example once a group enters the model, no matter how many times it exits or re-enters model through the path, it will be counted only once. Default is $\min(\text{dfmax} * 1.2, \text{bs})$ .
<code>eps</code>	convergence termination tolerance. Defaults value is $1e-8$ .
<code>maxit</code>	maximum number of outer-loop iterations allowed at fixed lambda value. Default is $3e8$ . If models do not converge, consider increasing <code>maxit</code> .
<code>delta</code>	the parameter $\delta$ in "hsvm" (Huberized squared hinge loss). Default is 1.
<code>intercept</code>	Whether to include intercept in the model. Default is TRUE.

## Details

Note that the objective function for "ls" least squares is

$$RSS/(2 * n) + \text{lambda} * \text{penalty};$$

for "hsvm" Huberized squared hinge loss, "sqsvm" Squared hinge loss and "logit" logistic regression, the objective function is

$$-\text{loglik}/n + \text{lambda} * \text{penalty}.$$

Users can also tweak the penalty by choosing different penalty factor.

For computing speed reason, if models are not converging or running slow, consider increasing `eps`, decreasing `nlambda`, or increasing `lambda.factor` before increasing `maxit`.

**Value**

An object with S3 class `gglasso`.

<code>call</code>	the call that produced this object
<code>b0</code>	intercept sequence of length <code>length(lambda)</code>
<code>beta</code>	a <code>p*length(lambda)</code> matrix of coefficients.
<code>df</code>	the number of nonzero groups for each value of <code>lambda</code> .
<code>dim</code>	dimension of coefficient matrix ( <code>ices</code> )
<code>lambda</code>	the actual sequence of <code>lambda</code> values used
<code>npasses</code>	total number of iterations (the most inner loop) summed over all <code>lambda</code> values
<code>jerr</code>	error flag, for warnings and errors, 0 if no error.
<code>group</code>	a vector of consecutive integers describing the grouping of the coefficients.

**Author(s)**

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

**References**

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
BugReport: <https://github.com/emeryyi/gglasso>

**See Also**

`plot.gglasso`

**Examples**

```
# load gglasso library
library(gglasso)

# load bardet data set
data(bardet)

# define group index
group1 <- rep(1:20,each=5)

# fit group lasso penalized least squares
m1 <- gglasso(x=bardet$x,y=bardet$y,group=group1,loss="ls")

# load colon data set
data(colon)

# define group index
```

```
group2 <- rep(1:20,each=5)

# fit group lasso penalized logistic regression
m2 <- gglasso(x=colon$x,y=colon$y,group=group2,loss="logit")
```

---

plot.cv.gglasso      *plot the cross-validation curve produced by cv.gglasso*

---

## Description

Plots the cross-validation curve, and upper and lower standard deviation curves, as a function of the lambda values used. This function is modified based on the plot.cv function from the glmnet package.

## Usage

```
## S3 method for class 'cv.gglasso'
plot(x, sign.lambda = 1, ...)
```

## Arguments

x	fitted <a href="#">cv.gglasso</a> object
sign.lambda	either plot against log(lambda) (default) or its negative if sign.lambda=-1.
...	other graphical parameters to plot

## Details

A plot is produced.

## Author(s)

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

## References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.

BugReport: <https://github.com/emeryyi/gglasso>

Friedman, J., Hastie, T., and Tibshirani, R. (2010), "Regularization paths for generalized linear models via coordinate descent," *Journal of Statistical Software*, 33, 1.

<http://www.jstatsoft.org/v33/i01/>

## See Also

[cv.gglasso](#).

**Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)

# 5-fold cross validation using group lasso
# penalized logistic regression
cv <- cv.gglasso(x=colon$x, y=colon$y, group=group, loss="logit",
pred.loss="misclass", lambda.factor=0.05, nfolds=5)

# make a CV plot
plot(cv)
```

---

plot.gglasso

*Plot solution paths from a "gglasso" object*


---

**Description**

Produces a coefficient profile plot of the coefficient paths for a fitted [gglasso](#) object.

**Usage**

```
## S3 method for class 'gglasso'
plot(x, group = FALSE, log.l = TRUE, ...)
```

**Arguments**

x	fitted <a href="#">gglasso</a> model
group	what is on the Y-axis. Plot the norm of each group if TRUE. Plot each coefficient if FALSE.
log.l	what is on the X-axis. Plot against the log-lambda sequence if TRUE. Plot against the lambda sequence if FALSE.
...	other graphical parameters to plot

**Details**

A coefficient profile plot is produced.

**Author(s)**

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

## References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
 BugReport: <https://github.com/emeryyi/gglasso>

## Examples

```
# load gglasso library
library(gglasso)

# load data set
data(bardet)

# define group index
group <- rep(1:20,each=5)

# fit group lasso
m1 <- gglasso(x=bardet$x,y=bardet$y,group=group,loss="ls")

# make plots
par(mfrow=c(1,3))
plot(m1) # plots the coefficients against the log-lambda sequence
plot(m1,group=TRUE) # plots group norm against the log-lambda sequence
plot(m1,log.l=FALSE) # plots against the lambda sequence
```

---

predict.cv.gglasso      *make predictions from a "cv.gglasso" object.*

---

## Description

This function makes predictions from a cross-validated gglasso model, using the stored "gglasso.fit" object, and the optimal value chosen for lambda.

## Usage

```
## S3 method for class 'cv.gglasso'
predict(object, newx, s = c("lambda.1se", "lambda.min"), ...)
```

## Arguments

object	fitted <a href="#">cv.gglasso</a> object.
newx	matrix of new values for x at which predictions are to be made. Must be a matrix. See documentation for <code>predict.gglasso</code> .
s	value(s) of the penalty parameter lambda at which predictions are required. Default is the value <code>s="lambda.1se"</code> stored on the CV object. Alternatively <code>s="lambda.min"</code> can be used. If s is numeric, it is taken as the value(s) of lambda to be used.

... not used. Other arguments to predict.

### Details

This function makes it easier to use the results of cross-validation to make a prediction.

### Value

The returned object depends on the ... argument which is passed on to the `predict` method for `gglasso` objects.

### Author(s)

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

### References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
BugReport: <https://github.com/emeryyi/gglasso>

### See Also

`cv.gglasso`, and `coef.cv.gglasso` methods.

### Examples

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)

# 5-fold cross validation using group lasso
# penalized logisitic regression
cv <- cv.gglasso(x=colon$x, y=colon$y, group=group, loss="logit",
pred.loss="misclass", lambda.factor=0.05, nfolds=5)

# the coefficients at lambda = lambda.min, newx = x[1,]
pre = predict(cv$gglasso.fit, newx = colon$x[1:10,],
s = cv$lambda.min, type = "class")
```



---

predict.gglasso      *make predictions from a "gglasso" object.*

---

### Description

Similar to other predict methods, this functions predicts fitted values and class labels from a fitted [gglasso](#) object.

### Usage

```
## S3 method for class 'gglasso'  
predict(object, newx, s = NULL, type = c("class", "link"), ...)
```

### Arguments

object	fitted <a href="#">gglasso</a> model object.
newx	matrix of new values for x at which predictions are to be made. Must be a matrix.
s	value(s) of the penalty parameter lambda at which predictions are required. Default is the entire sequence used to create the model.
type	type of prediction required: <ul style="list-style-type: none"><li>• Type "link", for regression it returns the fitted response; for classification it gives the linear predictors.</li><li>• Type "class", only valid for classification, it produces the predicted class label corresponding to the maximum probability.</li></ul>
...	Not used. Other arguments to predict.

### Details

s is the new vector at which predictions are requested. If s is not in the lambda sequence used for fitting the model, the predict function will use linear interpolation to make predictions. The new values are interpolated using a fraction of predicted values from both left and right lambda indices.

### Value

The object returned depends on type.

### Author(s)

Yi Yang and Hui Zou  
Maintainer: Yi Yang <yi.yang6@mcgill.ca>

### References

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
BugReport: <https://github.com/emeryyi/gglasso>

**See Also**

[coef](#) method

**Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)

# fit group lasso
m1 <- gglasso(x=colon$x,y=colon$y,group=group,loss="logit")

# predicted class label at x[10,]
print(predict(m1,type="class",newx=colon$x[10,]))

# predicted linear predictors at x[1:5,]
print(predict(m1,type="link",newx=colon$x[1:5,]))
```

---

```
print.gglasso      print a gglasso object
```

---

**Description**

Print the nonzero group counts at each lambda along the gglasso path.

**Usage**

```
## S3 method for class 'gglasso'
print(x, digits = max(3, getOption("digits") - 3), ...)
```

**Arguments**

x	fitted <a href="#">gglasso</a> object
digits	significant digits in printout
...	additional print arguments

**Details**

Print the information about the nonzero group counts at each lambda step in the [gglasso](#) object. The result is a two-column matrix with columns Df and Lambda. The Df column is the number of the groups that have nonzero within-group coefficients, the Lambda column is the the corresponding lambda.

### **Value**

a two-column matrix, the first columns is the number of nonzero group counts and the second column is Lambda.

### **Author(s)**

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### **References**

Yang, Y. and Zou, H. (2015), "A Fast Unified Algorithm for Computing Group-Lasso Penalized Learning Problems," *Statistics and Computing*. 25(6), 1129-1141.  
BugReport: <https://github.com/emeryyi/gglasso>

### **Examples**

```
# load gglasso library
library(gglasso)

# load data set
data(colon)

# define group index
group <- rep(1:20,each=5)

# fit group lasso
m1 <- gglasso(x=colon$x,y=colon$y,group=group,loss="logit")

# print out results
print(m1)
```

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