

# Package ‘SGL’

September 27, 2019

**Type** Package

**Title** Fit a GLM (or Cox Model) with a Combination of Lasso and Group Lasso Regularization

**Version** 1.3

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**Description** Fit a regularized generalized linear model via penalized maximum likelihood. The model is fit for a path of values of the penalty parameter. Fits linear, logistic and Cox models.

**License** GPL

**LazyLoad** yes

**Repository** CRAN

**NeedsCompilation** yes

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SGL-package	<i>Fit a GLM (or Cox Model) with a Combination of Lasso and Group Lasso Regularization</i>
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### Description

Fit a regularized generalized linear model via penalized maximum likelihood. The model is fit for a path of values of the penalty parameter. Fits linear, logistic and Cox models.

### Details

Package:	SGL
Type:	Package
Version:	1.0
Date:	2012-3-12
License:	GPL
LazyLoad:	yes

Only 4 functions: SGL cvSGL predictSGL plot.cvSGL

### Author(s)

Noah Simon, Jerome Friedman, Trevor Hastie, and Rob Tibshirani

Maintainer: Noah Simon <nrsimon@uw.edu>

### References

Simon, N., Friedman, J., Hastie T., and Tibshirani, R. (2011) *A Sparse-Group Lasso*,  
<http://faculty.washington.edu/nrsimon/SGLpaper.pdf>

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cvSGL	<i>Fit and Cross-Validate a GLM with a Combination of Lasso and Group Lasso Regularization</i>
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### Description

Fits and cross-validates a regularized generalized linear model via penalized maximum likelihood. The model is fit for a path of values of the penalty parameter, and a parameter value is chosen by cross-validation. Fits linear, logistic and Cox models.

**Usage**

```
cvSGL(data, index = rep(1, ncol(data$x)), type = "linear", maxit = 1000, thresh = 0.001,
min.frac = 0.05, nlam = 20, gamma = 0.8, nfold = 10, standardize = TRUE,
verbose = FALSE, step = 1, reset = 10, alpha = 0.95, lambdas = NULL,
foldid = NULL)
```

**Arguments**

data	For type="linear" should be a list with $x$ an input matrix of dimension $n$ -obs by $p$ -vars, and $y$ a length $n$ response vector. For type="logit" should be a list with $x$ , an input matrix, as before, and $y$ a length $n$ binary response vector. For type="cox" should be a list with $x$ as before, time, an $n$ -vector corresponding to failure/censor times, and status, an $n$ -vector indicating failure (1) or censoring (0).
index	A $p$ -vector indicating group membership of each covariate
type	model type: one of ("linear", "logit", "cox")
maxit	Maximum number of iterations to convergence
thresh	Convergence threshold for change in beta
min.frac	The minimum value of the penalty parameter, as a fraction of the maximum value
nlam	Number of lambda to use in the regularization path
gamma	Fitting parameter used for tuning backtracking (between 0 and 1)
nfold	Number of folds of the cross-validation loop
standardize	Logical flag for variable standardization (scaling) prior to fitting the model.
verbose	Logical flag for whether or not step number will be output
step	Fitting parameter used for initial backtracking step size (between 0 and 1)
reset	Fitting parameter used for taking advantage of local strong convexity in nesterov momentum (number of iterations before momentum term is reset)
alpha	The mixing parameter. $\alpha = 1$ is the lasso penalty.
lambdas	A user inputted sequence of lambda values for fitting. We recommend leaving this NULL and letting SGL self-select values
foldid	An optional user-pecified vector indicating the cross-validation fold in which each observation should be included. Values in this vector should range from 1 to $nfold$ . If left unspecified, SGL will randomly assign observations to folds

**Details**

The function runs SGL  $nfold+1$  times; the initial run is to find the lambda sequence, subsequent runs are used to compute the cross-validated error rate and its standard deviation.

**Value**

An object with S3 class "cv.SGL"

l1diff	An n1am vector of cross validated negative log likelihoods (squared error loss in the linear case, along the regularization path)
l1SD	An n1ame vector of approximate standard deviations of l1diff
lambdas	The actual list of lambda values used in the regularization path.
type	Response type (linear/logic/cox)
fit	A model fit object created by a call to SGL on the entire dataset
foldid	A vector indicating the cross-validation folds that each observation is assigned to
prevals	A matrix of prevalidated predictions for each observation, for each lambda-value

**Author(s)**

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**References**

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<http://faculty.washington.edu/nrsimon/SGLpaper.pdf>

**See Also**

SGL

**Examples**

```
set.seed(1)
n = 50; p = 100; size.groups = 10
index <- ceiling(1:p / size.groups)
X = matrix(rnorm(n * p), ncol = p, nrow = n)
beta = (-2:2)
y = X[,1:5] %*% beta + 0.1*rnorm(n)
data = list(x = X, y = y)
cvFit = cvSGL(data, index, type = "linear")
```

---

plot.cv.SGL

*plots the cross-validated error curve produced by cv.SGL*

---

**Description**

Plots the cross-validated error curve, and confidence bounds for each lambda in our regularization path.

**Usage**

```
## S3 method for class 'cv.SGL'  
plot(x, ...)
```

**Arguments**

```
x          fitted "cv.SGL" object  
...        additional arguments to be passed to plot
```

**Details**

A cross validated deviance plot is produced. More regularized models are to the right (less regularized to the left)

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<http://faculty.washington.edu/nrsimon/SGLpaper.pdf>

**See Also**

SGL and cv.SGL.

**Examples**

```
n = 50; p = 100; size.groups = 10  
index <- ceiling(1:p / size.groups)  
X = matrix(rnorm(n * p), ncol = p, nrow = n)  
beta = (-2:2)  
y = X[,1:5] %*% beta + 0.1*rnorm(n)  
data = list(x = X, y = y)  
cvFit = cvSGL(data, index, type = "linear")  
plot(cvFit)
```

---

predictSGL

*Outputs Predicted Responses from an SGL Model for New Observations*

---

**Description**

Outputs predicted response values for new user input observations at a specified lambda value

**Usage**

```
predictSGL(x, newX, lam)
```

**Arguments**

x	fitted "SGL" object
newX	covariate matrix for new observations whose responses we wish to predict
lam	the index of the lambda value for the model with which we desire to predict

**Details**

Predicted outcomes are given

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**References**

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<http://faculty.washington.edu/nrsimon/SGLpaper.pdf>

**See Also**

SGL and cvSGL.

**Examples**

```
n = 50; p = 100; size.groups = 10
index <- ceiling(1:p / size.groups)
X = matrix(rnorm(n * p), ncol = p, nrow = n)
beta = (-2:2)
y = X[,1:5] %*% beta + 0.1*rnorm(n)
data = list(x = X, y = y)
Fit = SGL(data, index, type = "linear")
X.new = matrix(rnorm(n * p), ncol = p, nrow = n)
predictSGL(Fit, X.new, 5)
```

---

print.SGL

*prints a summary of the SGL solution path*

---

**Description**

Prints a short summary of the SGL solution path.

**Usage**

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

**Arguments**

x	fitted "SGL" object
digits	significant digits in printout
...	additional print arguments

**Details**

The time of regression run, followed by a 2-column matrix with rows `lambdas` and `num.nonzero`. `lambdas` gives the lambda-value of each fit. `num.nonzero` gives the the number of non-zero coefficients.

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<http://faculty.washington.edu/nrsimon/SGLpaper.pdf>

**See Also**

SGL and `cv.SGL`.

**Examples**

```
n = 50; p = 100; size.groups = 10  
index <- ceiling(1:p / size.groups)  
X = matrix(rnorm(n * p), ncol = p, nrow = n)  
beta = (-2:2)  
y = X[,1:5] %*% beta + 0.1*rnorm(n)  
data = list(x = X, y = y)  
fit = SGL(data, index, type = "linear")  
print(fit)
```

---

SGL *Fit a GLM with a Combination of Lasso and Group Lasso Regularization*

---

### Description

Fit a regularized generalized linear model via penalized maximum likelihood. The model is fit for a path of values of the penalty parameter. Fits linear, logistic and Cox models.

### Usage

```
SGL(data, index, type = "linear", maxit = 1000, thresh = 0.001,
min.frac = 0.1, nlam = 20, gamma = 0.8, standardize = TRUE,
verbose = FALSE, step = 1, reset = 10, alpha = 0.95, lambdas = NULL)
```

### Arguments

data	For type="linear" should be a list with $x$ an input matrix of dimension $n$ -obs by $p$ -vars, and $y$ a length $n$ response vector. For type="logit" should be a list with $x$ , an input matrix, as before, and $y$ a length $n$ binary response vector. For type="cox" should be a list with $x$ as before, $time$ , an $n$ -vector corresponding to failure/censor times, and $status$ , an $n$ -vector indicating failure (1) or censoring (0).
index	A $p$ -vector indicating group membership of each covariate
type	model type: one of ("linear", "logit", "cox")
maxit	Maximum number of iterations to convergence
thresh	Convergence threshold for change in beta
min.frac	The minimum value of the penalty parameter, as a fraction of the maximum value
nlam	Number of lambda to use in the regularization path
gamma	Fitting parameter used for tuning backtracking (between 0 and 1)
standardize	Logical flag for variable standardization prior to fitting the model.
verbose	Logical flag for whether or not step number will be output
step	Fitting parameter used for initial backtracking step size (between 0 and 1)
reset	Fitting parameter used for taking advantage of local strong convexity in nesterov momentum (number of iterations before momentum term is reset)
alpha	The mixing parameter. $\alpha = 1$ is the lasso penalty. $\alpha = 0$ is the group lasso penalty.
lambdas	A user specified sequence of lambda values for fitting. We recommend leaving this NULL and letting SGL self-select values

### Details

The sequence of models along the regularization path is fit by accelerated generalized gradient descent.



**Value**

An object with S3 class "SGL"

beta	A $p$ by $n_{lam}$ matrix, giving the penalized MLEs for the $n_{lam}$ different models, where the index corresponds to the penalty parameter $\lambda$
lambdas	The actual sequence of $\lambda$ values used (penalty parameter)
type	Response type (linear/logic/cox)
intercept	For some model types, an intercept is fit
X.transform	A list used in <code>predict</code> which gives the empirical mean and variance of the $x$ matrix used to build the model
lambdas	A user specified sequence of $\lambda$ values for fitting. We recommend leaving this NULL and letting SGL self-select values

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**See Also**

`cv.SGL`

**Examples**

```
n = 50; p = 100; size.groups = 10
index <- ceiling(1:p / size.groups)
X = matrix(rnorm(n * p), ncol = p, nrow = n)
beta = (-2:2)
y = X[,1:5] %*% beta + 0.1*rnorm(n)
data = list(x = X, y = y)
fit = SGL(data, index, type = "linear")
```

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