

# Package ‘LassoNet’

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**Type** Package

**Title** 3CoSE Algorithm

**Version** 0.8.3

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**Author** Jonas Striaukas [aut, trl, cre] and Matthias Weber [aut]

**Maintainer** Jonas Striaukas <jonas.striaukas@gmail.com>

**Description** Contains functions to estimate a penalized regression model using 3CoSE algorithm, see Weber, Striaukas, Schumacher Binder (2018) <[doi:10.2139/ssrn.3211163](https://doi.org/10.2139/ssrn.3211163)>.

**License** GPL (>= 2)

**Imports** Rcpp (>= 0.11.5)

**Suggests** snowfall

**LinkingTo** Rcpp

**NeedsCompilation** yes

**Repository** CRAN

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LassoNet-package

*LassoNet: package for 3CoSE algorithm.*

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### Description

LassoNet contains functions to estimate a penalized regression model using 3CoSE algorithm described in the paper Weber, Striaukas, Schumacher and Binder (2018). The main function of the package is the function `lasso.net.grid`, see the example below.

### Details

Package: LassoNet  
Type: Package  
Version: 0.8.3  
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License: Open source

### Author(s)

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

### References

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

### See Also

[Rcpp](#), [glmnet](#)

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beta.update.net

*Updates  $\beta$  coefficients.*

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### Description

This function updates  $\beta$  for given penalty parameters.

### Usage

```
beta.update.net(x,y,beta,lambda1,lambda2,M1,n.iter,iscpp,tol)
```

**Arguments**

x	input data matrix of size $n \times p$ ; n - number of observations; p - number of covariates
y	response vector or size $n \times 1$
beta	initial value for $\beta$ ; default - zero vector of size $n \times 1$
lambda1	lasso penalty parameter
lambda2	network penalty parameter
M1	penalty matrix
n.iter	maximum number of iterations for $\beta$ step; default - 1e5
iscpp	binary choice for using cpp function in coordinate updates; 1 - use C++ (default), 0 - use R
tol	convergence tolerance level; default - 1e-6

**Details**

Updates the coefficient vector  $\beta$  given the data and penalty parameters  $\lambda_1$  and  $\lambda_2$ . Convergence criterion is defined as  $\sum_{i=1}^p |\beta_{i,j} - \beta_{i,j-1}| \leq \text{to}$ .

**Value**

beta	updated $\beta$ vector
convergence	binary variable; 1 - yes
steps	number of steps until convergence

**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**References**

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

**Examples**

```
p<-200
n<-100
beta.0=array(1,c(p,1))
x<-matrix(rnorm(n*p),n,p)
y<-rnorm(n,mean=0,sd=1)
lambda1<-1
lambda2<-1
M1<-diag(p)
updates<-beta.update.net(x, y, beta.0, lambda1, lambda2, M1)
```

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betanew\_lasso\_cpp      *C++ subroutine that updates  $\beta$  coefficients.*

---

### Description

This function updates  $\beta$  for given penalty parameters.

### Usage

```
betanew_lasso_cpp(xx, xy, beta, M, y, Lambda1, Lambda2, iter, tol)
```

### Arguments

xx	Bx matrix
xy	By vector
beta	initial value for $\beta$ ; default - zero vector of size $p \times 1$
M	penalty matrix
y	response vector or size $n \times 1$
Lambda1	lasso penalty parameter
Lambda2	network penalty parameter
iter	maximum number of iterations for $\beta$ step
tol	convergence tolerance level

### Details

See [beta.update.net](http://beta.update.net)

### Value

beta	updated $\beta$ vector
steps	number of steps until convergence

### Author(s)

Maintainer: Jonas Striaukas <[jonas.striaukas@gmail.com](mailto:jonas.striaukas@gmail.com)>

### References

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <[doi:10.2139/ssrn.3211163](https://doi.org/10.2139/ssrn.3211163)>

**Examples**

```
p<-200
n<-100
beta.0=array(1,c(p,1))
x<-matrix(rnorm(n*p),n,p)
y<-rnorm(n,mean=0,sd=1)
lambda1<-1
lambda2<-1
M1<-diag(p)
updates<-beta.update.net(x, y, beta.0, lambda1, lambda2, M1)
```

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fastols

*Fast least squares estimate.*

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**Description**

Computes least squares estimate in an efficient way.

**Usage**

```
fastols(y, x)
```

**Arguments**

y	dependent variable
x	response variable

**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**Examples**

```
p<-10
n<-100
x<-matrix(rnorm(n*p),n,p)
beta<-array(5, c(p,1))
y<-x%*%beta + rnorm(n,mean=0,sd=0.1)
fastols(y,x)
```

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get.BxBy                      *Computes decomposition elements.*

---

### Description

Computes matrices  $B_X^{ij}$  and  $B_Y^{ij}$  to speed up estimation of connection signs. These matrices are stored only for indices that have non zero entries in penalty matrix M.

### Usage

```
get.BxBy(x, y, M)
```

### Arguments

x	Input data matrix of size $n \times p$ , n - number of observations, p - number of covariates
y	y Response vector or size $n \times 1$
M	penalty matrix

### Details

Calculates matrices all for i and j indices that have non zero values in a given penalty matrix.

### Value

Bx	array of $B_X^{ij}$ stored matrices. $Bx[, k]$ are the k-th combination of i and j non zero entry in the penalty matrix M
By	array of $B_Y^{ij}$ stored matrices. $By[, k]$ are the k-th combination of i and j non zero entry in the penalty matrix M

### Author(s)

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

### References

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

### Examples

```
p<-200
n<-100
x<-matrix(rnorm(n*p),n,p)
y<-rnorm(n,mean=0,sd=1)
M<-diag(p)
get.BxBy(x, y, M)
```

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get.signs.M	<i>Vectorizes connection sign matrix.</i>
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**Description**

Stores a matrix of connection signs to a vector.

**Usage**

```
get.signs.M(MAT)
```

**Arguments**

MAT	matrix of connection signs that contains -1, 1 or 0
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**Value**

vec.out	vectorized MAT matrix
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**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

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get.xi	<i>Updates the estimates of the connection signs by running mini OLS models.</i>
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**Description**

Updates connection signs  $\hat{\xi}$ .

**Usage**

```
get.xi(Bx,By,beta,xi,M)
```

**Arguments**

Bx	Bx element
By	By element
beta	$\hat{\beta}$ estimated value
xi	$\hat{\xi}$ matrix estimated at the previous step
M	penalty matrix

**Value**

xi	$\hat{\xi}$ matrix
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**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**References**

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

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lasso.net.fixed      *Estimates coefficients over the grid values of penalty parameters.*

---

**Description**

See lasso.net.grid

**Usage**

```
lasso.net.fixed(x,y,beta.0,lambda1,lambda2,M1,n.iter,iscpp,tol)
```

**Arguments**

x	$n \times p$ input data matrix
y	response vector or size $n \times 1$
beta.0	initial value for $\beta$ ; default - zero vector of size $n \times 1$
lambda1	lasso penalty coefficient
lambda2	network penalty coefficient
M1	penalty matrix
n.iter	maximum number of iterations for $\beta$ updating; default - 1e5
iscpp	binary choice for using cpp function in coordinate updates; 1 - use C++ (default), 0 - use R.
tol	convergence in $\beta$ tolerance level; default - 1e-6

**Details**

Function loops through the grid of values of penalty parameters  $\lambda_1$  and  $\lambda_2$  until convergence is reached. Warm starts are stored for each iterator. The warm starts are stored once the coordinate updating converges.



**Value**

beta	Matrix of $\beta$ coefficients. Columns denote different $\lambda_1$ coefficients, rows - $\lambda_2$ coefficients
mse	Mean squared error value
iterations	matrix with stored number of steps for sign matrix to converge
update.steps	matrix with stored number of steps for $\beta$ updates to converge. (only stores the last values from connection signs iterations)
convergence.in.grid	matrix with stored values for convergence in $\beta$ coefficients. If at least one $\beta$ did not converge in sign matrix iterations, 0 (false) is stored, otherwise 1 (true)

**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**References**

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

**Examples**

```
p=200
n=100
beta.0=array(1,c(p,1))
x=matrix(rnorm(n*p),n,p)
y=rnorm(n,mean=0,sd=1)
lambda1=c(0,1)
lambda2=c(0,1)
M1=diag(p)
lasso.net.fixed(x, y, beta.0, lambda1, lambda2, M1)
```

---

lasso.net.grid	<i>Estimates coefficients and connection signs over the grid of values of penalty parameters <math>\lambda_1</math> and <math>\lambda_2</math>.</i>
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**Description**

Fits network regressions over the grid of values of penalty parameters  $\lambda_1$  and  $\lambda_2$ , stores connection signs, number of iterations until convergence and convergence outcome.

**Usage**

```
lasso.net.grid(x,y ,beta.0,lambda1,lambda2,M1,m.iter,n.iter,iscpp=TRUE,tol,alt.num)
```

**Arguments**

<code>x</code>	$n \times p$ input data matrix
<code>y</code>	response vector or size $n \times 1$
<code>beta.0</code>	initial value for $\beta$ . default - zero vector of size $n \times 1$
<code>lambda1</code>	lasso penalty coefficient
<code>lambda2</code>	network penalty coefficient
<code>M1</code>	penalty matrix
<code>m.iter</code>	maximum number of iterations for sign matrix updating; default - 100
<code>n.iter</code>	maximum number of iterations for $\beta$ updating; default - 1e5
<code>iscpp</code>	binary choice for using cpp function in coordinate updates; 1 - use C++ (default), 0 - use R
<code>tol</code>	convergence in $\beta$ tolerance level; default - 1e-6
<code>alt.num</code>	alt.num remaining iterations are stored; default - 12

**Details**

Fits network regression for the grid values of  $\lambda_1$  and  $\lambda_2$  using warm starts.

**Value**

<code>beta</code>	matrix of $\beta$ coefficients, columns are for different $\lambda_1$ parameters, rows $\lambda_2$ parameters
<code>mse</code>	mean squared error value
<code>M</code>	array of connection signs. $M[, i, j]$ is the connection sign matrix for j-th $\lambda_1$ value and i-th $\lambda_2$ value
<code>iterations</code>	matrix with stored number of steps for sign matrix to converge
<code>update.steps</code>	matrix with stored number of steps for $\beta$ updates to converge. (only stores the last values from connection signs iterations)
<code>convergence.in.M</code>	matrix with stored values for convergence in sign matrix
<code>convergence.in.grid</code>	matrix with stored values for convergence in $\beta$ coefficients. If at least one $\beta$ did not converge in sign matrix iterations, 0 (false) is stored, otherwise 1 (true)
<code>xi.conv</code>	array with stored connection signs changes in each iteration
<code>beta.alt</code>	array of coefficient vectors in case connection signs alternate

**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**References**

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

**Examples**

```
p=200
n=100
beta.0=array(1,c(p,1))
x=matrix(rnorm(n*p),n,p)
y=rnorm(n,mean=0,sd=1)
lambda1=c(0,1)
lambda2=c(0,1)
M1=diag(p)
lasso.net.grid(x, y, beta.0, lambda1, lambda2, M1)
```

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mat.to.laplacian	<i>Computes Laplacian matrix.</i>
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---

**Description**

Computes Laplacian matrix.

**Usage**

```
mat.to.laplacian(M1,type)
```

**Arguments**

M1	$p \times p$ matrix
type	Laplacian types: 1) "normalized" (default) - normalized Laplacian, 2) "combinatorial" - combinatorial Laplacian

**Value**

L	Laplacian
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**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

---

matrix.M.update	<i>Updates connection sign matrix.</i>
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**Description**

Updates M using relation  $(M)_{ij} = -\hat{\xi}_{ij}|(M_1)_{ij}$ .

**Usage**

```
matrix.M.update(M, xi)
```

**Arguments**

M	penalty matrix
xi	estimated $\hat{\xi}_{ij}$ matrix

**Details**

Updates M

**Value**

M	updated M
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**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**References**

Weber, M., Striaukas, J., Schumacher, M., Binder, H. "Network-Constrained Covariate Coefficient and Connection Sign Estimation" (2018) <doi:10.2139/ssrn.3211163>

**Examples**

```
p<-100
M<-diag(p)
xi<-matrix(rnorm(p*p), p, p)
matrix.M.update(M,xi)
```

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soft.thresh	<i>Soft thresholding operator.</i>
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**Description**

Soft thresholding operator.

**Usage**

```
soft.thresh(x, kappa)
```

**Arguments**

x	$\beta$ coordinate
kappa	$\kappa$ value in general or $\lambda_1$ for covariance updating

**Details**

Soft thresholding definition:  $S(x, \kappa) = \text{sign}(x)(|x| - \kappa)_+$

**Value**

x	value after applying soft thresholding operator
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**Author(s)**

Maintainer: Jonas Striaukas <jonas.striaukas@gmail.com>

**Examples**

```
kappa<-0.2  
x<-0.7  
soft.thresh(x, kappa)
```

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