

# Package ‘ScreenClean’

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**Title** Screen and clean variable selection procedures

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**Description** Routines for a collection of screen-and-clean type variable selection procedures, including UPS and GS.

**Imports** MASS, Matrix, quadprog

**License** GPL (>= 2)

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ScreenClean-package    *Screen and clean variable selection procedures, including UPS and GS.*

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### Description

Routines for a collection of screen-and-clean type variable selection procedures.

### Details

Package: ScreenClean  
Type: Package  
Version: 1.0.1  
Date: 2012-10-30  
License: GPL (>= 2)

### Note

In order to use ScreenClean, the data need to be normalized, to make the standard deviation of the noise to be 1, and the  $l_2$  norm of each length  $n$  predictor vector to be 1.

### Author(s)

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### References

Ji, P. and Jin, J. (2012). *UPS delivers optimal phase diagram in high dimensional variable selection*. Ann. Statist., 40(1), 73-103.

Jin, J., Zhang, C.-H. and Zhang, Q. (2012). *Optimality of Graphlet Screening in High Dimensional Variable Selection*. arXiv:1204.6452

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CleaningStep	<i>GC-step of the graphlet screening</i>
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**Description**

CleaningStep performs the cleaning step of the graphlet screening

**Usage**

```
CleaningStep(survivor, y.tilde, gram, lambda, uu)
```

**Arguments**

survivor	the result of the screening step, a logical vector.
y.tilde	$X'y$ , where X and y are the predictor matrix and the reponse vector.
gram	the thresholded sparse gram matrix
lambda	the tuning parameters of the cleaning step, whose optimal choice is tied to the sparse level.
uu	the tuning parameter of the cleaning step; its optimal choice has the intuition of the detected minimal signal strength.

**Value**

beta.gs	the estimated regression coefficient of the graphlet screening, a numeric vector
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**See Also**

[IterGS,ScreeningStep](#)

**Examples**

```
##See the demoGs.r
```

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FindAllCG	<i>Find all the connected subgraphs whose size <math>\leq lc</math></i>
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**Description**

FindAllCG uses FindCG iteratively, and lists all the connected subgraphs with no more than lc nodes

**Usage**

```
FindAllCG(adjacency.matrix, lc)
```

**Arguments**

`adjacency.matrix`  
 p by p adjacency matrix of an undirected graph; it must be symmetric.

`lc`  
 the maximal size of the connected subgraphs to be listed

**Value**

`cg.all`  
 A list, whose kth component is a matrix with k columns that lists all the connected subgraphs with k nodes.

**See Also**

[FindCG](#)

**Examples**

```
require(MASS)
require(Matrix)
p <- 10
Omega <- sparseMatrix(c(1:(p-1),2:p),c(2:p,1:(p-1)),x=1)
cg.all <- FindAllCG(Omega,3)
```

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 FindCG

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*Find the connected subgraphs with a certain number of nodes*


---

**Description**

FindCG is used to find all the connected subgraphs with a certain number of nodes.

**Usage**

```
FindCG(adjacency.matrix, cg.initial)
```

**Arguments**

`adjacency.matrix`  
 p by p adjacency matrix of an undirected graph. It must be symmetric.

`cg.initial`  
 It could be 1:p or a matrix, whose elements are positive integers from 1 to p. If it is a length p vector, FindCG converts it into a matrix with one column. For a matrix with k columns, FindCG reads its rows as th indices of a collection of connected subgraphs with k nodes.

**Value**

`cg.new`  
 If the input is a matrix with k columns and stores the indices of all the size k connected subgraphs, the output is a matrix with k+1 columns storing the indices of all the connected subgraphs with k+1 nodes.

**See Also**[FindAllCG](#)**Examples**

```

require(MASS)
require(Matrix)
p <- 10
Omega <- sparseMatrix(c(1:(p-1),2:p),c(2:p,1:(p-1)),x=1)
cg.2 <- FindCG(Omega,c(1:p))
cg.3 <- FindCG(Omega,cg.2)

```

IterGS

*Iterative graphlet screening procedure***Description**

The iterative graphlet screening procedure, main function of the package.

**Usage**

```

IterGS(y.tilde, gram, gram.bias, cg.all, sp, tau, nm, q0=0.1, scale = 1, max.iter = 3,
std.thresh = 1.05, beta.initial = NULL)

```

**Arguments**

<code>y.tilde</code>	$X'y$ where $X$ and $y$ are the predictor matrix and the response vector, respectively.
<code>gram</code>	the thresholded gram matrix
<code>gram.bias</code>	the bias of the thresholded gram matrix
<code>cg.all</code>	all the connected cg.alls of gram with size no more than nm.
<code>sp</code>	the expected sparse level
<code>tau</code>	the minimal signal strength to be detected
<code>nm</code>	the maximal size of the connected subgraphs considered in the screening step.
<code>q0</code>	the minimal screening parameter
<code>scale</code>	optional numerical parameter of the screening step. The default is 1
<code>max.iter</code>	the maximal number of iterations. The default is 3.
<code>std.thresh</code>	the threshold of the std change that stop the loop. The default is 1.05.
<code>beta.initial</code>	the initial estimate of beta in reducing the bias. The default is $uu * \text{sign}(y.tilde) * (\text{abs}(y.tilde) > uu)$ .

**Value**

IterGS returns a list with two elements

estimate	The iterative GS estimate of beta
n.iter	The number of iterations it takes

**Examples**

```
##See demoIterGs.r
```

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 PMLE

*Penalized MLE procedure used in the cleaning step*

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**Description**

Penalized MLE procedure used in the cleaning step, an inner function.

**Usage**

```
PMLE(gram, y, lambda, uu)
```

**Arguments**

gram	the sub gram matrix of the small scale quadratic problem.
y	the sub-vector of $y.tilde$
lambda	the tuning parameter of the cleaning step, tied to the sparse level.
uu	the tuning parameters of the cleaning step. It has the intuitive interpretation of the minimal signal strength to be detected.

**Value**

b	the estimate of the subvector of beta
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**See Also**

[CleaningStep](#)

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ScreeningStep	<i>GS-step of the graphlet screening</i>
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**Description**

ScreeningStep performs the cleaning step of the graphlet screening

**Usage**

```
ScreeningStep(y.tilde, gram, cg.all, nm, v, r, q0 = 0.1, scale = 1)
```

**Arguments**

y.tilde	$X'y$ , where X and y are the predictor matrix and the response vector.
gram	the regularized gram matrix
cg.all	a list whose kth element is a matrix of k columns. Its rows contain all the connected subgraph with k nodes.
nm	the maximal subgraph investigated in the screening step
v	an essential tuning parameter of graphlet screening, tied to the sparse level
r	an essential tuning parameter of graphlet screening, tied to the signal strength
q0	the minimal screening parameter
scale	$q(D, F) = q^{max}(D, F) * scale$ , default is scale=1

**Value**

survivor	A logical vector, where TRUE means retained as a potential signal.
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**Note**

When nm=1, it is just univariate thresholding, and thus the screening step of UPS.

**See Also**

[CleaningStep](#), [IterGS](#)

**Examples**

```
##See the demoGS.r
```

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ThresholdGram                    *Thresholds the gram matrix*

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### Description

Thresholds the gram matrix

### Usage

```
ThresholdGram(gram.full, delta = 1/log(dim(gram.full)[1]))
```

### Arguments

gram.full                    the gram matrix before the elementwise thresholding, a p by p symmetric matrix  
 delta                        the threshold, the default is 1/log(p)

### Value

A list with two elements

gram.sd                    the threhsolded gram matrix, a sparse matrix  
 gram.bias                   the difference of the orginal matrix and the threholded matrix

### Examples

```
p <-10
off.diag<-matrix(runif(p^2),p,p)
omega <- (off.diag+t(off.diag))*0.3
diag(omega) <- 1
omega.omega<-ThresholdGram(omega,0.3)
omega.omega$gram
omega.omega$gram.bias
```

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VectorizeBase                    *expresses the number i on the base as a vector*

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### Description

expresses the number i on the base as a vector, an inner function.

### Usage

```
VectorizeBase(i, base, length)
```

**Arguments**

i	the non-negative number to be converted
base	the base to be converted on
length	the length of the converted vector

**Value**

vector	A vector with the given length, whose elements can be read as the number i with the given base.
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