

# Package ‘gclm’

October 13, 2022

**Type** Package

**Title** Graphical Continuous Lyapunov Models

**Version** 0.0.1

**Description** Estimation of covariance matrices as solutions of continuous time Lyapunov equations. Sparse coefficient matrix and diagonal noise are estimated with a proximal gradient method for an  $l_1$ -penalized loss minimization problem. Varando G, Hansen NR (2020) <[arXiv:2005.10483](https://arxiv.org/abs/2005.10483)>.

**License** MIT + file LICENSE

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.1.0

**URL** <https://github.com/gherardovarando/gclm>

**BugReports** <https://github.com/gherardovarando/gclm/issues>

**Suggests** testthat

**NeedsCompilation** yes

**Author** Gherardo Varando [aut, cre, cph]  
(<<https://orcid.org/0000-0002-6708-1103>>),  
Niels Richard Hansen [aut] (<<https://orcid.org/0000-0003-3883-365X>>)

**Maintainer** Gherardo Varando <[gherardo.varando@gmail.com](mailto:gherardo.varando@gmail.com)>

**Repository** CRAN

**Date/Publication** 2020-06-04 08:40:07 UTC

## R topics documented:

B0 . . . . .	2
clyap . . . . .	2
gclm . . . . .	3
gclm.lowertri . . . . .	5
<b>Index</b>	<b>6</b>

---

B0	<i>Generate a naive stable matrix</i>
----	---------------------------------------

---

**Description**

Generate a naive stable matrix

**Usage**

B0(p)

**Arguments**

p                      dimension of the matrix

**Value**

a stable matrix with off-diagonal entries equal to 1 and diagonal entries equal to -p

---

clyap	<i>Solve continuous-time Lyapunov equations</i>
-------	---

---

**Description**

clyap solve the continuous-time Lyapunov equations

$$BX + XB' + C = 0$$

Using the Bartels-Stewart algorithm with Hessenberg–Schur decomposition. Optionally the Hessenberg-Schur decomposition can be returned.

**Usage**

clyap(B, C, Q = NULL, all = FALSE)

**Arguments**

B	Square matrix
C	Square matrix
Q	Square matrix, the orthogonal matrix used to transform the original equation
all	logical

**Details**

If the matrix  $Q$  is set then the matrix  $B$  is assumed to be in upper quasi-triangular form (Hessenberg-Schur canonical form), as required by LAPACK subroutine DTRSYL and  $Q$  is the orthogonal matrix associated with the Hessenberg-Schur form of  $B$ . Usually the matrix  $Q$  and the appropriate form of  $B$  are obtained by a first call to `clyap(B, C, all = TRUE)`

`clyap` uses lapack subroutines:

- DGEES
- DTRSYL
- DGEMM

**Value**

The solution matrix  $X$  if `all = FALSE`. If `all = TRUE` a list with components  $X$ ,  $B$  and  $Q$ . Where  $B$  and  $Q$  are the Hessenberg-Schur form of the original matrix  $B$  and the orthogonal matrix that performed the transformation.

**Examples**

```
B <- matrix(data = rnorm(9), nrow = 3)
## make B negative diagonally dominant, thus stable:
diag(B) <- - 3 * max(B)
C <- diag(runif(3))
X <- clyap(B, C)
## check X is a solution:
max(abs(B %*% X + X %*% t(B) + C))
```

---

gclm

*l1 penalized loss estimation for GCLM*


---

**Description**

Estimates a sparse continuous time Lyapunov parametrization of a covariance matrix using a lasso (L1) penalty.

**Usage**

```
gclm(
  Sigma,
  B = -0.5 * diag(ncol(Sigma)),
  C = rep(1, ncol(Sigma)),
  C0 = rep(1, ncol(Sigma)),
  loss = "loglik",
  eps = 0.01,
  alpha = 0.5,
  maxIter = 100,
  lambda = 0,
```

```

    lambdac = 0,
    job = 0
)

gclm.path(
  Sigma,
  lambdas = NULL,
  B = -0.5 * diag(ncol(Sigma)),
  C = rep(1, ncol(Sigma)),
  ...
)

```

### Arguments

Sigma	covariance matrix
B	initial B matrix
C	diagonal of initial C matrix
C0	diagonal of penalization matrix
loss	one of "loglik" (default) or "frobenius"
eps	convergence threshold
alpha	parameter line search
maxIter	maximum number of iterations
lambda	penalization coefficient for B
lambdac	penalization coefficient for C
job	integer 0,1,10 or 11
lambdas	sequence of lambda
...	additional arguments passed to gclm

### Details

gclm performs proximal gradient descent for the optimization problem

$$\operatorname{argmin} L(\Sigma(B, C)) + \lambda \rho(B) + \lambda_C \|C - C_0\|_F^2$$

subject to  $B$  stable and  $C$  diagonal, where  $\rho(B)$  is the l1 norm of the off-diagonal element of  $B$ .

gclm.path simply calls iteratively gclm with different lambda values. Warm start is used, that is in the  $i$ -th call to gclm the B and C matrices are initialized as the one obtained in the  $(i-1)$ th call.

### Value

for gclm: a list with the result of the optimization

for gclm.path: a list of the same length of lambdas with the results of the optimization for the different lambda values

**Examples**

```
x <- matrix(rnorm(50*20),ncol=20)
S <- cov(x)

## l1 penalized log-likelihood
res <- gclm(S, eps = 0, lambda = 0.1, lambdac = 0.01)

## l1 penalized log-likelihood with fixed C
res <- gclm(S, eps = 0, lambda = 0.1, lambdac = -1)

## l1 penalized frobenius loss
res <- gclm(S, eps = 0, lambda = 0.1, loss = "frobenius")
```

---

gclm.lowertri	<i>Recover lower triangular GCLM</i>
---------------	--------------------------------------

---

**Description**

Recover the only lower triangular stable matrix  $B$  such that Sigma ( $\Sigma$ ) is the solution of the associated continuous Lyapunov equation:

$$B\Sigma + \Sigma B' + C = 0$$

**Usage**

```
gclm.lowertri(Sigma, P = solve(Sigma), C = diag(nrow = nrow(Sigma)))
```

**Arguments**

Sigma	covariance matrix
P	the inverse of the covariance matrix
C	symmetric positive definite matrix

**Value**

A stable lower triangular matrix

# Index

B0, [2](#)

clyap, [2](#)

gclm, [3](#)

gclm.lowertri, [5](#)