## Package 'LqG'

April 27, 2022

Type Package

Title Robust Group Variable Screening Based on Maximum Lq-Likelihood Estimation

Version 0.1.0

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**Description** Produces a group screening procedure that is based on maximum Lq-likelihood estimation, to simultaneously account for the group structure and data contamination in variable screening. The methods are described in Li, Y., Li, R., Qin, Y., Lin, C., & Yang, Y. (2021) Robust Group Variable Screening Based on Maximum Lq-likelihood Estimation. Statistics in Medicine, 40:6818-6834.

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**Encoding** UTF-8

LazyData true

Imports base

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**Depends** R (>= 3.5.0)

NeedsCompilation no

**Repository** CRAN

Date/Publication 2022-04-27 08:20:09 UTC

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grsc.marg.MLqE

#### Description

Group screening by ranking utility of each group. The group effect is defined based on the cumulation of the maximum Lq-likelihood estimate of the regression using only one predictor each time within the group.

#### Usage

```
grsc.marg.MLqE(
X,
Y,
n = dim(X)[1],
p = dim(X)[2],
q = 0.9,
m,
group,
eps = 1e-06,
d = n/log(n)
)
```

#### Arguments

Х	A matrix of predictors.
Y	A vector of response.
n	A value of sample size
р	A value denoting the dimension of predictors
q	A value of distortion parameter of Lq function, default to 0.9.
m	A number of the predictor groups
group	A vector of consecutive integers describing the grouping of the coefficients (see example below).
eps	The iteration coverage criterion, default to 1e-06.
d	A value of the number of groups retained after screening, default to $n/log(n)$ .

#### Details

grsc.marg.MLqE obtains the group effect of each group for subsequential group screening, based on the cumulative marginal MLqE coefficients within the group. It can work when both the correlation within groups and between groups are small. If group size equals to 1, individual screening is conducted.

#### grsc.MLqE

#### Value

The grsc.marg.MLqE returns a list containing the following components:

beta.group	The vector of utility of each group, which is the criterion for the variable screen-
	ing procedure.

group.screened The vector of integers denoting the screened groups.

#### Examples

```
# This is an example of grsc.marg.MLqE with simulated data
data(LqG_SimuData)
X = LqG_SimuData$X
Y = LqG_SimuData$Y
n = dim(X)[1]
p = dim(X)[2]
m = 200
groups = rep(1:(p/5), each = 5)
result <- grsc.marg.MLqE(X = X,</pre>
                         Y = Y,
                         n = n,
                         p = p,
                          q = 0.9,
                         m = m,
                          group = groups,
                          eps = 1e-06,
                          d = 15)
result$beta.group
result$group.screened
```

grsc.MLqE

Group Screening based on Maximum Lq-likelihood Estimation

#### Description

Group screening by ranking utility of each group. The group effect is defined based on the maximum Lq-likelihood estimates of the regression using each group of variables.

#### Usage

```
grsc.MLqE(
X,
Y,
n = dim(X)[1],
q = 0.9,
m,
group,
eps = 1e-06,
d = n/log(n)
)
```

#### Arguments

Х	A matrix of predictors.
Υ	A vector of response.
n	A value of sample size
q	A value of distortion parameter of Lq function, default to 0.9.
m	A number of the predictor groups
group	A vector of consecutive integers describing the grouping of the coefficients (see example below).
eps	The iteration coverage criterion, default to 1e-06.
d	A value of the number of groups retained after screening, default to $n/\log(n)$ .

#### Details

grsc.MLqE obtains the group effect of each group for subsequential group screening, based on the maximum Lq-likelihood estimates of the regression using each group of variables. By inheriting the advantage of the MLqE in small or moderate sample situations, the method is more robust to heterogeneous data and heavy-tailed distributions. It can work when correlation is mild or large. If group size equals to 1, individual screening is conducted.

#### Value

The grsc.MLqE returns a list containing the following components:

beta.group	The vector of utility of each group, which is the criterion for the variable screen-
	ing procedure.

group.screened The vector of integers denoting the screened groups.

#### Examples

```
# This is an example of grsc.MLqE with simulated data
data(LqG_SimuData)
X = LqG_SimuData$X
Y = LqG_SimuData$Y
n = dim(X)[1]
m = 200
groups = rep(1:( dim(X)[2] / 5), each = 5)
result <- grsc.MLqE(X = X,</pre>
                    Y = Y,
                    n = n,
                    q = 0.9,
                    m = m,
                     group = groups,
                    eps = 1e-06,
                     d = 15)
result$beta.group
result$group.screened
```

LqG\_SimuData

#### Description

The dataset LqG\_SimuData contains n = 100 samples with p = 1000 predictors. The number of the groups m = 200.

#### Usage

LqG\_SimuData

#### Format

A data list containing 100 samples

MLqE.est

Maximum Lq-likelihood Estimation

#### Description

The iterative algorithm for MLqE of coefficients of regression using each group of variables.

#### Usage

```
MLqE.est(
X,
Y,
q = 0.9,
eps = 1e-06
)
```

#### Arguments

Х	The matrix of the predictor group.
Y	The vector of response.
q	The value of distortion parameter of Lq function, default to $0.9$ .
eps	The iteration coverage criterion, default to 1e-06.

#### Details

The estimating equation of MLqE is a weighted version of that of the classical maximum likelihood estimation (MLE) where the distortion parameter q determines the similarity between the Lq function and the log function. When q = 1, MLqE is equivalent to MLE. The closer q is to 1, the more sensitive the MLqE is to outliers. As for the selection of q, there is presently no general method. However, MLqE is generally less sensitive to data contamination than MLE (to different degrees) when q is smaller than 1. Here, the default value of q is 0.9. Distortion parameter q can also be determined according to sample size n, choices of  $q_n$  with  $|1 - q_n|$  between  $\frac{1}{n}$  and  $\frac{1}{\sqrt{n}}$  usually improves over the MLE.

#### Value

The MLqE.est returns a list containing the following components:

t	The integer specifying the number of the total iterations in the algorithm.
beta_hat	The vector of estimated coefficients.
sigma_hat	The value of the estimated variance.
OMEGA_hat	The matrix of the estimated weight.

#### Examples

```
# This is an example of grsc.marg.MLqE with simulated data
data(LqG_SimuData)
X = LqG_SimuData X
Y = LqG_SimuData Y
n = dim(X)[1]
p = dim(X)[2]
m = 200
groups = rep(1:(dim(X)[2] / 5), each = 5)
Xb = X[ , which( groups == 1)]
result = MLqE.est(Xb,
                  Υ,
                  q = 0.9,
                  eps = 1e-06)
result$beta_hat
result$sigma_hat
result$OMEGA_hat
result$t
```

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