# Package 'LassoGEE'

# September 24, 2020

# Description

Does k-fold cross-validation for LassoGEE to select tuning parameter value for longitudinal data with working independence structure.

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# Usage

```
cv.LassoGEE(
   X,
   y,
   id,
   family,
   method = c("CGD", "RWL"),
   scale.fix,
   scale.value,
   fold,
   lambda.vec,
   maxiter,
   tol
)
```

# Arguments

Χ	A design matrix of dimension (nm) * p.
У	A response vector of length m * n.
id	A vector for identifying subjects/clusters.
family	A family object: a list of functions and expressions for defining link and variance functions. Families supported here is same as in <b>PGEE</b> which are binomial, gaussian, gamma and poisson.
method	The algorithms that are available. "CGD" represents the I-CGD algorithm, and "RWL" represents re-weighted least square algorithm.
scale.fix	A logical variable; if true, the scale parameter is fixed at the value of scale.value. The default value is TRUE.
scale.value	If scale.fix = TRUE, this assignes a numeric value to which the scale parameter should be fixed. The default value is 1.
fold	The number of folds used in cross-validation.
lambda.vec	A vector of tuning parameters that will be used in the cross-validation.
maxiter	The number of iterations that is used in the estimation algorithm. The default value is 50.
tol	The tolerance level that is used in the estimation algorithm. The default value is $1e^{-3}$ .

# Value

An object class of cv.LassoGEE.

# References

Li, Y., Gao, X., and Xu, W. (2020). Statistical consistency for generalized estimating equation with  $L_1$  regularization.

## See Also

LassoGEE

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IC

Information Criterion for selecting the tuning parameter.

## **Description**

Information Criterion for a fitted LassoGEE object with the AIC, BIC, or GCV criteria.

## Usage

```
IC(obj, criterion = c("BIC", "AIC", "GCV", "AICc", "EBIC"))
```

# Arguments

obj A fitted LassoGEE object.

criterion The criterion by which to select the regularization parameter. One of "AIC", "BIC", "GCV", "AICc", or "EBIC"; default is "BIC".

#### Value

IC The calculated model selection criteria

#### References

Gao, X., and Yi, G. Y. (2013). Simultaneous model selection and estimation for mean and association structures with clustered binary data. Stat, 2(1), 102-118.

LassoGEE

Function to fit penalized GEE by I-CGD algorithm.

## **Description**

This function fits a  $L_1$  penalized GEE model to longitudinal data by I-CGD algorithm or reweighted least square algorithm.

## Usage

```
LassoGEE(
   X,
   y,
   id,
   family = binomial("probit"),
   lambda,
   corstr = "independence",
   method = c("CGD", "RWL"),
   beta.ini = NULL,
   R = NULL,
   scale.fix = TRUE,
   scale.value = 1,
   maxiter = 50,
```

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```
tol = 0.001,
silent = TRUE,
Mv = NULL
)
```

# **Arguments**

X A design matrix of dimension (nm) \* p.y A response vector of length m \* n.

id A vector for identifying subjects/clusters.

family A family object representing one of the built-in families. Families supported

here are the same as in **PGEE**, e.g, binomial, gaussian, gamma and poisson, and the corresponding link functions are supported, e.g, identity, and probit.

lambda A user supplied value for the penalization parameter.

corstr A character string that indicates the correlation structure among the repeated

measurements of a subject. Structures supported in LassoGEE are "AR-1", "exchangeable", "unstructured", and "independence". The default corstr type is

"independence".

method The algorithms that are available. "CGD" represents the I-CGD algorithm, and

"RWL" represents re-weighted least square algorithm.

beta.ini User specified initial values for regression parameters. The default value is NULL.

R User specified correlation matrix. The default value is NULL.

scale.fix A logical variable. The default value is TRUE, then the value of the scale param-

eter is fixed to scale.value.

scale.value If scale.fix = TRUE, a numeric value will be assigned to the fixed scale param-

eter. The default value is 1.

maxiter The maximum number of iterations used in the algorithm. The default value is

50.

tol The tolerance level used in the algorithm. The default value is 1e-3.

silent A logical variable; if false, the interation counts at each iteration of CGD are

printed. The default value is TRUE.

Mv If either "stat\_M\_dep", or "non\_stat\_M\_dep" is specified in corstr, then this

assigns a numeric value for Mv. Otherwise, the default value is NULL.

#### Value

A list containing the following components:

betaest return final estimation

beta\_all\_step return estimate in each iteration
inner.count iterative count in each stage
outer.iter iterate number of outer loop

#### References

Li, Y., Gao, X., and Xu, W. (2020). Statistical consistency for generalized estimating equation with  $L_1$  regularization.

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#### See Also

cv.LassoGEE

#### **Examples**

```
## Not run:
set.seed(123)
p <- 256
s \leftarrow ceiling(p^{1/3})
n \leftarrow ceiling(10 * s * log(p))
m < -4
# covariance matrix of p number of continuous covariates
X.sigma <- matrix(0, p, p)</pre>
  for (i in 1:p)
    X.sigma[i,] <- 0.5^{(abs((1:p)-i))}
}
# generate matrix of covariates
X \leftarrow as.matrix(rmvnorm(n*m, mean = rep(0,p), X.sigma))
# true regression parameter associated with the covariate
bt <- runif(s, 0.05, 0.5) # = rep(1/s,s)
beta.true <- c(bt,rep(0,p-s))
# intercept
beta_intercepts <- 0
# unstructure
tt <- runif(m*m,-1,1)
Rtmp <- t(matrix(tt, m,m))%*%matrix(tt, m,m)+diag(1,4)</pre>
R_{tr} \leftarrow diag(diag(Rtmp)^{-1/2})%*Rtmp%*%diag(diag(Rtmp)^{-1/2})
diag(R_tr) = round(diag(R_tr))
# library(SimCorMultRes)
# simulation of clustered binary responses
simulated_binary_dataset <- rbin(clsize = m, intercepts = beta_intercepts,</pre>
                                   betas = beta.true, xformula = ~X, cor.matrix = R_tr,
                                   link = "probit")
lambda <- 0.2* s *sqrt(log(p)/n)
data = simulated_binary_dataset$simdata
y = data$y
X = data$X
id = data$id
ptm <- proc.time()</pre>
nCGDfit = LassoGEE(X = X, y = y, id = id, family = binomial("probit"),
                  lambda = lambda, corstr = "unstructured")
proc.time() - ptm
betaest <- nCGDfit$betaest</pre>
## End(Not run)
```

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## **Description**

Print a summary of the results of cross-validation for a LassoGEE model.

## Usage

```
## S3 method for class 'cv.LassoGEE'
print(x, digits = NULL, ...)
```

#### **Arguments**

X	fitted 'cv.LassoGEE' object
digits	significant digits in printout
	additional print arguments

#### **Details**

A summary of the cross-validated fit is produced. print.cv.LassoGEE(object) will print the summary for a sequence of lambda.

#### References

Li, Y., Gao, X., and Xu, W. (2020). Statistical consistency for generalized estimating equation with  $L_1$  regularization.

#### See Also

 ${\tt LassoGEE}, and ~{\tt cv.LassoGEE}~methods.$ 

```
print.LassoGEE print a LassoGEE object
```

#### **Description**

Print a summary of the results of a LassoGEE model.

## Usage

```
## S3 method for class 'LassoGEE'
print(x, digits = NULL, ...)
```

# Arguments

```
x fitted 'LassoGEE' objectdigits significant digits in printoutadditional print arguments
```

## **Details**

A summary of the cross-validated fit is produced. print.cv.LassoGEE(object) will print the summary includes Working Correlation and Returned Error Value.

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# References

Li, Y., Gao, X., and Xu, W. (2020). Statistical consistency for generalized estimating equation with  $L_1$  regularization.

# See Also

LassoGEE, and cv.LassoGEE methods.

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