

Package ‘SIHR’

July 3, 2022

Type Package

Title Statistical Inference in High Dimensional Regression

Version 1.0.0

Author Prabrisha Rakshit, Zhenyu Wang, Zijian Guo, Tony Cai

Maintainer Zijian Guo <zijguo@stat.rutgers.edu>

Description Inference procedures in the high-dimensional setting for

- (1) linear functionals in generalized linear regression ('Cai et al.' (2019) <arXiv:1904.12891>, 'Guo et al.' (2020) <arXiv:2012.07133>, 'Cai et al.' (2021)),
- (2) individual treatment effects in generalized linear regression.
- (3) quadratic functionals in linear regression ('Guo et al.' (2019) <arXiv:1909.01503>)

License GPL-3

Encoding UTF-8

LazyData true

RoxygenNote 7.2.0

URL <https://github.com/prabrishar1/SIHR>

Imports CVXR,
glmnet,
stats

R topics documented:

ITE	1
LF	3
QF	5
Index	7

ITE	<i>Inference for difference of linear combinations of the regression vectors in high dimensional generalized linear regressions</i>
-----	---

Description

Computes the bias-corrected estimator of the difference of linear combinations of the regression vectors for the high dimensional generalized linear regressions and the corresponding standard error.

Usage

```

ITE(
  X1,
  y1,
  X2,
  y2,
  loading.mat,
  model = "linear",
  intercept = TRUE,
  intercept.loading = TRUE,
  lambda = NULL,
  mu = NULL,
  init.step = NULL,
  resol = 1.5,
  maxiter = 6,
  alpha = 0.05,
  verbose = TRUE
)

```

Arguments

X1	Design matrix for the first sample, of dimension $n_1 \times p$
y1	Outcome vector for the first sample, of length n_1
X2	Design matrix for the second sample, of dimension $n_2 \times p$
y2	Outcome vector for the second sample, of length n_1
loading.mat	Loading matrix, $nrow=p$, each column corresponds to a loading of interest
model	The high dimensional regression model, either linear or logistic or logistic alternative or probit
intercept	Should intercept(s) be fitted for the initial estimators (default = TRUE)
intercept.loading	Should intercept be included for the loading (default = TRUE)
lambda	lambda The tuning parameter in fitting model (default = NULL)
mu	The dual tuning parameter used in the construction of the projection direction (default = NULL)
init.step	The initial step size used to compute mu; if set to NULL it is computed to be the number of steps (maxiter) to obtain the smallest mu
resol	The factor by which mu is increased/decreased to obtain the smallest mu such that the dual optimization problem for constructing the projection direction converges (default = 1.5)
maxiter	Maximum number of steps along which mu is increased/decreased to obtain the smallest mu such that the dual optimization problem for constructing the projection direction converges (default = 6)
alpha	Level of significance to construct two-sided confidence interval (default = 0.05)
verbose	Should intermediate message(s) be printed (default = TRUE)

Value

<code>est.plugin.vec</code>	The vector of plugin(biased) estimators for the linear combination of regression coefficients, length of <code>ncol(loading.mat)</code> ; corresponding to different column in <code>loading.mat</code>
<code>est.debias.vec</code>	The vector of bias-corrected estimators for the linear combination of regression coefficients, length of <code>ncol(loading.mat)</code> ; corresponding to different column in <code>loading.mat</code>
<code>se.vec</code>	The vector of standard errors of the bias-corrected estimators, length of <code>ncol(loading.mat)</code> ; corresponding to different column in <code>loading.mat</code>
<code>ci.mat</code>	The matrix of two.sided confidence interval for the linear combination, of dimension <code>ncol(loading.mat) x 2</code> ; the row corresponding to different column in <code>loading.mat</code>

Examples

```

X1 = matrix(rnorm(100*10), nrow=100, ncol=10)
y1 = -0.5 + X1[,1] * 0.5 + X1[,2] * 1 + rnorm(100)
X2 = matrix(rnorm(90*10), nrow=90, ncol=10)
y2 = -0.4 + X2[,1] * 0.48 + X2[,2] * 1.1 + rnorm(90)
loading1 = c(1, 1, rep(0,8))
loading2 = c(-0.5, -1, rep(0,8))
loading.mat = cbind(loading1, loading2)
Est = ITE(X1, y1, X2, y2, loading.mat, model="linear")

## compute confidence intervals
ci(Est, alpha=0.05, alternative="two.sided")

## summary statistics
summary(Est)

```

LF

Inference for linear combination of the regression vector in high dimensional generalized linear regression

Description

Inference for linear combination of the regression vector in high dimensional generalized linear regression

Usage

```

LF(
  X,
  y,
  loading.mat,
  model = c("linear", "logistic", "logistic_alternative", "probit"),
  intercept = TRUE,
  intercept.loading = TRUE,
  lambda = NULL,
  mu = NULL,
  init.step = NULL,

```

```

    resol = 1.5,
    maxiter = 6,
    alpha = 0.05,
    verbose = TRUE
  )

```

Arguments

<code>x</code>	Design matrix, of dimension $n \times p$
<code>y</code>	Outcome vector, of length n
<code>loading.mat</code>	Loading matrix, $nrow=p$, each column corresponds to a loading of interest
<code>model</code>	The high dimensional regression model, either linear or logistic or <code>logistic_alternative</code> or <code>probit</code>
<code>intercept</code>	Should intercept be fitted for the initial estimator (default = TRUE)
<code>intercept.loading</code>	Should intercept be included for the loading (default = TRUE)
<code>lambda</code>	The tuning parameter in fitting model (default = NULL)
<code>mu</code>	The dual tuning parameter used in the construction of the projection direction (default = NULL)
<code>init.step</code>	The initial step size used to compute μ ; if set to NULL it is computed to be the number of steps (<code>maxiter</code>) to obtain the smallest μ
<code>resol</code>	The factor by which μ is increased/decreased to obtain the smallest μ such that the dual optimization problem for constructing the projection direction converges (default = 1.5)
<code>maxiter</code>	Maximum number of steps along which μ is increased/decreased to obtain the smallest μ such that the dual optimization problem for constructing the projection direction converges (default = 6)
<code>alpha</code>	Level of significance to construct two-sided confidence interval (default = 0.05)
<code>verbose</code>	Should intermediate message(s) be printed (default = TRUE)

Value

<code>est.plugin.vec</code>	The vector of plugin(biased) estimators for the linear combination of regression coefficients, length of <code>ncol(loading.mat)</code> ; each corresponding to a loading of interest
<code>est.debias.vec</code>	The vector of bias-corrected estimators for the linear combination of regression coefficients, length of <code>ncol(loading.mat)</code> ; each corresponding to a loading of interest
<code>se.vec</code>	The vector of standard errors of the bias-corrected estimators, length of <code>ncol(loading.mat)</code> ; each corresponding to a loading of interest
<code>ci.mat</code>	The matrix of two.sided confidence interval for the linear combination, of dimension <code>ncol(loading.mat) x 2</code> ; each row corresponding to a loading of interest
<code>proj.mat</code>	The matrix of projection directions; each column corresponding to a loading of interest

Examples

```
X = matrix(rnorm(100*10), nrow=100, ncol=10)
y = -0.5 + X[,1] * 0.5 + X[,2] * 1 + rnorm(100)
loading1 = c(1, 1, rep(0, 8))
loading2 = c(-0.5, -1, rep(0, 8))
loading.mat = cbind(loading1, loading2)
Est = LF(X, y, loading.mat, model="linear")

## compute confidence intervals
ci(Est, alpha=0.05, alternative="two.sided")

## summary statistics
summary(Est)
```

QF

Inference for quadratic forms of the regression vector in high dimensional linear and logistic regressions

Description

Inference for quadratic forms of the regression vector in high dimensional linear and logistic regressions

Usage

```
QF(
  X,
  y,
  G,
  A = NULL,
  model = c("linear", "logistic", "logistic_alternative", "probit"),
  intercept = TRUE,
  tau.vec = c(0.5, 1),
  lambda = NULL,
  mu = NULL,
  init.step = NULL,
  resol = 1.5,
  maxiter = 6,
  alpha = 0.05,
  verbose = TRUE
)
```

Arguments

X	Design matrix, of dimension $n \times p$
y	Outcome vector, of length n
G	The set of indices, G in the quadratic form
A	The matrix A in the quadratic form, of dimension $ G \times G $. If NULL A would be set as the $ G \times G $ submatrix of the population covariance matrix corresponding to the index set G (default = NULL)

<code>model</code>	The high dimensional regression model, either <code>linear</code> or <code>logistic</code> or <code>logistic_alternative</code> or <code>probit</code>
<code>intercept</code>	Should intercept be fitted for the initial estimator (default = TRUE)
<code>tau.vec</code>	The vector of enlargement factors for asymptotic variance of the bias-corrected estimator to handle super-efficiency (default = $c(0.5, 1)$)
<code>lambda</code>	The tuning parameter in fitting model (default = NULL)
<code>mu</code>	The dual tuning parameter used in the construction of the projection direction (default = NULL)
<code>init.step</code>	The initial step size used to compute <code>mu</code> ; if set to NULL it is computed to be the number of steps ($< \text{maxiter}$) to obtain the smallest <code>mu</code> such that the dual optimization problem for constructing the projection direction converges (default = NULL)
<code>resol</code>	Resolution or the factor by which <code>mu</code> is increased/decreased to obtain the smallest <code>mu</code> such that the dual optimization problem for constructing the projection direction converges (default = 1.5)
<code>maxiter</code>	maximum number of steps along which <code>mu</code> is increased/decreased to obtain the smallest <code>mu</code> such that the dual optimization problem for constructing the projection direction converges (default = 6)
<code>alpha</code>	Level of significance to construct two-sided confidence interval (default = 0.05)
<code>verbose</code>	Should intermediate message(s) be printed (default = TRUE)

Value

<code>est.plugin</code>	The plugin(biased) estimator for the quadratic form of the regression vector restricted to <code>G</code>
<code>est.debias</code>	The bias-corrected estimator of the quadratic form of the regression vector
<code>se.vec</code>	The vector of standard errors of the bias-corrected estimator, length of <code>tau.vec</code> ; corresponding to different values of <code>tau.vec</code>
<code>ci.mat</code>	The matrix of two.sided confidence interval for the quadratic form of the regression vector; row corresponds to different values of <code>tau.vec</code>
<code>proj</code>	The projection direction

Examples

```
X = matrix(rnorm(100*10), nrow=100, ncol=10)
y = X[,1] * 0.5 + X[,2] * 1 + rnorm(100)
G = c(1,2)
A = matrix(c(1.5, 0.8, 0.8, 1.5), nrow=2, ncol=2)
Est = QF(X, y, G, A, model="linear")
## compute confidence intervals
ci(Est, alpha=0.05, alternative="two.sided")

## summary statistics
summary(Est)
```

Index

ITE, [1](#)

LF, [3](#)

QF, [5](#)