

# Package ‘SIHR’

August 9, 2022

**Type** Package

**Title** Statistical Inference in High Dimensional Regression

**Version** 1.0.1

**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
<b>Index</b>	<b>8</b>

---

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 = FALSE,
  beta.init1 = NULL,
  beta.init2 = NULL,
  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 = FALSE)
beta.init1	The initial estimator of the regression vector for the 1st data (default = NULL)
beta.init2	The initial estimator of the regression vector for the 2nd data (default = NULL)
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>
<code>prob.debias.vec</code>	The vector of bias-corrected estimators after probability transformation, length of <code>ncol(loading.mat)</code> ; corresponding to different column in <code>loading.mat</code> . The value would be NULL for non-logistic model.
<code>prob.se.vec</code>	The vector of standard errors of the bias-corrected estimators after probability transformation, length of <code>ncol(loading.mat)</code> ; corresponding to different column in <code>loading.mat</code> . The value would be NULL for non-logistic model.

**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 = FALSE,
  beta.init = NULL,
  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 logistic_alternative or probit
<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 = FALSE)
<code>beta.init</code>	The initial estimator of the regression vector (default = NULL)
<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 mu; if set to NULL it is computed to be the number of steps ( <code>maxiter</code> ) to obtain the smallest mu
<code>resol</code>	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)
<code>maxiter</code>	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)
<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**

`est.plugin.vec` The vector of plugin(biased) estimators for the linear combination of regression coefficients, length of `ncol(loading.mat)`; 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, 0.5, 1),
  beta.init = NULL,
  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>G</code>	The set of indices, $G$ in the quadratic form
<code>A</code>	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 = $c(0, 0.5, 1)$ )
<code>beta.init</code>	The initial estimator of the regression vector (default = NULL)
<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 $\mu$ ; if set to NULL it is computed to be the number of steps ( $< \text{maxiter}$ ) to obtain the smallest $\mu$ such that the dual optimization problem for constructing the projection direction converges (default = NULL)
<code>resol</code>	Resolution or 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)
<code>maxiter</code>	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)
<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 $G$
<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](#)