

Package: bndovb (via r-universe)

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Title Bounding Omitted Variable Bias Using Auxiliary Data

Version 1.2

Description Functions to implement a Hwang(2021)

<[doi:10.2139/ssrn.3866876](https://doi.org/10.2139/ssrn.3866876)> estimator, which bounds an omitted variable bias using auxiliary data.

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Encoding UTF-8

LazyData true

Depends R (>= 2.10)

Roxygen list(markdown = TRUE)

RoxygenNote 7.1.1

Imports np, pracma, stats, utils, MASS, dplyr, factormodel, nnet,
doParallel, foreach, parallel

Suggests knitr, rmarkdown

VignetteBuilder knitr

Repository <https://yujunghwang.r-universe.dev>

RemoteUrl <https://github.com/yujunghwang/bndovb>

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Contents

auxdat_mecont	2
auxdat_medisc	2
auxdat_nome	3
bndovb	4
bndovbme	6
bndovbme_tuning	8
bndovb_tuning	11
maindat_mecont	13
maindat_medisc	13
maindat_nome	14

auxdat_mecont*A simulated auxiliary data to show how to use 'bndovbme' function with continuous proxy variables***Description**

A simulated auxiliary data to show how to use 'bndovbme' function with continuous proxy variables

Usage

```
auxdat_mecont
```

Format

A data frame with 3000 rows and 5 variables:

- w1** A common covariate in both main and auxiliary data
- x** A common covariate in both main and auxiliary data
- z1** A continuous proxy variable
- z2** A continuous proxy variable
- z3** A continuous proxy variable

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

auxdat_medisc*A simulated auxiliary data to show how to use 'bndovbme' function with discrete proxy variables***Description**

A simulated auxiliary data to show how to use 'bndovbme' function with discrete proxy variables

Usage

```
auxdat_medisc
```

Format

A data frame with 3000 rows and 5 variables:

- w1** A common covariate in both main and auxiliary data
- x** A common covariate in both main and auxiliary data
- z1** A discrete proxy variable
- z2** A discrete proxy variable
- z3** A discrete proxy variable

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

auxdat_nome

*A simulated auxiliary data to show how to use 'bndovb' function***Description**

A simulated auxiliary data to show how to use 'bndovb' function

Usage

auxdat_nome

Format

A data frame with 50000 rows and 3 variables:

- x1** An omitted variable in the main data
- x2** A common covariate in both main and auxiliary data
- x3** A common covariate in both main and auxiliary data

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

bndovb*bndovb***Description**

This function runs a two sample least squares when auxiliary data contains every right-hand side regressor and main data contains a dependent variable and every right-hand side regressor but one omitted variable.

Usage

```
bndovb(
  maindat,
  auxdat,
  depvar,
  ovar,
  comvar,
  method = 1,
  mainweights = NULL,
  auxweights = NULL,
  signres = NULL,
  ci = FALSE,
  nboot = 100,
  scale = -1/2,
  tau = 0.05,
  seed = 210823,
  display = TRUE
)
```

Arguments

<code>maindat</code>	Main data set. It must be a data frame.
<code>auxdat</code>	Auxiliary data set. It must be a data frame.
<code>depvar</code>	A name of a dependent variable in main dataset
<code>ovar</code>	A name of an omitted variable in main dataset which exists in auxiliary data
<code>comvar</code>	A vector of the names of common regressors existing in both main data and auxiliary data
<code>method</code>	CDF and Quantile function estimation method. Users can choose either 1 or 2. If the method is 1, the CDF and quantile function is estimated assuming a parametric normal distribution. If the method is 2, the CDF and quantile function is estimated using a nonparametric estimator in Li and Racine(2008) doi:10.1198/073500107000000250 , Li, Lin, and Racine(2013) doi:10.1080/07350015.2012.738955 . Default is 1.
<code>mainweights</code>	An optional weight vector for the main dataset. The length must be equal to the number of rows of 'maindat'.

auxweights	An optional weight vector for the auxiliary dataset. The length must be equal to the number of rows of 'auxdat'.
signres	An option to impose a sign restriction on a coefficient of an omitted variable. Set either NULL or pos or neg. Default is NULL. If NULL, there is no sign restriction. If 'pos', the estimator imposes an extra restriction that the coefficient of an omitted variable must be positive. If 'neg', the estimator imposes an extra restriction that the coefficient of an omitted variable must be negative.
ci	An option to compute an equal-tailed confidence interval. Default is FALSE. It may take some time to compute CI from bootstrap.
nboot	Number of bootstraps to compute the confidence interval. Default is 100.
scale	A tuning parameter for rescaled numerical bootstrap. The value must be between -1/2 and 0. (main data sample size)^scale is the tuning parameter epsilon_n in Hwang (2021). Default is -1/2 (that is, standard bootstrap).
tau	Significance level. (1-tau)% confidence interval is computed. Default is 0.05.
seed	Seed for random number generation. Default is 210823.
display	It must be either TRUE or FALSE. Whether to display progress and messages. Default is TRUE.

Value

Returns a list of 12 components :

hat_beta_l lower bound estimates of regression coefficients
hat_beta_u upper bound estimates of regression coefficients
mu_l lower bound estimate of $E[\text{ovar}^*\text{depvar}]$
mu_u upper bound estimate of $E[\text{ovar}^*\text{depvar}]$
hat_beta_l_cil (1-tau)% confidence interval lower bound for hat_beta_l
hat_beta_l_ciu (1-tau)% confidence interval upper bound for hat_beta_l
hat_beta_u_cil (1-tau)% confidence interval lower bound for hat_beta_u
hat_beta_u_ciu (1-tau)% confidence interval upper bound for hat_beta_u
mu_l_cil (1-tau)% confidence interval lower bound for mu_l
mu_l_ciu (1-tau)% confidence interval upper bound for mu_l
mu_u_cil (1-tau)% confidence interval lower bound for mu_u
mu_u_ciu (1-tau)% confidence interval upper bound for mu_u

Author(s)

Yujung Hwang, <yujungghwang@gmail.com>

References

Hwang, Yujung (2021) Bounding Omitted Variable Bias Using Auxiliary Data. Available at SSRN.[doi:10.2139/ssrn.3866876](#)

Examples

```
data(maindat_nome)
data(auxdat_nome)

bndovb(maindat=maindat_nome,auxdat=auxdat_nome,depvar="y",ovar="x1",comvar=c("x2","x3"),method=1)
```

bndovbme

bndovbme

Description

This function runs a two sample least squares when main data contains a dependent variable and every right hand side regressor but one omitted variable. The function requires an auxiliary data which includes every right hand side regressor but one omitted variable, and enough proxy variables for the omitted variable. When the omitted variable is continuous, the auxiliary data must contain at least two continuous proxy variables. When the omitted variable is discrete, the auxiliary data must contain at least three continuous proxy variables.

Usage

```
bndovbme(
  maindat,
  auxdat,
  depvar,
  pvar,
  ptype = 1,
  comvar,
  sbar = 2,
  mainweights = NULL,
  auxweights = NULL,
  normalize = TRUE,
  signres = NULL,
  ci = FALSE,
  nboot = 100,
  scale = -1/2,
  tau = 0.05,
  seed = 210823,
  display = TRUE
)
```

Arguments

<code>maindat</code>	Main data set. It must be a data frame.
<code>auxdat</code>	Auxiliary data set. It must be a data frame.
<code>depvar</code>	A name of a dependent variable in main dataset

pvar	A vector of the names of the proxy variables for the omitted variable. When proxy variables are continuous, the first proxy variable is used as an anchoring variable. When proxy variables are discrete, the first proxy variable is used for initialization (For details, see a documentation for "dproxyme" function).
ptype	Either 1 (continuous) or 2 (discrete). Whether proxy variables are continuous or discrete. Default is 1 (continuous).
comvar	A vector of the names of the common regressors existing in both main data and auxiliary data
sbar	A cardinality of the support of the discrete proxy variables. Default is 2. If proxy variables are continuous, this variable is irrelevant.
mainweights	An optional weight vector for the main dataset. The length must be equal to the number of rows of 'maindat'.
auxweights	An optional weight vector for the auxiliary dataset. The length must be equal to the number of rows of 'auxdat'.
normalize	Whether to normalize the omitted variable to have mean 0 and standard deviation 1. Set TRUE or FALSE. Default is TRUE. If FALSE, then the scale of the omitted variable is anchored with the first proxy variable in pvar list.
signres	An option to impose a sign restriction on a coefficient of an omitted variable. Set either NULL or pos or neg. Default is NULL. If NULL, there is no sign restriction. If 'pos', the estimator imposes an extra restriction that the coefficient of an omitted variable must be positive. If 'neg', the estimator imposes an extra restriction that the coefficient of an omitted variable must be negative.
ci	An option to compute an equal-tailed confidence interval. Default is FALSE. It may take some time to compute CI from bootstrap.
nboot	Number of bootstraps to compute the confidence interval. Default is 100.
scale	A tuning parameter for rescaled numerical bootstrap. The value must be between -1/2 and 0. (main data sample size) ^{scale} is the tuning parameter epsilon_n in Hwang (2021). Default is -1/2 (that is, standard bootstrap).
tau	Significance level. (1-tau)% confidence interval is computed. Default is 0.05.
seed	Seed for random number generation. Default is 210823.
display	It must be either TRUE or FALSE. Whether to display progress and messages. Default is TRUE.

Value

Returns a list of 4 components :

hat_beta_l lower bound estimates of regression coefficients
hat_beta_u upper bound estimates of regression coefficients
mu_l lower bound estimate of E[ovar*depvar]
mu_u upper bound estimate of E[ovar*depvar]
hat_beta_l_cil (1-tau)% confidence interval lower bound for hat_beta_l
hat_beta_l_ciu (1-tau)% confidence interval upper bound for hat_beta_l

```
hat_beta_u_cil (1-tau)% confidence interval lower bound for hat_beta_u
hat_beta_u_ciu (1-tau)% confidence interval upper bound for hat_beta_u
mu_l_cil (1-tau)% confidence interval lower bound for mu_l
mu_l_ciu (1-tau)% confidence interval upper bound for mu_l
mu_u_cil (1-tau)% confidence interval lower bound for mu_u
mu_u_ciu (1-tau)% confidence interval upper bound for mu_u
```

Author(s)

Yujung Hwang, <yujungghwang@gmail.com>

References

Hwang, Yujung (2021) Bounding Omitted Variable Bias Using Auxiliary Data. Available at SSRN.
[doi:10.2139/ssrn.3866876](https://doi.org/10.2139/ssrn.3866876)

Examples

```
## load example data
data(maindat_mecont)
data(auxdat_mecont)

## set ptype=1 for continuous proxy variables
pvar<-c("z1","z2","z3")
cvar<-c("x","w1")
bndovbme(maindat=maindat_mecont,auxdat=auxdat_mecont,depvar="y",pvar=pvar,ptype=1,comvar=cvar)

## set ptype=2 for discrete proxy variables
data(maindat_medisc)
data(auxdat_medisc)
bndovbme(maindat=maindat_medisc,auxdat=auxdat_medisc,depvar="y",pvar=pvar,ptype=2,comvar=cvar)
```

Description

This function computes an optimal tuning parameter to compute the confidence interval for bndovbme function. The function returns an optimal tuning parameter using double bootstrap procedure.

Usage

```
bndovbme_tuning(
  maindat,
  auxdat,
  depvar,
  pvar,
  ptype = 1,
  comvar,
  sbar = 2,
  mainweights = NULL,
  auxweights = NULL,
  normalize = TRUE,
  signres = NULL,
  nboot = 100,
  scalegrid = c(-1/2, -1/3, -1/4, -1/5, -1/6),
  tau = 0.05,
  seed = 210823,
  parallel = TRUE
)
```

Arguments

<code>maindat</code>	Main data set. It must be a data frame.
<code>auxdat</code>	Auxiliary data set. It must be a data frame.
<code>depvar</code>	A name of a dependent variable in main dataset
<code>pvar</code>	A vector of the names of the proxy variables for the omitted variable. When proxy variables are continuous, the first proxy variable is used as an anchoring variable. When proxy variables are discrete, the first proxy variable is used for initialization (For details, see a documentation for "dproxyme" function).
<code>ptype</code>	Either 1 (continuous) or 2 (discrete). Whether proxy variables are continuous or discrete. Default is 1 (continuous).
<code>comvar</code>	A vector of the names of the common regressors existing in both main data and auxiliary data
<code>sbar</code>	A cardinality of the support of the discrete proxy variables. Default is 2. If proxy variables are continuous, this variable is irrelevant.
<code>mainweights</code>	An optional weight vector for the main dataset. The length must be equal to the number of rows of 'maindat'.
<code>auxweights</code>	An optional weight vector for the auxiliary dataset. The length must be equal to the number of rows of 'auxdat'.
<code>normalize</code>	Whether to normalize the omitted variable to have mean 0 and standard deviation 1. Set TRUE or FALSE. Default is TRUE. If FALSE, then the scale of the omitted variable is anchored with the first proxy variable in <code>pvar</code> list.
<code>signres</code>	An option to impose a sign restriction on a coefficient of an omitted variable. Set either NULL or pos or neg. Default is NULL. If NULL, there is no sign restriction. If 'pos', the estimator imposes an extra restriction that the coefficient

of an omitted variable must be positive. If 'neg', the estimator imposes an extra restriction that the coefficient of an omitted variable must be negative.

<code>nboot</code>	Number of bootstraps to compute the confidence interval. Default is 100.
<code>scalegrid</code>	Tuning parameter grid to search. It must be a vector of numbers between -1/2 and 0. Default is <code>c(-1/2,-1/3,-1/4,-1/5,-1/6)</code> .
<code>tau</code>	Significance level. $(1-\tau)\%$ confidence interval is computed. Default is 0.05.
<code>seed</code>	Seed for random number generation. Default is 210823.
<code>parallel</code>	Either TRUE or FALSE. Whether to compute in parallel. Default is TRUE.

Value

Returns a list of 3 components :

optimal_scale An optimal scale parameter which gives coverage rates closest to $(1-\tau)$

cover_beta_l A matrix of coverage rates of the lower bound parameters under different scale parameters

cover_beta_u A matrix of coverage rates of the lower bound parameters under different scale parameters

Author(s)

Yujung Hwang, <yujunghwang@gmail.com>

References

Hwang, Yujung (2021) Bounding Omitted Variable Bias Using Auxiliary Data. Available at SSRN.[doi:10.2139/ssrn.3866876](https://doi.org/10.2139/ssrn.3866876)

Examples

```
## load example data
data(maindat_mecont)
data(auxdat_mecont)

## set ptype=1 for continuous proxy variables
pvar<-c("z1","z2","z3")
cvar<-c("x","w1")

# To shorten computation time, I set the number of bootstrap small in an example below.
# In practice, please set it a large number
bndovbme_tuning(maindat_mecont,auxdat_mecont,depvar="y",pvar=pvar,ptype=1,comvar=cvar,nboot=2)
```

bndovb_tuning	<i>bndovb_tuning</i>	
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Description

This function computes an optimal tuning parameter to compute the confidence interval for bndovb function. The function returns an optimal tuning parameter using double bootstrap procedure.

Usage

```
bndovb_tuning(
  maindat,
  auxdat,
  depvar,
  ovar,
  comvar,
  method = 1,
  mainweights = NULL,
  auxweights = NULL,
  signres = NULL,
  nboot = 100,
  scalegrid = c(-1/2, -1/3, -1/4, -1/5, -1/6),
  tau = 0.05,
  seed = 210823,
  parallel = TRUE
)
```

Arguments

<code>maindat</code>	Main data set. It must be a data frame.
<code>auxdat</code>	Auxiliary data set. It must be a data frame.
<code>depvar</code>	A name of a dependent variable in main dataset
<code>ovar</code>	A name of an omitted variable in main dataset which exists in auxiliary data
<code>comvar</code>	A vector of the names of common regressors existing in both main data and auxiliary data
<code>method</code>	CDF and Quantile function estimation method. Users can choose either 1 or 2. If the method is 1, the CDF and quantile function is estimated assuming a parametric normal distribution. If the method is 2, the CDF and quantile function is estimated using a nonparametric estimator in Li and Racine(2008) doi:10.1198/073500107000000250 , Li, Lin, and Racine(2013) doi:10.1080/07350015.2012.738955 . Default is 1.
<code>mainweights</code>	An optional weight vector for the main dataset. The length must be equal to the number of rows of 'maindat'.
<code>auxweights</code>	An optional weight vector for the auxiliary dataset. The length must be equal to the number of rows of 'auxdat'.

signres	An option to impose a sign restriction on a coefficient of an omitted variable. Set either NULL or pos or neg. Default is NULL. If NULL, there is no sign restriction. If 'pos', the estimator imposes an extra restriction that the coefficient of an omitted variable must be positive. If 'neg', the estimator imposes an extra restriction that the coefficient of an omitted variable must be negative.
nboot	Number of bootstraps to compute the confidence interval. Default is 100.
scalegrid	Tuning parameter grid to search. It must be a vector of numbers between -1/2 and 0. Default is c(-1/2,-1/3,-1/4,-1/5,-1/6).
tau	Significance level. (1-tau)% confidence interval is computed. Default is 0.05.
seed	Seed for random number generation. Default is 210823.
parallel	Either TRUE or FALSE. Whether to compute in parallel. Default is TRUE.

Value

Returns a list of 3 components :

optimal_scale An optimal scale parameter which gives coverage rates closest to (1-tau)

cover_beta_I A matrix of coverage rates of the lower bound parameters under different scale parameters

cover_beta_u A matrix of coverage rates of the lower bound parameters under different scale parameters

Author(s)

Yujung Hwang, <yujunghwang@gmail.com>

References

Hwang, Yujung (2021) Bounding Omitted Variable Bias Using Auxiliary Data. Available at SSRN.[doi:10.2139/ssrn.3866876](#)

Examples

```
data(maindat_nome)
data(auxdat_nome)

# To shorten computation time, I set the number of bootstrap small in an example below.
# In practice, please set it a large number
bndovb_tuning(maindat_nome,auxdat_nome,depvar="y",ovar="x1",comvar=c("x2","x3"),method=1,nboot=2)
```

maindat_mecont	<i>A simulated main data to show how to use 'bndovbme' function with continuous proxy variables</i>
----------------	---

Description

A simulated main data to show how to use 'bndovbme' function with continuous proxy variables

Usage

```
maindat_mecont
```

Format

A data frame with 3000 rows and 3 variables:

- w1 A common covariate in both main and auxiliary data
- x A common covariate in both main and auxiliary data
- y A dependent variable

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

maindat_medisc	<i>A simulated main data to show how to use 'bndovbme' function with discrete proxy variables</i>
----------------	---

Description

A simulated main data to show how to use 'bndovbme' function with discrete proxy variables

Usage

```
maindat_medisc
```

Format

A data frame with 3000 rows and 3 variables:

- w1 A common covariate in both main and auxiliary data
- x A common covariate in both main and auxiliary data
- y A dependent variable

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

maindat_nome

A simulated main data to show how to use 'bndovb' function

Description

A simulated main data to show how to use 'bndovb' function

Usage

maindat_nome

Format

A data frame with 100000 rows and 3 variables:

- x2** A common covariate in both main and auxiliary data
- x3** A common covariate in both main and auxiliary data
- y** A dependent variable

Source

This dataset was simulated by simulatePackageData.R in data-raw folder

Index

* datasets

auxdat_mecont, 2
auxdat_medisc, 2
auxdat_nome, 3
maindat_mecont, 13
maindat_medisc, 13
maindat_nome, 14

auxdat_mecont, 2
auxdat_medisc, 2
auxdat_nome, 3

bndovb, 4
bndovb_tuning, 11
bndovbme, 6
bndovbme_tuning, 8

maindat_mecont, 13
maindat_medisc, 13
maindat_nome, 14