Package 'bootStateSpace'

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|---|
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| Version 1.0.1 |
| Description Provides a streamlined and user-friendly framework for bootstrapping in state space models, particularly when the number of subjects/units (n) exceeds one, a scenario commonly encountered in social and behavioral sciences. For an introduction to state space models in social and behavioral sciences, refer to Chow, Ho, Hamaker, and Dolan (2010) <doi:10.1080 10705511003661553="">.</doi:10.1080> |
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coef.bootstatespace

Estimated Parameter Method for an Object of Class bootstatespace

Description

Estimated Parameter Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
coef(object, ...)
```

Arguments

object Object of Class bootstatespace. ... additional arguments.

Value

Returns a vector of estimated parameters.

Author(s)

Ivan Jacob Agaloos Pesigan

confint.bootstatespace

Confidence Intervals Method for an Object of Class bootstatespace

Description

Confidence Intervals Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
confint(object, parm = NULL, level = 0.95, type = "pc", ...)
```

extract 3

Arguments

object of Class bootstatespace.

parm a specification of which parameters are to be given confidence intervals, either

a vector of numbers or a vector of names. If missing, all parameters are consid-

ered.

level the confidence level required.

type Charater string. Confidence interval type, that is, type = "pc" for percentile;

type = "bc" for bias corrected.

... additional arguments.

Value

Returns a matrix of confidence intervals.

Author(s)

Ivan Jacob Agaloos Pesigan

extract Generic Function

Description

A generic function for extracting elements from objects.

Usage

```
extract(object, what)
```

Arguments

object An object.

what Character string.

Value

A value determined by the specific method for the object's class.

```
extract.bootstatespace
```

Extract Method for an Object of Class bootstatespace

Description

Extract Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
extract(object, what = NULL)
```

Arguments

object of Class bootstatespace.

what Character string. What specific matrix to extract. If what = NULL, extract all

available matrices.

Value

Returns a list. Each element of the list is a list of bootstrap estimates in matrix format.

Author(s)

Ivan Jacob Agaloos Pesigan

PBSSMFixed

Parametric Bootstrap for the State Space Model (Fixed Parameters)

Description

This function simulates data from a state-space model and fits the model using the dynr package. The process is repeated R times. It assumes that the parameters remain constant across individuals and over time. At the moment, the function only supports type = 0.

Usage

```
PBSSMFixed(
  R,
  path,
  prefix,
  n,
  time,
  delta_t = 1,
```

```
mu0,
sigma0_1,
alpha,
beta,
psi_l,
nu,
lambda,
theta_1,
type = 0,
x = NULL
gamma = NULL,
kappa = NULL,
mu0_fixed = FALSE,
sigma0_fixed = FALSE,
alpha_level = 0.05,
optimization_flag = TRUE,
hessian_flag = FALSE,
verbose = FALSE,
weight_flag = FALSE,
debug_flag = FALSE,
perturb_flag = FALSE,
xtol_rel = 1e-07,
stopval = -9999,
ftol_rel = -1,
ftol_abs = -1,
maxeval = as.integer(-1),
maxtime = -1,
ncores = NULL,
seed = NULL
```

Arguments

)

| R | Positive integer. Number of bootstrap samples. |
|----------|---|
| path | Path to a directory to store bootstrap samples and estimates. |
| prefix | Character string. Prefix used for the file names for the bootstrap samples and estimates. |
| n | Positive integer. Number of individuals. |
| time | Positive integer. Number of time points. |
| delta_t | Numeric. Time interval. The default value is 1.0 with an option to use a numeric value for the discretized state space model parameterization of the linear stochastic differential equation model. |
| mu0 | Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. |
| sigma0_l | Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance matrix of initial latent variable values $(\Sigma_{\eta 0})$. |
| alpha | Numeric vector. Vector of constant values for the dynamic model (α) . |

| beta | Numeric matrix. Transition matrix relating the values of the latent variables at the previous to the current time point (β) . |
|----------------|---|
| psi_l | Numeric matrix. Cholesky factorization (t(chol(psi))) of the covariance matrix of the process noise (Ψ) . |
| nu | Numeric vector. Vector of intercept values for the measurement model (ν) . |
| lambda | Numeric matrix. Factor loading matrix linking the latent variables to the observed variables (Λ) . |
| theta_l | Numeric matrix. Cholesky factorization ($t(chol(theta))$) of the covariance matrix of the measurement error (Θ). |
| type | Integer. State space model type. See Details for more information. |
| X | List. Each element of the list is a matrix of covariates for each individual i in n. The number of columns in each matrix should be equal to time. |
| gamma | Numeric matrix. Matrix linking the covariates to the latent variables at current time point (Γ) . |
| kappa | Numeric matrix. Matrix linking the covariates to the observed variables at current time point (κ). |
| mu0_fixed | Logical. If mu0_fixed = TRUE, fix the initial mean vector to mu0. If mu0_fixed = FALSE, mu0 is estimated. |
| sigma0_fixed | Logical. If sigma0_fixed = TRUE, fix the initial covariance matrix to tcrossprod(sigma0_1). If sigma0_fixed = FALSE, sigma0 is estimated. |
| alpha_level | Numeric vector. Significance level α . |
| optimization_f | |
| | a flag (TRUE/FALSE) indicating whether optimization is to be done. |
| hessian_flag | a flag (TRUE/FALSE) indicating whether the Hessian matrix is to be calculated. |
| verbose | a flag (TRUE/FALSE) indicating whether more detailed intermediate output during the estimation process should be printed |
| weight_flag | a flag (TRUE/FALSE) indicating whether the negative log likelihood function should be weighted by the length of the time series for each individual |
| debug_flag | a flag (TRUE/FALSE) indicating whether users want additional dynr output that can be used for diagnostic purposes |
| perturb_flag | a flag (TRUE/FLASE) indicating whether to perturb the latent states during estimation. Only useful for ensemble forecasting. |
| xtol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| stopval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ftol_rel | Stopping criteria option for parameter optimization. See dynr.model () for more details. |
| ftol_abs | Stopping criteria option for parameter optimization. See dynr.model () for more details. |
| maxeval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |

maxtime Stopping criteria option for parameter optimization. See dynr::dynr.model()

for more details.

ncores Positive integer. Number of cores to use. If ncores = NULL, use a single core.

Consider using multiple cores when number of bootstrap samples R is a large

value.

seed Random seed.

Details

Type 0:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad ext{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right)$$

where $\mathbf{y}_{i,t}$, $\eta_{i,t}$, and $\varepsilon_{i,t}$ are random variables and ν , Λ , and Θ are model parameters. $\mathbf{y}_{i,t}$ represents a vector of observed random variables, $\eta_{i,t}$ a vector of latent random variables, and $\varepsilon_{i,t}$ a vector of random measurement errors, at time t and individual i. ν denotes a vector of intercepts, Λ a matrix of factor loadings, and Θ the covariance matrix of ε .

An alternative representation of the measurement error is given by

$$oldsymbol{arepsilon}_{i,t} = oldsymbol{\Theta}^{rac{1}{2}} \mathbf{z}_{i,t}, \quad ext{with} \quad \mathbf{z}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \mathbf{I}
ight)$$

where $\mathbf{z}_{i,t}$ is a vector of independent standard normal random variables and $\left(\Theta^{\frac{1}{2}}\right)\left(\Theta^{\frac{1}{2}}\right)' = \Theta$. The dynamic structure is given by

$$\boldsymbol{\eta}_{i,t} = \boldsymbol{\alpha} + \boldsymbol{\beta} \boldsymbol{\eta}_{i,t-1} + \boldsymbol{\zeta}_{i,t}, \text{ with } \boldsymbol{\zeta}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Psi}\right)$$

where $\eta_{i,t}$, $\eta_{i,t-1}$, and $\zeta_{i,t}$ are random variables, and α , β , and Ψ are model parameters. Here, $\eta_{i,t}$ is a vector of latent variables at time t and individual i, $\eta_{i,t-1}$ represents a vector of latent variables at time t-1 and individual i, and $\zeta_{i,t}$ represents a vector of dynamic noise at time t and individual i. α denotes a vector of intercepts, β a matrix of autoregression and cross regression coefficients, and Ψ the covariance matrix of $\zeta_{i,t}$.

An alternative representation of the dynamic noise is given by

$$oldsymbol{\zeta}_{i,t} = oldsymbol{\Psi}^{rac{1}{2}} oldsymbol{\mathbf{z}}_{i,t}, \quad ext{with} \quad oldsymbol{\mathbf{z}}_{i,t} \sim \mathcal{N}\left(oldsymbol{0}, oldsymbol{\mathbf{I}}
ight)$$

where
$$\left(\Psi^{rac{1}{2}}
ight)\left(\Psi^{rac{1}{2}}
ight)'=\Psi.$$

Type 1:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{arepsilon}_{i,t}, \quad ext{with} \quad \boldsymbol{arepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, oldsymbol{\Theta}
ight).$$

The dynamic structure is given by

$$oldsymbol{\eta}_{i,t} = oldsymbol{lpha} + oldsymbol{eta} oldsymbol{\eta}_{i,t-1} + oldsymbol{\Gamma} \mathbf{x}_{i,t} + oldsymbol{\zeta}_{i,t}, \quad ext{with} \quad oldsymbol{\zeta}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, oldsymbol{\Psi}
ight)$$

where $\mathbf{x}_{i,t}$ represents a vector of covariates at time t and individual i, and Γ the coefficient matrix linking the covariates to the latent variables.

Type 2:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \kappa \mathbf{x}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right)$$

where κ represents the coefficient matrix linking the covariates to the observed variables. The dynamic structure is given by

$$oldsymbol{\eta}_{i,t} = oldsymbol{lpha} + oldsymbol{eta} oldsymbol{\eta}_{i,t-1} + \Gamma \mathbf{x}_{i,t} + oldsymbol{\zeta}_{i,t}, \quad ext{with} \quad oldsymbol{\zeta}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, oldsymbol{\Psi}
ight).$$

Value

Returns an object of class bootstatespace which is a list with the following elements:

call Function call.

args Function arguments.

thetahatstar Sampling distribution of $\hat{\theta}$.

vcov Sampling variance-covariance matrix of $\hat{\theta}$.

est Vector of estimated $\hat{\theta}$.

fun Function used ("PBSSMFixed").

method Bootstrap method used ("parametric").

Author(s)

Ivan Jacob Agaloos Pesigan

References

Chow, S.-M., Ho, M. R., Hamaker, E. L., & Dolan, C. V. (2010). Equivalence and differences between structural equation modeling and state-space modeling techniques. *Structural Equation Modeling: A Multidisciplinary Journal*, 17(2), 303–332. doi:10.1080/10705511003661553

See Also

Other Bootstrap for State Space Models Functions: PBSSMLinSDEFixed(), PBSSMOUFixed(), PBSSMVARFixed()

Examples

```
# prepare parameters
set.seed(42)
## number of individuals
n <- 5
## time points
time <- 50
delta_t <- 1
## dynamic structure
p <- 3
mu0 <- rep(x = 0, times = p)
sigma0 <- 0.001 * diag(p)</pre>
```

```
sigma0_l <- t(chol(sigma0))</pre>
alpha < - rep(x = 0, times = p)
beta <- 0.50 * diag(p)
psi <- 0.001 * diag(p)
psi_l <- t(chol(psi))</pre>
## measurement model
k <- 3
nu \leftarrow rep(x = 0, times = k)
lambda <- diag(k)</pre>
theta <- 0.001 * diag(k)
theta_1 \leftarrow t(chol(theta))
path <- tempdir()</pre>
pb <- PBSSMFixed(</pre>
  R = 10L, # use at least 1000 in actual research
  path = path,
  prefix = "ssm",
  n = n,
  time = time,
  delta_t = delta_t,
  mu0 = mu0,
  sigma0_1 = sigma0_1,
  alpha = alpha,
  beta = beta,
  psi_l = psi_l,
  nu = nu,
  lambda = lambda,
  theta_l = theta_l,
  type = 0,
  ncores = 1, # consider using multiple cores
  seed = 42
)
print(pb)
summary(pb)
confint(pb)
vcov(pb)
coef(pb)
print(pb, type = "bc") # bias-corrected
summary(pb, type = "bc")
confint(pb, type = "bc")
```

PBSSMLinSDEFixed

Parametric Bootstrap for the Linear Stochastic Differential Equation Model using a State Space Model Parameterization (Fixed Parameters)

Description

This function simulates data from a linear stochastic differential equation model using a state-space model parameterization and fits the model using the dynr package. The process is repeated R times. It assumes that the parameters remain constant across individuals and over time. At the moment, the function only supports type = \emptyset .

Usage

```
PBSSMLinSDEFixed(
  R,
 path,
 prefix,
  time,
  delta_t = 0.1,
 mu0,
  sigma0_1,
  iota,
  phi,
  sigma_1,
  nu,
  lambda,
  theta_1,
  type = 0,
  x = NULL
  gamma = NULL,
  kappa = NULL,
 mu0_fixed = FALSE,
  sigma0_fixed = FALSE,
  alpha_level = 0.05,
  optimization_flag = TRUE,
  hessian_flag = FALSE,
  verbose = FALSE,
 weight_flag = FALSE,
  debug_flag = FALSE,
  perturb_flag = FALSE,
  xtol_rel = 1e-07,
  stopval = -9999,
  ftol_rel = -1,
  ftol_abs = -1,
 maxeval = as.integer(-1),
 maxtime = -1,
 ncores = NULL,
  seed = NULL
)
```

Arguments

| R | Positive integer. Number of bootstrap samples. |
|----------------|---|
| path | Path to a directory to store bootstrap samples and estimates. |
| prefix | Character string. Prefix used for the file names for the bootstrap samples and estimates. |
| n | Positive integer. Number of individuals. |
| time | Positive integer. Number of time points. |
| delta_t | Numeric. Time interval (Δ_t) . |
| mu0 | Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. |
| sigma0_l | Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance matrix of initial latent variable values $(\Sigma_{\eta 0})$. |
| iota | Numeric vector. An unobserved term that is constant over time (ι). |
| phi | Numeric matrix. The drift matrix which represents the rate of change of the solution in the absence of any random fluctuations (Φ) . |
| sigma_l | Numeric matrix. Cholesky factorization (t(chol(sigma))) of the covariance matrix of volatility or randomness in the process (Σ). |
| nu | Numeric vector. Vector of intercept values for the measurement model (ν) . |
| lambda | Numeric matrix. Factor loading matrix linking the latent variables to the observed variables (Λ) . |
| theta_l | Numeric matrix. Cholesky factorization ($t(chol(theta))$) of the covariance matrix of the measurement error (Θ). |
| type | Integer. State space model type. See Details for more information. |
| X | List. Each element of the list is a matrix of covariates for each individual i in n. The number of columns in each matrix should be equal to time. |
| gamma | Numeric matrix. Matrix linking the covariates to the latent variables at current time point (Γ) . |
| kappa | Numeric matrix. Matrix linking the covariates to the observed variables at current time point (κ). |
| mu0_fixed | Logical. If mu0_fixed = TRUE, fix the initial mean vector to mu0. If mu0_fixed = FALSE, mu0 is estimated. |
| sigma0_fixed | Logical. If sigma0_fixed = TRUE, fix the initial covariance matrix to tcrossprod(sigma0_1). If sigma0_fixed = FALSE, sigma0 is estimated. |
| alpha_level | Numeric vector. Significance level α . |
| optimization_f | |
| | a flag (TRUE/FALSE) indicating whether optimization is to be done. |
| hessian_flag | a flag (TRUE/FALSE) indicating whether the Hessian matrix is to be calculated. |
| verbose | a flag (TRUE/FALSE) indicating whether more detailed intermediate output during the estimation process should be printed |
| weight_flag | a flag (TRUE/FALSE) indicating whether the negative log likelihood function should be weighted by the length of the time series for each individual |

| debug_flag | a flag (TRUE/FALSE) indicating whether users want additional dynr output that can be used for diagnostic purposes |
|--------------|---|
| perturb_flag | a flag (TRUE/FLASE) indicating whether to perturb the latent states during estimation. Only useful for ensemble forecasting. |
| xtol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| stopval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ftol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ftol_abs | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| maxeval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| maxtime | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ncores | Positive integer. Number of cores to use. If ncores = NULL, use a single core. Consider using multiple cores when number of bootstrap samples R is a large value. |
| seed | Random seed. |

Details

Type 0:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad ext{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right)$$

where $\mathbf{y}_{i,t}$, $\boldsymbol{\eta}_{i,t}$, and $\boldsymbol{\varepsilon}_{i,t}$ are random variables and $\boldsymbol{\nu}$, $\boldsymbol{\Lambda}$, and $\boldsymbol{\Theta}$ are model parameters. $\mathbf{y}_{i,t}$ represents a vector of observed random variables, $\boldsymbol{\eta}_{i,t}$ a vector of latent random variables, and $\boldsymbol{\varepsilon}_{i,t}$ a vector of random measurement errors, at time t and individual i. $\boldsymbol{\nu}$ denotes a vector of intercepts, $\boldsymbol{\Lambda}$ a matrix of factor loadings, and $\boldsymbol{\Theta}$ the covariance matrix of $\boldsymbol{\varepsilon}$.

An alternative representation of the measurement error is given by

$$oldsymbol{arepsilon}_{i,t} = oldsymbol{\Theta}^{rac{1}{2}} \mathbf{z}_{i,t}, \quad ext{with} \quad \mathbf{z}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \mathbf{I}
ight)$$

where $\mathbf{z}_{i,t}$ is a vector of independent standard normal random variables and $\left(\Theta^{\frac{1}{2}}\right)\left(\Theta^{\frac{1}{2}}\right)' = \Theta$. The dynamic structure is given by

$$\mathrm{d}oldsymbol{\eta}_{i,t} = \left(oldsymbol{\iota} + oldsymbol{\Phi}oldsymbol{\eta}_{i,t}
ight)\mathrm{d}t + oldsymbol{\Sigma}^{rac{1}{2}}\mathrm{d}\mathbf{W}_{i,t}$$

where ι is a term which is unobserved and constant over time, Φ is the drift matrix which represents the rate of change of the solution in the absence of any random fluctuations, Σ is the matrix of volatility or randomness in the process, and $\mathrm{d}W$ is a Wiener process or Brownian motion, which represents random fluctuations.

Type 1:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right).$$

The dynamic structure is given by

$$\mathrm{d} \boldsymbol{\eta}_{i,t} = \left(\boldsymbol{\iota} + \boldsymbol{\Phi} \boldsymbol{\eta}_{i,t} \right) \mathrm{d} t + \boldsymbol{\Gamma} \mathbf{x}_{i,t} + \boldsymbol{\Sigma}^{\frac{1}{2}} \mathrm{d} \mathbf{W}_{i,t}$$

where $\mathbf{x}_{i,t}$ represents a vector of covariates at time t and individual i, and Γ the coefficient matrix linking the covariates to the latent variables.

Type 2:

The measurement model is given by

$$\mathbf{y}_{i,t} = \mathbf{\nu} + \mathbf{\Lambda} \boldsymbol{\eta}_{i,t} + \kappa \mathbf{x}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad ext{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \mathbf{\Theta}\right)$$

where κ represents the coefficient matrix linking the covariates to the observed variables.

The dynamic structure is given by

$$d\boldsymbol{\eta}_{i:t} = (\boldsymbol{\iota} + \boldsymbol{\Phi} \boldsymbol{\eta}_{i:t}) dt + \Gamma \mathbf{x}_{i:t} + \boldsymbol{\Sigma}^{\frac{1}{2}} d\mathbf{W}_{i:t}.$$

State Space Parameterization:

The state space parameters as a function of the linear stochastic differential equation model parameters are given by

$$\boldsymbol{\beta}_{\Delta t_{l}} = \exp\left(\Delta t \boldsymbol{\Phi}\right)$$

$$oldsymbol{lpha}_{\Delta t_{L_{z}}} = oldsymbol{\Phi}^{-1} \left(oldsymbol{eta} - \mathbf{I}_{p}
ight) oldsymbol{\iota}$$

$$\operatorname{vec}\left(\mathbf{\Psi}_{\Delta t_{l_{i}}}\right) = \left[\left(\mathbf{\Phi} \otimes \mathbf{I}_{p}\right) + \left(\mathbf{I}_{p} \otimes \mathbf{\Phi}\right)\right] \left[\exp\left(\left[\left(\mathbf{\Phi} \otimes \mathbf{I}_{p}\right) + \left(\mathbf{I}_{p} \otimes \mathbf{\Phi}\right)\right] \Delta t\right) - \mathbf{I}_{p \times p}\right] \operatorname{vec}\left(\mathbf{\Sigma}\right)$$

where p is the number of latent variables and Δt is the time interval.

Value

Returns an object of class bootstatespace which is a list with the following elements:

call Function call.

args Function arguments.

thetahatstar Sampling distribution of $\hat{\theta}$.

vcov Sampling variance-covariance matrix of $\hat{\theta}$.

est Vector of estimated $\hat{\theta}$.

fun Function used ("PBSSMLinSDEFixed").

method Bootstrap method used ("parametric").

Author(s)

Ivan Jacob Agaloos Pesigan

References

Chow, S.-M., Ho, M. R., Hamaker, E. L., & Dolan, C. V. (2010). Equivalence and differences between structural equation modeling and state-space modeling techniques. *Structural Equation Modeling: A Multidisciplinary Journal*, 17(2), 303–332. doi:10.1080/10705511003661553

See Also

Other Bootstrap for State Space Models Functions: PBSSMFixed(), PBSSMOUFixed(), PBSSMVARFixed()

Examples

```
# prepare parameters
## number of individuals
n <- 5
## time points
time <- 50
delta_t <- 0.10
## dynamic structure
p <- 2
mu0 < -c(-3.0, 1.5)
sigma0 <- 0.001 * diag(p)
sigma0_l <- t(chol(sigma0))</pre>
iota <- c(0.317, 0.230)
phi <- matrix(</pre>
  data = c(
    -0.10,
    0.05,
    0.05,
    -0.10
  ),
  nrow = p
)
sigma <- matrix(</pre>
  data = c(
    2.79,
    0.06,
    0.06,
    3.27
  ),
  nrow = p
sigma_l <- t(chol(sigma))</pre>
## measurement model
k <- 2
nu \leftarrow rep(x = 0, times = k)
lambda <- diag(k)</pre>
theta <- 0.001 * diag(k)
theta_l <- t(chol(theta))</pre>
```

```
path <- tempdir()</pre>
pb <- PBSSMLinSDEFixed(</pre>
 R = 10L, # use at least 1000 in actual research
 path = path,
 prefix = "lse",
 n = n,
 time = time,
 delta_t = delta_t,
 mu0 = mu0,
 sigma0_1 = sigma0_1,
 iota = iota,
 phi = phi,
 sigma_l = sigma_l,
 nu = nu,
 lambda = lambda,
 theta_l = theta_l,
 type = 0,
 ncores = 1, # consider using multiple cores
 seed = 42
)
print(pb)
summary(pb)
confint(pb)
vcov(pb)
coef(pb)
print(pb, type = "bc") # bias-corrected
summary(pb, type = "bc")
confint(pb, type = "bc")
```

PBSSMOUFixed

Parametric Bootstrap for the Ornstein-Uhlenbeck Model using a State Space Model Parameterization (Fixed Parameters)

Description

This function simulates data from a Ornstein-Uhlenbeck (OU) model using a state-space model parameterization and fits the model using the dynr package. The process is repeated R times. It assumes that the parameters remain constant across individuals and over time. At the moment, the function only supports type = \emptyset .

Usage

```
PBSSMOUFixed(
R,
path,
prefix,
```

```
n,
  time,
 delta_t = 0.1,
 mu0,
  sigma0_1,
 mu,
 phi,
 sigma_1,
 nu,
 lambda,
  theta_l,
  type = 0,
 x = NULL,
  gamma = NULL,
  kappa = NULL,
 mu0_fixed = FALSE,
  sigma0_fixed = FALSE,
  alpha_level = 0.05,
 optimization_flag = TRUE,
 hessian_flag = FALSE,
 verbose = FALSE,
 weight_flag = FALSE,
 debug_flag = FALSE,
 perturb_flag = FALSE,
 xtol_rel = 1e-07,
 stopval = -9999,
  ftol_rel = -1,
  ftol_abs = -1,
 maxeval = as.integer(-1),
 maxtime = -1,
 ncores = NULL,
  seed = NULL
)
```

Arguments

| path Path to a directory to store bootstrap samples and estimates. | R | Positive integer. Number of bootstrap samples. |
|--|----------|---|
| estimates. n Positive integer. Number of individuals. time Positive integer. Number of time points. delta_t Numeric. Time interval (Δ_t) . mu0 Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. sigma0_1 Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance | path | Path to a directory to store bootstrap samples and estimates. |
| time Positive integer. Number of time points. delta_t Numeric. Time interval (Δ_t) . mu0 Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. sigma0_1 Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance | prefix | |
| delta_t Numeric. Time interval (Δ_t) . mu0 Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. sigma0_1 Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance | n | Positive integer. Number of individuals. |
| mu0 Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. sigma0_1 Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance | time | Positive integer. Number of time points. |
| sigma0_1 Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance | delta_t | Numeric. Time interval (Δ_t) . |
| • | mu0 | Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. |
| | sigma0_l | • |

| mu | Numeric vector. The long-term mean or equilibrium level (μ) . |
|----------------|---|
| phi | Numeric matrix. The drift matrix which represents the rate of change of the solution in the absence of any random fluctuations (Φ). It also represents the rate of mean reversion, determining how quickly the variable returns to its mean. |
| sigma_l | Numeric matrix. Cholesky factorization ($t(chol(sigma))$) of the covariance matrix of volatility or randomness in the process (Σ). |
| nu | Numeric vector. Vector of intercept values for the measurement model (ν) . |
| lambda | Numeric matrix. Factor loading matrix linking the latent variables to the observed variables (Λ) . |
| theta_l | Numeric matrix. Cholesky factorization (t(chol(theta))) of the covariance matrix of the measurement error (Θ) . |
| type | Integer. State space model type. See Details for more information. |
| х | List. Each element of the list is a matrix of covariates for each individual i in n. The number of columns in each matrix should be equal to time. |
| gamma | Numeric matrix. Matrix linking the covariates to the latent variables at current time point (Γ) . |
| kappa | Numeric matrix. Matrix linking the covariates to the observed variables at current time point (κ) . |
| mu0_fixed | Logical. If mu0_fixed = TRUE, fix the initial mean vector to mu0. If mu0_fixed = FALSE, mu0 is estimated. |
| sigma0_fixed | Logical. If sigma0_fixed = TRUE, fix the initial covariance matrix to tcrossprod(sigma0_1). If sigma0_fixed = FALSE, sigma0 is estimated. |
| alpha_level | Numeric vector. Significance level α . |
| optimization_f | |
| | a flag (TRUE/FALSE) indicating whether optimization is to be done. |
| hessian_flag | a flag (TRUE/FALSE) indicating whether the Hessian matrix is to be calculated. |
| verbose | a flag (TRUE/FALSE) indicating whether more detailed intermediate output during the estimation process should be printed |
| weight_flag | a flag (TRUE/FALSE) indicating whether the negative log likelihood function should be weighted by the length of the time series for each individual |
| debug_flag | a flag (TRUE/FALSE) indicating whether users want additional dynr output that can be used for diagnostic purposes |
| perturb_flag | a flag (TRUE/FLASE) indicating whether to perturb the latent states during estimation. Only useful for ensemble forecasting. |
| xtol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| stopval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ftol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ftol_abs | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |

| maxeval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
|---------|---|
| maxtime | Stopping criteria option for parameter optimization. See dynr.model () for more details. |
| ncores | Positive integer. Number of cores to use. If ncores = NULL, use a single core. Consider using multiple cores when number of bootstrap samples R is a large value. |
| seed | Random seed. |

Details

Type 0:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad ext{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right)$$

where $\mathbf{y}_{i,t}$, $\boldsymbol{\eta}_{i,t}$, and $\boldsymbol{\varepsilon}_{i,t}$ are random variables and $\boldsymbol{\nu}$, $\boldsymbol{\Lambda}$, and $\boldsymbol{\Theta}$ are model parameters. $\mathbf{y}_{i,t}$ represents a vector of observed random variables, $\boldsymbol{\eta}_{i,t}$ a vector of latent random variables, and $\boldsymbol{\varepsilon}_{i,t}$ a vector of random measurement errors, at time t and individual i. $\boldsymbol{\nu}$ denotes a vector of intercepts, $\boldsymbol{\Lambda}$ a matrix of factor loadings, and $\boldsymbol{\Theta}$ the covariance matrix of $\boldsymbol{\varepsilon}$.

An alternative representation of the measurement error is given by

$$oldsymbol{arepsilon}_{i,t} = oldsymbol{\Theta}^{rac{1}{2}} \mathbf{z}_{i,t}, \quad ext{with} \quad \mathbf{z}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \mathbf{I}\right)$$

where $\mathbf{z}_{i,t}$ is a vector of independent standard normal random variables and $\left(\Theta^{\frac{1}{2}}\right)\left(\Theta^{\frac{1}{2}}\right)' = \Theta$. The dynamic structure is given by

$$\mathrm{d} oldsymbol{\eta}_{i,t} = oldsymbol{\Phi} \left(oldsymbol{\eta}_{i,t} - oldsymbol{\mu}
ight) \mathrm{d} t + oldsymbol{\Sigma}^{rac{1}{2}} \mathrm{d} \mathbf{W}_{i,t}$$

where μ is the long-term mean or equilibrium level, Φ is the rate of mean reversion, determining how quickly the variable returns to its mean, Σ is the matrix of volatility or randomness in the process, and dW is a Wiener process or Brownian motion, which represents random fluctuations.

Type 1:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right).$$

The dynamic structure is given by

$$d\boldsymbol{\eta}_{i,t} = \boldsymbol{\Phi} \left(\boldsymbol{\eta}_{i,t} - \boldsymbol{\mu} \right) dt + \boldsymbol{\Gamma} \mathbf{x}_{i,t} + \boldsymbol{\Sigma}^{\frac{1}{2}} d\mathbf{W}_{i,t}$$

where $\mathbf{x}_{i,t}$ represents a vector of covariates at time t and individual i, and Γ the coefficient matrix linking the covariates to the latent variables.

Type 2:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\nu} + \boldsymbol{\Lambda} \boldsymbol{\eta}_{i,t} + \kappa \mathbf{x}_{i,t} + \boldsymbol{\varepsilon}_{i,t}, \quad \text{with} \quad \boldsymbol{\varepsilon}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, \boldsymbol{\Theta}\right)$$

where κ represents the coefficient matrix linking the covariates to the observed variables. The dynamic structure is given by

$$\mathrm{d} oldsymbol{\eta}_{i,t} = oldsymbol{\Phi} \left(oldsymbol{\eta}_{i,t} - oldsymbol{\mu}
ight) \mathrm{d} t + oldsymbol{\Gamma} \mathbf{x}_{i,t} + oldsymbol{\Sigma}^{rac{1}{2}} \mathrm{d} \mathbf{W}_{i,t}.$$

The OU model as a linear stochastic differential equation model:

The OU model is a first-order linear stochastic differential equation model in the form of

$$\mathrm{d}\eta_{i,t}=\left(\iota+\Phi\eta_{i,t}
ight)\mathrm{d}t+\mathbf{\Sigma}^{rac{1}{2}}\mathrm{d}\mathbf{W}_{i,t}$$
 where $\pmb{\mu}=-\Phi^{-1}\iota$ and, equivalently $\pmb{\iota}=-\Phi\pmb{\mu}.$

Value

Returns an object of class bootstatespace which is a list with the following elements:

call Function call.

args Function arguments.

thetahatstar Sampling distribution of $\hat{\theta}$.

vcov Sampling variance-covariance matrix of $\hat{\theta}$.

est Vector of estimated $\hat{\theta}$.

fun Function used ("PBSSMOUFixed").

method Bootstrap method used ("parametric").

Author(s)

Ivan Jacob Agaloos Pesigan

References

Chow, S.-M., Ho, M. R., Hamaker, E. L., & Dolan, C. V. (2010). Equivalence and differences between structural equation modeling and state-space modeling techniques. *Structural Equation Modeling: A Multidisciplinary Journal*, 17(2), 303–332. doi:10.1080/10705511003661553

See Also

Other Bootstrap for State Space Models Functions: PBSSMFixed(), PBSSMLinSDEFixed(), PBSSMVARFixed()

Examples

```
# prepare parameters
## number of individuals
n <- 5
## time points
time <- 50
delta_t <- 0.10
## dynamic structure
p <- 2
mu0 <- c(-3.0, 1.5)
sigma0 <- 0.001 * diag(p)
sigma0_1 <- t(chol(sigma0))
mu <- c(5.76, 5.18)
phi <- matrix(
   data = c(</pre>
```

```
-0.10,
    0.05,
    0.05,
    -0.10
  ),
  nrow = p
sigma <- matrix(</pre>
  data = c(
    2.79,
    0.06,
    0.06,
    3.27
  ),
  nrow = p
)
sigma_l <- t(chol(sigma))</pre>
## measurement model
k <- 2
nu \leftarrow rep(x = 0, times = k)
lambda <- diag(k)</pre>
theta <- 0.001 * diag(k)
theta_l <- t(chol(theta))</pre>
path <- tempdir()</pre>
pb <- PBSSMOUFixed(</pre>
  R = 10L, # use at least 1000 in actual research
  path = path,
  prefix = "ou",
  n = n,
  time = time,
  delta_t = delta_t,
  mu0 = mu0,
  sigma0_1 = sigma0_1,
  mu = mu,
  phi = phi,
  sigma_l = sigma_l,
  nu = nu,
  lambda = lambda,
  theta_1 = theta_1,
  type = 0,
  ncores = 1, # consider using multiple cores
  seed = 42
)
print(pb)
summary(pb)
confint(pb)
vcov(pb)
coef(pb)
print(pb, type = "bc") # bias-corrected
summary(pb, type = "bc")
confint(pb, type = "bc")
```

PBSSMVARFixed

Parametric Bootstrap for the Vector Autoregressive Model (Fixed Parameters)

Description

This function simulates data from a vector autoregressive model using a state-space model parameterization and fits the model using the dynr package. The process is repeated R times. It assumes that the parameters remain constant across individuals and over time. At the moment, the function only supports type = \emptyset .

Usage

```
PBSSMVARFixed(
 R,
 path,
  prefix,
  n,
  time,
 mu0,
  sigma0_l,
  alpha,
  beta,
  psi_l,
  type = 0,
  x = NULL
  gamma = NULL,
 mu0_fixed = FALSE,
  sigma0_fixed = FALSE,
  alpha_level = 0.05,
  optimization_flag = TRUE,
  hessian_flag = FALSE,
  verbose = FALSE,
  weight_flag = FALSE,
  debug_flag = FALSE,
  perturb_flag = FALSE,
  xtol_rel = 1e-07,
  stopval = -9999,
  ftol_rel = -1,
  ftol_abs = -1,
 maxeval = as.integer(-1),
 maxtime = -1,
 ncores = NULL,
  seed = NULL
)
```

Arguments

| R | Positive integer. Number of bootstrap samples. |
|----------------|---|
| path | Path to a directory to store bootstrap samples and estimates. |
| prefix | Character string. Prefix used for the file names for the bootstrap samples and estimates. |
| n | Positive integer. Number of individuals. |
| time | Positive integer. Number of time points. |
| mu0 | Numeric vector. Mean of initial latent variable values $(\mu_{\eta 0})$. |
| sigma0_l | Numeric matrix. Cholesky factorization (t(chol(sigma0))) of the covariance matrix of initial latent variable values ($\Sigma_{\eta 0}$). |
| alpha | Numeric vector. Vector of constant values for the dynamic model (α) . |
| beta | Numeric matrix. Transition matrix relating the values of the latent variables at the previous to the current time point (β) . |
| psi_l | Numeric matrix. Cholesky factorization (t(chol(psi))) of the covariance matrix of the process noise (Ψ). |
| type | Integer. State space model type. See Details for more information. |
| X | List. Each element of the list is a matrix of covariates for each individual i in n. The number of columns in each matrix should be equal to time. |
| gamma | Numeric matrix. Matrix linking the covariates to the latent variables at current time point (Γ) . |
| mu0_fixed | Logical. If mu0_fixed = TRUE, fix the initial mean vector to mu0. If mu0_fixed = FALSE, mu0 is estimated. |
| sigma0_fixed | Logical. If sigma0_fixed = TRUE, fix the initial covariance matrix to tcrossprod(sigma0_1). If sigma0_fixed = FALSE, sigma0 is estimated. |
| alpha_level | Numeric vector. Significance level α . |
| optimization_f | |
| | a flag (TRUE/FALSE) indicating whether optimization is to be done. |
| hessian_flag | a flag (TRUE/FALSE) indicating whether the Hessian matrix is to be calculated. |
| verbose | a flag (TRUE/FALSE) indicating whether more detailed intermediate output during the estimation process should be printed |
| weight_flag | a flag (TRUE/FALSE) indicating whether the negative log likelihood function should be weighted by the length of the time series for each individual |
| debug_flag | a flag (TRUE/FALSE) indicating whether users want additional dynr output that can be used for diagnostic purposes |
| perturb_flag | a flag (TRUE/FLASE) indicating whether to perturb the latent states during estimation. Only useful for ensemble forecasting. |
| xtol_rel | Stopping criteria option for parameter optimization. See dynr.model () for more details. |
| stopval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |

| ftol_rel | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
|----------|---|
| ftol_abs | Stopping criteria option for parameter optimization. See dynr.model () for more details. |
| maxeval | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| maxtime | Stopping criteria option for parameter optimization. See dynr::dynr.model() for more details. |
| ncores | Positive integer. Number of cores to use. If ncores = NULL, use a single core. Consider using multiple cores when number of bootstrap samples R is a large value. |
| seed | Random seed. |

Details

Type 0:

The measurement model is given by

$$\mathbf{y}_{i,t} = oldsymbol{\eta}_{i,t}$$

where $\mathbf{y}_{i,t}$ represents a vector of observed variables and $\boldsymbol{\eta}_{i,t}$ a vector of latent variables for individual i and time t. Since the observed and latent variables are equal, we only generate data from the dynamic structure.

The dynamic structure is given by

$$oldsymbol{\eta}_{i,t} = oldsymbol{lpha} + oldsymbol{eta} oldsymbol{\eta}_{i,t-1} + oldsymbol{\zeta}_{i,t}, \quad ext{with} \quad oldsymbol{\zeta}_{i,t} \sim \mathcal{N}\left(oldsymbol{0}, oldsymbol{\Psi}
ight)$$

where $\eta_{i,t}$, $\eta_{i,t-1}$, and $\zeta_{i,t}$ are random variables, and α , β , and Ψ are model parameters. Here, $\eta_{i,t}$ is a vector of latent variables at time t and individual i, $\eta_{i,t-1}$ represents a vector of latent variables at time t-1 and individual i, and $\zeta_{i,t}$ represents a vector of dynamic noise at time t and individual i. α denotes a vector of intercepts, β a matrix of autoregression and cross regression coefficients, and Ψ the covariance matrix of $\zeta_{i,t}$.

An alternative representation of the dynamic noise is given by

$$oldsymbol{\zeta}_{i,t} = oldsymbol{\Psi}^{rac{1}{2}} oldsymbol{\mathbf{z}}_{i,t}, \quad ext{with} \quad oldsymbol{\mathbf{z}}_{i,t} \sim \mathcal{N}\left(oldsymbol{0}, oldsymbol{\mathbf{I}}
ight)$$

where
$$\left(\Psi^{rac{1}{2}}
ight)\left(\Psi^{rac{1}{2}}
ight)'=\Psi.$$

Type 1:

The measurement model is given by

$$\mathbf{y}_{i,t} = \boldsymbol{\eta}_{i,t}.$$

The dynamic structure is given by

$$oldsymbol{\eta}_{i,t} = oldsymbol{lpha} + oldsymbol{eta} oldsymbol{\eta}_{i,t-1} + oldsymbol{\Gamma} \mathbf{x}_{i,t} + oldsymbol{\zeta}_{i,t}, \quad ext{with} \quad oldsymbol{\zeta}_{i,t} \sim \mathcal{N}\left(\mathbf{0}, oldsymbol{\Psi}
ight)$$

where $\mathbf{x}_{i,t}$ represents a vector of covariates at time t and individual i, and Γ the coefficient matrix linking the covariates to the latent variables.

Value

Returns an object of class bootstatespace which is a list with the following elements:

```
call Function call.

args Function arguments.

thetahatstar Sampling distribution of \hat{\theta}.

vcov Sampling variance-covariance matrix of \hat{\theta}.

est Vector of estimated \hat{\theta}.

fun Function used ("PBSSMVARFixed").

method Bootstrap method used ("parametric").
```

Author(s)

Ivan Jacob Agaloos Pesigan

References

Chow, S.-M., Ho, M. R., Hamaker, E. L., & Dolan, C. V. (2010). Equivalence and differences between structural equation modeling and state-space modeling techniques. *Structural Equation Modeling: A Multidisciplinary Journal*, 17(2), 303–332. doi:10.1080/10705511003661553

See Also

Other Bootstrap for State Space Models Functions: PBSSMFixed(), PBSSMLinSDEFixed(), PBSSMOUFixed()

Examples

```
# prepare parameters
## number of individuals
n <- 5
## time points
time <- 50
## dynamic structure
p <- 3
mu0 < -rep(x = 0, times = p)
sigma0 < -0.001 * diag(p)
sigma0_l <- t(chol(sigma0))</pre>
alpha < - rep(x = 0, times = p)
beta <- 0.50 * diag(p)
psi <- 0.001 * diag(p)
psi_l <- t(chol(psi))</pre>
path <- tempdir()</pre>
pb <- PBSSMVARFixed(</pre>
  R = 10L, # use at least 1000 in actual research
  path = path,
  prefix = "var",
  n = n,
```

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```
time = time,
  mu0 = mu0,
  sigma0_1 = sigma0_1,
  alpha = alpha,
  beta = beta,
  psi_l = psi_l,
  type = 0,
  ncores = 1, # consider using multiple cores
  seed = 42
)
print(pb)
summary(pb)
confint(pb)
vcov(pb)
coef(pb)
print(pb, type = "bc") # bias-corrected
summary(pb, type = "bc")
confint(pb, type = "bc")
```

Description

Print Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
print(x, alpha = NULL, type = "pc", digits = 4, ...)
```

Arguments

| X | Object of Class bootstatespace. |
|--------|---|
| alpha | Numeric vector. Significance level α . If alpha = NULL, use the argument alpha used in x. |
| type | Charater string. Confidence interval type, that is, type = "pc" for percentile; type = "bc" for bias corrected. |
| digits | Digits to print. |
| | additional arguments. |

Value

Prints a matrix of estimates, standard errors, number of bootstrap replications, and confidence intervals.

Author(s)

Ivan Jacob Agaloos Pesigan

 $\verb|summary.bootstatespace| \\$

Summary Method for an Object of Class bootstatespace

Description

Summary Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
summary(object, alpha = NULL, type = "pc", digits = 4, ...)
```

Arguments

| object | Object of Class bootstatespace. |
|--------|---|
| alpha | Numeric vector. Significance level $\alpha.$ If alpha = NULL, use the argument alpha used in object. |
| type | Charater string. Confidence interval type, that is, type = "pc" for percentile; type = "bc" for bias corrected. |
| digits | Digits to print. |
| | additional arguments. |

Value

Returns a matrix of estimates, standard errors, number of bootstrap replications, and confidence intervals.

Author(s)

Ivan Jacob Agaloos Pesigan

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 ${\tt vcov.bootstatespace}$

Sampling Variance-Covariance Matrix Method for an Object of Class bootstatespace

Description

Sampling Variance-Covariance Matrix Method for an Object of Class bootstatespace

Usage

```
## S3 method for class 'bootstatespace'
vcov(object, ...)
```

Arguments

 $object \qquad Object \ of \ Class \ bootstate space.$

... additional arguments.

Value

Returns the variance-covariance matrix of estimates.

Author(s)

Ivan Jacob Agaloos Pesigan

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