

Package ‘HeckmanStan’

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Type Package

Title Heckman Selection Models Based on Bayesian Analysis

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Description Implements Heckman selection models using a Bayesian approach via 'Stan' and compares the performance of normal, Student's t, and contaminated normal distributions in addressing complexities and selection bias (Heeju Lim, Victor E. Lachos, and Victor H. Lachos, Bayesian analysis of flexible Heckman selection models using Hamiltonian Monte Carlo, 2025, under submission).

Imports rstan (>= 2.26.23), mvtnorm (>= 1.2-3), loo, stats

License GPL-3

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NeedsCompilation no

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Description

'geraHeckman()' generates a random sample from the Heckman selection model (Normal, Student-t or CN).

Usage

```
geraHeckman(x, w, beta, gamma, sigma2, rho, nu, family = "T")
```

Arguments

x	A covariate matrix for the response y.
w	A covariate matrix for the missing indicator cc.
beta	Values for the beta vector.
gamma	Values for the gamma vector.
sigma2	Value for the variance.
rho	Value for the dependence between the response and missing value.
nu	When using the t- distribution, the initial value for the degrees of freedom.
family	The distribution family to be used (Normal, T, or CN).

Value

Return an object with the response (y) and missing values (cc).

References

Lachos, V. H., Prates, M. O., & Dey, D. K. (2021). Heckman selection-t model: Parameter estimation via the EM-algorithm. *Journal of Multivariate Analysis*, 184, 104737.

Examples

```
n <- 100
rho <- .6
cens <- 0.25
nu <- 4
set.seed(20200527)
w <- cbind(1,runif(n,-1,1),rnorm(n))
x <- cbind(w[,1:2])

family <- "T"
c <- qt(cens, df=nu)

sigma2 <- 1
```

```

beta <- c(1,0.5)
gamma<- c(1,0.3,-.5)
gamma[1] <- -c*sqrt(sigma2)

data <- geraHeckman(x,w,beta,gamma,sigma2,rho,nu,family=family)

```

HeckmanStan	<i>Fit the Heckman Selection Stan model using the Normal, Student-t or Contaminated Normal distributions.</i>
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Description

‘HeckmanStan()’ fits the Heckman selection model using a Bayesian approach to address sample selection bias.

Usage

```

HeckmanStan(
  y,
  x,
  w,
  cc,
  family = "CN",
  init = "random",
  thin = 5,
  chains = 1,
  iter = 10,
  warmup = 5
)

```

Arguments

y	A response vector.
x	A covariate matrix for the response y.
w	A covariate matrix for the missing indicator cc.
cc	A missing indicator vector (1=observed, 0=missing) .
family	The distribution family to be used (Normal, T, or CN).
init	Parameters specifies the initial values for model parameters.
thin	An Interval at which samples are retained from the MCMC process to reduce autocorrelation.
chains	The number of chains to run during the MCMC sampling. Running multiple chains is useful for checking convergence.
iter	The total number of iterations for the MCMC sampling, determining how many samples will be drawn.
warmup	The number of initial iterations that will be discarded as the algorithm stabilizes before collecting samples.

Value

An object of class HeckmanStan, which is a list containing two elements:

- `list[[1]]`: Includes inference results from the Stan model, along with EAIC and EBIC.
- `list[[2]]`: Includes the HPC confidence intervals, along with LOOIC, WAIC, and CPO.

Examples

```
#####
# Simulation
#####
library(mvtnorm)
n<- 100
w<- cbind(1,rnorm(n),rnorm(n))
x<- cbind(w[,1:2])
family="CN"
sigma2<- 1
rho<-0.7
beta<- c(1,0.5)
gamma<- c(1,0.3,-.5)
nu=c(0.1,0.1)
data<-geraHeckman(x,w,beta,gamma,sigma2,rho,nu,family=family)
y<-data$y
cc<-data$cc
# Fit Heckman Normal Stan model
fit.n_stan <- HeckmanStan(y, x, w, cc, family="Normal"
                        , thin = 5, chains = 1, iter = 10000, warmup = 1000)
qoi=c("beta","gamma","sigma_e","sigma2", "rho","EAIC","EBIC")
print(fit.n_stan[[1]],par=qoi)
print(fit.n_stan[[2]])

require(rstan)
plot(fit.n_stan[[1]], pars=qoi)
plot(fit.n_stan[[1]], plotfun="hist", pars=qoi)
plot(fit.n_stan[[1]], plotfun="trace", pars=qoi)
plot(fit.n_stan[[1]], plotfun = "rhat")
```

Description

This dataset is an extract from the 2001 Medical Expenditure Panel Survey (MEPS), providing information on ambulatory expenditures and various demographic and health-related variables. It has been used for illustrative examples by Cameron and Trivedi (2009, Chapter 16).

Usage

```
data(MEPS2001)
```

Format

A data frame with 3,328 observations on the following 22 variables.

educ Education status
age Age
income Income
female Gender
vgood Self-reported health status, very good
good Self-reported health status, good
hospexp Hospital expenditures
totchr Total number of chronic diseases
ffs Family support
dhospexp Dummy variable for hospital expenditures
age2 Age squared
agefem Interaction between age and gender
fairpoor Self-reported health status, fair or poor
year01 Year of survey
instype Type of insurance
ambexp Ambulatory expenditures
lambexp Log of ambulatory expenditures
blhisp Ethnicity
instype_s1 Insurance type, version 1
dambexp Dummy variable for ambulatory expenditures
lnambx Log-transformed ambulatory expenditures
ins Insurance status

Source

2001 Medical Expenditure Panel Survey by the Agency for Healthcare Research and Quality.

References

Cameron, C.A. and Trivedi, P.K. (2009). *Microeconometrics Using Stata*. College Station, TX: Stata Press.

Examples

```
data(MEPS2001)  
head(MEPS2001)
```

PSID1976

*Panel Study of Income Dynamics 1976 Extract***Description**

Cross-section data originating from the 1976 Panel Study of Income Dynamics (PSID). The dataset includes demographic and economic characteristics of married women and their husbands, and is commonly used for analyzing female labor force participation.

Usage

```
data(PSID1976)
```

Format

A data frame with 753 observations on the following 22 variables.

age age of the woman

city dummy for living in a city

college dummy for college education (woman)

education years of education (woman)

experience years of labor market experience

feducation father's years of education

fincome family income in 1,000s

hage husband's age

hcollege dummy for husband's college education

heducation husband's years of education

hhours husband's weekly working hours

hours woman's weekly working hours

hwage husband's log hourly wage

meducation mother's years of education

oldkids number of children older than 6

participation dummy for woman's labor force participation

repwage replacement wage (predicted wage if not employed)

tax marginal tax rate

unemp state unemployment rate

wage log hourly wage of the woman

youngkids number of children 6 or younger

References

Mroz, T. A. (1987). The sensitivity of an empirical model of married women's hours of work to economic and statistical assumptions. *Econometrica*, 55(4), 765–799.

Examples

```
data(PSID1976)  
head(PSID1976)
```

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