# Package 'SumcaVer1'

July 21, 2024

Type Package

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mspe_FH_Boot	MSPE estimation in FH model using double-phase bootstrap method. Calculate the mspe of Fay-Herriot model in SAE using double-phase bootstrap method.

## **Description**

MSPE estimation in FH model using double-phase bootstrap method. Calculate the mspe of Fay-Herriot model in SAE using double-phase bootstrap method.

#### Usage

```
mspe_FH_Boot(m, p, X, beta, A, D, B1, B2, R)
```

### **Arguments**

m	number of small areas
р	number of fixed model parameters
Χ	covariates
beta	regression coefficients
A	variance of area-specific random effects
D	sampling variance
B1	number of first-phase bootstrap method
B2	number of second-phase bootstrap method
R	number of simulation runs

#### Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.Boot1.Final: return mspe of small area predictor using the bootstrap method 1

mspe.Boot2.Final: return mspe of small area predictor using the bootstrap method 2

RB.Boot1: return relative bias (RB) of mspe of small area predictor using the bootstrap method 1 RB.Boot2: return relative bias (RB) of mspe of small area predictor using the bootstrap method 2

```
mspe_FH_Boot(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,2.5,20,20,10)
```

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mspe_FH_PR	MSPE estimation in FH model using Prasad-Rao method. Calculate the mspe of Fay-Herriot model in SAE using Prasad-Rao method.

## Description

 $MSPE\ estimation\ in\ FH\ model\ using\ Prasad-Rao\ method.\ Calculate\ the\ mspe\ of\ Fay-Herriot\ model\ in\ SAE\ using\ Prasad-Rao\ method.$ 

## Usage

```
mspe_FH_PR(m, p, X, beta, A, D, R)
```

## Arguments

m	number of small areas
р	number of fixed model parameters
X	Covariates
beta	regression coefficients
A	variance of area-specific random effects
D	sampling variance
R	number of simulation runs

## Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.PR.Final: return mspe of small area predictor using the Prasad-Rao method

RB.PR: return relative bias (RB) of mspe of small area predictor using the Prasad-Rao method

```
{\tt mspe\_FH\_PR(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,2.5,10)}
```

4 mspe\_FH\_Sumca

mspe_FH_Sumca	MSPE estimation in FH model using SUMCA method. Calculate the mspe of Fay-Herriot model in SAE using Sumca method.

## Description

MSPE estimation in FH model using SUMCA method. Calculate the mspe of Fay-Herriot model in SAE using Sumca method.

## Usage

```
mspe_FH_Sumca(m, p, X, beta, A, D, K, R)
```

## Arguments

m	number of small areas
p	number of fixed model parameters
Χ	covariates
beta	regression coefficients
A	variance of area-specific random effects
D	sampling variance
K	number of Monte Carlo for the SUMCA method
R	number of simulation runs

## Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.Sumca.Final: return mspe of small area predictor using the SUMCA method

RB.SUMCA: return relative bias (RB) of mspe of small area predictor using the SUMCA method

```
{\tt mspe\_FH\_Sumca(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,2.5,10,10)}
```

```
mspe_LOGISTIC_HealthData_BOOT
```

MSPE estimation in mixed logistic model (Health Insurance data) using bootstrap method. Calculate the mspe of mixed logistic model (Health Insurance data) using bootstrap method.

## Description

MSPE estimation in mixed logistic model (Health Insurance data) using bootstrap method. Calculate the mspe of mixed logistic model (Health Insurance data) using bootstrap method.

## Usage

```
mspe_LOGISTIC_HealthData_BOOT(
    m,
    p,
    n.new,
    y.new,
    cum.n.new,
    Xi,
    yi.tem,
    X.tem,
    county.tem,
    B
)
```

## Arguments

m	number of domains
p	number of complete model parameters
n.new	sample size of each domain
y.new	response variable
cum.n.new	Cummulaticve sum of n
Xi	covariates
yi.tem	response variable for each individual
X.tem	Individual level covariates
county.tem	county
В	number of bootstrap iterations

## Value

Par: return estimation of model parameters

Mu.hat: return prediction of domain parameters

mspe.boot: return mspe of small area (domain) predictor using the bootstrap method

sq.mspe.boot: return square root of mspe of small area predictor for non-zero domains using the bootstrap method

## **Examples**

```
\label{eq:mspe_logistic_HealthData_Boot(20,3,c(2,1,2,2,1,2,3,1,1,3,1,3,2,3,3,1,2,1,3,3),c(3,4,2,2,3,3,4,3,4,1,4,1,3,5,4,7,1,3,1,2),c(2,3,5,7,8,10,13,14,15,18,19,22,24,27,30,31,33,34,37,40),\\ matrix(runif(60,0,1),nrow=20,byrow=TRUE),sample(c(0,1),replace=TRUE,40),\\ matrix(c(runif(40,7,10),runif(40,14,22),runif(40,2,4)),nrow=40,byrow=FALSE),\\ rep(1:20,each=2),10)
```

```
mspe_LOGISTIC_HealthData_JLW
```

MSPE estimation in mixed logistic model (Health Insurance data) using jackknife method. Calculate the mspe of mixed logistic model (Health Insurance data) using jackknife method.

## **Description**

MSPE estimation in mixed logistic model (Health Insurance data) using jackknife method. Calculate the mspe of mixed logistic model (Health Insurance data) using jackknife method.

#### Usage

```
mspe_LOGISTIC_HealthData_JLW(
   m,
   p,
   n.new,
   y.new,
   Xi,
   yi.tem,
   cum.n.new,
   county.tem,
   X.tem
)
```

#### **Arguments**

```
m number of domains
p number of complete model parameters
n.new sample size of each domain
y.new response variable
Xi covariates for each domain
```

```
yi.tem response variable for each individual
```

cum.n.new Cummulative sum of n

county.tem county

X. tem Individual level covariates

#### Value

Par: return estimation of model parameters

Mu.hat: return prediction of domain parameters

mspe.JLW: return mspe of small area (domain) predictor using the jackknife method

sq.mspe.JLW: return square root of mspe of small area predictor for non-zero domains using the jackknife method

## **Examples**

```
mspe_LOGISTIC_HealthData_SUMCA
```

MSPE estimation in mixed logistic model (Health Insurance data) using SUMCA method. Calculate the mspe of mixed logistic model (Health Insurance data) using SUMCA method.

#### **Description**

MSPE estimation in mixed logistic model (Health Insurance data) using SUMCA method. Calculate the mspe of mixed logistic model (Health Insurance data) using SUMCA method.

```
mspe_LOGISTIC_HealthData_SUMCA(
    m,
    p,
    n.new,
    y.new,
    Xi,
    cum.n.new,
    yi.tem,
    X.tem,
    county.tem,
    K
)
```

p number of complete model parameters

n.new sample size of each domain

y.new response variable

Xi covariates

cum.n.new Cummulative sum of n

yi.tem response variable for each individual

X. tem Individual level covariates

county.tem county

K number of Monte Carlo for the SUMCA method

#### Value

Par: return estimation of model parameters

Mu.hat: return prediction of domain parameters

mspe.Sumca: return mspe of small area (domain) predictor using the SUMCA method

sq.mspe.Sumca: return square root of mspe of small area predictor for non-zero domains using the SUMCA method

## **Examples**

```
\label{eq:mspe_logistic_HealthData_SUMCA} $(20,3,c(2,1,2,2,1,2,3,1,1,3,1,3,2,3,3,1,2,1,3,3),c(3,4,2,2,3,3,4,3,4,1,4,1,3,5,4,7,1,3,1,2), \\ matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(2,3,5,7,8,10,13,14,15,18,19,22,24,27,30,31,33,34,37,40),sample(c(0,1),replace=TRUE,40), \\ matrix(c(runif(40,7,10),runif(40,14,22),runif(40,2,4)),nrow=40,byrow=FALSE),rep(1:20,each=2),10)
```

mspe\_MS\_LOGISTIC\_JLW

Model selection MSPE estimation in mixed logistic model using jackknife method. Calculate the model selection mspe of mixed logistic model using jackknife method.

## Description

Model selection MSPE estimation in mixed logistic model using jackknife method. Calculate the model selection mspe of mixed logistic model using jackknife method.

```
mspe_MS_LOGISTIC_JLW(m, p, ni, X, beta, A, R)
```

m	number of small areas
р	number of complete model parameters
ni	sample size of each small area
Χ	covariates for the complete model
beta	regression coefficients of the complete model
Α	variance of area-specific random effects
R	number of simulation runs

#### Value

Par1: return estimation of model parameters of the complete model

Par2: return estimation of model parameters of the reduced model

MSPE: return empirical MSPE of small area predictor

mspe.JLW: return mspe of small area predictor using the jackknife method

RB.JLW: return relative bias (RB) of mspe of small area predictor using the jackknife method

BIC: return BIC of the complete and reduced models

## **Examples**

```
\label{eq:mspe_MS_LOGISTIC_JLW} $$  \mbox{mspe_MS_LOGISTIC_JLW}(20,3,2,$$  \mbox{matrix}(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,2) $$
```

```
mspe_MS_LOGISTIC_SUMCA
```

Model selection MSPE estimation in mixed logistic model using SUMCA method. Calculate the model selection mspe of mixed logistic model using SUMCA method.

## Description

Model selection MSPE estimation in mixed logistic model using SUMCA method. Calculate the model selection mspe of mixed logistic model using SUMCA method.

```
mspe_MS_LOGISTIC_SUMCA(m, p, ni, X, beta, A, K, R)
```

m	number of small areas
р	number of complete model parameters
ni	sample size of each small area
Χ	covariates for the complete model
beta	regression coefficients of the complete model
Α	variance of area-specific random effects
K	number of Monte Carlo for the SUMCA method
R	number of simulation runs

#### Value

Par1: return estimation of model parameters of the complete model

Par2: return estimation of model parameters of the reduced model

MSPE: return empirical MSPE of small area predictor

mspe.Sumca: return mspe of small area predictor using the SUMCA method

RB.SUMCA: return relative bias (RB) of mspe of small area predictor using the SUMCA method

BIC: return BIC of the complete and reduced models

## **Examples**

```
\label{eq:mspe_MS_LOGISTIC_SUMCA(20,3,2,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,5,5)} \\
```

mspe_PMS_FH_DHM	Post model selection MSPE estimation in FH model using Datta-Hall-
	Mandal method. Calculate the post-model selection mspe of Fay-
	Herriot model using Datta-Hall-Mandal method.
	Trender model using Dania Transact memora.

## Description

Post model selection MSPE estimation in FH model using Datta-Hall-Mandal method. Calculate the post-model selection mspe of Fay-Herriot model using Datta-Hall-Mandal method.

```
mspe_PMS_FH_DHM(m, p, X, beta, A, D, R)
```

m		number of small areas
р		number of fixed model parameters
Χ		covariates
bet	a	regression coefficients
Α		variance of area-specific random effects
D		sampling variance
R		number of simulation runs

#### Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.DHM.Final: return mspe of small area predictor using the Datta-Hall-Mandal method

RB.DHM: return relative bias (RB) of mspe of small area predictor using the Datta-Hall-Mandal method

Rate: return the probability of rejection (nominal level= 0.2)

## Examples

```
mspe_PMS_FH_DHM(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),
c(1,1,1),10,2.5,10)
```

mspe\_PMS\_FH\_SUMCA

Post model selection MSPE estimation in FH model using SUMCA method. Calculate the post-model selection mspe of Fay-Herriot model using SUMCA method.

## Description

Post model selection MSPE estimation in FH model using SUMCA method. Calculate the postmodel selection mspe of Fay-Herriot model using SUMCA method.

```
mspe_PMS_FH_SUMCA(m, p, X, beta, A, D, K, R)
```

m	number of small areas
p	number of fixed model parameters
X	covariates
beta	regression coefficients
A	variance of area-specific random effects
D	sampling variance
K	number of Monte Carlo for the SUMCA method
R	number of simulation runs

#### Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.Sumca.Final: return mspe of small area predictor using the SUMCA method

RB.SUMCA: return relative bias (RB) of mspe of small area predictor using the SUMCA method

## **Examples**

```
mspe_PMS_FH_SUMCA(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),10,2.5,10,10)
```

mspe\_PMS\_Mis\_FH\_DHM

Post model selection MSPE estimation in FH model with mean mis-specification using Datta-Hall-Mandal method. Calculate the post-model selection mspe of Fay-Herriot model with mean misspecification using Datta-Hall-Mandal method.

## Description

Post model selection MSPE estimation in FH model with mean mis-specification using Datta-Hall-Mandal method. Calculate the post-model selection mspe of Fay-Herriot model with mean mis-specification using Datta-Hall-Mandal method.

```
mspe_PMS_Mis_FH_DHM(m, p, X, beta1, beta2, A, D, R)
```

m	number of small areas
р	number of fixed model parameters
X	covariates
beta1	regression coefficients
beta2	regression coefficients
Α	variance of area-specific random effects
D	sampling variance
R	number of simulation runs

#### Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.DHM.Final: return mspe of small area predictor using the Datta-Hall-Mandal method

RB.DHM: return relative bias (RB) of mspe of small area predictor using the Datta-Hall-Mandal

method

Rate: return the probability of rejection (nominal level= 0.2)

#### **Examples**

```
mspe_PMS_Mis_FH_DHM(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE),c(1,1,1),c(1,1,1),10,2.5,10)
```

mspe\_PMS\_Mis\_FH\_SUMCA Post model selection MSPE estimation in FH model with mean misspecification using SUMCA method. Calculate the post-model selection mspe of Fay-Herriot model with mean mis-specification using SUMCA method.

## Description

Post model selection MSPE estimation in FH model with mean mis-specification using SUMCA method. Calculate the post-model selection mspe of Fay-Herriot model with mean mis-specification using SUMCA method.

```
mspe_PMS_Mis_FH_SUMCA(m, p, X, beta1, beta2, A, D, K, R)
```

m	number of small	areas
111	number of sinan	arcas

p number of fixed model parameters

X covariates

beta1 regression coefficient beta2 regression coefficient

A variance of area-specific random effects

D sampling variance

K number of Monte Carlo for the SUMCA method

R number of simulation runs

#### Value

Par: return estimation of model parameters

MSPE.TRUE.Final: return empirical MSPE of small area predictor

mspe.Sumca.Final: return mspe of small area predictor using the SUMCA method

RB.SUMCA: return relative bias (RB) of mspe of small area predictor using the SUMCA method

```
\label{eq:mspe_PMS_Mis_FH_SUMCA} $$ mspe_PMS_Mis_FH_SUMCA(20,3,matrix(runif(60,0,1),nrow=20,byrow=TRUE), c(1,1,1),c(1,1,1),10,2.5,10,10) $$
```

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