Package 'msaeDB'

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Type Package

Title Difference Benchmarking for Multivariate Small Area Estimation

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Description Implements Benchmarking Method for Multivariate Small Area Estimation under Fay Herriot Model. Multivariate Small Area Estimation (MSAE) is a development of Univariate Small Area Estimation that considering the correlation among response variables and borrowing the strength from related areas and auxiliary variables to increase the effectiveness of sample size, the multivariate model in this package is based on multivariate model 1 proposed by Roberto Benavent and Domingo Morales (2016) <doi:10.1016/j.csda.2015.07.013>. Benchmarking in Small Area Estimation is a modification of Small Area Estimation model to guarantee that the aggregate weighted mean of the county predictors equals the corresponding weighted mean of survey estimates. Difference Benchmarking is the simplest benchmarking method but widely used by multiplying empirical best linear unbiased prediction (EBLUP) estimator by the common adjustment factors (J.N.K Rao and Isabel Molina, 2015).

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Encoding UTF-8

LazyData true

RoxygenNote 7.1.1

URL https://github.com/zazaperwira/msaeDB

BugReports https://github.com/zazaperwira/msaeDB/issues

Suggests knitr, rmarkdown, covr

VignetteBuilder knitr

Imports MASS, magic, stats

Depends R (>= 2.10)

NeedsCompilation no

Repository CRAN

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R topics documented:

`datamsaeDB` *Sample Data for Multivariate Small Area Estimation with Difference Benchmarking*

Description

Dataset to simulate difference benchmarking of Multivariate Fay Herriot model

This data is generated base on multivariate Fay Herriot model by these following steps:

1. Generate explanatory variables `X1` and `X2`. Take $\mu_{X1} = \mu_{X1} = 10$, σ_{X11}=1, σ_{X2}=2, and ρ_x= 1/2.
   Sampling error `e` is generated with the following σ_{e11} = 0.15, σ_{e22} = 0.25, σ_{e33} = 0.35, and ρ_e = 1/2.
   For random effect `u`, we set σ_{u11}= 0.2, σ_{u22}= 0.6, and σ_{u33}= 1.8.
   For the weight we generate `w1 w2 w3` by set the `w1 ~ U(25,30)` , `w2 ~ U(25,30)`, `w3 ~ U(25,30)`
   Calculate direct estimation `Y1 Y2 Y3` where $Y_i = X * \beta + u_i + e_i$
2. Then combine the direct estimations `Y1 Y2 Y3`, explanatory variables `X1 X2`, weights `w1 w2 w3`, and sampling varians covarians `v1 v12 v13 v2 v23 v3` in a dataframe then named as datamsaeDB

Usage

```
datamsaeDB
```

Format

A data frame with 30 rows and 14 variables:

Y1 Direct Estimation of Y1

Y2 Direct Estimation of Y2

Y3 Direct Estimation of Y3

X1 Auxiliary variable of X1

X2 Auxiliary variable of X2

w1 Known proportion of units in small areas of Y1

w2 Known proportion of units in small areas of Y2

w3 Known proportion of units in small areas of Y3

v1 Sampling Variance of Y1

v12 Sampling Covariance of Y1 and Y2

v13 Sampling Covariance of Y1 and Y3

v2 Sampling Variance of Y2

v23 Sampling Covariance of Y2 and Y3

v3 Sampling Variance of Y3

`datamsaeDBns` *Sample Data for Multivariate Small Area Estimation with Difference Benchmarking with clustering*

Description

Dataset to simulate difference benchmarking of Multivariate Fay Herriot model for non-sampled area using clustering This data is generated base on multivariate Fay Herriot model by these following steps:

1. Generate explanatory variables `X1` and `X2`. Take μ_{X1} = μ_{X1} = 10, σ_{X11}=1, σ_{X2}=2, and ρ_x= 1/2.
   Sampling error `e` is generated with the following σ_{e11} = 0.15, σ_{e22} = 0.25, σ_{e33} = 0.35, and ρ_e = 1/2.
   For random effect `u`, we set σ_{u11}= 0.2, σ_{u22}= 0.6, and σ_{u33}= 1.8.
   For the weight we generate `w1 w2 w3` by set the `w1 ~ U(25,30)` , `w2 ~ U(25,30)`, `w3 ~ U(25,30)`
   Calculate direct estimation `Y1 Y2 Y3` where $Y_i = X * \beta + u_i + e_i$
   `cl1 cl2 cl3` were obtained using K-Means clustering from the explanatory variables.
2. Then combine the direct estimations `Y1 Y2 Y3`, explanatory variables `X1 X2`, weights `w1 w2 w3`, and sampling varians covarians `v1 v12 v13 v2 v23 v3` in a data frame then named as datamsaeDB

Usage

```
datamsaeDBns
```

Format

A data frame with 30 rows and 18 variables:

clY1 cluster information of Y1

clY2 cluster information of Y2

clY3 cluster information of Y3

nonsample A column with logical values, TRUE if the area is non-sampled

Y1 Direct Estimation of Y1

Y2 Direct Estimation of Y2

Y3 Direct Estimation of Y3

X1 Auxiliary variable of X1

X2 Auxiliary variable of X2

w1 Known proportion of units in small areas of Y1

w2 Known proportion of units in small areas of Y2

w3 Known proportion of units in small areas of Y3

v1 Sampling Variance of Y1

v12 Sampling Covariance of Y1 and Y2

v13 Sampling Covariance of Y1 and Y3

v2 Sampling Variance of Y2

v23 Sampling Covariance of Y2 and Y3

v3 Sampling Variance of Y3

`msaedb` *EBLUPs under Multivariate Fay Herriot Model with Difference Benchmarking*

Description

This function produces EBLUPs, MSE, and aggregation of Multivariate SAE with Difference Benchmarking

Usage

```
msaedb(
  formula,
  vardir,
  weight,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations,if it is included in data frame so it is the vector with the name of sampling variances.if it is not, it is a data frame of sampling variance in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1)(r),var(r)
weight	Known proportion of units in small areas, where $\sum_{d=1}^{D} W_{rd} = 1$. d = 1 ... D is the number of small areas, and r = 1 ... R is the number of response variables
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

MSAE_Eblup	A dataframe with the values of the EBLUPs estimators
MSE_Eblup	A dataframe with the values of estimated mean square errors of EBLUPs estimators
randomEffect	A dataframe with the values of the random effect estimators
Rmatrix	A block diagonal matrix composed of sampling errors
fit	A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations
- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

difference_benchmarking
: a list containing the following objects:

- Estimation : A dataframe with the value of Benchmarked EBLUPs estimators
- Aggregation : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations
- MSE_DB : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators
- g.4a : First component of g4 in difference benchmarking MSE estimation formula
- g.4b : Second component of g4 in difference benchmarking MSE estimation formula

Examples

```
##load dataset
data(datamsaeDB)

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X2
#Y3 ~ X1

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X2,
               f3 = Y3~X1)
vardir = c("v1","v12","v13","v2","v23","v3")
weight = c("w1","w2","w3")
msaeDB <- msaedb(formula, vardir, weight, data=datamsaeDB)

##Do not use parameter 'data'
formula = list(f1 = datamsaeDB$Y1~datamsaeDB$X1+datamsaeDB$X2,
               f2 = datamsaeDB$Y2~datamsaeDB$X2,
               f3 = datamsaeDB$Y3~datamsaeDB$X1)
vardir = datamsaeDB[,c("v1","v12","v13","v2","v23","v3")]
weight = datamsaeDB[,c("w1","w2","w3")]
msaeDB_d <- msaedb(formula, vardir, weight)

msaeDB$MSAE_Eblup       #to see EBLUP Estimators
msaeDB$MSE_Eblup        #to see estimated MSE of EBLUP estimators
msaeDB$difference_benchmarking$Estimation   #to see Benchmarked EBLUP Estimators
msaeDB$difference_benchmarking$MSE_DB     #to see estimated MSE of Benchmarked EBLUP Estimators
msaeDB$difference_benchmarking$Aggregation  #to see the aggregation of, benchmarking.
```

`msaedbns` *EBLUPs under Multivariate Fay Herriot Model with Difference Benchmarking for non-sampled area*

Description

This function produces EBLUPs, MSE, and aggregation of Multivariate SAE with Difference Benchmarking for non-sampled area

Usage

```
msaedbns(
  formula,
  vardir,
  weight,
  cluster,
```

```
  nonsample,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations included in data frame as the vector with the name of sampling variances in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1
weight	Known proportion of units in small areas, where $\sum_{d=1}^{D} W_{rd} = 1$. d = 1 ... D is the number of small areas, and r = 1 ... R is the number of response variables
cluster	cluster information
nonsample	A column with logical values, TRUE if the area is non-sampled
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

MSAE_Eblup_sampled
: A dataframe with the values of the EBLUPs estimators for sampled areas

MSAE_Eblup_all
: A dataframe with the values of the EBLUPs estimators for all areas

MSE_Eblup_sampled
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for sampled areas

MSE_Eblup_all
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for all areas

randomEffect_sampled
: a dataframe with the values of the random effect estimators for sampled areas

randomEffect_all
: a dataframe with the values of the random effect estimators for all areas

Rmatrix_sampled
: a block diagonal matrix composed of sampling errors for sampled areas

fit
: A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations

- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

`difference_benchmarking`
a list containing the following objects:

- Estimation_sampled : A dataframe with the values of benchmarked EBLUPs estimators for sampled areas
- Estimation_all : A dataframe with the values of benchmarked EBLUPs estimators for all areas
- Aggregation_sampled : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations for sampled areas
- Aggregation_all : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations for all areas
- MSE_DB_sampled : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators for sampled areas
- MSE_DB_all : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators for all areas
- g.4a : First component of g4 in difference benchmarking MSE estimation formula
- g.4b : Second component of g4 in difference benchmarking MSE estimation formula

Examples

```
##load dataset
data(datamsaeDBns)
#Note : Make sure yout dataset does not contain NA Values
#       you can set 0 in Direct estinations and vardir for non-sampled areas.


#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X1 + X2
#Y3 ~ X1 + X2

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X1+X2,
               f3 = Y3~X1+X2)
vardir = c("v1","v12","v13","v2","v23","v3")
weight = c("w1","w2","w3")
cluster = c("clY1","clY2","clY3")
nonsample = "nonsample"
msaeDBns <- msaedbns(formula,vardir, weight,cluster, nonsample, data=datamsaeDBns)
```

msaefh *EBLUPs under Multivariate Fay Herriot Model*

Description

This function produces EBLUPs, MSE of Multivariate SAE

Usage

```
msaefh(
  formula,
  vardir,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations,if it is included in data frame so it is the vector with the name of sampling variances.if it is not, it is a data frame of sampling variance in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1)(r),var(r)
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

MSAE_Eblup	A dataframe with the values of the EBLUPs estimators
MSE_Eblup	A dataframe with the values of estimated mean square errors of EBLUPs estimators
randomEffect	A dataframe with the values of the random effect estimators
Rmatrix	A block diagonal matrix composed of sampling errors
fit	A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations

- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

Examples

```
##load dataset
data(datamsaeDB)

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X2
#Y3 ~ X1

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X2,
               f3 = Y3~X1)
vardir = c("v1","v12","v13","v2","v23","v3")
msaeFH <- msaefh(formula, vardir, data=datamsaeDB)

#Do not use parameter 'data'
formula = list(f1 = datamsaeDB$Y1~datamsaeDB$X1+datamsaeDB$X2,
               f2 = datamsaeDB$Y2~datamsaeDB$X2,
               f3 = datamsaeDB$Y3~datamsaeDB$X1)
vardir = datamsaeDB[,c("v1","v12","v13","v2","v23","v3")]
msaeFH_d <- msaefh(formula, vardir)

msaeFH$MSAE_Eblup       #to see EBLUP Estimators
msaeFH$MSE_Eblup        #to see estimated MSE of EBLUP estimators
```

`msaefhns` *EBLUPs under Multivariate Fay Herriot Model for non-sampled area*

Description

This function produces EBLUPs and MSE of Multivariate SAE with Difference Benchmarking for non-sampled area

Usage

```
msaefhns(
  formula,
  vardir,
  cluster,
  nonsample,
  samevar = FALSE,
```

```
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations included in data frame as the vector with the name of sampling variances in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1
cluster	cluster information
nonsample	A column with logical values, TRUE if the area is non-sampled
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

MSAE_Eblup_sampled
: A dataframe with the values of the EBLUPs estimators for sampled areas

MSAE_Eblup_all
: A dataframe with the values of the EBLUPs estimators for all areas

MSE_Eblup_sampled
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for sampled areas

MSE_Eblup_all
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for all areas

randomEffect_sampled
: a dataframe with the values of the random effect estimators for sampled areas

randomEffect_all
: a dataframe with the values of the random effect estimators for all areas

Rmatrix_sampled
: a block diagonal matrix composed of sampling errors for sampled areas

fit
: A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations
- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

Examples

```
##load dataset
data(datamsaeDBns)
#Note : Make sure yout dataset does not contain NA Values
#       you can set 0 in Direct estinations and vardir for non-sampled areas.

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X1 + X2
#Y3 ~ X1 + X2

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X1+X2,
               f3 = Y3~X1+X2)
vardir = c("v1","v12","v13","v2","v23","v3")
cluster = c("clY1","clY2","clY3")
nonsample = "nonsample"
msaeFHns <- msaefhns(formula,vardir, cluster, nonsample, data=datamsaeDBns)
```

---

saedb *EBLUPs under Univariate Fay Herriot Model with Difference Benchmarking*

Description

This function produces EBLUPs, MSE, and aggregation of Univariate SAE with Difference Benchmarking

Usage

```
saedb(
  formula,
  vardir,
  weight,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations,if it is included in data frame so it is the vector with the name of sampling variances.if it is not, it is a data frame of sampling variance in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1)(r),var(r)

weight	Known proportion of units in small areas, where $\sum_{d=1}^{D} W_{rd}$ = 1 . d = 1...D is the number of small areas, and r = 1...R is the number of response variables
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

SAE_Eblup	A dataframe with the values of the EBLUPs estimators
MSE_Eblup	A dataframe with the values of estimated mean square errors of EBLUPs estimators
randomEffect	A dataframe with the values of the random effect estimators
Rmatrix	A block diagonal matrix composed of sampling errors
fit	A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations
- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

difference_benchmarking
: a list containing the following objects:

- Estimation : A dataframe with the value of Benchmarked EBLUPs estimators
- Aggregation : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations
- MSE_DB : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators
- g.4a : First component of g4 in difference benchmarking MSE estimation formula
- g.4b : Second component of g4 in difference benchmarking MSE estimation formula

Examples

```
##load dataset
data(datamsaeDB)

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
```

```
#Y2 ~ X2
#Y3 ~ X1

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X2,
               f3 = Y3~X1)
vardir = c("v1","v12","v13","v2","v23","v3")
#Note : in real data for univariate SAE, if you does not have the values of covariances,
#       set covariancse as zero in the dataframe

weight = c("w1","w2","w3")
saeDB <- saedb(formula, vardir, weight, data=datamsaeDB)

#to calculate only one response variable
saeDB1 <- saedb(formula=list(f1=Y1~X1+X2),vardir ="v1", weight="w1",data=datamsaeDB )

##Do not use parameter 'data'
formula = list(f1 = datamsaeDB$Y1~datamsaeDB$X1+datamsaeDB$X2,
               f2 = datamsaeDB$Y2~datamsaeDB$X2,
               f3 = datamsaeDB$Y3~datamsaeDB$X1)
vardir = datamsaeDB[,c("v1","v12","v13","v2","v23","v3")]
#Note : in real data for univariate SAE, if you does not have the values of covariances,
#       set covariancse as zero in the dataframe
weight = datamsaeDB[,c("w1","w2","w3")]
saeDB_d <- saedb(formula, vardir, weight = weight)


saeDB$SAE_Eblup        #to see EBLUP Estimators
saeDB$MSE_Eblup        #to see estimated MSE of EBLUP estimators
saeDB$difference_benchmarking$Estimation   #to see Benchmarked EBLUP Estimators
saeDB$difference_benchmarking$MSE_DB     #to see estimated MSE of Benchmarked EBLUP Estimators
saeDB$difference_benchmarking$Aggregation  #to see the aggregation of, benchmarking
```

---

saedbns *EBLUPs under Univariate Fay Herriot Model with Difference Benchmarking for non-sampled area*

Description

This function produces EBLUPs, MSE, and aggregation of Univariate SAE with Difference Benchmarking for non-sampled area

Usage

```
saedbns(
  formula,
  vardir,
  weight,
  cluster,
  nonsample,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations included in data frame as the vector with the name of sampling variances in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1
weight	Known proportion of units in small areas, where $\sum_{d=1}^{D} W_{rd} = 1$. d = 1 ... D is the number of small areas, and r = 1 ... R is the number of response variables
cluster	cluster information
nonsample	A column with logical values, TRUE if the area is non-sampled
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

SAE_Eblup_sampled
: A dataframe with the values of the EBLUPs estimators for sampled areas

SAE_Eblup_all
: A dataframe with the values of the EBLUPs estimators for all areas

MSE_Eblup_sampled
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for sampled areas

MSE_Eblup_all
: A dataframe with the values of estimated mean square errors of EBLUPs estimators for all areas

randomEffect_sampled
: a dataframe with the values of the random effect estimators for sampled areas

randomEffect_all
: a dataframe with the values of the random effect estimators for all areas

Rmatrix_sampled
: a block diagonal matrix composed of sampling errors for sampled areas

`fit` A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations
- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

`difference_benchmarking`
a list containing the following objects:

- Estimation_sampled : A dataframe with the values of benchmarked EBLUPs estimators for sampled areas
- Estimation_all : A dataframe with the values of benchmarked EBLUPs estimators for all areas
- Aggregation_sampled : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations for sampled areas
- Aggregation_all : The aggregation of benchmarked EBLUPs estimators, EBLUPs estimators and direct estimations for all areas
- MSE_DB_sampled : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators for sampled areas
- MSE_DB_all : A dataframe with the values of estimated mean square errors of benchmarked EBLUPs estimators for all areas
- g.4a : First component of g4 in difference benchmarking MSE estimation formula
- g.4b : Second component of g4 in difference benchmarking MSE estimation formula

Examples

```
##load dataset
data(datamsaeDBns)
#Note : Make sure yout dataset does not contain NA Values
#       you can set 0 in Direct estinations and vardir for non-sampled areas.

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X1 + X2
#Y3 ~ X1 + X2

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X1+X2,
               f3 = Y3~X1+X2)
vardir = c("v1","v12","v13","v2","v23","v3")
weight = c("w1","w2","w3")
cluster = c("clY1","clY2","clY3")
```

```
nonsample = "nonsample"
saeDBns <- saedbns(formula,vardir, weight,cluster, nonsample, data=datamsaeDBns)

#to calculate only one response variable
saeDB1 <- saedbns(formula=list(f1=Y1~X1+X2),vardir ="v1", weight="w1",
         cluster = "clY1",nonsample = "nonsample",data=datamsaeDBns )
```

---

saefh *EBLUPs under Univariate Fay Herriot Model*

Description

This function produces EBLUPs, MSE of Univariate SAE

Usage

```
saefh(formula, vardir, samevar = FALSE, MAXITER = 100, PRECISION = 1e-04, data)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations,if it is included in data frame so it is the vector with the name of sampling variances.if it is not, it is a data frame of sampling variance in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1)(r),var(r)
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

SAE_Eblup	A dataframe with the values of the EBLUPs estimators
MSE_Eblup	A dataframe with the values of estimated mean square errors of EBLUPs estimators
randomEffect	A dataframe with the values of the random effect estimators
Rmatrix	A block diagonal matrix composed of sampling errors
fit	A list containing the following objects:

- method : The fitting method (this function is using "REML")
- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations

- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

Examples

```
##load dataset
data(datamsaeDB)

#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X2
#Y3 ~ X1

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X2,
               f3 = Y3~X1)
vardir = c("v1","v12","v13","v2","v23","v3")
#Note : in real data for univariate SAE, if you does not have the values of covariances,
#       set covariancse as zero in the dataframe

saeFH <- saefh(formula, vardir, data=datamsaeDB)

#to calculate only one response variable
saeFH1 <- saefh(formula=list(f1=Y1~X1+X2),vardir ="v1",data=datamsaeDB )

##Do not use parameter 'data'
formula = list(f1 = datamsaeDB$Y1~datamsaeDB$X1+datamsaeDB$X2,
               f2 = datamsaeDB$Y2~datamsaeDB$X2,
               f3 = datamsaeDB$Y3~datamsaeDB$X1)
vardir = datamsaeDB[,c("v1","v12","v13","v2","v23","v3")]
#Note : in real data for univariate SAE, if you does not have the values of covariances,
#       set covariancse as zero in the dataframe
saeFH_d <- saefh(formula, vardir)

saeFH$SAE_Eblup       #to see EBLUP Estimators
saeFH$MSE_Eblup        #to see estimated MSE of EBLUP estimators
```

---

saefhns *EBLUPs under Univariate Fay Herriot Model for non-sampled area*

Description

This function produces EBLUPs, MSE, and aggregation of Univariate SAE for non-sampled area

Usage

```
saefhns(
  formula,
  vardir,
  cluster,
  nonsample,
  samevar = FALSE,
  MAXITER = 100,
  PRECISION = 1e-04,
  data
)
```

Arguments

formula	List of formula that describe the fitted model
vardir	Sampling variances of direct estimations included in data frame as the vector with the name of sampling variances in order : var1,cov12,.,cov1r,var2,cov23,.,cov2r,.,cov(r-1
cluster	cluster information
nonsample	A column with logical values, TRUE if the area is non-sampled
samevar	Whether the variances of the data are same or not. Logical input with default FALSE
MAXITER	Maximum number of iteration in Fisher-scoring algorithm with default 100
PRECISION	Limit of Fisher-scoring convergence tolerance with default 1e-4
data	The data frame

Value

This function returns a list of the following objects:

SAE_Eblup_sampled	A dataframe with the values of the EBLUPs estimators for sampled areas
SAE_Eblup_all	A dataframe with the values of the EBLUPs estimators for all areas
MSE_Eblup_sampled	A dataframe with the values of estimated mean square errors of EBLUPs estimators for sampled areas
MSE_Eblup_all	A dataframe with the values of estimated mean square errors of EBLUPs estimators for all areas
randomEffect_sampled	a dataframe with the values of the random effect estimators for sampled areas
randomEffect_all	a dataframe with the values of the random effect estimators for all areas
Rmatrix_sampled	a block diagonal matrix composed of sampling errors for sampled areas
fit	A list containing the following objects:

- method : The fitting method (this function is using "REML")

- convergence : The convergence result of Fisher-scoring algorithm (Logical Value)
- iterations : The number of Fisher-Scoring algorithm iterations
- estcoef : A dataframe with the estimated model coefficient, standard error,t statistics, p-values of the significance of each coefficient
- refvar : A dataframe with estimated random effect variances
- informationFisher : A matrix of information fisher from Fisher-scoring algorithm

Examples

```
##load dataset
data(datamsaeDBns)
#Note : Make sure yout dataset does not contain NA Values
#       you can set 0 in Direct estinations and vardir for non-sampled areas.


#Compute Fitted model for Y1, Y2, and Y3
#Y1 ~ X1 + X2
#Y2 ~ X1 + X2
#Y3 ~ X1 + X2

##Using parameter 'data'
formula = list(f1 = Y1~X1+X2,
               f2 = Y2~X1+X2,
               f3 = Y3~X1+X2)
vardir = c("v1","v12","v13","v2","v23","v3")
cluster = c("clY1","clY2","clY3")
nonsample = "nonsample"
saeFHns <- saefhns(formula,vardir,cluster, nonsample, data=datamsaeDBns)

#to calculate only one response variable
saeFHns1 <- saefhns(formula=list(f1=Y1~X1+X2),vardir ="v1", cluster = "clY1",
            nonsample = "nonsample",data=datamsaeDBns )
```

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