Package 'stratallo'

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Title Optimum Sample Allocation in Stratified Sampling Schemes

Version 2.0.1

Description Functions in this package provide solution to classical problem in survey methodology - an optimum sample allocation in stratified sampling schemes. In this context, the optimal allocation is in the classical Tschuprov-Neyman's sense and it satisfies additional either lower or upper bounds restrictions imposed on sample sizes in strata. There are few different algorithms available to use, and one them is based on popular sample allocation method that applies Neyman allocation to recursively reduced set of strata.

This package also provides the function that computes a solution to the minimum sample size allocation problem, which is a minor modification of the classical optimium sample allocation. This problems lies in the determination of a vector of strata sample sizes that minimizes total sample size, under assumed fixed level of the pi-estimator's variance. As in the case of the classical optimal allocation, the problem of minimum sample size allocation can be complemented by imposing upper bounds constraints on sample sizes in strata.

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Author Wojciech Wójciak [aut, cre], Jacek Wesołowski [sad], Robert Wieczorkowski [ctb]

Maintainer Wojciech Wójciak <wojciech.wojciak@gmail.com>

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stratallo-package Functions for Optimal Sample Allocation in Stratified Sampling Schemes

Description

Optimal Sample Allocation in Stratified Sampling Schemes

Author(s)

Wojciech Wójciak <wojciech.wojciak@gmail.com>

References

Wesołowski, J., Wieczorkowski, R., Wójciak, W. (2021), Optimality of the recursive Neyman allocation, *Journal of Survey Statistics and Methodology*, doi: 10.1093/jssam/smab018, doi: 10.48550/ arXiv.2105.14486

Wójciak, W. (2022), Minimum sample size allocation in stratified sampling under constraints on variance and strata sample sizes, doi: 10.48550/arXiv.2204.04035

dopt

Wójciak, W. (2019), Optimal allocation in stratified sampling schemes, *MSc Thesis*, Warsaw University of Technology, Warsaw, Poland. http://home.elka.pw.edu.pl/~wwojciak/msc_optimal_allocation.pdf

Sarndal, C.-E., Swensson, B., and Wretman, J. (1992), *Model Assisted Survey Sampling*, New York, NY: Springer.

dopt

Optimum Sample Allocation in Stratified Sampling Schemes

Description

[Stable]

A classical problem in survey methodology in stratified sampling is an optimum sample allocation problem. This problem is formulated as determination of a vector of strata sample sizes that minimizes the variance of the pi-estimator of the population total of a given study variable, under constraint on total sample size.

The dopt() function solves the problem of optimum sample allocation under either lower or upper bounds constraints, optionally imposed on strata sample sizes. The allocation computed is valid for all stratified sampling schemes for which the variance of the stratified pi-estimator is of the form

$$D(x_1, ..., x_H) = a_1^2 / x_1 + ... + a_H^2 / x_H - b,$$

where H denotes total number of strata, $x_1, ..., x_H$ are the strata sample sizes, and b, $a_w > 0$ do not depend on $x_w, w = 1, ..., H$.

The dopt() function makes use of the following allocation algorithms: rNa, sga, sgaplus, coma for optimal sample allocation under upper bounds constraints only, and LrNa for optimal sample allocation under lower bounds constraints only. The rNa, sga, and coma are described in Wesołowski et al. (2021), the sgaplus in Wójciak (2019), and the LrNa is introduced in Wójciak (2022). If no inequality constraints are added, then no any special algorithm is used as the allocation is given as a closed form expression, known as Neyman allocation

$$x_w = a_w * n/(a_1 + \dots + a_H), w = 1, \dots, H.$$

Usage

dopt(n, a, m = NULL, M = NULL, M_method = "rna")

Arguments

n	(number) total sample size. A strictly positive scalar.
a	(numeric) parameters $a_1,, a_H$ of variance function D. Strictly positive numbers.
m	(numeric or NULL) lower bounds constraints optionally imposed on strata sample sizes. If different than NULL, it is then required that n >= sum(m). Must be assigned with NULL (default) if M is not NULL. Strictly positive numbers.
Μ	(numeric or NULL) upper bounds constraints optionally imposed on strata sample sizes. If different than NULL, it is then required that $n \le sum(M)$. Must be assigned with NULL (default) if m is not NULL. Strictly positive numbers.
M_method	(string) the name of the underlying algorithm to be used for computing a sample allo- cation under one-sided upper bounds constraints (Case I). One of the following: "rna" (default), "sga", "sgaplus", "coma".

Details

The dopt() function computes

 $argminD(x_1, ..., x_H),$

under the equality constraint imposed on total sample size

 $x_1 + \ldots + x_H = n.$

Optionally, one of the following set of one-sided inequality constraints can be added

$$x_w \ll M_w, w = 1, \dots, H, (CaseI)$$

or

$$x_w >= m_w, w = 1, ..., H, (CaseII)$$

where n > 0 denotes overall sample size, $m_w > 0$, and $M_w > 0, w = 1, ..., H$, are the lower and upper bounds respectively, imposed on sample strata sizes.

User of dopt() can choose whether the inequality constraints will be added to the optimization problem or not. This is achieved with the proper use of m and M arguments of the function. In case of no inequality constraints to be added, m and M must be both specified as NULL (default). If only upper bounds constraints should be added (Case I), user specifies these bounds with M argument, while leaving m as NULL. Finally, if only lower bounds constraints should be added (Case II), user specifies these bounds with m argument, while leaving M as NULL. At the moment, dopt() function does not allow to add box-constraints simultaneously.

For the Case I, there are four different underlying algorithms available to use. These are abbreviated as: "rNa" (rna_one_sided()), "sga" (sga()), "sgaplus" (sgaplus()), "coma" (coma()). Functions names that perform given algorithms are given in the brackets. See its help page for more details. For the Case II, the "rNa" (rna_one_sided()) is used.

dopt

Value

Numeric vector with optimal sample allocation in strata.

Note

For simple random sampling without replacement design in each stratum, parameters of the variance function D are $b = N_1 * S_1^2 + ... + N_H * S_H^2$, and $a_w = N_w * S_w$, where $N_w, S_w, w = 1, ..., H$, are strata sizes and standard deviations of a study variable in strata respectively.

References

Wesołowski, J., Wieczorkowski, R., Wójciak, W. (2021), Optimality of the recursive Neyman allocation, *Journal of Survey Statistics and Methodology*, doi: 10.1093/jssam/smab018, doi: 10.48550/ arXiv.2105.14486

Wójciak, W. (2022), Minimum sample size allocation in stratified sampling under constraints on variance and strata sample sizes, doi: 10.48550/arXiv.2204.04035

Wójciak, W. (2019), Optimal allocation in stratified sampling schemes, *MSc Thesis*, Warsaw University of Technology, Warsaw, Poland. http://home.elka.pw.edu.pl/~wwojciak/msc_optimal_allocation.pdf

Sarndal, C.-E., Swensson, B., and Wretman, J. (1992), *Model Assisted Survey Sampling*, New York, NY: Springer.

See Also

nopt(), rna_one_sided(), sga(), sgaplus(), coma().

a <- c(3000, 4000, 5000, 2000)

Examples

```
bounds <- c(100, 90, 70, 80)
# Only lower bounds.
dopt(n = 340, a = a, m = bounds)
dopt(n = 400, a = a, m = bounds)
dopt(n = 600, a = a, m = bounds)
# Only upper bounds.
dopt(n = 190, a = a, M = bounds)
dopt(n = 300, a = a, M = bounds)
dopt(n = 340, a = a, M = bounds)
# Example of execution-time comparison of different algorithms
# using bench R package.
## Not run:
N <- pop969[, "N"]
S <- pop969[, "S"]</pre>
```

```
a <- N * S
nfrac <- seq(0.01, 0.9, 0.05)
n <- setNames(as.integer(nfrac * sum(N)), nfrac)</pre>
lapply(
 n,
 function(ni) {
   bench::mark(
      dopt(ni, a, M = N, M_method = "rna"),
      dopt(ni, a, M = N, M_method = "sga"),
      dopt(ni, a, M = N, M_method = "sgaplus"),
      dopt(ni, a, M = N, M_method = "coma"),
      iterations = 200
    )[c(1, 3)]
 }
)
## End(Not run)
```

dopt_upper

Algorithms for Optimum Sample Allocation in Stratified Sampling Under Upper Bounds Constraints

Description

[Stable]

Internal functions that implement the optimal sample allocation algorithms: sga, sgaplus and coma. Functions from this family compute

$$argminD(x_1, ..., x_H) = a_1^2/x_1 + ... + a_H^2/x_H - b_H^2$$

under the equality constraint imposed on total sample size

$$x_1 + \dots + x_H = n,$$

and upper bounds constraints imposed on strata sample sizes

$$x_w <= M_w, w = 1, ..., H.$$

Here, H denotes total number of strata, $x_1, ..., x_H$ are the strata sample sizes, and $n > 0, b, a_w > 0$, $M_w > 0, w = 1, ..., H$ are given numbers.

The sga(), sgaplus() and coma() are internal implementations of the algorithms in subject, and hence, users should not use any of these functions directly. Instead, the dopt() should be used.

Usage

sga(n, a, M)
sgaplus(n, a, M)
coma(n, a, M)

dopt_upper

Arguments

n	(number) total sample size. A strictly positive scalar.
а	(numeric) parameters $a_1,, a_H$ of variance function D. Strictly positive numbers.
М	(numeric) upper bounds constraints imposed on strata sample sizes. It is required that n <= sum(M). Strictly positive numbers.

Value

Numeric vector with optimal sample allocations in strata.

Functions

- sga: implementation of the Stenger-Gabler type algorithm SGa, described in Wesołowski et al. (2021) and in Stenger and Gabler (2005).
- sgaplus: implementation of the modified Stenger-Gabler type algorithm, described in Wójciak (2019) as Sequential Allocation (version 1) algorithm.
- coma: implementation of the Change of Monotonicity Algorithm, or coma, described in Wesołowski et al. (2021).

Note

For simple random sampling without replacement design in each stratum, parameters of the variance function D are $b = N_1 * S_1^2 + ... + N_H * S_H^2$, and $a_w = N_w * S_w$, where $N_w, S_w, w = 1, ..., H$, are strata sizes and standard deviations of a study variable in strata respectively.

References

Wesołowski, J., Wieczorkowski, R., Wójciak, W. (2021), Optimality of the recursive Neyman allocation, *Journal of Survey Statistics and Methodology*, doi: 10.1093/jssam/smab018, doi: 10.48550/ arXiv.2105.14486

Stenger, H., Gabler, S. (2005), Combining random sampling and census strategies - Justification of inclusion probabilities equal 1, *Metrika*, 61, 137-156

Wójciak, W. (2019), Optimal allocation in stratified sampling schemes, *MSc Thesis*, Warsaw University of Technology, Warsaw, Poland. http://home.elka.pw.edu.pl/~wwojciak/msc_optimal_allocation.pdf

See Also

dopt(), rna_one_sided().

Examples

a <- c(3000, 4000, 5000, 2000)
M <- c(100, 90, 70, 80)
sga(n = 190, a = a, M = M)
sgaplus(n = 190, a = a, M = M)
coma(n = 190, a = a, M = M)</pre>

h_get_which_violated Get Proper Version of which_violated Function.

Description

[Stable]

Internal function that prepares a simple 2-arguments wrapper of base::which() that checks whether its first argument exceeds the second one. Both arguments are numeric. This excess is hard coded in the returned wrapper function and it is defined either as >= ("greater or equal") or <= ("lower or equal"), depending on the value of the geq flag.

Usage

```
h_get_which_violated(geq = TRUE)
```

Arguments geq

(flag) if TRUE, then "greater or equal" condition is set. Otherwise, "less then or equal" is set.

Value

2-arguments function that checks whether its first argument exceeds the second one. Both arguments must be numeric.

See Also

rna_one_sided().

Examples

```
which_violated <- stratallo:::h_get_which_violated()
which_violated(1:3, 3:1)
which_violated <- stratallo:::h_get_which_violated(geq = FALSE)
which_violated(1:3, 3:1)</pre>
```

nopt

Description

[Stable]

User function that determines fixed strata sample sizes that minimize total sample size, under assumed level of the variance of the stratified pi-estimator of the total and optional one-sided upper bounds imposed on strata sample sizes. The algorithm used by nopt() is described in Wójciak (2022). The allocation computed is valid for all stratified sampling schemes for which the variance of the stratified pi-estimator is of the form

$$D(x_1, ..., x_H) = a_1^2/x_1 + ... + a_H^2/x_H - b_H^2$$

where H denotes total number of strata, $x_1, ..., x_H$ are the strata sample sizes, and $b, a_w > 0$ do not depend on $x_w, w = 1, ..., H$.

Usage

nopt(D, a, b, M = NULL)

Arguments

D	(number) variance equality constraint value. A strictly positive scalar.
а	(numeric) parameters $a_1,, a_H$ of variance function D. Strictly positive numbers.
b	(number) parameter b of variance function D .
Μ	(numeric or NULL) upper bounds constraints optionally imposed on strata sample sizes. If different than NULL, it is then required that $D \ge sum(a/M) - b > 0$. Strictly positive numbers.

Details

The nopt() function computes

 $argminn(x_1, ..., x_H) = x_1 + ... + x_H,$

under the equality constraint imposed on the variance

$$a_1^2/x_1 + \dots + a_H^2/x_H - b = D.$$

Optionally, the following set of one-sided inequality constraints can be added

$$x_w <= M_w, w = 1, ..., H,$$

where D > 0 is a given number and $M_w > 0, w = 1, ..., H$, are the upper bounds, imposed on sample strata sizes.

Value

Numeric vector with optimal sample allocation in strata.

Note

For simple random sampling without replacement design in each stratum, parameters of the variance function D are $b = N_1 * S_1^2 + ... + N_H * S_H^2$, and $a_w = N_w * S_w$, where $N_w, S_w, w = 1, ..., H$, are strata sizes and standard deviations of a study variable in strata respectively.

References

Wójciak, W. (2022), Minimum sample size allocation in stratified sampling under constraints on variance and strata sample sizes, doi: 10.48550/arXiv.2204.04035

See Also

rna_one_sided(), dopt().

Examples

a <- c(3000, 4000, 5000, 2000) M <- c(100, 90, 70, 80) nopt(1017579, a = a, b = 579, M = M)

pop10_mM

Example Population with 10 Strata and Lower and Upper Bounds

Description

Artificial example population with 10 strata defined by strata sizes and standard deviations of some study variable. Additionally, the lower and upper bounds have been specified for the sample in strata.

Usage

pop10_mM

Format

A matrix with 10 rows and 5 columns: N (strata sizes), S (standard deviations of a variable under study in strata), m (lower bounds), M (upper bounds).

pop507

Description

Artificial example population with 507 strata defined by strata sizes and standard deviations of some study variable.

Usage

pop507

Format

A matrix with 507 rows and 2 columns: N with strata sizes and S standard deviations of a variable under study in strata.

pop969

Example Population with 969 Strata

Description

Artificial example population with 969 strata defined by strata sizes and standard deviations of some study variable.

Usage

pop969

Format

A matrix with 969 rows and 2 columns: N with strata sizes and S standard deviations of a variable under study in strata.

rna_one_sided

Description

[Stable]

Internal function that implements the recursive Neyman optimal allocation algorithms, rNa and LrNa, described in Wesołowski et al. (2021) and Wójciak (2022) respectively. The rna_one_sided() should not be used directly. Instead, user function dopt() or nopt() should be used.

The rna_one_sided() function computes

$$argminD(x_1, ..., x_H) = a_1^2/x_1 + ... + a_H^2/x_H - b_H^2/x_H$$

under the equality constraint imposed on total sample size

$$x_1 + \ldots + x_H = n.$$

Here, H denotes total number of strata, $x_1, ..., x_H$ are the strata sample sizes, and n > 0, $b, a_w > 0, w = 1, ..., H$, are given numbers.

Optionally, one of the following set of one-sided inequality constraints can be added

$$x_w <= M_w, w = 1, ..., H, (CaseI)$$

or

$$x_w >= m_w, w = 1, ..., H, (CaseII)$$

where $m_w > 0$ and $M_w > 0, w = 1, ..., H$ are lower and upper bounds respectively, imposed on sample strata sizes.

User of rna_one_sided() can choose whether the inequality constraints will be added to the optimization problem or not. This is achieved with the proper use of bounds and upper arguments of the function. In case of no inequality constraints to be added, bounds must be specified as NULL (default). If any bounds should be imposed on sample strata sizes, user must specify these with bounds argument. For the Case I of the upper bounds, upper flag must be set to TRUE (default) and then the rna_one_sided() performs the rNa. For the Case II of lower bounds, upper flag must be set to FALSE and then the rna_one_sided() performs the LrNa. The upper flag is ignored when bounds is NULL. If no inequality constraints are added, the allocation is given as a closed form expression, known as Neyman allocation

$$x_w = a_w * n/(a_1 + \dots + a_H), w = 1, \dots, H.$$

Usage

```
rna_one_sided(n, a, bounds = NULL, upper = TRUE)
```

Arguments

n	(number) total sample size. A strictly positive scalar.
а	(numeric) parameters $a_1,, a_H$ of variance function D. Strictly positive numbers.
bounds	(numeric or NULL) optional lower or upper bounds constraints imposed on strata sample sizes. If bounds is not NULL, it is required that n <= sum(bounds) in case of upper = TRUE, and n >= sum(bounds), in case of upper = FALSE. Strictly positive num- bers.
upper	(flag) should values of bounds be treated as one-sided upper bounds constraints (de- fault)? Otherwise, they are treated as lower bounds.

Value

Numeric vector with optimal sample allocations in strata.

Note

For simple random sampling without replacement design in each stratum, parameters of the variance function D are $b = N_1 * S_1^2 + ... + N_H * S_H^2$, and $a_w = N_w * S_w$, where $N_w, S_w, w = 1, ..., H$, are strata sizes and standard deviations of a study variable in strata respectively.

The rNa and LrNa are kind of more general versions of popular recursive Neyman allocation procedure that is described in Remark 12.7.1 in Sarndal et al. (1992). It is a procedure of optimal sample allocation in stratified sampling scheme with simple random sampling without replacement design in each stratum while not exceeding strata sizes.

References

Wesołowski, J., Wieczorkowski, R., Wójciak, W. (2021), Optimality of the recursive Neyman allocation, *Journal of Survey Statistics and Methodology*, doi: 10.1093/jssam/smab018, doi: 10.48550/ arXiv.2105.14486

Wójciak, W. (2022), Minimum sample size allocation in stratified sampling under constraints on variance and strata sample sizes, doi: 10.48550/arXiv.2204.04035

Sarndal, C.-E., Swensson, B., and Wretman, J. (1992), *Model Assisted Survey Sampling*, New York, NY: Springer.

See Also

dopt(), nopt(), sga(), sgaplus(), coma().

Examples

```
a <- c(3000, 4000, 5000, 2000)
bounds <- c(100, 90, 70, 80)
rna_one_sided(n = 190, a = a, bounds = bounds)</pre>
```

var_tst

Description

[Stable]

Compute the variance of the stratified pi-estimator of the population total, that is of the following generic form

$$D(x_1, ..., x_H) = a_1^2 / x_1 + ... + a_H^2 / x_H - b,$$

where H denotes total number of strata, $x_1, ..., x_H$ are the strata sample sizes, and $b, a_w > 0$ do not depend on $x_w, w = 1, ..., H$.

Usage

var_tst(x, a, b)
var_tst_si(x, N, S)

Arguments

x	(numeric) sample allocations in strata. Strictly positive numbers.
а	(numeric) parameters $a_1,, a_H$ of variance function D. Strictly positive numbers.
b	(numeric) parameter b of variance function D .
Ν	(numeric) strata sizes. Strictly positive numbers.
S	(numeric) strata standard deviations of a given study variable. Strictly positive numbers.

Value

Value of the variance D for a given allocation vector x.

var_tst

Functions

• var_tst_si: computes variance of stratified pi-estimator of the total for simple random sampling without replacement design in each stratum. Under this design, parameters of the variance function D take the following form

$$a_w = N_w * S_w, w = 1, ..., H,$$

and

$$b = N_1 * S_1^2 + \dots + N_H * S_H^2,$$

where $N_w, S_w, w = 1, ..., H$, are strata sizes and standard deviations of a study variable in strata respectively.

References

Sarndal, C.-E., Swensson, B., and Wretman, J. (1992), *Model Assisted Survey Sampling*, Chapter 3.7 *Stratified Sampling*, New York, NY: Springer.

Examples

N <- c(3000, 4000, 5000, 2000) S <- rep(1, 4) M <- c(100, 90, 70, 80) opt <- dopt(n = 190, a = N * S, M = M) var_tst_si(x = opt, N, S)

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