Package ‘simPop’

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Description Tools and methods to simulate populations for surveys based on auxiliary data. The tools include model-based methods, calibration and combinatorial optimization algorithms, see Templ, Kowarik and Meindl (2017) <doi:10.18637/jss.v079.i10>) and Templ (2017) <doi:10.1007/978-3-319-50272-4>. The package was developed with support of the International Household Survey Network, DFID Trust Fund TF011722 and funds from the World bank.
License GPL (>= 2)
LazyLoad yes
ByteCompile TRUE
Collate '0classes.R' 'addKnownMargins.R' 'addWeights.r' 'calibPop.R'
  'calibSample.R' 'calibVars.R' 'contingencyWt.R' 'correctHeap.R'
  'fitGPD.R' 'getBreaks.R' 'getCat.R' 'ipu.r' 'ipu2.r' 'meanWt.R'
  'printFunctions.R' 'quantileWt.R' 'sampHH.R' 'RcppExports.R'
  'sileTools.R' 'sileTools2.R' 'simAnnealingDT.R'
  'simCategorical.R' 'simComponents.R' 'simContinuous.R'
  'simEU.SILC.R' 'simGPD.R' 'simInitSpatial.R' 'simple_dis.R'
  'simPop-package.R' 'simRelation.R' 'simStructure.R'
  'spBwplot.R' 'spBwplotStats.R' 'spCdf.R' 'spCdfplot.R'
  'spMosaic.R' 'spPredict.R' 'spSample.R' 'spTable.R'
  'specifyInput.R' 'sprague.R' 'tableWt.R' 'utility.R' 'utils.R'
  'whipple.R' 'dataSets.R' 'zzz.R'
**Description**

The production of synthetic datasets has been proposed as a statistical disclosure control solution to generate public use files out of protected data, and as a tool to create “augmented datasets” to serve as input for micro-simulation models. Synthetic data have become an important instrument for *ex-ante* assessments of policies’ impact. The performance and acceptability of such a tool relies heavily on the quality of the synthetic populations, i.e., on the statistical similarity between the synthetic and the true population of interest.

**Details**

Multiple approaches and tools have been developed to generate synthetic data. These approaches can be categorized into three main groups: synthetic reconstruction, combinatorial optimization, and model-based generation.

The package: **simPop** is a user-friendly R-package based on a modular object-oriented concept. It provides a highly optimized S4 class implementation of various methods, including calibration by iterative proportional fitting and simulated annealing, and modeling or data fusion by logistic regression.

The following applications further shows the methods and package: We firstly demonstrated the use of **simPop** by creating a synthetic population of Austria based on the European Statistics of Income and Living Conditions (Alfons et al., 2011) including the evaluation of the quality of the generated population. In this contribution, the mathematical details of functions **simStructure**, **simCategorical**, **simContinuous** and **simComponents** are given in detail. The disclosure risk of this synthetic population has been evaluated in (Templ and Alfons, 2012) using large-scale simulation studies.

Employer-employee data were created in Templ and Filzmoser (2014) whereby the structure of companies and employees are considered.
Finally, the R package simPop is presented in full detail in Templ et al. (2017). In this paper - the main reference to this work - all functions and the S4 class structure of the package are described in detail. For beginners, this paper might be the starting point to learn about the methods and package.

Package: simPop
Type: Package
Version: 1.0.0
Date: 20017-08-07
License: GPL (>= 2)

Author(s)
Bernhard Meindl, Matthias Templ, Andreas Alfons, Alexander Kowarik,
Maintainer: Matthias Templ <matthias.templ@gmail.com>

References

Examples

```r
## we use synthetic eusilcS survey sample data
## included in the package to simulate a population

## create the structure
data(eusilcS)

## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
## in the following, nr_cpus are selected automatically
simPop <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
simPop <- simCategorical(simPop, additional=c("pl030", "pb220a"), method="multinom", nr_cpus=1)
simPop
class(simPop)
regModel = ~rb090+hsize+pl030+pb220a
```
## multinomial model with random draws

eusilcM <- simContinuous(simPop, additional="netIncome",
                          regModel = regModel,
                          upper=200000, equidist=FALSE, nr_cpus=1)
class(eusilcM)

## this is already a basic synthetic population, but
## many other functions in the package might now
## be used for fine-tuning, adding further variables,
## evaluating the quality, adding finer geographical details,
## using different methods, calibrating surveys or populations, etc.
## -- see Templ et al. (2017) for more details.

---

### addKnownMargins

add known margins/totals

**Description**

add known margins/totals for a combination of variables for the population to an object of class `simPopObj`.

**Usage**

```r
addKnownMargins(inp, margins)
```

**Arguments**

- `inp`: a `simPopObj` containing population and household survey data as well as optionally margins in standardized format.
- `margins`: a `data.frame` containing for a combination of unique variable levels for `n` variables the number of known occurrences in the population. The numbers must be listed in the last column of data.frame 'margins' while the characteristics must be listed in the first 'n' columns.

**Details**

The function takes a `data.frame` containing known marginals/totals for some variables that must exist in the population (stored in slot 'pop' of input object 'inp') and updates slot 'table' of the input object. This slot finally contains the known totals. Households are drawn from the data and new ID's are generated for the new households.

**Value**

an object of class `simPopObj` with updated slot 'table'.
AddWeights<-

Author(s)
Bernhard Meindl

References

Examples
data(eusilcS)
data(eusilcP)

## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
inp <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
inp <- simCategorical(inp, additional=c("pl030", "pb220a"), method="multinom", nr_cpus=1)
margins <- as.data.frame(
  xtabs(rep(1, nrow(eusilcP)) ~ eusilcP$region + eusilcP$gender + eusilcP$citizenship))
colnames(margins) <- c("db040", "rb090", "pb220a", "freq")
inp <- addKnownMargins(inp, margins)
str(inp)

AddWeights<- Methods for function AddWeights

Description
allows to modify sampling weights of an dataObj or simPopObj-object. As input the output of calibSample must be used.

Usage
AddWeights(object) <- value

## S4 replacement method for signature 'dataObj'
AddWeights(object) <- value

## S4 replacement method for signature 'simPopObj'
AddWeights(object) <- value

Arguments
object an object of class dataObj or simPopObj.
value a numeric vector of suitable length
Examples

data(eusilcS)
data(totalsRG)
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")

## approx. 20 seconds ...
addWeights(inp) <- calibSample(inp, totalsRG)

---

calibPop

Calibration of 0/1 weights by Simulated Annealing

Description

A Simulated Annealing Algorithm for calibration of synthetic population data available in a simPopObj object. The aims is to find, given a population, a combination of different households which optimally satisfy, in the sense of an acceptable error, a given table of specific known marginals. The known marginals are also already available in slot 'table' of the input object 'inp'.

Usage

calibPop(
inp,  
split,  
temp = 1,  
eps.factor = 0.05,  
maxiter = 200,  
temp.cooldown = 0.9,  
factor.cooldown = 0.85,  
min.temp = 10^-3,  
nr_cpus = NULL,  
sizefactor = 2,  
memory = TRUE,  
choose.temp = TRUE,  
choose.temp.factor = 0.2,  
scale.redraw = 0.5,  
observe.times = 50,  
observe.break = 0.05,  
verbose = FALSE
)

Arguments

inp an object of class simPopObj with slot 'table' being non-null! (see addKnownMargins).

split given strata in which the problem will be split. Has to correspond to a column population data (slot 'pop' of input argument 'inp'). For example split = c("region"), problem will be split for different regions. Parallel computing is performed automatically, if possible.
temp starting temperatur for simulated annealing algorithm
eps.factor a factor (between 0 and 1) specifying the acceptance error. For example eps.factor = 0.05 results in an acceptance error for the objective function of 0.05*sum(totals).
maxiter maximum iterations during a temperature step.
temp.cooldown a factor (between 0 and 1) specifying the rate at which temperature will be reduced in each step.
factor.cooldown a factor (between 0 and 1) specifying the rate at which the number of permutations of households, in each iteration, will be reduced in each step.
min.temp minimal temperature at which the algorithm will stop.
rnr_cpu if specified, an integer number defining the number of cpus that should be used for parallel processing.
sizefactor the factor for inflating the population before applying 0/1 weights
memory if TRUE simulated annealing is applied in less memory intensive way. Is especially usefull if factor or population is large. For this option simulated annealing is not entirely implemented in C++, therefore it might be slower than option memory=FALSE.
choose.temp if TRUE temp will be rescaled according to eps and choose.temp.factor. eps is defined by the product between eps.factor and the sum over the target population margins, see addKnownMargins. Only used if memory=TRUE.
choose.temp.factor number between (0,1) for rescaling temp for simulated annealing. temp redefined by max(temp,eps*choose.temp.factor). Can be usefull if simulated annealing is split into subgroups with considerably different population sizes. Only used if choose.temp=TRUE and memory=TRUE.
scale.redraw Only used if memory=TRUE. Number between (0,1) scaling the number of households that need to be drawn and discarded in each iteration step. The number of individuals currently selected through simulated annealing is substracted from the sum over the target population margins added to inp via addKnownMargins. This difference is divided by the median household size resulting in an estimated number of households that the current synthetic population differs from the population margins (~redraw_gap). The next iteration will then adjust the number of households to be drawn or discarded (redraw) according to max(ceiling(redraw-redraw_gap*scale.redraw),1) or max(ceiling(redraw+redraw_gap*scale.redraw),1) respectively. This keeps the number of individuals in the synthetic population relatively stable regarding the population margins. Otherwise the synthetic population might be considerably larger or smaller then the population margins, through selection of many large or small households.
observe.times Only used if memory=TRUE. Number of times the new value of the objective function is saved. If observe.times=0 values are not saved.
observe.break Only used if memory=TRUE. When objective value has been saved observe.times times the coefficient of variation is calculated over saved values; if the coefficient of variation falls below observe.break simulated annealing terminates. This repeats for each new set of observe.times new values of the objective function. Can help save run time if objective value does not improve much. Disable this termination by either setting observe.times=0 or observe.break=0.
verbose boolean variable: if TRUE some additional verbose output is provided, however only if split is NULL. Otherwise the computation is performed in parallel and no useful output can be provided.

Details

Calibrates data using simulated annealing. The algorithm searches for a (near) optimal combination of different households, by swapping households at random in each iteration of each temperature level. During the algorithm as well as for the output the optimal (or so far best) combination will be indicated by a logical vector containing only 0s (not included) and 1s (included in optimal selection). The objective function for simulated annealing is defined by the sum of absolute differences between target marginals and synthetic marginals (=marginals of synthetic dataset). The sum of target marginals can at most be as large as the sum of target marginals. For every factor-level in “split”, data must at least contain as many entries of this kind as target marginals.

Possible donors are automatically generated within the procedure.

The number of cpus are selected automatically in the following manner. The number of cpus is equal the number of strata. However, if the number of cpus is less than the number of strata, the number of cpus - 1 is used by default. This should be the best strategy, but the user can also overwrite this decision.

Value

Returns an object of class simPopObj with an updated population listed in slot ‘pop’.

Author(s)

Bernhard Meindl, Johannes Gussenbauer and Matthias Templ

References


Examples

data(eusilcS) # load sample data
data(eusilcP) # population data

## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsizer="hsize", strata="db040", weight="db090")
simPop <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
simPop <- simCategorical(simPop, additional=c("pl030", "pb220a"), method="multinom", nr_cpus=1)

# add margins
margins <- as.data.frame(
  xtabs(rep(1, nrow(eusilcP)) ~ eusilcP$region + eusilcP$gender + eusilcP$citizenship))
colnames(margins) <- c("db040", "rb090", "pb220a", "freq")
simPop <- addKnownMargins(simPop, margins)
simPop_adj2 <- calibPop(simPop, split="db040", split="db040", verbose="true")
```r
temp=1, eps.factor=0.1,
memory=TRUE, nr_cpus = 1)

# apply simulated annealing
## long computation time
simPop_adj <- calibPop(simPop, split="db040", temp=1, eps.factor=0.1, memory=FALSE)
```

---

calibSample  

**Calibrate sample weights**

**Description**

Calibrate sample weights according to known marginal population totals. Based on initial sample weights, the so-called g-weights are computed by generalized raking procedures.

**Details**

The methods return a list containing both the g-weights (slot `g_weights`) as well as the final weights (slot `final_weights`) (initial sampling weights adjusted by the g-weights).

**Methods**

The function provides methods with the following signatures.

Argument 'inp' must be an object of class `data.frame`, `dataObj` or `simPopObj` and the totals must be specified in either objects of class `table` or `data.frame`. If argument 'totals' is a data.frame it must be provided in a way that in the first columns n-columns the combinations of variables are listed. In the last column, the frequency counts must be specified. Furthermore, variable names of all but the last column must be available also from the sample data specified in argument 'inp'. If argument 'total' is a table (e.g. created with function `tableWt`), it must be made sure that the dimnames match the variable names (and levels) of the specified input data set.

**Note**

`list(signature(inp="df_or_dataObj_or_simPopObj", totals="dataFrame_or_Table","..."))` This is a faster implementation of parts of `calib` from package `sampling`. Note that the default calibration method is raking and that the truncated linear method is not yet implemented.

**Author(s)**

Andreas Alfons and Bernhard Meindl
References


Examples

data(eusilcS)
eusilcS$agecut <- cut(eusilcS$age, 7)
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="rb090")

## for simplicity, we are using population data directly from the sample, but you get the idea
totals1 <- tableWt(eusilcS[, c("agecut","rb090")], weights=eusilcS$rb050)
totals2 <- tableWt(eusilcS[, c("rb090","agecut")], weights=eusilcS$rb050)
totals3 <- tableWt(eusilcS[, c("rb090","agecut","db040")], weights=eusilcS$rb050)
totals4 <- tableWt(eusilcS[, c("agecut","db040","rb090")], weights=eusilcS$rb050)

weights1 <- calibSample(inp, totals1)
totals1.df <- as.data.frame(totals1)
weights1.df <- calibSample(inp, totals1.df)
identical(weights1, weights1.df)

# we can also use a data.frame and an optional weight vector as input
df <- as.data.frame(inp@data)
w <- inp@data[[inp@weight]]
weights1.x <- calibSample(df, totals1.df, w=inp@data[[inp@weight]])
identical(weights1, weights1.x)

weights2 <- calibSample(inp, totals2)
totals2.df <- as.data.frame(totals2)
weights2.df <- calibSample(inp, totals2.df)
identical(weights2, weights2.df)

## Not run:
## approx 10 seconds computation time ...
weights3 <- calibSample(inp, totals3)
totals3.df <- as.data.frame(totals3)
weights3.df <- calibSample(inp, totals3.df)
identical(weights3, weights3.df)

## approx 10 seconds computation time ...
weights4 <- calibSample(inp, totals4)
totals4.df <- as.data.frame(totals4)
weights4.df <- calibSample(inp, totals4.df)
identical(weights4, weights4.df)

## End(Not run)
calibVars

Construct a matrix of binary variables for calibration

Description

Construct a matrix of binary variables for calibration of sample weights according to known marginal population totals. The following methods are implemented:

- calibVars.default(x)
- calibVars.matrix(x)
- calibVars.matrix(x)
- calibVars.data.frame(x)

Usage

calibVars(x)

Arguments

x a vector that can be interpreted as factor, or a matrix or data.frame consisting of such variables.

Value

A matrix of binary variables that indicate membership to the corresponding factor levels.

Author(s)

Bernhard Meindl and Andreas Alfons

References


See Also

calibSample

Examples

data(eusilcS)
# default method
aux <- calibVars(eusilcS$rb090)
head(aux)
# data.frame method
aux <- calibVars(eusilcS[, c("db040", "rb090")])
head(aux)
Description

Compute (weighted) pairwise contingency coefficients.

Usage

contingencyWt(x, ...)

Arguments

x for the default method, a vector that can be interpreted as factor. For the matrix and data.frame methods, the columns should be interpretable as factors.

... for the generic function, arguments to be passed down to the methods, otherwise ignored.

Details

The function tableWt is used for the computation of the corresponding pairwise contingency tables. The following methods are implemented:

- contingencyWt.default(x, y, weights = NULL, ...)
- contingencyWt.matrix(x, weights = NULL, ...)
- contingencyWt.data.frame(x, weights = NULL, ...)

Additional parameters are:

- y: a vector that can be interpreted as factor (for the default method)
- weights: an optional numeric vector containing sample weights

Value

For the default method, the (weighted) contingency coefficient of x and y.

For the matrix and data.frame method, a matrix of (weighted) pairwise contingency coefficients for all combinations of columns. Elements below the diagonal are NA.

Author(s)

Andreas Alfons and Stefan Kraft

References

correctHeaps

Correct age heaping

Description

Correct for age heaping using truncated (log-)normal distributions

Usage

correctHeaps(x, heaps = "10year", method = "lnorm", start = 0, fixed = NULL)

Arguments

x       numeric vector
heaps   • 5year: heaps are assumed to be every 5 years (0,5,10,...)
         • 10year: heaps are assumed to be every 10 years (0,10,20,...)
method  a character specifying the algorithm used to correct the age heaps. Allowed values are
         • lnorm: drawing from a truncated log-normal distribution. The required parameters are estimated using original input data.
         • norm: drawing from a truncated normal distribution. The required parameters are estimated using original input data.
         • unif: random sampling from a (truncated) uniform distribution
start   a numeric value for the starting of the 5 or 10 year sequences (e.g. 0, 5 or 10)
fixed   numeric index vector with observation that should not be changed

See Also

tableWt

Examples

data(eusilcS)

## default method
contingencyWt(eusilcS$pl030, eusilcS$pb220a, weights = eusilcS$rb050)

## data.frame method
basic <- c("age", "rb090", "hsize", "pl030", "pb220a")
contingencyWt(eusilcS[, basic], weights = eusilcS$rb050)
Details

Age heaping can cause substantial bias in important measures and thus age heaping should be corrected.

For method “lnorm”, a truncated log-normal is fit to the whole age distribution. Then for each age heap (at 0, 5, 10, 15, ...) random numbers of a truncated log-normal (with lower and upper bound) is drawn in the interval +− 2 around the heap (rounding of degree 2) using the inverse transformation method. A ratio of randomly chosen observations on an age heap are replaced by these random draws. For the ratio the age distribution is chosen, whereas on an age heap (e.g. 5) the arithmetic means of the two neighboring ages are calculated (average counts on age 4 and age 6 for age heap equals 5, for example). The ratio on, e.g. age equals 5 is then given by the count on age 5 divided by this mean. This is done for any age heap at (0, 5, 10, 15, ...).

Method “norm” replace the draws from truncated log-normals to draws from truncated normals. It depends on the age distrubution (if right-skewed or not) if method “lnorm” or “norm” should be used. Many distributions with heaping problems are right-skewed.

Method “unif” draws the mentioned ratio of observations on truncated uniform distributions around the age heaps.

Repeated calls of this function mimics multiple imputation, i.e. repeating this procedure m times provides m imputed datasets that properly reflect the uncertainty from imputation.

Value

a numeric vector without age heaps

Author(s)

Matthias Templ, Bernhard Meindl, Alexander Kowarik

References


Examples

## create some artificial data
age <- rlnorm(10000, meanlog=2.466869, sdlog=1.652772)
age <- round(age[age < 93])
barplot(table(age))

## artificially introduce age heaping and correct it:
# heaps every 5 years
year5 <- seq(0, max(age), 5)
age5 <- sample(c(age, age[age %in% year5]))
c5 <- rep("darkgrey", length(unique(age)));
c5[year5+1] <- "yellow"
barplot(table(age5), col=c5)
barplot(table(correctHeaps(age5, heaps="5year", method="lnorm"), col=c5))
Correct a specific age heap in a vector containing age in years

Usage

```
correctSingleHeap(
  x,
  heap,
  before = 2,
  after = 2,
  method = "lnorm",
  fixed = NULL
)
```

Arguments

- **x**
  - numeric vector representing age in years (integers)
- **heap**
  - numeric or integer vector of length 1 specifying the year for which a heap should be corrected
- **before**
  - numeric or integer vector of length 1 specifying the number of years before the heap that may be used to correct the heap. This input will be rounded!
- **after**
  - numeric or integer vector of length 1 specifying the number of years after the heap that may be used to correct the heap. This input will be rounded!
- **method**
  - a character specifying the algorithm used to correct the age heaps. Allowed values are
    - **lnorm**: drawing from a truncated log-normal distribution. The required parameters are estimated using original input data.
norm: drawing from a truncated normal distribution. The required parameters are estimated using original input data.

• unif: random sampling from a (truncated) uniform distribution

fixed numeric index vector with observation that should not be changed

Value

a numeric vector without age heaps

Author(s)

Matthias Templ, Bernhard Meindl, Alexander Kowarik

Examples

```r
## create some artificial data
age <- rlnorm(10000, meanlog=2.466869, sdlog=1.652772)
age <- round(age[age < 93])
barplot(table(age))

## artificially introduce an age heap for a specific year
## and correct it
age23 <- c(age, rep(23, length=sum(age==23)))
cc23 <- rep("darkgrey", length(unique(age)))
cc23[24] <- "yellow"
barplot(table(age23), col=cc23)
barplot(table(correctSingleHeap(age23, heap=23, before=2, after=3, method="lnorm"), col=cc23))
barplot(table(correctSingleHeap(age23, heap=23, before=5, after=5, method="lnorm"), col=cc23))

# the first 5 observations should be unchanged
barplot(table(correctSingleHeap(age23, heap=23, before=5, after=5, method="lnorm", fixed=1:5)), col=cc23)
```

Description

Objects of this class contain information on a population or survey.

Objects from the Class

Objects can be created by calls of the form `new("dataObj",...)` but are usually automatically created when using `simStructure`.

Author(s)

Bernhard Meindl and Matthias Templ
See Also

simPopObj

Examples

showClass("dataObj")

## show method, generate an object of class dataObj first
data(eusilcS)
inp <- specifyInput(data=eusilcS, hhid="db030", weight="rb050", strata="db040")
## shows some basic information:
inp

eusilc13puf  Synthetic EU-SILC 2013 survey data

Description

This data set is synthetically generated from real Austrian EU-SILC (European Union Statistics on Income and Living Conditions) data 2013.

Format

A data frame with 13513 observations on the following 62 variables.

- `db030` integer; the household ID.
- `hsize` integer; the number of persons in the household.
- `db040` factor; the federal state in which the household is located (levels Burgenland, Carinthia, Lower Austria, Salzburg, Styria, Tyrol, Upper Austria, Vienna and Vorarlberg).
- `age` integer; the person’s age.
- `rb090` factor; the person’s gender (levels male and female).
- `pid` personal ID
- `weight` sampling weights
- `pl031` factor; the person’s economic status (levels 1 = working full time, 2 = working part time, 3 = unemployed, 4 = pupil, student, further training or unpaid work experience or in compulsory military or community service, 5 = in retirement or early retirement or has given up business, 6 = permanently disabled or/and unfit to work or other inactive person, 7 = fulfilling domestic tasks and care responsibilities).
- `pb220a` factor; the person’s citizenship (levels AT, EU and Other).
- `pb190` for details, see Eurostat’s code book
- `pe040` for details, see Eurostat’s code book
- `pl111` for details, see Eurostat’s code book
- `pgrossIncomeCat` for details, see Eurostat’s code book
pgrossIncome for details, see Eurostat’s code book
hgrossIncomeCat for details, see Eurostat’s code book
hgrossIncome for details, see Eurostat’s code book
hgrossminusCat for details, see Eurostat’s code book
hgrossminus for details, see Eurostat’s code book
py010g for details, see Eurostat’s code book
py021g for details, see Eurostat’s code book
py050g for details, see Eurostat’s code book
py080g for details, see Eurostat’s code book
py090g for details, see Eurostat’s code book
py100g for details, see Eurostat’s code book
py110g for details, see Eurostat’s code book
py120g for details, see Eurostat’s code book
py130g for details, see Eurostat’s code book
py140g for details, see Eurostat’s code book
hy040g for details, see Eurostat’s code book
hy050g for details, see Eurostat’s code book
hy060g for details, see Eurostat’s code book
hy070g for details, see Eurostat’s code book
hy080g for details, see Eurostat’s code book
hy090g for details, see Eurostat’s code book
hy100g for details, see Eurostat’s code book
hy110g for details, see Eurostat’s code book
hy120g for details, see Eurostat’s code book
hy130g for details, see Eurostat’s code book
hy140g for details, see Eurostat’s code book
rb250 for details, see Eurostat’s code book
p119000 for details, see Eurostat’s code book
p038003f for details, see Eurostat’s code book
p118000i for details, see Eurostat’s code book
aktivi for details, see Eurostat’s code book
erwintensneu for details, see Eurostat’s code book
rb050 for details, see Eurostat’s code book
pb040 for details, see Eurostat’s code book
hb030 for details, see Eurostat’s code book
px030 for details, see Eurostat’s code book
rx030 for details, see Eurostat’s code book
The data set consists of 5977 households and is used as sample data in some of the examples in package simPop. Note that it is included for illustrative purposes only. The sample weights do not reflect the true population sizes of Austria and its regions.

62 variables of the original survey are simulated for this example data set. The variable names are rather cryptic codes, but these are the standardized names used by the statistical agencies. Furthermore, the variables hsize, age and netIncome are not included in the standardized format of EU-SILC data, but have been derived from other variables for convenience.

Author(s)
Matthias Templ

Source
This is a synthetic data set based on Austrian EU-SILC data from 2013. The original sample was provided by Statistics Austria.

References

Examples
data(eusilc13puf)
str(eusilc13puf)
Synthetic EU-SILC data

Description

This data set is synthetically generated from real Austrian EU-SILC (European Union Statistics on Income and Living Conditions) data.

Format

A data frame with 58,654 observations on the following 28 variables:

- **hid**: integer; the household ID.
- **region**: factor; the federal state in which the household is located (levels Burgenland, Carinthia, Lower Austria, Salzburg, Styria, Tyrol, Upper Austria, Vienna and Vorarlberg).
- **hsize**: integer; the number of persons in the household.
- **eqsize**: numeric; the equivalized household size according to the modified OECD scale.
- **eqIncome**: numeric; a simplified version of the equivalized household income.
- **pid**: integer; the personal ID.
- **id**: the household ID combined with the personal ID. The first five digits represent the household ID, the last two digits the personal ID (both with leading zeros).
- **age**: integer; the person's age.
- **gender**: factor; the person's gender (levels male and female).
- **ecoStat**: factor; the person's economic status (levels 1 = working full time, 2 = working part time, 3 = unemployed, 4 = pupil, student, further training or unpaid work experience or in compulsory military or community service, 5 = in retirement or early retirement or has given up business, 6 = permanently disabled or/and unfit to work or other inactive person, 7 = fulfilling domestic tasks and care responsibilities).
- **citizenship**: factor; the person's citizenship (levels AT, EU and Other).
- **py010n**: numeric; employee cash or near cash income (net).
- **py050n**: numeric; cash benefits or losses from self-employment (net).
- **py090n**: numeric; unemployment benefits (net).
- **py100n**: numeric; old-age benefits (net).
- **py110n**: numeric; survivor's benefits (net).
- **py120n**: numeric; sickness benefits (net).
- **py130n**: numeric; disability benefits (net).
- **py140n**: numeric; education-related allowances (net).
- **hy040n**: numeric; income from rental of a property or land (net).
- **hy050n**: numeric; family/children related allowances (net).
- **hy070n**: numeric; housing allowances (net).
hy080n numeric; regular inter-household cash transfer received (net).
hy090n numeric; interest, dividends, profit from capital investments in unincorporated business (net).
hy110n numeric; income received by people aged under 16 (net).
hy130n numeric; regular inter-household cash transfer paid (net).
hy145n numeric; repayments/receipts for tax adjustment (net).
main logical; indicates the main income holder (i.e., the person with the highest income) of each household.

details

The data set is used as population data in some of the examples in package simFrame. Note that it is included for illustrative purposes only. It consists of 25,000 households, hence it does not represent the true population sizes of Austria and its regions.

Only a few of the large number of variables in the original survey are included in this example data set. Some variable names are different from the standardized names used by the statistical agencies, as the latter are rather cryptic codes. Furthermore, the variables hsize, eqsize, eqIncome and age are not included in the standardized format of EU-SILC data, but have been derived from other variables for convenience. Moreover, some very sparse income components were not included in the generation of this synthetic data set. Thus the equivalized household income is computed from the available income components.

source

This is a synthetic data set based on Austrian EU-SILC data from 2006. The original sample was provided by Statistics Austria.

references


examples

data(eusilcP)
summary(eusilcP)
Format

A data frame with 11725 observations on the following 18 variables.

- **db030**: integer; the household ID.
- **hsize**: integer; the number of persons in the household.
- **db040**: factor; the federal state in which the household is located (levels Burgenland, Carinthia, Lower Austria, Salzburg, Styria, Tyrol, Upper Austria, Vienna and Vorarlberg).
- **age**: integer; the person’s age.
- **rb090**: factor; the person’s gender (levels male and female).
- **pl030**: factor; the person’s economic status (levels 1 = working full time, 2 = working part time, 3 = unemployed, 4 = pupil, student, further training or unpaid work experience or in compulsory military or community service, 5 = in retirement or early retirement or has given up business, 6 = permanently disabled or/and unfit to work or other inactive person, 7 = fulfilling domestic tasks and care responsibilities).
- **pb220a**: factor; the person’s citizenship (levels AT, EU and Other).
- **netIncome**: numeric; the personal net income.
- **py010n**: numeric; employee cash or near cash income (net).
- **py050n**: numeric; cash benefits or losses from self-employment (net).
- **py090n**: numeric; unemployment benefits (net).
- **py100n**: numeric; old-age benefits (net).
- **py110n**: numeric; survivor’s benefits (net).
- **py120n**: numeric; sickness benefits (net).
- **py130n**: numeric; disability benefits (net).
- **py140n**: numeric; education-related allowances (net).
- **db090**: numeric; the household sample weights.
- **rb050**: numeric; the personal sample weights.

Details

The data set consists of 4641 households and is used as sample data in some of the examples in package simPopulation. Note that it is included for illustrative purposes only. The sample weights do not reflect the true population sizes of Austria and its regions. The resulting population data is about 100 times smaller than the real population size to save computation time.

Only a few of the large number of variables in the original survey are included in this example data set. The variable names are rather cryptic codes, but these are the standardized names used by the statistical agencies. Furthermore, the variables hsize, age and netIncome are not included in the standardized format of EU-SILC data, but have been derived from other variables for convenience.

Source

This is a synthetic data set based on Austrian EU-SILC data from 2006. The original sample was provided by Statistics Austria.
References


Examples

```r
data(eusilcS)
summary(eusilcS)
```

---

**getBreaks**  
*Compute break points for categorizing (semi-)continuous variables*

**Description**

Compute break points for categorizing continuous or semi-continuous variables using (weighted) quantiles. This is a utility function that is useful for writing custom wrapper functions such as `simEUSILC`.

**Usage**

```r
getBreaks(
x,  
weights = NULL,  
zeros = TRUE,  
lower = NULL,  
upper = NULL,  
equidist = TRUE,  
probs = NULL,  
strata = NULL
)
```

**Arguments**

- **x**: a numeric vector to be categorized.
- **weights**: an optional numeric vector containing sample weights.
- **zeros**: a logical indicating whether `x` is semi-continuous, i.e., contains a considerable amount of zeros. See “Details” on how this affects the behavior of the function.
- **lower**, **upper**: optional numeric values specifying lower and upper bounds other than minimum and maximum of `x`, respectively.
- **equidist**: a logical indicating whether the (positive) break points should be equidistant or whether there should be refinements in the lower and upper tail (see “Details”).
- **probs**: a numeric vector of probabilities with values in [0, 1] giving quantiles to be used as (positive) break points. If supplied, this is preferred over `equidist`.
- **strata**: an optional vector specifying a strata variable (e.g., household ids), if specified, the mean of `x` (and also of `weights` if specified) is computed within each strata before calculating the breaks.
getCat

Details

If `equidist` is `TRUE`, the behavior is as follows. If `zeros` is `TRUE` as well, the 0%, 10%, ... , 90% quantiles of the negative values and the 10%, 20%, ... , 100% of the positive values are computed. These quantiles are then used as break points together with 0. If `zeros` is not `TRUE`, on the other hand, the 0%, 10%, ..., 100% quantiles of all values are used.

If `equidist` is not `TRUE`, the behavior is as follows. If `zeros` is not `TRUE`, the 1%, 5%, 10%, 20%, 40%, 60%, 80%, 90%, 95% and 99% quantiles of all values are used for the inner part of the data (instead of the equidistant 10%, ..., 90% quantiles). If `zeros` is `TRUE`, these quantiles are only used for the positive values while the quantiles of the negative values remain equidistant.

Note that duplicated values among the quantiles are discarded and that the minimum and maximum are replaced with `lower` and `upper`, respectively, if these are specified.

The (weighted) quantiles are computed with the function `quantileWt`.

Value

A numeric vector of break points.

Author(s)

Andreas Alfons and Bernhard Meindl

See Also

`getCat`, `quantileWt`

Examples

data(eusilcS)

# semi-continuous variable, positive break points equidistant
getBreaks(eusilcS$netIncome, weights=eusilcS$rb050)

# semi-continuous variable, positive break points not equidistant
getBreaks(eusilcS$netIncome, weights=eusilcS$rb050, equidist = FALSE)

getCat

Categorize (semi-)continuous variables

Description

Categorize continuous or semi-continuous variables. This is a utility function that is useful for writing custom wrapper functions such as `simEUSILC`.  
getCat

Usage

getCat(x, breaks, zeros = TRUE, right = FALSE)

Arguments

x
  a numeric vector to be categorized.
breaks
  a numeric vector of two or more break points.
zeros
  a logical indicating whether x is semi-continuous, i.e., contains a considerable amount of zeros. See “Details” on how this affects the behavior of the function.
right
  logical; if zeros is not TRUE, this indicates whether the intervals should be closed on the right (and open on the left) or vice versa.

Details

If zeros is TRUE, 0 is added to the break points and treated as its own factor level. Consequently, intervals for negative values are left-closed and right-open, whereas intervals for positive values are left-open and right-closed.

Value

A factor containing the categories.

Author(s)

Andreas Alfons

See Also

getBreaks, cut

Examples

data(eusilcS)

## semi-continuous variable
breaks <- getBreaks(eusilcS$netIncome,
  weights=eusilcS$rb050, equidist = FALSE)
netIncomeCat <- getCat(eusilcS$netIncome, breaks)
summary(netIncomeCat)
get_set-methods

Extract and modify variables from population or sample data stored in an object of class simPopObj-class.

Description

Using samp samp<- it is possible to extract or rather modify variables of the sample data within slot data in slot sample of the simPopObj-class-object. Using pop pop<- it is possible to extract or rather modify variables of the synthetic population within in slot data in slot sample of the simPopObj-class-object.

Arguments

- obj: An object of class simPopObj-class
- var: variable name or index for the variable in slot 'samp' of object with the slot name to be accessed. If NULL, the entire dataset (sample or population) is returned.
- value: Content replacing whatever the variable in slot var in obj currently holds.

Value

Returns an object of class simPopObj-class with the appropriate replacement.

Author(s)

Bernhard Meindl

See Also

simPopObj-class, pop, pop<-, samp<-, manageSimPopObj

Examples

data(eusilcS)
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
simPopObj <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))

## get/set variables in sample-object of simPopObj
head(samp(simPopObj, var="age"))
samp(simPopObj, var="newVar") <- 1
head(samp(simPopObj, var="newVar"))
## deleting is also possible
samp(simPopObj, var="newvar") <- NULL
head(samp(simPopObj, var="newvar"))
## extract multiple variables
head(samp(simPopObj, var=c("db030","db040")))
## get/set variables in pop-object of simPopObj

```r
head(pop(simPopObj, var="age"))
pop(simPopObj, var="newVar") <- 1
head(pop(simPopObj, var="newVar"))
```

## deleting is also possible

```r
pop(simPopObj, var="newvar") <- NULL
head(pop(simPopObj, var="newvar"))
```

## extract multiple variables

```r
head(pop(simPopObj, var=c("db030","db040")))
```

---

### GhanaS

**Synthetic GLSS survey data**

---

### Description

This data set is synthetically generated from real GLSS (Ghana Living Standards Survey) data.

### Format

A data frame with 36970 observations on the following 14 variables.

- **hhid** integer; the household ID.
- **hsize** integer; the number of persons in the household.
- **region** factor; the region in which the household is located (levels western, central, greater accra, volta, eastern, ashanti, brong ahafo, northern, upper east and upper west).
- **clust** factor; the enumeration area.
- **age** integer; the person’s age.
- **sex** factor; the person’s sex (levels male and female).
- **relate** factor; the relationship with the household head (levels head, spouse, child, grandchild, parent/parentlaw, son/daughterlaw, other relative, adopted child, househelp and non_relative).
- **nation** factor; the person’s nationality (levels ghanaian birth, ghanaian naturalise, burkinabe, malian, nigerian, ivorian, togolese, liberian, other ecowas, other africa and other).
- **ethnic** factor; the person’s ethnicity (levels akan, all other tribes, ewe, ga-dangbe, grusi, guan, gurma, mande and mole-dagbani).
- **religion** factor; the person’s religion (levels catholic, anglican, presbyterian, methodist, pentecostal, spiritualist, other christian, moslem, traditional, no religion and other).
- **highest_degree** factor; the person’s highest degree of education (levels none, mlsc, bece, voc/comm, teacher trng a, teacher trng b, gce ’o’ level, ssce, gce ’a’ level, tech/prof cert, tech/prof dip, hnd, bachelor, masters, doctorate and other).
occupation factor; the person’s occupation (levels armed forces and other security personnel, clerks, craft and related trades workers, elementary occupations, legislators, senior officials and managers, none, plant and machine operators and assemblers, professionals, service workers and shop and market sales workers, skilled agricultural and fishery workers, and technicians and associate professionals).

income numeric; the person’s annual income.

weight numeric; the sample weights.

Details
The data set consists of 8700 households and is used as sample data in some of the examples in package simPopulation. Note that it is included for illustrative purposes only. The sample weights do not reflect the true population sizes of Ghana and its regions. The resulting population data is about 100 times smaller than the real population size to save computation time.

Only some of the variables in the original survey are included in this example data set. Furthermore, categories are aggregated for certain variables due to the large number of possible outcomes in the original survey data.

Source
This is a synthetic data set based on GLSS data from 2006. The original sample was provided by Ghana Statistical Service.

References

Examples
```r
data(ghanaS)
summary(ghanaS)
```

---

ipu iterative proportional updating

Description
adjust sampling weights to given totals based on household-level and/or individual level constraints

Usage
```r
ipu(inp, con, hid = NULL, eps = 1e-07, verbose = FALSE)
```
Arguments

inp a data.frame or data.table containing household ids (optionally), counts for household and/or personal level attributes that should be fitted.

con named list with each list element holding a constraint total with list-names relating to column-names in inp.

hid character vector specifying the variable containing household-ids within inp or NULL if such a variable does not exist.

eps number specifying convergence limit

verbose if TRUE, ipu will print some progress information.

Author(s)

Bernhard Meindl

Examples

# basic example
inp <- as.data.frame(matrix(0, nrow=8, ncol=6))
colnames(inp) <- c("hhid", "hh1", "hh2", "p1", "p2", "p3")
inp$hhid <- 1:8
inp$hh1[1:3] <- 1
inp$hh2[4:8] <- 1
inp$p1 <- c(1, 1, 2, 1, 0, 1, 2, 1)
in$p2 <- c(1, 0, 1, 0, 2, 1, 1, 1)
in$p3 <- c(1, 1, 0, 2, 1, 0, 2, 0)
con <- list(hh1=35, hh2=65, p1=91, p2=65, p3=104)
res <- ipu(inp=inp, hid="hhid", con=con, verbose=FALSE)

# more sophisticated
# load sample and population data
data(eusilcS)
data(eusilcP)

# variable generation and preparation
eusilcS$hsize <- factor(eusilcS$hsize)
# make sure, factor levels in sample and population match
eusilcP$region <- factor(eusilcP$region, levels = levels(eusilcS$db040))
eusilcP$gender <- factor(eusilcP$gender, levels = levels(eusilcS$rb090))
eusilcP$hsize <- factor(eusilcP$hsize, levels = levels(eusilcS$hsize))

# generate input matrix
# we want to adjust to variable "db040" (region) as household variables and
# variable "rb090" (gender) as individual information
library(data.table)
samp <- data.table(eusilcS)
pop <- data.table(eusilcP)
setkeyv(samp, "db030")
hh <- samp[!duplicated(samp$db030),]
hhpop <- pop[!duplicated(pop$hid),]
# reg contains for each region the number of households
reg <- data.table(model.matrix(~db040 +0, data=hh))
# hsize contains for each household size the number of households
hsize <- data.table(model.matrix(~factor(hsize) +0, data=hh))

# aggregate persons-level characteristics per household
# gender contains for each household the number of males and females
gender <- data.table(model.matrix(~db030+rb090 +0, data=samp))
setkeyv(gender, "db030")
gender <- gender[, lapply(.SD, sum), by = key(gender)]

# bind together and use it as input
inp <- cbind(reg, hsize, gender)

# the totals we want to calibrate to
con <- c(
  as.list(xtabs(rep(1, nrow(hhpop)) ~ hhpop$region)),
  as.list(xtabs(rep(1, nrow(hhpop)) ~ hhpop$hsize)),
  as.list(xtabs(rep(1, nrow(eusilcP)) ~ eusilcP$gender))
)
# we need to have the same names as in ‘inp’
names(con) <- setdiff(names(inp), "db030")

# run ipu and check results
res <- ipu(inp=inp, hid="db030", con=con, verbose=TRUE)
is <- sapply(2:(ncol(res)-1), function(x) {
  sum(res[,x]*res$weights)
})
data.frame(required=unlist(con), is=is)

---

**ipu2**

**Iterative Proportional Updating**

**Description**

Adjust sampling weights to given totals based on household-level and/or individual level constraints.

**Usage**

```r
ipu2(
  dat,
  hid = NULL,
  conP = NULL,
  conH = NULL,
  epsP = 1e-06,
  epsH = 0.01,
```
verbose = FALSE,
w = NULL,
bound = 4,
maxIter = 200,
meanHH = TRUE,
allPthenH = TRUE,
returnNA = TRUE,
looseH = FALSE,
numericalWeighting = NULL,
check_hh_vars = TRUE,
conversion_messages = FALSE
)

Arguments

dat a data.table containing household ids (optionally), base weights (optionally), household and/or personal level variables (numerical or categorical) that should be fitted.

hid name of the column containing the household-ids within dat or NULL if such a variable does not exist.

conP list or (partly) named list defining the constraints on person level. The list elements are contingency tables in array representation with dimnames corresponding to the names of the relevant calibration variables in dat. If a numerical variable is to be calibrated, the respective list element has to be named with the name of that numerical variable. Otherwise the list element should NOT be named.

conH list or (partly) named list defining the constraints on household level. The list elements are contingency tables in array representation with dimnames corresponding to the names of the relevant calibration variables in dat. If a numerical variable is to be calibrated, the respective list element has to be named with the name of that numerical variable. Otherwise the list element should NOT be named.

epsP numeric value or list (of numeric values and/or arrays) specifying the convergence limit(s) for conP. The list can contain numeric values and/or arrays which must appear in the same order as the corresponding constraints in conP. Also, an array must have the same dimensions and dimnames as the corresponding constraint in conP.

epsH numeric value or list (of numeric values and/or arrays) specifying the convergence limit(s) for conH. The list can contain numeric values and/or arrays which must appear in the same order as the corresponding constraints in conH. Also, an array must have the same dimensions and dimnames as the corresponding constraint in conH.

verbose if TRUE, some progress information will be printed.

w name if the column containing the base weights within dat or NULL if such a variable does not exist. In the latter case, every observation in dat is assigned a starting weight of 1.
**bound**

numeric value specifying the multiplier for determining the weight trimming boundary if the change of the base weights should be restricted, i.e. if the weights should stay between $1/\text{bound} \cdot w$ and $\text{bound} \cdot w$.

**maxIter**

numeric value specifying the maximum number of iterations that should be performed.

**meanHH**

if TRUE, every person in a household is assigned the mean of the person weights corresponding to the household. If "geometric", the geometric mean is used rather than the arithmetic mean.

**allPthenH**

if TRUE, all the person level calibration steps are performed before the household level calibration steps (and meanHH, if specified). If FALSE, the household level calibration steps (and meanHH, if specified) are performed after every person level calibration step. This can lead to better convergence properties in certain cases but also means that the total number of calibration steps is increased.

**returnNA**

if TRUE, the calibrated weight will be set to NA in case of no convergence.

**looseH**

if FALSE, the actual constraints conH are used for calibrating all the hh weights. If TRUE, only the weights for which the lower and upper thresholds defined by conH and epsH are exceeded are calibrated. They are however not calibrated against the actual constraints conH but against these lower and upper thresholds, i.e. $\text{conH} - \text{conH} \cdot \text{epsH}$ and $\text{conH} + \text{conH} \cdot \text{epsH}$.

**numericalWeighting**

If NULL computeLinear from the package survey sd will be used.

**check hh vars**

If TRUE check for non-unique values inside of a household for variables in household constraints

**conversion_messages**

show a message, if inputs need to be reformatted. This can be useful for speed optimizations if ipu2 is called several times with similar inputs (for example bootstrapping)

**Details**

This function implements the weighting procedure described here.

**conP** and **conH** are contingency tables, which can be created with xtabs. The dimnames of those tables should match the names and levels of the corresponding columns in dat.

**maxIter**, **epsP** and **epsH** are the stopping criteria. **epsP** and **epsH** describe relative tolerances in the sense that

$$1 - \text{epsP} < \frac{w_{i+1}}{w_i} < 1 + \text{epsP}$$

will be used as convergence criterium. Here $i$ is the iteration step and $w_i$ is the weight of a specific person at step $i$.

The algorithm performs best if all variables occurring in the constraints (conP and conH) as well as the household variable are coded as factor-columns in dat. Otherwise, conversions will be necessary which can be monitored with the conversion_messages argument. Setting check hh vars to FALSE can also increase the performance of the scheme.
**Value**

The function will return the input data dat with the calibrated weights `calibWeight` as an additional column as well as attributes. If no convergence has been reached in `maxIter` steps, and `returnNA` is `TRUE` (the default), the column `calibWeights` will only consist of NAs. The attributes of the table are attributes derived from the `data.table` class as well as the following.

- `converged` Did the algorithm converge in `maxIter` steps?
- `iterations` The number of iterations performed.
- `conP`, `conH`, `epsP`, `epsH` See Arguments.
- `conP_adj`, `conH_adj` Adjusted versions of `conP` and `conH`.
- `formP`, `formH` Formulas that were used to calculate `conP_adj` and `conH_adj` based on the output table.

**Author(s)**

Alexander Kowarik, Gregor de Cillia

**See Also**

`ipu`

**Examples**

```r
library(data.table)
data(eusilcS)
setDT(eusilcS)
eusilcS <- eusilcS[, list(db030, hsize, db040, age, rb090, netIncome, db090, rb050)]
```

```r
## rename columns
setnames(eusilcS, "rb090", "gender")
setnames(eusilcS, "db040", "state")
setnames(eusilcS, "db030", "household")
setnames(eusilcS, "rb050", "weight")
```

```r
## some recoding
# generate age groups
eusilcS[, agegroup := cut(age, c(-Inf, 10*1:9, Inf), right = FALSE)]
# some recoding of netIncome for reasons of simplicity
eusilcS[is.na(netIncome), netIncome := 0]
eusilcS[netIncome < 0, netIncome := 0]
# set hsize to 1,...,5+
eusilcS[, hsize := cut(hsize, c(0:4, Inf), labels = c(1:4, "5+"))]
# treat households as a factor variable
eusilcS[, household := as.factor(household)]
```

```r
## example for base weights assuming a simple random sample of households stratified per region
eusilcS[, regSamp := .N, by = state]
eusilcS[, regPop := sum(weight), by = state]
eusilcS[, baseWeight := regPop/regSamp]
```

```r
## constraints on person level
# age
```
kishFactor

conP1 <- xtabs(weight ~ agegroup, data = eusilcS)
# gender by region
conP2 <- xtabs(weight ~ gender + state, data = eusilcS)
# personal net income by gender
conP3 <- xtabs(weight*netIncome ~ gender, data = eusilcS)

## constraints on household level
conH1 <- xtabs(weight ~ hsize + state, data = eusilcS, subset = !duplicated(household))

# array of convergence limits for conH1
epsH1 <- conH1
epsH1[1:4,] <- 0.005
epsH1["5+",] <- 0.2

# without array epsH1
calibweights1 <- ipu2(eusilcS, hid = "household",
    conP = list(conP1, conP2, netIncome = conP3),
    conH = list(conH1),
    epsP = list(1e-06, 1e-06, 1e-03),
    epsH = 0.01,
    bound = NULL, verbose = TRUE, maxIter = 200)

# with array epsH1, base weights and bound
calibweights2 <- ipu2(eusilcS, hid = "household",
    conP = list(conP1, conP2),
    conH = list(conH1),
    epsP = 1e-06,
    epsH = list(epsH1),
    w = "baseWeight",
    bound = 4, verbose = TRUE, maxIter = 200)

# show an adjusted version of conP and the original
attr(calibweights2, "conP_adj")
attr(calibweights2, "conP")

---

kishFactor

Kish Factor

Description

Compute the kish factor for a specific weight vector

Usage

kishFactor(w)

Arguments

w

a numeric vector with weights
manageSimPopObj

**Value**

The function will return the kish factor

**Author(s)**

Alexander Kowarik

**Examples**

```r
kishFactor(rep(1,10))
kishFactor(rlnorm(10))
```

**Description**

This function allows to get or set variables in slots `pop` and `sample` of `simPopObj`-objects. This is a utility function that is useful for writing custom wrapper functions.

**Usage**

```r
manageSimPopObj(x, var, sample = FALSE, set = FALSE, values = NULL)
```

**Arguments**

- `x` an object of class `simPopObj`.
- `var` character vector of length 1; variable name that should be set or extracted.
- `sample` a logical indicating whether `var` should be extracted/set from slot `sample` (TRUE) or slot `pop` (FALSE).
- `set` logical; if TRUE, argument ‘values’ is set to either the sample or population data stored in ‘x’, depending on argument ‘sample’. If FALSE, the desired variable given by ‘var’ is returned from either the sample or the pop slot of ‘x’.
- `values` vector; if ‘set’ is TRUE, then this vector is used to update the variable of sample or population data depending of choice of argument ‘sample’.

**Value**

An object of class `simPopObj` (if ‘set’ is TRUE) or a vector (if ‘set’ is FALSE).

**Author(s)**

Bernhard Meindl and Matthias Templ
### quantileWt

**Weighted sample quantiles**

### Description

Compute quantiles taking into account sample weights. The following methods are implemented:

- `quantileWt.default(x, weights=NULL, probs=seq(0,1,0.25), na.rm=TRUE, ...)`
- `quantileWt.dataObj(x, vars, probs=seq(0,1,0.25), na.rm=TRUE, ...)`

Additional parameters are:

- `weights` an optional numeric vector containing sample weights.
- `vars` a character vector of length 1 specifying a variable name that is available in the data-slot of `x` and which is used for the calculation.
- `probs` a numeric vector of probabilities with values in $[0, 1]$.
- `na.rm` a logical indicating whether any NA or NaN values should be removed from `x` before the quantiles are computed. Note that the default is TRUE, contrary to the function `quantile`.

### Usage

```r
quantileWt(x, ...)
```

### Arguments

- `x` a numeric vector.
- `...` for the generic function `quantileWt` additional arguments to be passed to methods. Additional arguments not included in the definition of the methods are currently ignored.

### Details

If weights are not specified then `quantile(x, probs, na.rm=na.rm, names=FALSE, type=1)` is used for the computation.

Note probabilities outside $[0, 1]$ cause an error.

### Value

A vector of the (weighted) sample quantiles.
sampHH

Sample households from given microdata.

Description

The function samples households from microdata containing personal and household information.

Usage

sampHH(pop, sizefactor = 1, hid = "hid", strata = "region", hsize = NULL)

Arguments

pop data frame containing households and persons
sizefactor factor of how many times the initial population should be resampled
hid string specifying the name of the household-id variable in the data.
strata can be used to sample within strata.
hsize string specifying the name of the household size variable in the data.

details

households are drawn from the data and new ID’s are generated for the new households.

Value

the data frame of new households.
Author(s)
Bernhard Meindl, Matthias Templ and Johannes Gussenbauer

References

Examples
```r
data(eusilcP)
pop <- eusilcP
colnames(pop)[3] <- "hhsize"

system.time(x1 <- sampHH(pop, strata="region", hsize="hhsize"))
dim(x1)

## approx. 10 second computation time ...

system.time(x1 <- sampHH(pop, sizefactor=4, strata="region", hsize="hhsize"))
dim(x1)

system.time(x2 <- sampHH(pop, strata=NULL, hsize="hhsize"))

pop <- pop[,which(colnames(pop) == "hhsize")]

system.time(y1 <- sampHH(pop, strata="region", hsize=NULL))

system.time(y2 <- sampHH(pop, strata=NULL, hsize=NULL))
```

---

**silcTools2**

*Utility functions for EU-SILC data*

**Description**
Various utility functions mainly used for simulating EU-SILC data

**Usage**
```r
loadSILC(
    file = NULL,
    filed = NULL,
    filer = NULL,
    filep = NULL,
    fileh = NULL,
    year = 2013,
    country = "Austria"
)

mergeSILC(filed, filer, fileh, filep)
```
checkCol(x, y)

chooseSILCvars(
  x,
  vars = c("db030", "db040", "rb030", "rb080", "rb090", "pl031", "pb220a", "py010g", "py021g", "py050g", "py080g", "py090g", "py100g", "py110g", "py120g", "py130g", "py140g", "hy040g", "hy050g", "hy060g", "hy070g", "hy080g", "hy090g", "hy110g", "hy110g", "hy120g", "hy130g", "hy140g", "db090", "rb050", "pb190", "pe040", "pl051", "pl111", "rb010"),
  country = NULL
)

modifySILC(x, country = "Austria")

Arguments

- **file**: data set in R binary format, csv or sav (SPSS) of merged EU-SILC data.
- **filed**: data set including the household register information
- **filer**: data set including the personal register information
- **filep**: data set including the personal information
- **fileh**: data set including the household information
- **year**: year of origin
- **country**: country
- **x**: public-use file (for checkCol function) or orginal data
- **y**: scientific-use file (for checkCol function)
- **vars**: variables to be selected for function chooseSILCvars

Details

Collection of functions to import, select and modify data EU-SILC data. Either file (merged data) or single files have to be provided for loadSILC().

Author(s)

Matthias Templ

Examples

```r
## Not run:
x <- loadSILC("new_workfile.RData")
filed <- "zielvar_d_eurostat2013.sav"
filer <- "zielvar_r_eurostat2013.sav"
filep <- "zielvar_p_eurostat2013.sav"
fileh <- "zielvar_h_eurostat2013.sav"
suf4 <- loadSILC(filed = filed,
  filer = filer,
```
Simulate categorical variables of population data

**Description**

Simulate categorical variables of population data. The household structure of the population data needs to be simulated beforehand.
Usage

```r
simCategorical(
  simPopObj,
  additional,
  method = c("multinom", "distribution", "ctree", "cforest", "ranger"),
  limit = NULL,
  censor = NULL,
  maxit = 500,
  MaxNWts = 1500,
  eps = NULL,
  nr_cpus = NULL,
  regModel = NULL,
  seed = 1,
  verbose = FALSE,
  by = "strata"
)
```

Arguments

- **simPopObj**: a `simPopObj` containing population and household survey data as well as optionally margins in standardized format.
- **additional**: a character vector specifying additional categorical variables available in the sample object of `simPopObj` that should be simulated for the population data.
- **method**: a character string specifying the method to be used for simulating the additional categorical variables. Accepted values are "multinom" (estimation of the conditional probabilities using multinomial log-linear models and random draws from the resulting distributions) or "distribution" (random draws from the observed conditional distributions of their multivariate realizations). "ctree" for using Classification trees "cforest" for using random forest (implementation in package party) "ranger" for using random forest (implementation in package ranger).
- **limit**: if `method` is "multinom", this can be used to account for structural zeros. If only one additional variable is requested, a named list of lists should be supplied. The names of the list components specify the predictor variables for which to limit the possible outcomes of the response. For each predictor, a list containing the possible outcomes of the response for each category of the predictor can be supplied. The probabilities of other outcomes conditional on combinations that contain the specified categories of the supplied predictors are set to 0. If more than one additional variable is requested, such a list of lists can be supplied for each variable as a component of yet another list, with the component names specifying the respective variables.
- **censor**: if `method` is "multinom", this can be used to account for structural zeros. If only one additional variable is requested, a named list of lists or data.frames should be supplied. The names of the list components specify the categories that should be censored. For each of these categories, a list or data.frame containing levels of the predictor variables can be supplied. The probability of the specified categories is set to 0 for the respective predictor levels. If more
than one additional variable is requested, such a list of lists or data.frames can be supplied for each variable as a component of yet another list, with the component names specifying the respective variables.

maxit, MaxNWts control parameters to be passed to `multinom` and `nnet`. See the help file for `nnet`.

eps a small positive numeric value, or NULL (the default). In the former case and if method is "multinom", estimated probabilities smaller than this are assumed to result from structural zeros and are set to exactly 0.

nr_cpus if specified, an integer number defining the number of cpus that should be used for parallel processing.

regModel allows to specify the variables or model that is used when simulating additional categorical variables. The following choices are available if different from NULL.

- 'basic' only the basic household variables (generated with `simStructure`) are used.
- 'available' all available variables (that are common in the sample and the synthetic population such as previously generated varaibles) excluding id-variables, strata variables and household sizes are used for the modelling. This parameter should be used with care because all factors are automatically used as factors internally.
- formula-object Users may also specify a specific formula (class 'formula') that will be used. Checks are performed that all required variables are available.

If method 'distribution' is used, it is only possible to specify a vector of length one containing one of the choices described above. If parameter ‘regModel’ is NULL, only basic household variables are used in any case.

seed optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.

verbose set to TRUE if additional print output should be shown.

by defining which variable to use as split up variable of the estimation. Defaults to the strata variable.

Details

The number of cpus are selected automatically in the following manner. The number of cpus is equal the number of strata. However, if the number of cpus is less than the number of strata, the number of cpus - 1 is used by default. This should be the best strategy, but the user can also overwrite this decision.

Value

An object of class `simPopObj` containing survey data as well as the simulated population data including the categorical variables specified by argument `additional`.
simComponents

Note

The basic household structure needs to be simulated beforehand with the function simStructure.

Author(s)

Bernhard Meindl, Andreas Alfons, Stefan Kraft, Alexander Kowarik, Matthias Templ

References


See Also

simStructure, simRelation, simContinuous, simComponents

Examples

data(eusilcS) # load sample data

## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
## in the following, nr_cpus are selected automatically
simPop <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
simPop <- simCategorical(simPop, additional=c("pl030", "pb220a"), method="multinom", nr_cpus=1)
simPop

---

**simComponents**  
*Simulate components of continuous variables of population data*

**Description**

Simulate components of continuous variables of population data by resampling fractions from survey data. The continuous variable to be split and any categorical conditioning variables need to be simulated beforehand.

**Usage**

```r
simComponents(
  simPopObj,  
  total = "netIncome",  
  components = c("py010n", "py050n", "py090n", "py100n", "py110n", "py120n", "py130n", "py140n"),
)```

conditional = c(getCatName(total), "pl030"),
replaceEmpty = c("sequential", "min"),
seed
}

**Arguments**

- **simPopObj** a `simPopObj`-object.
- **total** a character string specifying the continuous variable of dataP that should be split into components. Currently, only one variable can be split at a time.
- **components** a character vector specifying the components in dataS that should be simulated for the population data.
- **conditional** an optional character vector specifying categorical conditioning variables for resampling. The fractions occurring in dataS are then drawn from the respective subsets defined by these variables.
- **replaceEmpty** a character string; if conditional specifies at least two conditioning variables, this determines how replacement cells for empty subsets in the sample are obtained. If "sequential", the conditioning variables are browsed sequentially such that replacement cells have the same value in one conditioning variable and minimum Manhattan distance in the other conditioning variables. If no such cells exist, replacement cells with minimum overall Manhattan distance are selected. The latter is always done if this is "min" or only one conditioning variable is used.
- **seed** optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.

**Value**

An object of class `simPopObj` containing survey data as well as the simulated population data including the components of the continuous variable specified by total and components.

**Note**

The basic household structure, any categorical conditioning variables and the continuous variable to be split need to be simulated beforehand with the functions `simStructure`, `simCategorical` and `simContinuous`.

**Author(s)**

Stefan Kraft and Andreas Alfons and Bernhard Meindl

**References**


**See Also**

simStructure, simCategorical, simContinuous, simEUSILC

**Examples**

data(eusilcS)

```r
## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsiz="hsize",
                     strata="db040", weight="db090")
simPopObj <- simStructure(data=inp, method="direct",
                         basicHVars=c("age", "rb090", "hsize", "pl030", "pb220a"))
simPopObj <- simContinuous(simPopObj, additional = "netIncome",
                          regModel = ~rb090+hsize+pl030+pb220a+hsize,
                          method="multinom", upper=200000, equidist=FALSE, nr_cpus=1)

# categorize net income for use as conditioning variable
sIncome <- manageSimPopObj(simPopObj, var="netIncome", sample=TRUE, set=FALSE)
sWeight <- manageSimPopObj(simPopObj, var="rb050", sample=TRUE, set=FALSE)
pIncome <- manageSimPopObj(simPopObj, var="netIncome", sample=FALSE, set=FALSE)
breaks <- getBreaks(x=unlist(sIncome), w=unlist(sWeight), upper=Inf, equidist=FALSE)
simPopObj <- manageSimPopObj(simPopObj, var="netIncomeCat", sample=TRUE,
                              values=getCat(x=unlist(sIncome), breaks))
simPopObj <- manageSimPopObj(simPopObj, var="netIncomeCat", sample=FALSE,
                              values=getCat(x=unlist(pIncome), breaks))

# simulate net income components
simPopObj <- simComponents(simPopObj=simPopObj, total="netIncome",
                          components=c("py010n","py050n","py090n","py100n","py110n","py120n","py130n","py140n"),
                          conditional = c("netIncomeCat", "pl030"), replaceEmpty = "sequential", seed=1 )

class(simPopObj)
```

---

**simContinuous**

*Simulate continuous variables of population data*

**Description**

Simulate continuous variables of population data using multinomial log-linear models combined with random draws from the resulting categories or (two-step) regression models combined with random error terms. The household structure of the population data and any other categorical predictors need to be simulated beforehand.
Usage

simContinuous(
    simPopObj, additional = "netIncome",
    method = c("multinom", "lm", "poisson"),
    zeros = TRUE,
    breaks = NULL,
    lower = NULL,
    upper = NULL,
    equidist = TRUE,
    probs = NULL,
    gpd = TRUE,
    threshold = NULL,
    est = "moments",
    limit = NULL,
    censor = NULL,
    log = TRUE,
    const = NULL,
    alpha = 0.01,
    residuals = TRUE,
    keep = TRUE,
    maxit = 500,
    MaxNWts = 1500,
    tol = .Machine$double.eps^0.5,
    nr_cpus = NULL,
    eps = NULL,
    regModel = "basic",
    byHousehold = NULL,
    imputeMissings = FALSE,
    seed,
    verbose = FALSE,
    by = "strata"
)

Arguments

simPopObj a simPopObj holding household survey data, population data and optionally some margins.
additional a character string specifying the additional continuous variable of dataS that should be simulated for the population data. Currently, only one additional variable can be simulated at a time.
method a character string specifying the method to be used for simulating the continuous variable. Accepted values are "multinom", for using multinomial log-linear models combined with random draws from the resulting categories, "lm", for using (two-step) regression models combined with random error terms and "poisson" for using Poisson regression for count variables.
zeros a logical indicating whether the variable specified by additional is semi-continuous, i.e., contains a considerable amount of zeros. If TRUE and method is "multinom",...
a separate factor level for zeros in the response is used. If TRUE and method is "lm", a two-step model is applied. The first step thereby uses a log-linear or multinomial log-linear model (see “Details”).

breaks
an optional numeric vector; if multinomial models are computed, this can be used to supply two or more break points for categorizing the variable specified by additional. If NULL, break points are computed using weighted quantiles.

lower, upper
optional numeric values; if multinomial models are computed and breaks is NULL, these can be used to specify lower and upper bounds other than minimum and maximum, respectively. Note that if method is "multinom" and gpd is TRUE (see below), upper defaults to Inf.

equidist
logical; if method is "multinom" and breaks is NULL, this indicates whether the (positive) default break points should be equidistant or whether there should be refinements in the lower and upper tail (see getBreaks).

probs
numeric vector with values in [0, 1]; if method is "multinom" and breaks is NULL, this gives probabilities for quantiles to be used as (positive) break points. If supplied, this is preferred over equidist.

gpd
logical; if method is "multinom", this indicates whether the upper tail of the variable specified by additional should be simulated by random draws from a (truncated) generalized Pareto distribution rather than a uniform distribution.

threshold
a numeric value; if method is "multinom", values for categories above threshold are drawn from a (truncated) generalized Pareto distribution.

est
a character string; if method is "multinom", the estimator to be used to fit the generalized Pareto distribution.

limit
an optional named list of lists; if multinomial models are computed, this can be used to account for structural zeros. The names of the list components specify the predictor variables for which to limit the possible outcomes of the response. For each predictor, a list containing the possible outcomes of the response for each category of the predictor can be supplied. The probabilities of other outcomes conditional on combinations that contain the specified categories of the supplied predictors are set to 0. Currently, this is only implemented for more than two categories in the response.

censor
an optional named list of lists or dataframes; if multinomial models are computed, this can be used to account for structural zeros. The names of the list components specify the categories that should be censored. For each of these categories, a list or dataframe containing levels of the predictor variables can be supplied. The probability of the specified categories is set to 0 for the respective predictor levels. Currently, this is only implemented for more than two categories in the response.

log
logical; if method is "lm", this indicates whether the linear model should be fitted to the logarithms of the variable specified by additional. The predicted values are then back-transformed with the exponential function. See “Details” for more information.

const
numeric; if method is "lm" and log is TRUE, this gives a constant to be added before log transformation.
alpha numeric; if method is "lm", this gives trimming parameters for the sample data. Trimming is thereby done with respect to the variable specified by additional. If a numeric vector of length two is supplied, the first element gives the trimming proportion for the lower part and the second element the trimming proportion for the upper part. If a single numeric is supplied, it is used for both. With NULL, trimming is suppressed.

residuals logical; if method is "lm", this indicates whether the random error terms should be obtained by draws from the residuals. If FALSE, they are drawn from a normal distribution (median and MAD of the residuals are used as parameters).

keep logical; if multinomial models are computed, this indicates whether the simulated categories should be stored as a variable in the resulting population data. If TRUE, the corresponding column name is given by additional with postfix "Cat".

maxit, MaxNWts control parameters to be passed to multinom and nnet. See the help file for nnet.

tol if method is "lm" and zeros is TRUE, a small positive numeric value or NULL. When fitting a log-linear model within a stratum, factor levels may not exist in the sample but are likely to exist in the population. However, the coefficient for such factor levels will be 0. Therefore, coefficients smaller than tol in absolute value are replaced by coefficients from an auxiliary model that is fit to the whole sample. If NULL, no auxiliary log-linear model is computed and no coefficients are replaced.

nr_cpus if specified, an integer number defining the number of cpus that should be used for parallel processing.

eps a small positive numeric value, or NULL (the default). In the former case and if (multinomial) log-linear models are computed, estimated probabilities smaller than this are assumed to result from structural zeros and are set to exactly 0.

regModel allows to specify the model that should be for the simulation of the additional continuous variable. The following choices are possible:

- 'basic': only the basic household-variables (generated with simStructure) are used.
- 'available': all available variables (that are common in the sample and the synthetic population (e.g. previously generated variables) are used for the modeling. Should be used with care because all variables are automatically used as factors!
- formula-object: Users may also specify a specific formula (class 'formula') that will be used. Checks are performed that all required variables are available.

byHousehold if NULL, simulated values are used as is. If either 'sum', 'mean' or 'random' is specified, the values are aggregated and each member of the household gets the same value (mean, sum or a random value) assigned.

imputeMissings if TRUE, missing values in variables that are used for the underlying model are imputed using hock-deck.

seed optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.
verbose (logical) if TRUE, additional output is written to the prompt by defining which variable to use as split up variable of the estimation. Defaults to the strata variable.

Details

If method is "lm", the behavior for two-step models is described in the following.

If zeros is TRUE and log is not TRUE or the variable specified by additional does not contain negative values, a log-linear model is used to predict whether an observation is zero or not. Then a linear model is used to predict the non-zero values.

If zeros is TRUE, log is TRUE and const is specified, again a log-linear model is used to predict whether an observation is zero or not. In the linear model to predict the non-zero values, const is added to the variable specified by additional before the logarithms are taken.

If zeros is TRUE, log is TRUE, const is NULL and there are negative values, a multinomial log-linear model is used to predict negative, zero and positive observations. Categories for the negative values are thereby defined by breaks. In the second step, a linear model is used to predict the positive values and negative values are drawn from uniform distributions in the respective classes.

If zeros is FALSE, log is TRUE and const is NULL, a two-step model is used if there are non-positive values in the variable specified by additional. Whether a log-linear or a multinomial log-linear model is used depends on the number of categories to be used for the non-positive values, as defined by breaks. Again, positive values are then predicted with a linear model and non-positive values are drawn from uniform distributions.

The number of cpus are selected automatically in the following manner. The number of cpus is equal the number of strata. However, if the number of cpus is less than the number of strata, the number of cpus - 1 is used by default. This should be the best strategy, but the user can also overwrite this decision.

Value

An object of class simPopObj containing survey data as well as the simulated population data including the continuous variable specified by additional and possibly simulated categories for the desired continous variable.

Note

The basic household structure and any other categorical predictors need to be simulated beforehand with the functions simStructure and simCategorical, respectively.

Author(s)

Bernhard Meindl, Andreas Alfons, Alexander Kowarik (based on code by Stefan Kraft)

References

**simEUSILC**


See Also

*simStructure, simCategorical, simComponents, simEUSILC*

### Examples

```r
data(eusilcS)
## approx. 20 seconds computation time
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
simPop <- simStructure(data=inp, method="direct", 
    basicHHvars=c("age", "rb090", "hsize", "pl030", "pb220a"))
regModel = ~rb090+hsize+pl030+pb220a
# multinomial model with random draws
eusilcM <- simContinuous(simPop, additional="netIncome",
    regModel = regModel,  
    upper=200000, equidist=FALSE, nr_cpus=1)
class(eusilcM)

# two-step regression
eusilcT <- simContinuous(simPop, additional="netIncome",  
    regModel = "basic",  
    method = "lm")
class(eusilcT)
```

---

**simEUSILC**

*Simulate EU-SILC population data*

**Description**

Simulate population data for the European Statistics on Income and Living Conditions (EU-SILC).

**Usage**

```r
simEUSILC(
    dataS,  
    hid = "db030",  
    wh = "db090",  
```

```
wp = "rb050",
hsize = NULL,
strata = "db040",
pid = NULL,
age = "age",
gender = "rb090",
categorizeAge = TRUE,
breaksAge = NULL,
categorical = c("pl030", "pb220a"),
iccme = "netIncome",
method = c("multinom", "twostep"),
breaks = NULL,
lower = NULL,
upper = NULL,
equidist = TRUE,
probs = NULL,
gpd = TRUE,
threshold = NULL,
est = "moments",
const = NULL,
alpha = 0.01,
residuals = TRUE,
components = c("py010n", "py050n", "py090n", "py100n", "py110n", "py120n", "py130n", "py140n"),
conditional = c(getCatName(income), "pl030"),
keep = TRUE,
maxit = 500,
MaxNWts = 1500,
tol = .Machine$double.eps^0.5,
seed
)

Arguments

dataS a data.frame containing EU-SILC survey data.
hid a character string specifying the column of dataS that contains the household ID.
wh a character string specifying the column of dataS that contains the household sample weights.
wp a character string specifying the column of dataS that contains the personal sample weights.
hsize an optional character string specifying a column of dataS that contains the household size. If NULL, the household sizes are computed.
strata a character string specifying the column of dataS that define strata. Note that this is currently a required argument and only one stratification variable is supported.
pid an optional character string specifying a column of dataS that contains the personal ID.
age  a character string specifying the column of dataS that contains the age of the persons (to be used for setting up the household structure).

gender  a character string specifying the column of dataS that contains the gender of the persons (to be used for setting up the household structure).

categorizeAge  a logical indicating whether age categories should be used for simulating additional categorical and continuous variables to decrease computation time.

breaksAge  numeric; if categorizeAge is TRUE, an optional vector of two or more break points for constructing age categories, otherwise ignored.

categorical  a character vector specifying additional categorical variables of dataS that should be simulated for the population data.

income  a character string specifying the variable of dataS that contains the personal income (to be simulated for the population data).

method  a character string specifying the method to be used for simulating personal income. Accepted values are "multinom" (for using multinomial log-linear models combined with random draws from the resulting categories) and "twostep" (for using two-step regression models combined with random error terms).

breaks  if method is "multinom", an optional numeric vector of two or more break points for categorizing the personal income. If missing, break points are computed using weighted quantiles.

lower, upper  numeric values; if method is "multinom" and breaks is NULL, these can be used to specify lower and upper bounds other than minimum and maximum, respectively. Note that if gpd is TRUE (see below), upper defaults to Inf.

equidist  logical; if method is "multinom" and breaks is NULL, this indicates whether the (positive) default break points should be equidistant or whether there should be refinements in the lower and upper tail (see getBreaks).

probs  numeric vector with values in \([0, 1]\); if method is "multinom" and breaks is NULL, this gives probabilities for quantiles to be used as (positive) break points. If supplied, this is preferred over equidist.

gpd  logical; if method is "multinom", this indicates whether the upper tail of the personal income should be simulated by random draws from a (truncated) generalized Pareto distribution rather than a uniform distribution.

threshold  a numeric value; if method is "multinom", values for categories above threshold are drawn from a (truncated) generalized Pareto distribution.

est  a character string; if method is "multinom", the estimator to be used to fit the generalized Pareto distribution.

const  numeric; if method is "twostep", this gives a constant to be added before log transformation.

alpha  numeric; if method is "twostep", this gives trimming parameters for the sample data. Trimming is thereby done with respect to the variable specified by additional. If a numeric vector of length two is supplied, the first element gives the trimming proportion for the lower part and the second element the trimming proportion for the upper part. If a single numeric is supplied, it is used for both. With NULL, trimming is suppressed.
residuals logical; if method is "twostep", this indicates whether the random error terms should be obtained by draws from the residuals. If FALSE, they are drawn from a normal distribution (median and MAD of the residuals are used as parameters).

components a character vector specifying the income components in dataS (to be simulated for the population data).

conditional an optional character vector specifying categorical conditioning variables for resampling of the income components. The fractions occurring in dataS are then drawn from the respective subsets defined by these variables.

keep a logical indicating whether variables computed internally in the procedure (such as the original IDs of the corresponding households in the underlying sample, age categories or income categories) should be stored in the resulting population data.

maxit, MaxNWts control parameters to be passed to multinom and nnet. See the help file for nnet.

tol if method is "twostep", a small positive numeric value or NULL (see simContinuous).

seed optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.

Value

An object of class simPopObj containing the simulated EU-SILC population data as well as the underlying sample.

Note

This is a wrapper calling simStructure, simCategorical, simContinuous and simComponents.

Author(s)

Andreas Alfons and Stefan Kraft and Bernhard Meindl

See Also

simStructure, simCategorical, simContinuous, simComponents

Examples

data(eusilcS) # load sample data

## long computation time
# multinomial model with random draws
eusilcM <- simEUSILC(eusilcS, upper = 200000, equidist = FALSE)
summary(eusilcM)

# two-step regression
eusilcT <- simEUSILC(eusilcS, method = "twostep")
**simInitSpatial**

**Generation of smaller regions given an existing spatial variable and a table.**

**Description**

This function allows to manipulate an object of class `simPopObj` in a way that a new variable containing smaller regions within an already existing broader region is generated. The distribution of the smaller region within the broader region is respected.

**Usage**

```r
simInitSpatial(
  simPopObj,
  additional,
  region,
  tspatialP = NULL,
  tspatialHH = NULL,
  eps = 0.05,
  maxIter = 100,
  nr_cpus = NULL,
  seed = 1,
  verbose = FALSE
)
```

**Arguments**

- `simPopObj`: an object of class `simPopObj`.
- `additional`: a character vector of length one holding the variable name of the variable containing smaller geographical units. This variable name must be available as a column in input argument `tspatial`.
- `region`: a character vector of length one holding the variable name of the broader region. This variable must be available in the input `tspatial` as well as in the sample and population slots of input `simPopObj`.
- `tspatialP`: a data.frame (or data.table) containing three columns. The broader region (with the variable name being the same as in input `region`), the smaller geographical units (with the variable name being the same as in input `additional`) and a third column containing a numeric vector holding counts of persons. This argument or `tspatialHH` has to be provided.
- `tspatialHH`: a data.frame (or data.table) containing three columns. The broader region (with the variable name being the same as in input `region`), the smaller geographical units (with the variable name being the same as in input `additional`) and a third column containing a numeric vector holding counts of households. This argument or `tspatialP` has to be provided.
**Details**

The distributional information must be contained in an input table that holds combinations of characteristics of the broader region and the smaller regions as well as population counts (which may be available from a census).

**Value**

An object of class `simPopObj` with an additional variable in the synthetic population slot.

**Author(s)**

Bernhard Meindl and Alexander Kowarik

**References**


**Examples**

```r
data(eusilcS)
data(eusilcP)
library(data.table)

# no districts are available in the population, so we have to generate those
# we randomly assign districts within "region" in the eusilc population data
# each hh has the same district
simulate_districts <- function(inp) {
  hhid <- "hid"
  region <- "region"

  a <- inp[!duplicated(inp[,hhid]),c(hhid, region)]
  spl <- split(a, a[,region])
  regions <- unique(inp[,region])

  tmpres <- lapply(1:length(spl), function(x) {
    codes <- paste(x, 1:sample(3:9,1), sep="")
    spl[[x]]$district <- sample(codes, nrow(spl[[x]]), replace=TRUE)
  })

  tmpres[[1]]
}
```
simple_dis

Simple generation of new variables

Description

Fast simulation of new variables based on univariate distributions

Usage

univariate.dis(puf, data, additional, weights, value = "data", fNA = NA)

conditional.dis(}
simple_dis

puf, data, additional, conditional, weights, value = "data", fNA = NA
}

Arguments

puf data for which one additional column specified by function argument 'additional' is simulated
data donor data
additional name of variable to be simulated
weights sampling weights from data
value if "data" then the puf including the additional variable is returned, otherwise only the simulated vector.
fNA only used with missing values if another code as NA should be used
conditional conditioning variable

Details

Function uni.distribution: random draws from the weighted univariate distribution of the original data
Function conditional.dis: random draws from the weighted conditional distribution (conditioned on a factor variable)
This are simple functions to produce structural variables, variables that should have the same categories as given ones. For more advanced methods see simCategorical()

Author(s)

Lydia Spies, Matthias Templ

See Also

simCategorical

Examples

## we don't have original data, so let's use eusilc
data(eusilc13puf)
data(eusilcS)
v1 <- univariate.dis(eusilcS, eusilc13puf, additional = "db040",
weights = "rb050", value = "vector")
table(v1)
table(eusilc13puf$db040)
## we don't have original data, so let's use eusilc
##data(eusilc13puf)
##data(eusilcS)
##v1 <- conditional.dis(eusilcS, eusilc13puf, additional = "pb190",
## conditional = "db040", weights = "rb050")
##table(v1) / sum(table(v1))
##table(eusilc13puf$pb190) / sum(table(eusilc13puf$pb190))

---

###simPopObj-class

Class "simPopObj"

####Description

An object that is used throughout the package containing information on the sample (in slot `sample`),
the population (slot `pop`) and optionally some margins in form of a table (slot `table`).

####Objects from the Class

Objects are automatically created in function `simStructure`.

####Author(s)

Bernhard Meindl and Matthias Templ

####See Also

dataObj

####Examples

```r
showClass("simPopObj")

## show method: generate an object of class simPop first
data(eusilcS)
inp <- specifyInput(data=eusilcS, hhid="db030", hhsiz="hsize", strata="db040", weight="db090")
eusilcP <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
class(eusilcP)
## shows some basic information:
eusilcP
```
simRelation

Simulate categorical variables of population data

Description

Simulate categorical variables of population data taking relationships between household members into account. The household structure of the population data needs to be simulated beforehand using `simStructure()`.

Usage

```r
simRelation(
  simPopObj,
  relation = "relate",
  head = "head",
  direct = NULL,
  additional,
  limit = NULL,
  censor = NULL,
  maxit = 500,
  MaxNWts = 2000,
  eps = NULL,
  nr_cpus = NULL,
  seed = 1,
  regModel = NULL,
  verbose = FALSE,
  method = c("multinom", "ctree", "cforest", "ranger"),
  by = "strata"
)
```

Arguments

- `simPopObj` a `simPopObj` containing population and household survey data as well as optionally margins in standardized format.
- `relation` a character string specifying the columns of `dataS` and `dataP`, respectively, that define the relationships between the household members.
- `head` a character string specifying the category of the variable given by `relation` that identifies the household head.
- `direct` a character string specifying categories of the variable given by `relation`. Simulated individuals with those categories directly inherit the values of the additional variables from the household head. The default is `NULL` such that no individuals directly inherit value from the household head.
- `additional` a character vector specifying additional categorical variables of `dataS` that should be simulated for the population data.
This can be used to account for structural zeros. If only one additional variable is requested, a named list of lists should be supplied. The names of the list components specify the predictor variables for which to limit the possible outcomes of the response. For each predictor, a list containing the possible outcomes of the response for each category of the predictor can be supplied. The probabilities of other outcomes conditional on combinations that contain the specified categories of the supplied predictors are set to 0. If more than one additional variable is requested, such a list of lists can be supplied for each variable as a component of yet another list, with the component names specifying the respective variables.

Censor

This can be used to account for structural zeros. If only one additional variable is requested, a named list of lists or data.frames should be supplied. The names of the list components specify the categories that should be censored. For each of these categories, a list or data.frame containing levels of the predictor variables can be supplied. The probability of the specified categories is set to 0 for the respective predictor levels. If more than one additional variable is requested, such a list of lists or data.frames can be supplied for each variable as a component of yet another list, with the component names specifying the respective variables.

Control parameters to be passed to \texttt{nnet::multinom()} and \texttt{nnet::nnet()}. See the help file for \texttt{nnet::nnet()}.

A small positive numeric value, or \texttt{NULL} (the default). In the former case, estimated probabilities smaller than this are assumed to result from structural zeros and are set to exactly 0.

If specified, an integer number defining the number of cpus that should be used for parallel processing.

Optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.

Allows to specify the variables or model that is used when simulating additional categorical variables. The following choices are available if different from \texttt{NULL}.

- "basic": only the basic household variables (generated with \texttt{simStructure()}) are used.
- "available": all available variables (that are common in the sample and the synthetic population such as previously generated variables) excluding id-variables, strata variables and household sizes are used for the modeling. This parameter should be used with care because all factors are automatically used as factors internally.
- formula-object: users may also specify a formula (class 'formula') that will be used. Checks are performed that all required variables are available. If parameter \texttt{regModel} is \texttt{NULL}, only basic household variables are used in any case.

Set to \texttt{TRUE} if additional print output should be shown.

A character string specifying the method to be used for simulating the additional categorical variables. Accepted values are

- "multinom": estimation of the conditional probabilities using multinomial log-linear models and random draws from the resulting distributions
• "ctree": for using Classification trees
• "cforest": for using random forest (implementation in package party)
• "ranger": for using random forest (implementation in package ranger)
by defining which variable to use as split up variable of the estimation. Defaults to the strata variable.

Details

The values of a new variable are simulated in three steps, where the second step is optional. First, the values of the household heads are simulated with multinomial log-linear models. Second, individuals directly related to the corresponding household head (as specified by the argument direct) inherit the value of the latter. Third, the values of the remaining individuals are simulated with multinomial log-linear models in which the value of the respective household head is used as an additional predictor.

The number of cpus are selected automatically in the following manner. The number of cpus is equal the number of strata. However, if the number of cpus is less than the number of strata, the number of cpus - 1 is used by default. This should be the best strategy, but the user can also overwrite this decision.

Value

An object of class simPopObj containing survey data as well as the simulated population data including the categorical variables specified by additional.

Note

The basic household structure needs to be simulated beforehand with the function simStructure().

Author(s)

Andreas Alfons and Bernhard Meindl

See Also

simStructure(), simCategorical(), simContinuous(), simComponents()

Examples

data(ghanaS) # load sample data
samp <- specifyInput(
  data = ghanaS,
  hhid = "hhid",
  strata = "region",
  weight = "weight"
)
ghanaP <- simStructure(
  data = samp,
  method = "direct",
  basicHHvars = c("age", "sex", "relate")
)
Simulate the household structure of population data

Description

Simulate basic categorical variables that define the household structure (typically variables such as household ID, age and gender) of population data by resampling from survey data.

Usage

```
simStructure(
  dataS,
  method = c("direct", "multinom", "distribution"),
  basicHHvars,
  seed = 1
)
```

Arguments

dataS: an object of class dataObj containing household survey data that is usually generated with `specifyInput`.

method: a character string specifying the method to be used for simulating the household sizes. Accepted values are "direct" (estimation of the population totals for each combination of stratum and household size using the Horvitz-Thompson estimator), "multinom" (estimation of the conditional probabilities within the strata using a multinomial log-linear model and random draws from the resulting distributions), or "distribution" (random draws from the observed conditional distributions within the strata).

basicHHvars: a character vector specifying important variables for the household structure that need to be available in dataS. Typically variables such as age or sex may be used.

seed: optional; an integer value to be used as the seed of the random number generator, or an integer vector containing the state of the random number generator to be restored.
Value

An object of class simPopObj containing the simulated population household structure as well as the underlying sample that was provided as input.

Note

The function sample is used, which gives results incompatible with those from < 2.2.0 and produces a warning the first time this happens in a session.

Author(s)

Bernhard Meindl and Andreas Alfons

References


See Also

simCategorical, simContinuous, simComponents, simEUSILC

Examples

data(eusilcS)
inp <- specifyInput(data=eusilcS, hhid="db030", hhsizel="hsize", strata="db040", weight="db090")
eusilcP <- simStructure(data=inp, method="direct", basicHHvars=c("age", "rb090"))
class(eusilcP)
eusilcP

---

**spBwplotStats**  
*Weighted box plot statistics*

Description

Compute the statistics necessary for producing box-and-whisker plots of continuous or semi-continuous variables, taking into account sample weights.

Usage

spBwplotStats(x, weights = NULL, coef = 1.5, zeros = TRUE, do.out = TRUE)
Arguments

- **x**: a numeric vector.
- **weights**: an optional numeric vector containing sample weights.
- **coef**: a numeric value that determines the extension of the whiskers.
- **zeros**: a logical indicating whether the variable specified by `additional` is semi-continuous, i.e., contains a considerable amount of zeros. If TRUE, the (weighted) box plot statistics are computed for the non-zero data points only and the number of zeros is returned, too.
- **do.out**: a logical indicating whether data points that lie beyond the extremes of the whiskers should be returned.

Details

The function `quantileWt` is used for the computation of (weighted) quantiles. The median is computed together with the first and the third quartile, which form the box. If `range` is positive, the whiskers extend to the most extreme data points that have a distance to the box of no more than `coef` times the interquartile range. For `coef = 0`, the whiskers mark the minimum and the maximum of the sample, whereas a negative value causes an error.

Value

A list of class "spBwplotStats" with the following components:

- **stats**: A vector of length 5 containing the (weighted) statistics for the construction of a box plot.
- **n**: if `weights` is NULL, the number of non-missing and, if `zeros` is TRUE, non-zero data points. Otherwise the sum of the weights of the corresponding points.
- **nzero**: if `zeros` is TRUE and `weights` is NULL, the number of zeros. If `zeros` is TRUE and `weights` is not NULL, the sum of the weights of the zeros. If `zeros` is not TRUE, this is NULL.
- **out**: if `do.out`, the values of any data points that lie beyond the extremes of the whiskers.

Author(s)

Stefan Kraft and Andreas Alfons

See Also

- **spBwplot**, for producing (weighted) box plots of continuous or semi-continuous variables.
- **quantileWt**, for the computation of (weighted) sample quantiles.
- **boxplot.stats**, for the unweighted statistics for box plots (not considering semi-continuous variables).
Examples

```r
data(eusilcS)

## semi-continuous variable
spBwplotStats(eusilcS$netIncome,
              weights=eusilcS$rb050, do.out = FALSE)
```

---

**spCdf**  
*(Weighted empirical) cumulative distribution function*

Description

Compute a (weighted empirical) cumulative distribution function for survey or population data. For survey data, sample weights are taken into account.

Usage

```r
spCdf(x, weights = NULL, approx = FALSE, n = 10000)
```

Arguments

- `x`: a numeric vector.
- `weights`: an optional numeric vector containing sample weights.
- `approx`: a logical indicating whether an approximation of the cumulative distribution function should be computed.
- `n`: a single integer value; if `approx` is `TRUE`, this specifies the number of points at which the approximation takes place (see `approx`).

Details

Sample weights are taken into account by adjusting the step height. To be precise, the weighted step height for an observation is defined as its weight divided by the sum of all weights \( \frac{w_i}{\sum_{j=1}^{n} w_j} \).

If requested, the approximation is performed using the function `approx`.

Value

A list of class "spCdf" with the following components:

- `x`: a numeric vector containing the \( x \)-coordinates.
- `y`: a numeric vector containing the \( y \)-coordinates.
- `approx`: a logical indicating whether the coordinates represent an approximation.

Author(s)

Andreas Alfons and Stefan Kraft
**specifyInput**

**References**


**See Also**

spCdfplot, ecdf, approx

**Examples**

```r
data(eusilcS)
cdfS <- spCdf(eusilcS$netIncome, weights = eusilcS$rb050)
plot(cdfS, type="s")
```

**Description**

create an object of class 'dataObj' required for further processing

**Usage**

```r
specifyInput(
  data,
  hhid,
  hhsize = NULL,
  pid = NULL,
  weight = NULL,
  strata = NULL,
  population = FALSE
)
```

**Arguments**

- **data**
  - a data.frame or data.table featuring sample data.
- **hhid**
  - character vector of length 1 specifying variable containing household ids within slot data.
- **hhsize**
  - character vector of length 1 specifying variable containing household sizes within slot data. If NULL, household sizes are automatically calculated.
- **pid**
  - character vector of length 1 specifying variable containing person ids within slot data. If NULL, person ids are automatically calculated.
weight character vector of length 1 specifying variable holding sampling weights within slot data.

strata character vector of length 1 specifying variable name within slot data of variable holding information on strata, e.g. regions or NULL if such variable does not exist.

population TRUE/FALSE vector of length 1 specifying if the data object is a sample or a population NULL if such variable does not exist.

Author(s)
Bernhard Meindl

References

Examples
data(eusilcS)
inp <- specifyInput(data=eusilcS, hhid="db030", weight="rb050", strata="db040")
class(inp)
inp

---

**spMosaic**  
*Mosaic plots of expected and realized population sizes*

**Description**

Create mosaic plots of expected (i.e., estimated) and realized (i.e., simulated) population sizes.

**Usage**

```
spMosaic(x, method = c("split", "color"), ...)
```

**Arguments**

- `x`  
  An object of class "spTable" created using function `spTable`.

- `method`  
  A character string specifying the plot method. Possible values are "split" to plot the expected population sizes on the left hand side and the realized population sizes on the right hand side, and "color"

- `...`  
  If `method` is "split", further arguments to be passed to `cotabplot`. If `method` is "color", further arguments to be passed to `strucplot`
Details

If method is "split", the two tables of expected and realized population sizes are combined into a single table, with an additional conditioning variable indicating expected and realized values. A conditional plot of this table is then produced using `cotabplot`.

Author(s)

Andreas Alfons and Bernhard Meindl

References


See Also

`spTable`, `cotabplot`, `strucplot`

Examples

```r
set.seed(1234) # for reproducibility
data(eusilcS) # load sample data
samp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize",
                      strata="db040", weight="db090")
eusilcP <- simStructure(data=samp, method="direct", basicHHvars=c("age","rb090"))
abb <- c("B","LA","Vi","C","St","UA","Sa","T","Vo")
tab <- spTable(eusilcP, select=c("rb090", "db040", "hsize"))

# expected and realized population sizes
spMosaic(tab, method = "split",
         labeling=labeling_border(abbreviate=c(db040=TRUE)))

# realized population sizes colored according to relative
# differences with expected population sizes
spMosaic(tab, method = "color",
         labeling=labeling_border(abbreviate=c(db040=TRUE)))
```

---

### sprague

**Sprague index (multipliers)**

**Description**

Using the Sprague multipliers, the age counts are estimated for each year having 5-years interval data as input.
Usage

sprague(x)

Arguments

x numeric vector of age counts in five-year intervals

Details

The input is population counts of age classes 0-4, 5-9, 10-14, ..., 77-74, 75-79, 80+.

Value

Population counts for age 0, 1, 2, 3, 4, ..., 78, 79, 80+.

Author(s)

Matthias Templ

References

G. Calot and J.-P. Sardon. Methodology for the calculation of Eurostat’s demographic indicators. Detailed report by the European Demographic Observatory

See Also

whipple

Examples

## example from the world bank
x <- data.frame(age=as.factor(c(
  "0-4", "5-9", "10-14", "15-19", "20-24", 
  "50-54", "55-59", "60-64", "65-69", "70-74", "75-79", "80+"
)),
  pop=c(1971990, 2095820, 2157190, 2094110, 2116580, 2003840, 1785690,
  1502990, 1214170, 796934, 627551, 530305, 488014,
  364498, 259029, 158047, 125941)
)

s <- sprague(x[,2])
s

all.equal(sum(s), sum(x[,2]))
**spTable**

**Cross tabulations of expected and realized population sizes.**

**Description**

Compute contingency tables of expected (i.e., estimated) and realized (i.e., simulated) population sizes. The expected values are obtained with the Horvitz-Thompson estimator.

**Usage**

```r
spTable(inp, select)
```

**Arguments**

- `inp`: an object of class `simPopObj` containing household survey and simulated population data.
- `select`: character; vector defining the columns in slots 'pop' and 'sample' of argument 'input' that should be used for tabulation.

**Details**

The contingency tables are computed with `tableWt`.

**Value**

A list of class "spTable" with the following components:

- `expected`: the contingency table estimated from the survey data.
- `realized`: the contingency table computed from the simulated population data.

**Note**

Sampling weights are automatically used from the input object 'inp'!

**Author(s)**

Andreas Alfons and Bernhard Meindl

**See Also**

`spMosaic`, `tableWt`
Examples

set.seed(1234)  # for reproducibility
data(eusilcS)   # load sample data
samp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize",
                      strata="db040", weight="db090")
eusilcP <- simStructure(data=samp, method="direct", basicHHvars=c("age", "rb090"))
res <- spTable(eusilcP, select = c("age", "rb090"))
class(res)
res

---

tableWt

Weighted cross tabulation

Description

Compute contingency tables taking into account sample weights.

Usage

tableWt(x, weights = NULL, useNA = c("no", "ifany", "always"))

Arguments

- **x**: a vector that can be interpreted as a factor, or a matrix or data.frame whose columns can be interpreted as factors.
- **weights**: an optional numeric vector containing sample weights.
- **useNA**: a logical indicating whether to include extra NA levels in the table.

Details

For each combination of the variables in x, the weighted number of occurrence is computed as the sum of the corresponding sample weights. If weights are not specified, the function `table` is applied.

Value

The (weighted) contingency table as an object of class `table`, an array of integer values.

Author(s)

Andreas Alfons and Stefan Kraft

See Also

table, contingencyWt
Examples

data(eusilcS)
tableWt(eusilcS[, c("hsize", "db040")], weights = eusilcS$rb050)
tableWt(eusilcS[, c("rb090", "pb220a")], weights = eusilcS$rb050, 
  useNA = "ifany")

Description

Population characteristics Region times Gender from Austria.
Using samp samp<- it is possible to extract or rather modify variables of the sample data within 
slot data in slot sample of the simPopObj-class-object. Using pop pop<- it is possible to extract or rather modify variables of the synthetic population within in slot data in slot sample of the simPopObj-class-object.

Format

totalsRG: A data frame with 18 observations on the following 3 variables.

  list("rb090") gender; a factor with levels female male
  list("db040") region; a factor with levels Burgenland Carinthia Lower Austria, Salzburg Styria Tyrol Upper Austria Vienna Vorarlberg
  list("Freq") totals; a numeric vector

totalsRGtab: a two-dimensional table holding the same information

totalsRG: A data frame with 18 observations on the following 3 variables.

  list("rb090") gender; a factor with levels female male
  list("db040") region; a factor with levels Burgenland Carinthia Lower Austria, Salzburg Styria Tyrol Upper Austria Vienna Vorarlberg
  list("Freq") totals; a numeric vector

totalsRGtab: a two-dimensional table holding the same information

Details

Population totals Region times Gender for Austria 2006
Population characteristics Region times Gender from Austria.

Source

Examples

```r
data(totalsRG)
totalsRG
data(totalsRGtab)
totalsRGtab
data(totalsRG)
totalsRG
data(totalsRGtab)
totalsRGtab
```

utility

Utility measures

Description

Various utility measures that basically compares two data sets.

Usage

```r
utility(
  x,
  y,
  type = c("all", "compareColumns", "compareRows", "compareRowsHH", "compareNA"),
  hhid = NULL
)
```

```r
utilityModal(x, y, varx, vary = NULL)
```

```r
utilityIndicator(x, y)
```

Arguments

- **x**: a data.frame, typically the original data set. For `utilityIndicator` this should be a vector of length 1.
- **y**: a data.frame, typically the corresponding synthetic data set. For `utilityIndicator` this should be a vector of length 1.
- **type**: which measure
  - `compareColumns`: compares the intersection of variables
  - `compareRows`: compares the number of rows
  - `compareRowsHH`: compares the number of households
  - `compareNA`: compares the number of missings
- **hhid**: index or name of variable containing the household ID
- **varx**: name or index of a variable in data.frame `x`
- **vary**: NULL or name or index of a variable in data.frame `y` corresponding to variable `varx` in data.frame `x`. If NULL, the names of the selected variable should be the same in both `x` and `y`. 
Value

the measure(s) of interest

Functions

• utility: comparisons of two data sets
  • utilityModal: comparison of number of categories
  • utilityIndicator: difference between two values

Author(s)

Matthias Templ, Maxime Bergeaut

Examples

data(eusilcS)
data(eusilcP)
## for fast calculations, took a subsample
eusilcP <- eusilcP[1:15000, ]
utility(eusilcS, eusilcP)
data(eusilcS)
data(eusilcP)
utilityModal(eusilcS, eusilcP, "age")
utilityModal(eusilcS, eusilcP, "pl030", "ecoStat")
data(eusilcS)
data(eusilcP)
m1 <- meanWt(eusilcS$age, eusilcS$rb050)
m2 <- mean(eusilcP$age)
utilityIndicator(m1, m2)

weighted_estimators

Weighted mean, variance, covariance matrix and correlation matrix

Description

Compute mean, variance, covariance matrix and correlation matrix, taking into account sample weights.

• meanWt: a simple wrapper that calls mean(x, na.rm=na.rm) if weights is missing and weighted.mean(x, w=weights, na.rm=na.rm) otherwise. Implemented methods for this generic are:
  – meanWt.default(x, weights, na.rm=TRUE,...)
  – meanWt.dataObj(x, vars, na.rm=TRUE,...)
• varWt: calls var(x, na.rm=na.rm) if weights is missing. Implemented methods for this generic are:
  – varWt.default(x, weights, na.rm=TRUE,...)
  – varWt.dataObj(x, vars, na.rm=TRUE,...)
• **covWt** and **covWt**: always remove missing values pairwise and call **cov** and **cor**, respectively, if **weights** is missing. Implemented methods for these generics are:

  - `covWt.default(x,y,weights,...)`
  - `covWt.matrix(x,weights,...)`
  - `covWt.data.frame(x,weights,...)`
  - `covWt.dataObj(x,vars,...)`
  - `corWt.default(x,y,weights,...)`
  - `corWt.matrix(x,weights,...)`
  - `corWt.data.frame(x,weights,...)`
  - `corWt.dataObj(x,vars,...)`

The additional parameters are now described:

• **y**: a numeric vector. If missing, this defaults to **x**.
• **vars**: a character vector of variable names that should be used for the calculation.
• **na.rm**: a logical indicating whether any **NA** or **NaN** values should be removed from **x** before computation. Note that the default is **TRUE**.
• **weights**: an optional numeric vector containing sample weights.

**Usage**

```r
meanWt(x, ...)
varWt(x, ...)
covWt(x, ...)
corWt(x, ...)
```

**Arguments**

- **x** for `meanWt` and `varWt`, a numeric vector or an object of class `dataObj`. For `covWt` and `corWt`, a numeric vector, matrix, `data.frame` or `dataObj`. In case of a `dataObj`, weights are automatically used from the S4-object itself.

- **...** for the generic functions `covWt` and `corWt`, additional arguments to be passed to methods. Additional arguments not included in the definition of the methods are ignored.

**Value**

For `meanWt`, the (weighted) mean.
For `varWt`, the (weighted) variance.
For `covWt`, the (weighted) covariance matrix or, for the default method, the (weighted) covariance.
For `corWt`, the (weighted) correlation matrix or, for the default method, the (weighted) correlation coefficient.
whipple

Note

meanWt, varWt, covWt and corWt all make use of slot weights of the input object if the dataObj-method is used.

Author(s)

Stefan Kraft and Andreas Alfons

See Also

mean, weighted.mean, var, cov, cor

Examples

data(eusilcS)
meanWt(eusilcS$netIncome, weights=eusilcS$rb050)
sqrt(varWt(eusilcS$netIncome, weights=eusilcS$rb050))

# dataObj-methods
inp <- specifyInput(data=eusilcS, hhid="db030", hhsize="hsize", strata="db040", weight="db090")
meanWt(inp, vars="netIncome")
sqrt(varWt(inp, vars="netIncome"))
corWt(inp, vars=c("age", "netIncome"))
covWt(inp, vars=c("age", "netIncome"))

whipple

Whipple index (original and modified)

Description

The function calculates the original and modified Whipple index to evaluate age heaping.

Usage

whipple(x, method = "standard", weight = NULL)

Arguments

x numeric vector holding the age of persons
method “standard” or “modified” Whipple index.
weight numeric vector holding the weights of each person
Details

The original Whipple’s index is obtained by summing the number of persons in the age range between 23 and 62, and calculating the ratio of reported ages ending in 0 or 5 to one-fifth of the total sample. A linear decrease in the number of persons of each age within the age range is assumed. Therefore, low ages (0-22 years) and high ages (63 years and above) are excluded from analysis since this assumption is not plausible.

When the digits 0 and 5 are not reported in the data, the original Whipple index varies between 0 and 100, 100 if no preference for 0 or 5 is within the data. When only the digits 0 and 5 are reported in the data it reaches a to a maximum of 500.

For the modified Whipple index, age heaping is calculated for all ten digits (0-9). For each digit, the degree of preference or avoidance can be determined for certain ranges of ages, and the modified Whipple index then is given by the absolute sum of these (indices - 1). The index is scaled between 0 and 1, therefore it is 1 if all age values end with the same digit and 0 it is distributed perfectly equally.

Value

The original or modified Whipple index.

Author(s)

Matthias Templ, Alexander Kowarik

References


See Also

sprague

Examples

```r
#Equally distributed
age <- sample(1:100, 5000, replace=TRUE)
whipple(age)
whipple(age,method="modified")

# Only 5 and 10
age5 <- sample(seq(0,100,by=5), 5000, replace=TRUE)
whipple(age5)
whipple(age5,method="modified")

# Only 10
age10 <- sample(seq(0,100,by=10), 5000, replace=TRUE)
whipple(age10)
whipple(age10,method="modified")
```
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