# Package 'PracTools'

August 4, 2020

Type Package
Title Tools for Designing and Weighting Survey Samples
Version 1.2.2
<b>Date</b> 2020-07-12
Author Richard Valliant, Jill A. Dever, Frauke Kreuter
Maintainer Richard Valliant <valliant@umich.edu></valliant@umich.edu>
<b>Description</b> Functions and datasets to support Valliant, Dever, and Kreuter, "Practical Tools for Designing and Weighting Survey Samples" (2nd edition, 2018). Contains functions for sample size calculation for survey samples using stratified or clustered one-, two-, and three-stage sample designs. Other functions compute variance components for multistage designs and sample sizes in two-phase designs. A number of example data sets are included.
<b>Suggests</b> doBy, foreign, lpSolve, plyr, pps, Rcpp, reshape, sampling, samplingbook, survey, knitr, rmarkdown
License GPL (>= 2)
LazyLoad yes
VignetteBuilder knitr
RoxygenNote 7.1.1
NeedsCompilation no
<b>Depends</b> R (>= $2.10$ )
Repository CRAN
<b>Date/Publication</b> 2020-08-03 22:50:10 UTC
R topics documented:
BW2stagePPS       3         BW2stagePPSe       4         BW2stageSRS       6         BW3stagePPS       8         BW3stagePPSe       9         clusOpt2       12         clusOpt2fixedPSU       14

Index

clusOpt3	15
clusOpt3fixedPSU	17
CVcalc2	19
CVcalc3	20
deff	21
deffCR	24
deffH	26
deffK	27
deffS	28
Domainy1y2	29
dub	30
gamEst	32
gammaFit	33
- НМТ	34
hospital	35
labor	36
MDarea.pop	37
mibrfss	38
nCont	40
nContMoe	41
nDep2sam	
nDomain	44
nhis	45
nhis.large	
nhispart	48
nLogOdds	50
nProp	51
nProp2sam	52
nPropMoe	53
NRadjClass	55
NRFUopt	56
nWilson	58
pclass	59
smho.N874	60
smho98	
strAlloc	63
ThirdGrade	65
wtdvar	
	<b>67</b>

BW2stagePPS 3

BW2stagePPS	Relvariance components for 2-stage sample

# **Description**

Compute components of relvariance for a sample design where primary sampling units (PSUs) are selected with probability proportional to size (*pps*) and elements are selected via simple random sampling (*srs*). The input is an entire sampling frame.

### Usage

```
BW2stagePPS(X, pp, psuID)
```

# **Arguments**

X data vector; length is the number of elements in the population.

pp vector of one-draw probabilities for the PSUs; length is number of PSUs in

population.

psuID vector of PSU identification numbers. This vector must be as long as X. Each

element in a given PSU should have the same value in psuID. PSUs must be in

the same order as in X.

### Details

BW2stagePPS computes the between and within population relvariance components appropriate for a two-stage sample in which PSUs are selected with varying probabilities and with replacement. Elements within PSUs are selected by simple random sampling. The components are appropriate for approximating the relvariance of the probability-with-replacement (*pwr*)-estimator of a total when the same number of elements are selected within each sample PSU. The function requires that an entire frame of PSUs and elements be input. (Use BW2stagePPSe if only a sample of PSUs and elements are available.)

#### Value

List object with values:

B2 between PSU unit relvariance
W2 within PSU unit relvariance
unit relvar unit relvariance for population

B2+W2 sum of between and within relvariance estimates

k ratio of  $B^2 + W^2$  to unit relvariance

delta measure of homogeneity with PSUs estimated as  $B^2/(B^2+W^2)$ 

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

4 BW2stagePPSe

#### References

Cochran, W.G. (1977, pp.308-310). Sampling Techniques. New York: John Wiley & Sons.

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

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.3.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

BW2stagePPSe, BW2stageSRS, BW3stagePPS, BW3stagePPSe

# **Examples**

```
data(MDarea.pop)
MDsub <- MDarea.pop[1:100000,]</pre>
    # Use PSU and SSU variables to define psu's
pp.PSU <- table(MDsub$PSU) / nrow(MDsub)</pre>
pp.SSU <- table(MDsub$SSU) / nrow(MDsub)</pre>
    # components with psu's defined by the PSU variable
BW2stagePPS(MDsub$y1, pp=pp.PSU, psuID=MDsub$PSU)
    # components with psu's defined by the SSU variable
BW2stagePPS(MDsub$y1, pp=pp.SSU, psuID=MDsub$SSU)
    # Use census tracts and block groups to define psu's
trtBG <- 10*MDsub$TRACT + MDsub$BLKGROUP
pp.trt <- table(MDsub$TRACT) / nrow(MDsub)</pre>
pp.BG <- table(trtBG) / nrow(MDsub)</pre>
    # components with psu's defined by tracts
BW2stagePPS(MDsub$ins.cov, pp=pp.trt, psuID=MDsub$TRACT)
    # components with psu's defined by block groups
BW2stagePPS(MDsub$ins.cov, pp=pp.BG, psuID=trtBG)
```

BW2stagePPSe

Estimated relvariance components for 2-stage sample

# **Description**

Estimate components of relvariance for a sample design where primary sampling units (PSUs) are selected with *pps* and elements are selected via *srs*. The input is a sample selected in this way.

# Usage

```
BW2stagePPSe(Ni, ni, X, psuID, w, m, pp)
```

BW2stagePPSe 5

# Arguments

Ni	vector of number of elements in the population of each sample PSU; length is the number of PSUs in the sample.
ni	vector of number of sample elements in each sample PSU; length is the number of PSUs in the sample. PSUs must be in the same order in ni and in X.
Χ	data vector for sample elements; length is the number of elements in the sample. These must be in PSU order. PSUs must be in the same order in ni and in X.
psuID	vector of PSU identification numbers. This vector must be as long as X. Each element in a given PSU should have the same value in psuID.
W	vector of full sample weights. This vector must be as long as X. Vector must be in the same order as X.
m	number of sample PSUs
pp	vector of 1-draw probabilities for the PSUs. The length of this vector is the number of PSUs in the sample. Vector must be in the same order as Ni and ni.

#### **Details**

BW2stagePPSe computes the between and within population variance and relvariance components appropriate for a two-stage sample in which PSUs are selected with varying probabilities and with replacement. Elements within PSUs are selected by simple random sampling. The number of elements selected within each sample PSU can vary but must be at least two. The estimated components are appropriate for approximating the relvariance of the *pwr*-estimator of a total when the same number of elements are selected within each sample PSU. This function can also be used if PSUs are selected by *srswr* by appropriate definition of pp.

#### Value

List with values:

Vpsu estimated between PSU unit variance
Vssu estimated within PSU unit variance
B estimated between PSU unit relvariance
W estimated within PSU unit relvariance

k estimated ratio of B+W to estimated unit relvariance of the analysis variable

delta intraclass correlation estimated as B/(B+W)

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Cochran, W.G. (1977, pp.308-310). Sampling Techniques. New York: John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.4.1). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

6 BW2stageSRS

# See Also

BW2stagePPS, BW2stageSRS, BW3stagePPS, BW3stagePPSe

# **Examples**

```
## Not run:
require(sampling)
require(reshape)
                       # has function that allows renaming variables
data(MDarea.pop)
Ni <- table(MDarea.pop$TRACT)</pre>
m <- 20
probi <- m*Ni / sum(Ni)</pre>
    # select sample of clusters
sam <- cluster(data=MDarea.pop, clustername="TRACT", size=m, method="systematic",</pre>
                 pik=probi, description=TRUE)
    # extract data for the sample clusters
samclus <- getdata(MDarea.pop, sam)</pre>
samclus <- rename(samclus, c(Prob = "pi1"))</pre>
    # treat sample clusters as strata and select srswor from each
s <- strata(data = as.data.frame(samclus), stratanames = "TRACT",</pre>
            size = rep(50,m), method="srswor")
# extracts the observed data
samdat <- getdata(samclus,s)</pre>
samdat <- rename(samdat, c(Prob = "pi2"))</pre>
    # extract pop counts for PSUs in sample
pick <- names(Ni) %in% sort(unique(samdat$TRACT))</pre>
Ni.sam <- Ni[pick]
pp <- Ni.sam / sum(Ni)
wt <- 1/samdat$pi1/samdat$pi2</pre>
BW2stagePPSe(Ni = Ni.sam, ni = rep(50,20), X = samdat$y1,
            psuID = samdat$TRACT, w = wt,
            m = 20, pp = pp)
## End(Not run)
```

BW2stageSRS

Relvariance components for 2-stage sample

# Description

Compute components of relvariance for a sample design where primary sampling units (PSUs) and elements are selected via *srs*. The input is an entire sampling frame.

# Usage

```
BW2stageSRS(X, psuID)
```

BW2stageSRS 7

# **Arguments**

X data vector; length is the number of elements in the population.

psuID vector of PSU identification numbers. This vector must be as long as X. Each

element in a given PSU should have the same value in psuID. PSUs must be in

the same order as in X.

#### **Details**

BW2stageSRS computes the between and within population relvariance components appropriate for a two-stage sample in which PSUs are selected via *srs* (either with or without replacement). Elements within PSUs are assumed to be selected by *srswor*. The same number of elements is assumed to be selected within each sample PSU. The function requires that an entire frame of PSUs and elements be input. (Use BW2stagePPSe if only a sample of PSUs and elements are available.)

#### Value

List with values:

B2 between PSU unit relvariance
W2 within PSU unit relvariance
unit relvar
unit relvariance for population

 $\mathsf{B2}\text{+}\mathsf{W2} \qquad \qquad B^2+W^2$ 

k ratio of  $B^2 + W^2$  to unit relvariance

delta full intraclass correlation estimated as  $B^2/(B^2+W^2)$ 

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

### References

Cochran, W.G. (1977, chap. 11). Sampling Techniques. New York: John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.2.1). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

BW2stagePPS, BW2stagePPSe, BW3stagePPSe

```
data(MDarea.pop)
MDsub <- MDarea.pop[1:100000,]
    # psu's are defined by PSU variable
BW2stageSRS(abs(MDsub$Hispanic-2), psuID=MDsub$PSU)
    # psu's are defined by SSU variable
BW2stageSRS(abs(MDsub$Hispanic-2), psuID=MDsub$SSU)</pre>
```

8 BW3stagePPS

BW3stagePPS	Relvariance components for 3-stage sample	

# **Description**

Compute components of relvariance for a sample design where primary sampling units (PSUs) are selected with *ppswr* and secondary sampling units (SSUs) and elements within SSUs are selected via *srs*. The input is an entire sampling frame.

# Usage

```
BW3stagePPS(X, pp, psuID, ssuID)
```

# **Arguments**

Χ	data vector; length is the number of element	s in the population.
---	--	----------------------

pp vector of one-draw probabilities for the PSUs; length is number of PSUs in

population.

psuID vector of PSU identification numbers. This vector must be as long as X. Each

element in a given PSU should have the same value in psuID. PSUs must be in

the same order as in X.

ssuID vector of SSU identification numbers. This vector must be as long as X. Each

element in a given SSU should have the same value in ssuID. PSUs and SSUs must be in the same order as in X. ssuID should have the form psuIDII(ssuID

within PSU).

#### **Details**

BW3stagePPS computes the between and within population relvariance components appropriate for a three-stage sample in which PSUs are selected with varying probabilities and with replacement. SSUs and elements within SSUs are selected by simple random sampling. The components are appropriate for approximating the relvariance of the *pwr*-estimator of a total when the same number of SSUs are selected within each PSU, and the same number of elements are selected within each sample SSU. The function requires that an entire sampling frame of PSUs and elements be input. (Use BW2stagePPSe if only a sample of PSUs, SSUs, and elements is available.)

#### Value

#### List with values:

В	between PSU	unit relvariance
D	DCLWCCII I SO	unit icivariance

W within PSU unit relvariance computed as if the sample were two-stage

W2 unit relvariance among SSU totals

was unit relvariance among elements within PSU/SSUs

unit relvar unit relvariance for population

BW3stagePPSe 9

k1	ratio of $B^2 + W^2$ to unit relvariance
k2	ratio of $W_2^2 + W_3^2$ to unit relvariance
delta1	homogeneity measure among elements within PSUs estimated as $B^2/(B^2+W^2)$
delta2	homogeneity measure among elements within SSUs estimated as $W_2^2/(W_2^2+W_3^2)$

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Hansen, M.H., Hurwitz, W.N., and Madow, W.G. (1953, chap. 9, p.211). Sample Survey Methods and Theory, Vol.I. John Wiley & Sons.

Saerndal, C.E., Swensson, B., and Wretman, J. (1992, p.149). *Model Assisted Survey Sampling*. Springer.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.2.4). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

BW2stagePPS, BW2stagePPSe, BW2stageSRS, BW3stagePPSe

### **Examples**

BW3stagePPSe

Estimated relvariance components for 3-stage sample

# Description

Estimate components of relvariance for a sample design where primary sampling units (PSUs) are selected with probability proportional to size with replacement (*ppswr*) and secondary sampling units (SSUs) and elements within SSUs are selected via simple random sampling (*srs*). The input is a sample selected in this way.

BW3stagePPSe

# Usage

```
BW3stagePPSe(dat, v, Ni, Qi, Qij, m)
```

# **Arguments**

dat	data frame for sample elements with PSU and SSU identifiers, weights, and analysis variable(s). The data frame should be sorted in hierarchical order: by PSU and SSU within PSU. Required names for columns: psuID = PSU identifier; ssuID = SSU identifier. These must be unique, i.e., numbering should not restart within each PSU. Setting ssuID = psuIDll(ssuID within PSU) is a method of doing this. w1i = vector of weights for PSUs; w2ij = vector of weights for SSUs (PSU weight*SSU weight within PSU); w = full sample weight
V	Name or number of column in data frame dat with variable to be analyzed.
Ni	m-vector of number of SSUs in the population in the sample PSUs; m is number of sample PSUs.
Qi	m-vector of number of elements in the population in the sample PSUs
Qij	vector of numbers of elements in the population in the sample SSUs
m	number of sample PSUs

# **Details**

BW3stagePPSe computes the between and within population relvariance components appropriate for a three-stage sample in which PSUs are selected with varying probabilities and with replacement. SSUs and elements within SSUs are selected by simple random sampling. The estimated components are appropriate for approximating the relvariance of the *pwr*-estimator of a total when the same number of SSUs are selected within each PSU, and the same number of elements are selected within each sample SSU.

# Value

# List with values:

Vpsu	estimated between PSU unit variance
Vssu	estimated second-stage unit variance among SSU totals
Vtsu	estimated third-stage unit variance
В	estimated between PSU unit relvariance
W	estimated within PSU unit relvariance computed as if the sample were two-stage
k1	estimated ratio of B+W to estimated unit relvariance of the analysis variable
W2	estimated unit relvariance among SSU totals
W3	estimated third-stage unit relvariance among elements within PSU/SSUs
k2	estimated ratio of W2+W3 to estimated unit relvariance of the analysis variable
delta1	homogeneity measure among elements within PSUs estimated as $B^2/(B^2+W^2)$
delta2	homogeneity measure among elements within SSUs estimated as $W_2^2/(W_2^2+W_3^2)$

BW3stagePPSe 11

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Hansen, M.H., Hurwitz, W.N., and Madow, W.G. (1953, chap. 9, sect. 10). *Sample Survey Methods and Theory*, Vol.II. New York: John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.4.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

BW2stagePPS, BW2stagePPSe, BW2stageSRS, BW3stagePPS

```
## Not run:
    # select 3-stage sample from Maryland population
data(MDarea.pop)
MDpop <- MDarea.pop
require(sampling)
require(reshape)
                       # has function that allows renaming variables
    # make counts of SSUs and elements per PSU
xx <- do.call("rbind",list(by(1:nrow(MDpop),MDpop$SSU,head,1)))</pre>
pop.tmp <- MDpop[xx,]</pre>
Ni <- table(pop.tmp$PSU)
Qi <- table(MDarea.pop$PSU)</pre>
Qij <- table(MDpop$SSU)</pre>
m <- 30
                 # no. of PSUs to select
probi <- m*Qi / sum(Qi)</pre>
    # select sample of clusters
sam <- cluster(data=MDpop, clustername="PSU", size=m, method="systematic",</pre>
                pik=probi, description=TRUE)
    # extract data for the sample clusters
samclus <- getdata(MDpop, sam)</pre>
samclus <- rename(samclus, c(Prob = "p1i"))</pre>
samclus <- samclus[order(samclus$PSU),]</pre>
    # treat sample clusters as strata and select srswor of block groups from each
    # identify psu IDs for 1st instance of each ssuID
xx <- do.call("rbind",list(by(1:nrow(samclus),samclus$SSU,head,1)))</pre>
SSUs <- cbind(PSU=samclus$PSU[xx], SSU=samclus$SSU[xx])</pre>
    # select 2 SSUs per tract
n < -2
s <- strata(data = as.data.frame(SSUs), stratanames = "PSU",</pre>
             size = rep(n,m), method="srswor")
s \leftarrow rename(s, c(Prob = "p2i"))
    # extract the SSU data
    # s contains selection probs of SSUs, need to get those onto data file
SSUsam <- SSUs[s$ID_unit, ]</pre>
SSUsam <- cbind(SSUsam, s[, 2:3])
    # identify rows in PSU sample that correspond to sample SSUs
```

12 clusOpt2

```
tmp <- samclus$SSU %in% SSUsam$SSU</pre>
SSUdat <- samclus[tmp,]</pre>
SSUdat <- merge(SSUdat, SSUsam[, c("p2i","SSU")], by="SSU")</pre>
    # select srswor from each sample SSU
n.SSU <- m*n
s <- strata(data = as.data.frame(SSUdat), stratanames = "SSU",</pre>
             size = rep(50,n.SSU), method="srswor")
s \leftarrow rename(s, c(Prob = "p3i"))
samclus <- getdata(SSUdat, s)</pre>
del <- (1:ncol(samclus))[dimnames(samclus)[[2]] %in% c("ID_unit", "Stratum")]</pre>
samclus <- samclus[, -del]</pre>
    # extract pop counts for PSUs in sample
pick <- names(Qi) %in% sort(unique(samclus$PSU))</pre>
Qi.sam <- Qi[pick]</pre>
    # extract pop counts of SSUs for PSUs in sample
pick <- names(Ni) %in% sort(unique(samclus$PSU))</pre>
Ni.sam <- Ni[pick]
    # extract pop counts for SSUs in sample
pick <- names(Qij) %in% sort(unique(samclus$SSU))</pre>
Qij.sam <- Qij[pick]</pre>
    # compute full sample weight and wts for PSUs and SSUs
wt <- 1 / samclus$p1i / samclus$p2i / samclus$p3i</pre>
w1i <- 1 / samclus$p1i
w2ij <- 1 / samclus$p1i / samclus$p2i
samdat <- data.frame(psuID = samclus$PSU, ssuID = samclus$SSU,</pre>
                      w1i = w1i, w2ij = w2ij, w = wt,
                       samclus[, c("y1","y2","y3","ins.cov", "hosp.stay")])
BW3stagePPSe(dat=samdat, v="y1", Ni=Ni.sam, Qi=Qi.sam, Qij=Qij.sam, m)
## End(Not run)
```

clusOpt2

Compute optimal sample sizes for a two-stage sample

# Description

Compute the sample sizes that minimize the variance of the *pwr*-estimator of a total in a two-stage sample.

#### Usage

```
clusOpt2(C1, C2, delta, unit.rv, k=1, CV0=NULL, tot.cost=NULL, cal.sw)
```

# Arguments

C1	unit cost per primary sampling unit (PSU)
C2	unit cost per element
delta	homogeneity measure $\delta$
unit.rv	unit relvariance

clusOpt2

k	ratio of $B^2 + W^2$ to unit relvariance
CV0	target CV
tot.cost	total budget for variable costs
cal.sw	specify type of optimum: 1 = find optimal m. opt for fixed total budget; 2 = find optimal m. opt for target CV0

#### **Details**

clus0pt2 will compute  $m_{opt}$  and  $\bar{n}_{opt}$  for a two-stage sample which uses simple random sampling at each stage or ppswr at the first stage and srs at the second.

# Value

List with values:

C1 unit cost per PSU

C2 unit cost per element

delta homogeneity measure

unit relvar unit relvariance

k ratio of  $B^2 + W^2$  to unit relvariance

cost total budget for variable costs,  $C-C_0$  if cal. sw=1; or computed cost if cal. sw=2

m. opt optimum number of sample PSUs

n.opt optimum number of sample elements per PSU

CV computed CV if cal.sw=1; or target CV if cal.sw=2

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Hansen, M.H., Hurwitz, W.N., and Madow, W.G. (1953, chap. 6, sect. 16). *Sample Survey Methods and Theory*, Vol.I. John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.3.1). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
clusOpt2fixedPSU, clusOpt3, clusOpt3fixedPSU
```

```
# optimum for a fixed total budget
clusOpt2(C1=750, C2=100, delta=0.05, unit.rv=1, k=1, tot.cost=100000, cal.sw=1)
    # optimum for a target CV
clusOpt2(C1=750, C2=100, delta=0.01, unit.rv=1, k=1, CV0=0.05, cal.sw=2)
```

14 clusOpt2fixedPSU

clusOpt2fixedPSU	Optimal number of sample elements per PSU in a two-stage sample when the sample of PSUs is fixed

# Description

Compute the optimum number of sample elements per primary sampling unit (PSU) for a fixed set of PSUs

# Usage

```
clusOpt2fixedPSU(C1, C2, m, delta, unit.rv, k=1, CV0=NULL, tot.cost, cal.sw)
```

# Arguments

C1	unit cost per PSU
C2	unit cost per element
m	number of sample PSU's (fixed)
delta	homogeneity measure
unit.rv	unit relvariance
k	ratio of $B^2 + W^2$ to unit relvariance
CV0	target CV
tot.cost	total budget for variable costs
cal.sw	specify type of optimum: $1 = \text{find optimal } \bar{n} \text{ for fixed total budget; } 2 = \text{find optimal } \bar{n} \text{ for target CV0}$

# **Details**

clusOpt2fixedPSU will compute  $\bar{n}_{opt}$  for a two-stage sample which uses simple random sampling at each stage or ppswr at the first stage and srs at the second. The PSU sample is fixed.

# Value

List with values:

C1	unit cost per PSU
C2	unit cost per element
m	number of (fixed) sample PSUs
delta	homogeneity measure
unit relvar	unit relvariance
k	ratio of $B^2 + W^2$ to unit relvariance
cost	total budget for variable costs, $C-C_0$ if cal. sw=1; or computed cost if cal. sw=2
n	optimum number of sample elements per PSU
CV	computed CV if cal.sw=1; or target CV if cal.sw=2

clusOpt3

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.3.3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
clusOpt2, clusOpt3, clusOpt3fixedPSU
```

# **Examples**

clusOpt3

Compute optimal sample sizes for a three-stage sample

# **Description**

Compute the sample sizes that minimize the variance of the *pwr*-estimator of a total in a three-stage sample.

# Usage

```
clusOpt3(unit.cost, delta1, delta2, unit.rv, k1=1, k2=1, CV0=NULL, tot.cost=NULL, cal.sw)
```

# **Arguments**

unit.cost	vector with three components for unit costs: C1 = unit cost per primary sampling unit (PSU); C2 = unit cost per secondary sampling unit (SSU); C3 = unit cost per element
delta1	homogeneity measure among elements within PSUs
delta2	homogeneity measure among elements within SSUs
unit.rv	population unit relvariance
k1	ratio of $B^2 + W^2$ to the population unit relvariance
k2	ratio of $W_2^2 + W_3^2$ to the population unit relvariance
CV0	target CV
tot.cost	total budget for variable costs
cal.sw	specify type of optimum: $1 = \text{find optimal m.opt}$ for fixed total budget; $2 = \text{find optimal m.opt}$ for target CV0

16 clusOpt3

# **Details**

clusOpt3 will compute  $m_{opt}$ ,  $\bar{n}_{opt}$ , and  $\bar{q}_{opt}$  for a three-stage sample which uses simple random sampling at each stage or *ppswr* at the first stage and *srs* at the second and third stages.

# Value

List with values:

C1	unit cost per PSU
C2	unit cost per SSU
C3	unit cost per element
delta1	homogeneity measure among elements within PSUs
delta2	homogeneity measure among elements within SSUs
unit relvar	unit relvariance
k1	ratio of $B^2 + W^2$ to the population unit relvariance
k2	ratio of $W_2^2 + W_3^2$ to the population unit relvariance
cost	total budget for variable costs if cal.sw=1; or computed cost if cal.sw=2
m.opt	optimum number of sample PSUs
n.opt	optimum number of sample SSUs per PSU
q.opt	optimum number of sample elements per SSU
CV	achieved CV if cal.sw=1 or target CV if cal.sw=2

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Hansen, M.H., Hurwitz, W.N., and Madow, W.G. (1953, p. 225). Sample Survey Methods and Theory, Vol. II. John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.3.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

clusOpt2, clusOpt2fixedPSU, clusOpt3fixedPSU

clusOpt3fixedPSU 17

clusOpt3fixedPSU	Compute optimal number of sample secondary sampling units (SSUs) and elements per SSU for a fixed set of primary sampling units (PSUs) in a three-stage sample
	in a inree-stage sample

# Description

Compute the sample sizes that minimize the variance of the *pwr*-estimator of a total in a three-stage sample when the PSU sample is fixed.

# Usage

# **Arguments**

unit.cost	3-vector of unit costs: C1 = unit cost per PSU; C2 = unit cost per SSU; C3 = unit cost per element
m	number of sample PSUs (fixed)
delta1	homogeneity measure among elements within PSUs
delta2	homogeneity measure among elements within SSUs
unit.rv	unit relvariance
k1	ratio of $B^2 + W^2$ to unit relvariance
k2	ratio of $W_2^2 + W_3^2$ to unit relvariance
CV0	target CV
tot.cost	total budget for variable costs, including PSU costs
cal.sw	specify type of optimum: $1 = \text{find optimal m.opt}$ for fixed total budget; $2 = \text{find optimal m.opt}$ for target $CV0$

# **Details**

clusOpt3 will compute  $\bar{n}_{opt}$  and  $\bar{q}_{opt}$  for a three-stage sample which uses simple random sampling at each stage or ppswr at the first stage and srs at the second and third stages. The set of sample PSUs is assumed to be fixed. "Variable costs" in tot.cost includes the budget for all costs that vary with the number of sample PSUs, SSUs, and elements, i.e.,  $C_1m + C_2m\bar{n} + C_3m\bar{n}\bar{q}$ .

# Value

List with values:

C1	unit cost per PSU
C2	unit cost per SSU
C3	unit cost per element

18 clusOpt3fixedPSU

m	number of sample PSUs (fixed)
delta1	homogeneity measure among elements within PSUs
delta2	homogeneity measure among elements within SSUs
unit relvar	unit relvariance
k1	ratio of $B^2 + W^2$ to unit relvariance
k2	ratio of $W_2^2 + W_3^2$ to unit relvariance
cost	<pre>budget constraint, tot.cost if cal.sw=1; computed cost if cal.sw=2</pre>
n	optimum number of sample SSUs per PSU
q	optimum number of sample elements per SSU
CV	achieved CV, used if cal.sw=1; or target CV, used if cal.sw=2
CV check	computed CV based on optimal sample sizes; used only if cal.sw=2

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Hansen, M.H., Hurwitz, W.N., and Madow, W.G. (1953, p. 225). Sample Survey Methods and Theory, Vol. II. John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.3.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
clusOpt2, clusOpt2fixedPSU, clusOpt3
```

```
# optima for a fixed total budget
clusOpt3fixedPSU(unit.cost=c(500, 100, 120), m=100, delta1=0.01, delta2=0.05, unit.rv=1,
    k1=1, k2=1, tot.cost=500000,cal.sw=1)
# optima for a target CV
clusOpt3fixedPSU(unit.cost=c(500, 100, 120), m=100, delta1=0.01, delta2=0.05, unit.rv=1,
    k1=1, k2=1, CV0=0.05,cal.sw=2)
```

CVcalc2

CVcalc2

Coefficient of variation of an estimated total in a 2-stage sample

### **Description**

Compute the coefficient of variation of an estimated total in a two-stage design. Primary sampling units (PSUs) can be selected either with probability proportional to size (*pps*) or with equal probability. Elements are selected via simple random sampling (*srs*).

### Usage

```
CVcalc2(V=NULL, m=NULL, nbar=NULL, k=1, delta=NULL, Bsq=NULL, Wsq=NULL)
```

# **Arguments**

V unit relvariance of	analysis	variable	in the	population
-----------------------	----------	----------	--------	------------

m number of sample PSUs

nbar number of sample elements per PSU k ratio of  $B^2 + W^2$  to V. Default value is 1.

delta measure of homogeneity equal to  $B^2/(B^2+W^2)$ 

Bsq unit relvariance of PSU totals Wsq within PSU relvariance

# **Details**

CVcalc2 computes the coefficient of variation of an estimated total for a two-stage sample. PSUs can be selected either with varying probabilities and with replacement or with equal probabilities and with replacement. Elements within PSUs are selected by simple random sampling. The CV formula is appropriate for approximating the relvariance of the probability-with-replacement (pwr)-estimator of a total when the same number of elements is selected within each sample PSU. See Sections 9.2.1–9.2.3 of Valliant, Dever, and Kreuter (2013) for details of formulas.

#### Value

Value of the coefficient of variation of an estimated total

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Cochran, W.G. (1977, pp.308-310). Sampling Techniques. New York: John Wiley & Sons.

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

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.2.1). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

20 CVcalc3

# See Also

CVcalc3

# **Examples**

```
CVcalc2(V=1, m=20 , nbar=5, k=1, delta=0.05)
CVcalc2(V=10, m=20 , nbar=5, k=1, delta=0.5)
CVcalc2(V=2.5, m=20 , nbar=5, k=2, Bsq=1, Wsq=4)
```

CVcalc3

Coefficient of variation of an estimated total in a 3-stage sample

# **Description**

Compute the coefficient of variation of an estimated total in a three-stage design. Primary sampling units (PSUs) can be selected either with probability proportional to size (*pps*) or with equal probability. Secondary units and elements within SSUs are selected via simple random sampling (*srs*).

# Usage

```
CVcalc3(V=NULL, m=NULL, nbar=NULL, qbar=NULL, k1=1, k2=1, delta1=NULL, delta2=NULL, Bsq=NULL, Wsq=NULL, W2sq=NULL, W3sq=NULL)
```

# **Arguments**

V	unit relvariance of analysis variable in the population
m	number of sample PSUs
nbar	number of sample secondary units per PSU
qbar	number of sample elements per SSU
k1	ratio of $B^2 + W^2$ to $V$ . Default value is 1.
k2	ratio of $W_2^2 + W_3^2$ to $V$ . Default value is 1.
delta1	measure of homogeneity between PSUs equal to $B^2/(B^2+W^2)$
delta2	measure of homogeneity between SSUs within PSUs, equal to $W_2^2/(W_2^2+W_3^2)$
Bsq	unit relvariance of PSU totals, equal to population variance of totals divided by $\bar{t}_U^2$
Wsq	within PSU relvariance of elements
W2sq	unit SSU relvariance
W3sq	unit element relvariance

deff 21

#### **Details**

CVcalc3 computes the coefficient of variation of an estimated total for a three-stage sample. PSUs can be selected either with varying probabilities and with replacement or with equal probabilities and with replacement. SSUs and elements within SSUs are selected by simple random sampling. The CV formula is appropriate for approximating the relvariance of the probability-with-replacement (pwr)-estimator of a total when the same number of SSUs is selected in each PSU and the same number of elements is selected within each sample SSU. See Sect. 9.2.4 of Valliant, Dever, and Kreuter (2013) for details of formulas.

#### Value

Value of the coefficient of variation of an estimated total

### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Cochran, W.G. (1977, pp.308-310). Sampling Techniques. New York: John Wiley & Sons.

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

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 9.2.4). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

CVcalc3

# **Examples**

```
CVcalc3(V=1, m=20 , nbar=5, qbar=10, delta1=0.02, delta2=0.10)
CVcalc3(V=1, m=20 , nbar=5, qbar=10, delta1=0.02, delta2=0.09)
CVcalc3(V=2, m=20 , nbar=5, qbar=10, k1=5, k2=10, Bsq=1, Wsq=9, W2sq=2 , W3sq=18 )
```

deff

Design effects of various types

# **Description**

Compute the Kish, Henry, Spencer, or Chen-Rust design effects.

# Usage

```
deff(w, x=NULL, y=NULL, p=NULL, strvar=NULL, clvar=NULL, Wh=NULL, nest=FALSE, type)
```

22 deff

#### **Arguments**

W	vector of weights for a sample
Х	matrix of covariates used to construct a GREG estimator of the total of $y$ . This matrix does not include the intercept. Used only for Henry $deff$ .
у	vector of the sample values of an analysis variable
p	vector of 1-draw selection probabilities, i.e., the probability that each unit would be selected in a sample of size 1. Used only for Spencer <i>deff</i> .
strvar	vector of stratum identifiers; equal in length to that of w. Used only for Chen-Rust <i>deff</i> .
clvar	vector of cluster identifiers; equal in length to that of w. Used only for Chen-Rust <i>deff</i> .
Wh	vector of the proportions of elements that are in each stratum; length is number of strata. Used only for Chen-Rust <i>deff</i> .
nest	Are cluster IDs numbered within strata (TRUE or FALSE)? If TRUE, cluster IDs can be restarted within strata, e.g., 1,2,3,1,2,3,
type	type of allocation; must be one of "kish", "henry", "spencer", "cr"

# **Details**

deff calls one of deffK, deffH, deffS, or deffCR depending on the value of the type parameter. The Kish design effect is the ratio of the variance of an estimated mean in stratified simple random sampling without replacement (stsrswor) to the variance of the estimated mean in srswor, assuming that all stratum unit variances are equal. In that case, proportional allocation with equal weighting is optimal. deffK equals 1 + relvar(w) where relvar is relvariance of the vector of survey weights. This measure is not appropriate in samples where unequal weighting is more efficient than equal weighting.

The Henry design effect is the ratio of the variance of the general regression (GREG) estimator of a total of y to the variance of the estimated total in srswr. Calculations for the Henry deff are done as if the sample is selected in a single-stage and with replacement. Varying selection probabilities can be used. The model for the GREG is assumed to be  $y = \alpha + \beta x + \epsilon$ , i.e., the model has an intercept.

The Spencer design effect is the ratio of the variance of the *pwr*-estimator of the total of *y*, assuming that a single-stage sample is selected with replacement, to the variance of the total estimated in *srswr*. Varying selection probabilities can be used.

The Chen-Rust *deff* accounts for stratification, clustering, and unequal weights, but does not account for the use of any auxiliary data in the estimator of a mean. The Chen-Rust *deff* returned here is appropriate for stratified, two-stage sampling.

#### Value

Numeric design effect for types kish, henry, spencer. For type cr a list with components: strata components

Matrix with deff's due to weighting, clustering, and stratification for each stra-

overall deff Design effect for full sample accounting for weighting, clustering, and stratification deff 23

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Chen, S. and Rust, K. (2017). An Extension of Kish's Formula for Design Effects to Two- and Three-Stage Designs with Stratification. *Journal of Survey Statistics and Methodology*, 5(2), 111-130.

Henry, K.A., and Valliant, R. (2015). A Design Effect Measure for Calibration Weighting in Single-stage Samples. *Survey Methodology*, 41, 315-331.

Kish, L. (1965). Survey Sampling. New York: John Wiley & Sons.

Kish, L. (1992). Weighting for unequal Pi. Journal of Official Statistics, 8, 183-200.

Park, I., and Lee, H. (2004). Design Effects for the Weighted Mean and Total Estimators under Complex Survey Sampling. *Survey Methodology*, 30, 183-193.

Spencer, B. D. (2000). An Approximate Design Effect for Unequal Weighting When Measurements May Correlate With Selection Probabilities. Survey Methodology, 26, 137-138.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 14). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
deffK, deffH, deffS, deffCR
```

```
# has function that allows renaming variables
require(reshape)
require(sampling)
set.seed(-500398777)
    # generate population using HMT function
pop.dat <- as.data.frame(HMT())</pre>
mos <- pop.dat$x
pop.dat$prbs.1d <- mos / sum(mos)</pre>
    # select pps sample
n <- 80
pk <- n * pop.dat*prbs.1d
sam <- UPrandomsystematic(pk)</pre>
sam <- sam==1
sam.dat <- pop.dat[sam, ]</pre>
dsgn.wts <- 1/pk[sam]</pre>
deff(w=dsgn.wts, type="kish")
deff(w=dsgn.wts, y=sam.dat$y, p=sam.dat$prbs.1d, type="spencer")
deff(w=dsgn.wts, x=sam.dat$x, y=sam.dat$y, type="henry")
data(MDarea.pop)
Ni <- table(MDarea.pop$TRACT)</pre>
m <- 10
```

24 deffCR

```
probi <- m*Ni / sum(Ni)</pre>
    # select sample of clusters
set.seed(-780087528)
sam <- cluster(data=MDarea.pop, clustername="TRACT", size=m, method="systematic",</pre>
                 pik=probi, description=TRUE)
    # extract data for the sample clusters
samclus <- getdata(MDarea.pop, sam)</pre>
samclus <- rename(samclus, c(Prob = "pi1"))</pre>
    # treat sample clusters as strata and select srswor from each
nbar <- 4
s <- strata(data = as.data.frame(samclus), stratanames = "TRACT",</pre>
             size = rep(nbar,m), method="srswor")
    # extracts the observed data
samdat <- getdata(samclus,s)</pre>
samdat <- rename(samdat, c(Prob = "pi2"))</pre>
    # add a fake stratum ID
H <- 2
nh <- m * nbar / H
stratum <- NULL
for (h in 1:H){
    stratum <- c(stratum, rep(h,nh))</pre>
wt <- 1/(samdat$pi1*samdat$pi2) * runif(m*nbar)</pre>
samdat \leftarrow cbind(subset(samdat, select = -c(Stratum)), stratum, wt)
deff(w = samdat$wt, y=samdat$y2, strvar = samdat$stratum, clvar = samdat$TRACT, Wh=NULL, type="cr")
```

deffCR

Chen-Rust design effect

# **Description**

Chen-Rust design effect for stratified, clustered, two-stage samples

# Usage

```
deffCR(w, strvar=NULL, clvar=NULL, Wh=NULL, nest=FALSE, y)
```

# **Arguments**

W	vector of weights for a sample
strvar	vector of stratum identifiers; equal in length to that of w.
clvar	vector of cluster identifiers; equal in length to that of w.
Wh	vector of the proportions of elements that are in each stratum; length is number of strata.
nest	Are cluster IDs numbered within strata (TRUE or FALSE)? If TRUE, cluster IDs can be restarted within strata, e.g., $1,2,3,1,2,3,$
у	vector of the sample values of an analysis variable

deffCR 25

#### **Details**

The Chen-Rust *deff* accounts for stratification, clustering, and unequal weights, but does not account for the use of any auxiliary data in the estimator of a mean. The Chen-Rust *deff* returned here is appropriate for stratified, two-stage sampling.

# Value

A list with components:

strata components

Matrix with deff's due to weighting, clustering, and stratification for each stra-

tun

overall deff Design effect for full sample accounting for weighting, clustering, and stratifi-

cation

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Chen, S. and Rust, K. (2017). An Extension of Kish's Formula for Design Effects to Two- and Three-Stage Designs with Stratification. *Journal of Survey Statistics and Methodology*, 5(2), 111-130.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 14). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
deff, deffH, deffK, deffS
```

```
require(sampling)
require(reshape)
data(MDarea.pop)
Ni <- table(MDarea.pop$TRACT)</pre>
m <- 10
probi <- m*Ni / sum(Ni)</pre>
    # select sample of clusters
set.seed(-780087528)
sam <- sampling::cluster(data=MDarea.pop, clustername="TRACT", size=m, method="systematic",</pre>
                 pik=probi, description=TRUE)
    # extract data for the sample clusters
samclus <- getdata(MDarea.pop, sam)</pre>
samclus <- rename(samclus, c(Prob = "pi1"))</pre>
    # treat sample clusters as strata and select srswor from each
nbar <- 4
s <- sampling::strata(data = as.data.frame(samclus), stratanames = "TRACT",</pre>
             size = rep(nbar,m), method="srswor")
```

26 deffH

```
# extracts the observed data
samdat <- getdata(samclus,s)
samdat <- rename(samdat, c(Prob = "pi2"))
    # add a fake stratum ID
H <- 2
nh <- m * nbar / H
stratum <- NULL
for (h in 1:H){
    stratum <- c(stratum, rep(h,nh))
}
wt <- 1/(samdat$pi1*samdat$pi2) * runif(m*nbar)
samdat <- cbind(subset(samdat, select = -c(Stratum)), stratum, wt)
deffCR(w = samdat$wt, strvar = samdat$stratum, clvar = samdat$TRACT, Wh=NULL, y=samdat$y2)</pre>
```

deffH

Henry design effect for pps sampling and GREG estimation of totals

### **Description**

Compute the Henry design effect for single-stage samples when a general regression estimator is used for a total.

# Usage

```
deffH(w, y, x)
```

# **Arguments**

w vector of inverses of selection probabilities for a sample
 y vector of the sample values of an analysis variable
 x matrix of covariates used to construct a GREG estimator of the total of y. This matrix does not include the intercept.

# Details

The Henry design effect is the ratio of the variance of the general regression (GREG) estimator of a total of y to the variance of the estimated total in srswr. Calculations for the Henry deff are done as if the sample is selected in a single-stage and with replacement. Varying selection probabilities can be used. The model for the GREG is assumed to be  $y = \alpha + \beta x + \epsilon$ , i.e., the model has an intercept.

### Value

numeric design effect

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

deffK 27

# References

Henry, K.A., and Valliant, R. (2015). A Design Effect Measure for Calibration Weighting in Single-stage Samples. *Survey Methodology*, 41, 315-331.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 14). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
deff, deffCR, deffK, deffS
```

# **Examples**

```
set.seed(-500398777)
    # generate population using HMT function
pop.dat <- as.data.frame(HMT())
mos <- pop.dat$x
pop.dat$prbs.1d <- mos / sum(mos)
    # select pps sample
require(sampling)
n <- 80
pk <- n * pop.dat$prbs.1d
sam <- UPrandomsystematic(pk)
sam <- sam==1
sam.dat <- pop.dat[sam, ]
dsgn.wts <- 1/pk[sam]
deffH(w=dsgn.wts, y=sam.dat$y, x=sam.dat$x)</pre>
```

deffK

Kish design effect

# **Description**

Compute the Kish design effect due to having unequal weights.

# Usage

deffK(w)

# **Arguments**

W

vector of inverses of selection probabilities for a sample

### **Details**

The Kish design effect is the ratio of the variance of an estimated mean in stratified simple random sampling without replacement (stsrswor) to the variance of the estimated mean in srswor, assuming that all stratum unit variances are equal. In that case, proportional allocation with equal weighting is optimal. deffK equals 1 + relvar(w) where relvar is relvariance of the vector of survey weights. This measure is not appropriate in samples where unequal weighting is more efficient than equal weighting.

28 deffS

# Value

numeric design effect

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

```
Kish, L. (1965). Survey Sampling. New York: John Wiley & Sons.
```

Kish, L. (1992). Weighting for unequal Pi. Journal of Official Statistics, 8, 183-200.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 14). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
deff, deffCR, deffH, deffS
```

# **Examples**

```
data(nhis)
w <- nhis$svywt
deffK(w)</pre>
```

deffS

Spencer design effect for pps sampling

# **Description**

Compute the Spencer design effect for single-stage samples selected with probability proportional to a measure of size.

# Usage

```
deffS(p, w, y)
```

# **Arguments**

p	vector of 1-draw selection probabilities, i.e., the probability that each unit would be selected in a sample of size 1.
W	vector of inverses of selection probabilities for a sample
У	vector of the sample values of an analysis variable

# **Details**

The Spencer design effect is the ratio of the variance of the pwr-estimator of the total of y, assuming that a single-stage sample is selected with replacement, to the variance of the total estimated in srswr. Varying selection probabilities can be used.

Domainy1y2 29

# Value

numeric design effect

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Park, I., and Lee, H. (2004). Design Effects for the Weighted Mean and Total Estimators under Complex Survey Sampling. *Survey Methodology*, 30, 183-193.

Spencer, B. D. (2000). An Approximate Design Effect for Unequal Weighting When Measurements May Correlate With Selection Probabilities. *Survey Methodology*, 26, 137-138.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 14). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
deff, deffCR, deffH, deffK
```

# **Examples**

```
set.seed(-500398777)
    # generate population using HMT function
pop.dat <- as.data.frame(HMT())
mos <- pop.dat$x
pop.dat$prbs.1d <- mos / sum(mos)
    # select pps sample
require(sampling)
n <- 80
pk <- n * pop.dat$prbs.1d
sam <- UPrandomsystematic(pk)
sam <- sam==1
sam.dat <- pop.dat[sam, ]
dsgn.wts <- 1/pk[sam]
deffS(p=sam.dat$prbs.1d, w=dsgn.wts, y=sam.dat$y)</pre>
```

Domainy1y2

Domain data

# **Description**

Domainy1y2 is a small data file used for an exercise in sample size calculations.

# Usage

```
data(Domainy1y2)
```

30 dub

# **Format**

A data frame with 30 observations on 2 variables.

```
y1 an artificial variable
```

y2 an artificial variable

# References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# **Examples**

```
data(Domainy1y2)
str(Domainy1y2)
summary(Domainy1y2)
```

dub

Sample sizes for a double sampling design

# Description

Compute samples sizes at each phase of a two-phase design where strata are created using the first phase.

# Usage

```
dub(c1, c2, Ctot, Nh, Sh, Yh.bar)
```

# **Arguments**

c1	cost per unit in phase-1
c2	cost per unit in phase-2
Ctot	Total variable cost

Nh Vector of stratum population counts or proportions
Sh Vector of stratum population standard deviations

Yh.bar Vector of stratum population means

#### **Details**

Compute the first and second phase sample sizes for a double sampling design. A first phase sample is selected by simple random sampling (*srs*). Strata are formed based on information collected in the first phase. The Neyman allocation to strata of the second phase sample is computed ignoring costs. Optimal total sample sizes are computed for the first and second phases, given per-unit costs for the first and second phases and a fixed total budget for both phases combined.

dub 31

# Value

A list object with elements:

V1	Variance component associated with phase-1
V2	Variance component associated with phase-2
n1	Phase-1 sample size
n2	Total phase-2 sample across all strata
"n2/n1"	Fraction that phase-2 is of phase-1

ney.alloc Vector of stratum sample sizes for phase-2 sample

Vopt Variance of mean with the calculated phase-1 and phase-2 sample sizes

nsrs Size of an *srs* that has cost Ctot, assuming each unit costs c2

Vsrs Variance of mean in an srs of cost Ctot, assuming each unit costs c2

Vratio Ratio of Vopt to Vsrs
Ctot Input value of total cost

cost.chk Computed value of phase-1 plus phase-2 sample with optimal sample sizes;

should agree with Ctot

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

### References

Cochran W (1977, sect. 12.3) Sampling Techniques. New York: John Wiley & Sons, Inc.

Neyman J (1938) Contribution to the theory of sampling human populations. *Journal of the American Statistical Association*, 33(201), 101-116.

Valliant, R., Dever, J., Kreuter, F. (2013, sect. 17.5.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

```
Wh <- rep(0.25,4)
Ph <- c(0.02,0.12,0.37,0.54)
Sh <- sqrt(Ph*(1-Ph))
c1 <- 10
c2 <- 50
Ctot <- 20000
dub(c1, c2, Ctot, Nh=Wh, Sh, Yh.bar=Ph)
```

32 gamEst

gamEst

Estimate variance model parameter  $\gamma$ 

# Description

Regresses a y on a set of covariates X where  $Var_M(y) = \sigma^2 x^{\gamma}$  and then regresses the squared residuals on log(x) to estimate  $\gamma$ .

# Usage

```
gamEst(X1, x1, y1, v1)
```

# **Arguments**

X1	matrix of predictors in the linear model for y1
x1	vector of $x$ 's for individual units in the assumed specification of $Var_{M}(y)$
y1	vector of dependent variables for individual units
v1	vector proportional to $Var_M(y)$

# **Details**

The function gamEst estimates the power  $\gamma$  in a model where the variance of the errors is proportional to  $x^{\gamma}$  for some covariate x. Values of  $\gamma$  are typically in [0,2]. The function is iteratively called by gammaFit, which is normally the function that an analyst should use.

# Value

The estimate of  $\gamma$ .

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
gammaFit
```

gammaFit 33

# **Examples**

```
data(hospital)
x <- hospital$x
y <- hospital$y

X <- cbind(sqrt(x), x)
gamEst(X1 = X, x1 = x, y1 = y, v1 = x)</pre>
```

gammaFit

Iteratively estimate variance model parameter  $\gamma$ 

# **Description**

Iteratively computes estimate of  $\gamma$  in a model with  $E_M(y) = x^T \beta$  and  $Var_M(y) = \sigma^2 x^{\gamma}$ .

### Usage

```
gammaFit(X, x, y, maxiter = 100, show.iter = FALSE, tol = 0.001)
```

# **Arguments**

X matrix of predictors in the linear model for y

x vector of x's for individual units in the assumed specification of  $Var_M(y)$ 

y vector of dependent variables for individual units

maxiter maximum number of iterations allowed

show.iter should values of  $\gamma$  be printed of each iteration? TRUE or FALSE

tol size of relative difference in  $\hat{\gamma}$ 's between consecutive iterations used to deter-

mine convergence. Algorithm terminates when relative difference is less than

tol.

#### **Details**

The function gammaFit estimates the power  $\gamma$  in a model where the variance of the errors is proportional to  $x^{\gamma}$  for some covariate x. Values of  $\gamma$  are typically in [0,2]. The function calls gamEst.

# Value

A list with the components:

g.hat estimate of  $\gamma$  when iterative procedure stopped

converged TRUE or FALSE depending on whether convergence was obtained

steps number of steps used by the algorithm

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

34 HMT

# References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
gamEst
```

# **Examples**

```
data(hospital)
x <- hospital$x
y <- hospital$y

X <- cbind(sqrt(x), x)
gammaFit(X = X, x = x, y = y, maxiter=100, tol=0.001)</pre>
```

**HMT** 

Generate an HMT population

# **Description**

Generate a population that follows the model in Hansen, Madow, and Tepping (1983)

# Usage

```
HMT(N=5000, H=10)
```

# **Arguments**

N	population size
Н	number of strata

# **Details**

HMT generates a population based on the model:  $E(y) = \alpha + \beta x$ ,  $Var(y) = \sigma^2 x^{3/2}$ . Both x and y have gamma distributions. Strata are formed to have approximately the same total of x.

# Value

N x 3 matrix with columns:

```
strat stratum ID

x auxiliary variable x

y analysis variable y
```

hospital 35

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Hansen, M.H., Madow, W.G., and Tepping, B.J. (1983). An evaluation of model-dependent and probability sampling inferences in sample surveys. *Journal of the American Statistical Association*, 78, 776-793.

# **Examples**

```
# generate HMT population with 1000 units and 5 strata and plot results pop <- HMT(N=1000, H=5) plot(pop[,"x"],pop[,"y"])
```

hospital

Hospital Data

# **Description**

The hospital data file is a national sample of short-stay hospitals with fewer than 1000 beds.

#### **Usage**

```
data(hospital)
```

# **Format**

A data frame with 393 observations on the following 2 variables.

- y Number of patients discharged by the hospital in January 1968
- x Number of inpatient beds in the hospital

#### **Details**

The hospital data are from the National Hospital Discharge Survey conducted by the U.S. National Center for Health Statistics. The survey collects characteristics of inpatients discharged from non-Federal short-stay hospitals in the United States. This population is from the January 1968 survey and contains observations on 393 hospitals.

#### Source

National Center for Health Statistics Hospital Discharge Survey of 1968.

#### References

Herson, J. (1976). An Investigation of Relative Efficiency of Least-Squares Prediction to Conventional Probability Sampling Plans. *Journal of the American Statistical Association*, 71, 700-703.

36 labor

# **Examples**

```
data(hospital)
str(hospital)
```

labor

Labor force population

### **Description**

A clustered population of persons extracted from the September 1976 Current Population Survey (CPS)

# Usage

```
data(labor)
```

#### **Format**

A data frame with 478 persons on the following variables:

```
h stratum
cluster cluster (or segment) number
person person number
age age of person
agecat age category (1 = 19 years and under; 2 = 20-24; 3 = 25-34; 4 = 35-64; 5 = 65 years and over)
race race (1 = non-Black; 2 = Black)
sex Gender (1=Male; 2=Female)
HoursPerWk Usual number of hours worked per week
WklyWage Usual amount of weekly wages (in 1976 U.S. dollars)
```

y An artificial variable generated to follow a model with a common mean. Persons in the same cluster are correlated. Persons in different clusters are uncorrelated under the model.

# **Details**

This population is a clustered population of 478 persons extracted from the September 1976 Current Population Survey (CPS) in the United States. The clusters are compact geographic areas used as one of the stages of sampling in the CPS and are typically composed of about 4 nearby households. The elements within clusters for this illustrative population are individual persons.

# Source

Current Population Survey, https://www.census.gov/programs-surveys/cps.html

MDarea.pop 37

## **Examples**

```
data(labor)
str(labor)
table(labor$h)
hist(labor$WklyWage)
```

MDarea.pop

Maryland area population

## **Description**

An artificial population of census tracts, block groups, and persons

# Usage

```
data(MDarea.pop)
```

#### **Format**

A data frame with 403,997 persons on the following variables:

PSU A grouping of block groups (BLKGROUP) which has about 5050 persons

SSU A grouping of block groups which has about 1010 persons

TRACT A geographic area defined by the Census Bureau. Tracts generally have between 1,500 and 8,000 people but have a much wider range in Anne Arundel county.

BLKGROUP Block group. A geographic area defined by the Census Bureau. Block groups generally have between 600 and 3,000 people.

Hispanic Hispanic ethnicity (1=Hispanic; 2=Non-Hispanic)

Gender (1=Male; 2=Female)

Age 23 level age category (1 = Under 5 years; 2 = 5 to 9 years; 3 = 10 to 14 years; 4 = 15 to 17 years; 5 = 18 and 19 years; 6 = 20 years; 7 = 21 years; 8 = 22 to 24 years; 9 = 25 to 29 years; 10 = 30 to 34 years; 11 = 35 to 39 years; 12 = 40 to 44 years; 13 = 45 to 49 years; 14 = 50 to 54 years; 15 = 55 to 59 years; 16 = 60 and 61 years; 17 = 62 to 64 years; 18 = 65 and 66 years; 19 = 67 to 69 years; 20 = 70 to 74 years; 21 = 75 to 79 years; 22 = 80 to 84 years; 23 = 85 years and over)

person Counter for person within tract/block group/Hispanic/Gender/Age combination

- y1 Artificial continuous variable
- y2 Artificial continuous variable
- y3 Artificial continuous variable

ins.cov Medical coverage (0 = person does not have medical insurance coverage; 1 = person has medical insurance coverage)

hosp. stay Overnight hospital stay (0 = person did not have an overnight hospital stay in last 12 months; 1 = person did have an overnight hospital stay in last 12 months)

38 mibrfss

#### **Details**

A dataset of 403,997 persons based on the 2000 decennial U.S. Census for Anne Arundel County in the state of Maryland. Person records were generated based on counts from the 2000 census. Individual values for each person were generated using models. Groupings to form the variables PSU and SSU were done after sorting the census file by tract and block group within tract.

## Source

2000 U.S. decennial census, http://www.census.gov/main/www/cen2000.html

## **Examples**

```
data(MDarea.pop)
str(MDarea.pop)
table(MDarea.pop$TRACT)
table(MDarea.pop$TRACT, MDarea.pop$Hispanic)
```

mibrfss

Michigan Behavioral Risk Factor Surveillance Survey

## Description

Demographic and health related variables from a U.S. household survey in the state of Michigan

# Usage

```
data(mibrfss)
```

# Format

A data frame with 2485 observations on persons for the following 21 variables.

```
SMOKE100 Smoked 100 or more cigarettes in lifetime (1 = Yes; 2 = No)
```

BMICAT3 Body mass index category (1 = Neither overweight nor obese (BMI < 25); 2 = Overweight (25 < = BMI < 30); 3 = Obese (BMI > 30))

AGECAT Age group (1 = 18-24 years; 2 = 25-34 years; 3 = 35-44 years; 4 = 45-54 years; 5 = 55-64 years; 6 = 65+)

GENHLTH General health (self-reported) (1 = Excellent; 2 = Very good; 3 = Good; 4 = Fair; 5 = Poor)

PHYSACT Physical activity: In last month participated in activities such as running, calisthenics, golf, gardening, or walking for exercise (1 = Yes; 2 = No)

HIGHBP High blood pressure: Have you ever been told by a doctor, nurse, or other health professional that you have high blood pressure? (1 = Yes; 2 = No)

ASTHMA Asthma: Have you ever been told by a doctor, nurse, or other health professional that you have asthma? (1 = Yes; 2 = No)

HISPANIC Hispanic ethnicity (1 = Yes; 2 = No; 7 = Missing)

mibrfss 39

```
WEIGHT Body weight in pounds
```

```
GENDER Gender (1 = Male; 2 = Female)
```

CELLPHON Has a wireless phone (1 = Yes; 2 = No)

INETHOME Has access to the Internet at home (1 = Yes; 2 = No)

WEBUSE How often do you use the Internet at home? Would you say, at least once a day, five to six times a week, two to four times a week, about once a week, less than once a week, or have you not used the Internet in the last month? (1 = At least once a day; 2 = 5-6 times a week; 3 = 2-4 times a week; 4 = About once a week; 5 = Less than once a week; 6 = Not in the last month)

```
RACECAT Race (1 = White; 2 = African American; 3 = Other)
```

- EDCAT Education level (1 = Did not graduate high school; 2 = Graduated high school; 3 = Attended college or technical school; 4 = Graduated from college or technical school)
- INCOMC3 Income category (1 = Less than \$15000; 2 = \$15000 to less than \$25000; 3 = \$25000 to less than \$35000; 4 = \$35000 to less than \$50000; 5 = \$50000 or more)
- DIABETE2 Diabetes: Have you ever been told by a doctor, nurse, or other health professional that you have diabetes? (1 = Yes; 2 = No)
- CHOLCHK Cholesterol check: Blood cholesterol is a fatty substance found in the blood. Have you ever had your blood cholesterol checked? (1 = Yes; 2 = No)

BMI Body mass index (continuous)

- BINGE2 Binge drinking: At risk for binge drinking based on alcohol consumption responses (1 = Yes; 2 = No)
- ARTHRIT Arthritis: Have you ever been told by a doctor, nurse, or other health professional that you have some form of arthritis, rheumatoid arthritis, gout, lupus, or fibromyalgia, or have joint symptoms of arthritis? (1 = Yes; 2 = No; 3 = Don't know, not sure, or refused)

## **Details**

The Michigan Behavioral Risk Factor Surveillance Survey (MIBRFSS) is part of a national state-by-state system of surveys used to monitor health conditions in the U.S. Data are collected through telephone household interviews. Demographic variables and a few health related variables are included in this subset. The mibrfss data set contains observations on 2845 persons and is extracted from the 2003 U.S. survey. The file contains only persons 18 years and older.

#### Source

Michigan Behavioral Risk Factor Surveillance Survey of 2003 sponsored by the U.S. Center for Disease Control. http://www.cdc.gov/BRFSS

#### See Also

nhis, nhis.large

40 nCont

## **Examples**

```
data(mibrfss)
str(mibrfss)
table(mibrfss$SMOKE100, useNA = "always")
table(mibrfss$BMICAT3, useNA="always")
```

nCont

Compute a simple random sample size for an estimated mean

# **Description**

Compute a simple random sample size using either a target coefficient of variation,  $CV_0$ , or target variance,  $V_0$ , for an estimated mean.

## Usage

```
nCont(CV0=NULL, V0=NULL, S2=NULL, ybarU=NULL, N=Inf, CVpop=NULL)
```

## **Arguments**

CV0	target value of coefficient of variation of $\bar{y_s}$
V0	target value of variance of $\bar{y_s}$

S2 unit (population) variance

ybarU population mean of target variable

N number of units in finite population

CVpop unit (population) coefficient of variation

## **Details**

If  $CV_0$  is the desired target, then the unit CV, CVpop, or the population mean and variance, ybarU and S2, must also be provided. If  $V_0$  is the constrained value, then S2 must be also be included in the function call.

## Value

numeric sample size

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

nContMoe 41

## See Also

```
nLogOdds, nProp, nPropMoe, nWilson
```

# **Examples**

```
nCont(CV0=0.05, CVpop=2)
nCont(CV0=0.05, CVpop=2, N=500)
nCont(CV0=0.10/1.645, CVpop=1)

# Compute sample size for a ratio estimator in smho98 population
# Identify large units to select with certainty first
data(smho98)
cert <- smho98[,"BEDS"] > 2000
tmp <- smho98[!cert, ]
tmp <- tmp[tmp[, "BEDS"] > 0, ]

x <- tmp[,"BEDS"]
y <- tmp[, "EXPTOTAL"]
m <- lm(y ~ 0 + x, weights = 1/x)
ybarU <- mean(y)
S2R <- sum(m$residuals^2/(length(x)-1))
nCont(CV0=0.15, S2=S2R, ybarU=ybarU)</pre>
```

nContMoe

Compute a simple random sample size for an estimated mean of a continuous variable based on margin of error

# **Description**

Compute a simple random sample size using a margin of error specified as the half-width of a normal approximation confidence interval or the half-width relative to the population mean.

#### Usage

```
nContMoe(moe.sw, e, alpha=0.05, CVpop=NULL, S2=NULL, ybarU=NULL, N=Inf)
```

# Arguments

moe.sw	switch for setting desired margin of error (1 = CI half-width on the mean; 2 = CI half-width on the mean divided by $\bar{y}_U$ )
е	desired margin of error; either $e=z_{1-\alpha/2}\sqrt{V(\bar{y}_s)}$ or $e=z_{1-\alpha/2}CV(\bar{y}_s)$
alpha	1 - (confidence level)
CVpop	unit (population) coefficient of variation
S2	population variance of the target variable
ybarU	population mean of target variable
N	number of units in finite population

nDep2sam

## **Details**

If moe.sw=1, then S2 must be provided. If moe.sw=2, then either (i) CVpop or (ii) S2 and ybarU must be provided.

#### Value

numeric sample size

## Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

## References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

## See Also

```
nCont, nLogOdds, nProp, nPropMoe, nWilson
```

## **Examples**

```
nContMoe(moe.sw=1, e=0.05, alpha=0.05, S2=2)
nContMoe(moe.sw=1, e=0.05, alpha=0.05, S2=2, N=200)
nContMoe(moe.sw=2, e=0.05, alpha=0.05, CVpop=2)
nContMoe(moe.sw=2, e=0.05, alpha=0.05, CVpop=2, N=200)
nContMoe(moe.sw=2, e=0.05, alpha=0.05, S2=4, ybarU=2)
```

nDep2sam

Simple random sample size for difference in means

## **Description**

Compute a simple random sample size for estimating the difference in means when samples overlap

# Usage

```
nDep2sam(S2x, S2y, g, r, rho, alt, del, sig.level=0.05, pow=0.80)
```

# Arguments

S2x	unit variance of analysis variable $x$ in sample 1
S2y	unit variance of analysis variable y in sample 2
g	proportion of sample 1 that is in the overlap with sample 2
r	ratio of the size of sample 1 to that of sample 2
rho	unit-level correlation between x and y

nDep2sam 43

alt	should the test be 1-sided or 2-sided; allowable values are alt="one.sided" or alt="two.sided".
del	size of the difference between the means to be detected
sig.level	significance level of the hypothesis test
pow	desired power of the test

## **Details**

nDep2sam computes sample sizes in two groups that are required for testing whether the difference in group means is significant. The power of the test is one of the input parameters. The samples have a specified proportion of units in common. Both samples are assumed to be selected via simple random sampling.

## Value

#### List with values:

n1	sample size in group 1
n2	sample size in group 2
S2x.S2y	unit variances in groups 1 and 2
delta	difference in group means to be detected
gamma	proportion of sample 1 that is in the overlap with sample 2
r	ratio of the size of sample 1 to that of sample 2
rho	unit-level correlation between analysis variables in groups 1 and
alt	type of test: one-sided or two-sided
sig.level	significance level of test

# Author(s)

power

Richard Valliant, Jill A. Dever, Frauke Kreuter

power of the test

# References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 4). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

2

Woodward, M. (1992). Formulas for Sample Size, Power, and Minimum Detectable Relative Risk in Medical Studies. *The Statistician*, 41, 185-196.

#### See Also

nProp2sam

44 nDomain

## **Examples**

```
nDep2sam(S2x=200, S2y=200,
g=0.75, r=1, rho=0.9,
alt="one.sided", del=5,
sig.level=0.05, pow=0.80)
```

nDomain Compute a simple random sample size for an estimated mean or total for a domain

# **Description**

Compute a simple random sample size using either a target coefficient of variation,  $CV_0(d)$ , or target variance,  $V_0(d)$ , for an estimated mean or total for a domain.

# Usage

```
nDomain(CV0d=NULL, V0d=NULL, S2d=NULL, ybarUd=NULL, N=Inf, CVpopd=NULL, Pd, est.type)
```

# Arguments

CV0d	target value of coefficient of variation of estimated domain mean or total
V0d	target value of variance of estimated domain mean or total
S2d	unit (population) variance for domain units
ybarUd	population mean of target variable for domain units
N	number of units in full finite population (not just the domain population)
CVpopd	unit (population) coefficient of variation for domain units
Pd	proportion of units in the population that are in the domain
est.type	type of estimate; allowable values are "mean" or "total"

# Details

If CV0d is the desired target, then the unit CV, CVpopd, or the domain population mean and variance, ybarUd and S2d, must also be provided. If V0d is the constrained value, then ybarUd must be also be included in the function call. CV0d will then be computed as sqrt(V0d)/ybarUd.

## Value

numeric sample size

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

nhis 45

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, sec. 3.5.2). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
nCont, nLogOdds, nProp, nPropMoe, nWilson
```

#### **Examples**

```
nDomain(CV0d=0.05, N=Inf, CVpopd=1, Pd=0.5, est.type="total")
nDomain(CV0d=0.05, N=Inf, CVpopd=1, Pd=0.5, est.type="mean")
nDomain(V0d=50, ybarUd=50, S2d=100, N=Inf, Pd=0.5, est.type="total")
nDomain(CV0d=0.05, ybarUd=50, S2d=100, N=Inf, Pd=0.5, est.type="total")
nDomain(CV0d=0.05, ybarUd=50, S2d=100, N=Inf, Pd=0.5, est.type="mean")
```

nhis

National Health Interview Survey: Demographic variables

#### **Description**

Demographic variables from a U.S. national household survey

## Usage

```
data(nhis)
```

#### **Format**

A data frame with 3,911 observations on the following 16 variables.

```
ID Identification variable
```

stratum Sample design stratum

psu Primary sampling unit, numbered within each stratum (1,2)

svywt survey weight

sex Gender (1 = male; 2 = female)

age Age, continuous

age\_r Recoded age (3 = 18-24 years; 4 = 25-44 years; 5 = 45-64 years; 6 = 65-69 years; 7 = 70-74 years; 8 = 75 years and older)

hisp Hispanic ethnicity (1 = Hispanic; 2 = Non-Hispanic)

marital Marital status (1 = Separated; 2 = Divorced; 3 = Married; 4 = Single/never married; 5 = Widowed; 9 = Unknown marital status)

parents Parent(s) of sample person present in the family (1 = Mother, no father; 2 = Father, no mother; 3 = Mother and father; 4 = Neither mother nor father)

parents\_r Parent(s) of sample person present in the family recode (1 = Yes; 2 = No)

46 nhis.large

```
educ Education (1 = 8th grade or less; 2 = 9-12th grade, no high school diploma; 3 = High school graduate; 4 = General education development (GED) degree recipient; 5 = Some college, no degree; 6 = Associate's degree, technical or vocational; 7 = Associate's degree, academic program; 8 = Bachelor's degree (BA, BS, AB, BBA); 9 = Master's, professional, or doctoral degree)
```

educ\_r Education recode (1 = High school, general education development degree (GED), or less; 2 = Some college; 3 = Bachelor's or associate's degree; 4 = Master's degree & higher)

```
race Race (1 = White; 2 = Black; 3 = Other)
resp Respondent (0 = nonrespondent; 1 = respondent)
```

#### **Details**

The National Health Interview Survey (NHIS) is used to monitor health conditions in the U.S. Data are collected through personal household interviews. Only demographic variables are included in this subset which was collected in 2003. The nhis data set contains observations for 3,911 persons. The file contains only persons 18 years and older.

## Source

National Health Interview Survey of 2003 conducted by the U.S. National Center for Health Statistics. http://www.cdc.gov/nchs/nhis.htm

#### See Also

```
nhis.large
```

## **Examples**

```
data(nhis)
str(nhis)
table(nhis$sex,nhis$age_r)
```

nhis.large

National Health Interview Survey: Demographic and health variables

# **Description**

Demographic and health related variables from a U.S. national household survey

#### Usage

```
data(nhis.large)
```

nhis.large 47

#### **Format**

```
A data frame with 21,588 observations on the following 18 variables.
```

ID Identification variable

stratum Sample design stratum

psu Primary sampling unit, numbered within each stratum (1,2)

svywt survey weight

sex Gender (1 = male; 2 = female)

age.grp Age group (1 = < 18 years; 2 = 18-24 years; 3 = 25-44 years; 4 = 45-64 years; 5 = 65+)

hisp Hispanic ethnicity (1 = Hispanic; 2 = Non-Hispanic White; 3 = Non-Hispanic Black; 4 = Non-Hispanic All other race groups)

parents Parents present in the household (1 = mother, father, or both present; 2 = neither present)

educ Highest level of education attained (1 = High school graduate, graduate equivalence degree, or less; 2 = Some college; 3 = Bachelor's or associate's degree; 4 = Master's degree or higher; NA = missing)

race Race (1 = White; 2 = Black; 3 = All other race groups)

inc.grp Family income group (1 = < \$20K; 2 = \$20000-\$24999; 3 = \$25000-\$34999; 4 = \$35000-\$44999; 5 = \$45000-\$54999; 6 = \$55000-\$64999; 7 = \$65000-\$74999; 8 = \$75K+; NA = missing)

delay.med Delayed medical care in last 12 months because of cost (1 = Yes; 2 = No; NA = missing)

hosp. stay Had an overnight hospital stay in last 12 months (1 = Yes; 2 = No; NA = missing)

doc.visit During 2 WEEKS before interview, did person see a doctor or other health care professional at a doctor's office, a clinic, an emergency room, or some other place? (excluding overnight hospital stay)? (1 = Yes; 2 = No)

medicaid Covered by medicaid, a governmental subsidy program for the poor (1 = Yes; 2 = No; NA = missing)

notcov Not covered by any type of health insurance (1 = Yes; 2 = No; NA = missing)

doing.lw What was person doing last week? (1 = Working for pay at a job or business; 2 = With a job or business but not at work; 3 = Looking for work; 4 = Working, but not for pay, at a job or business; 5 = Not working and not looking for work; NA = missing)

limited Is the person limited in any way in any activities because of physical, mental or emotional problems? (1 = Limited in some way; 2 = Not limited in any way; NA = missing)

#### **Details**

The National Health Interview Survey (NHIS) is used to monitor health conditions in the U.S. Data are collected through personal household interviews. Demographic variables and a few health related variables are included in this subset. The nhis.large data set contains observations on 21,588 persons extracted from the 2003 U.S. NHIS survey. The file contains only persons 18 years and older.

#### Source

National Health Interview Survey of 2003 conducted by the U.S. National Center for Health Statistics. http://www.cdc.gov/nchs/nhis.htm

48 nhispart

## See Also

nhis

## **Examples**

```
data(nhis.large)
str(nhis.large)
table(nhis.large$stratum, nhis.large$psu)
table(nhis.large$delay.med, useNA="always")
```

nhispart

National Health Interview Survey data from 2003: socioeconomic variables

## **Description**

Socioeconomic variables from a U.S. national household survey

# Usage

```
data(nhispart)
```

#### **Format**

A data frame with 3,924 observations on the following variables.

HHX Household identification variable

PX Person identifier within household

STRATUM Sample design stratum

PSU Primary sampling unit, numbered within each stratum (1,2)

WTFA survey weight

SEX Gender (1 = male; 2 = female)

AGE\_P Age of persons; values are 18-85 (85 includes age 85 and older)

R\_AGE1 Age group (3 = 18-24 years; 4 = 25-44 years; 5 = 45-64 years; 6 = 65-69 years; 7 = 70-74 years; 8 = 75 years and over)

ORIGIN\_I Hispanic ethnicity (1 = Hispanic; 2 = Non-Hispanic)

RACERP12 Race grouped (1 = White only; 2 = Black/African American only; 3 = American Indian or Alaska native (AIAN) only; 4 = Asian only; 5 = Race group not releasable; 6 = Multiple race)

MRACRPI2 Race detailed (1 = White; 2 = Black/African American; 3 = Indian (American), Alaska Native; 9 = Asian Indian; 10 = Chinese; 11 = Filipino; 15 = Other Asian; 16 = Primary race not releasable; 17 = Multiple race, no primary race selected)

RACRECI2 White/Black (1 = White; 2 Black; 3 All other race groups)

nhispart 49

R\_MARITL Marital status (1 = Married - spouse in household; 2 = Married - spouse not in household; 3 = Married - unknown whether spouse in household; 4 = Widowed; 5 = Divorced; 6 = Separated; 7 = Never married; 8 = Living with partner; 9 = Unknown marital status)

- CDCMSTAT CDC marital status (1 = Mother, no father; 2 = Father, no mother; 3 = Mother and father; 4 = Neither mother nor father)
- INCGRP Total combined family income group (1 = 0-\$4999; 2 = \$5000-\$9999; 3 = \$10000-\$14999; 4 = \$15000-\$19999; 5 = \$20000-\$24999; 6 = \$25000-\$34999; 7 = \$35000-\$44999; 8 = \$45000-\$54999; 9 = \$55000-\$64999; 10 = \$65000-\$74999; 11 = \$75000 and over; 12 = \$20000 or more (no detail); 13 = Less than \$20000 (no detail); 97 = Refused; 98 = Not ascertained; 99 = Don't know)
- PARENTS Parent(s) present in the family (1 = Mother, no father; 2 = Father, no mother; 3 = Mother and father; 4 = Neither mother nor father)
- EDUC\_R1 Highest level of education attained (1 = Less than high school graduate; 3 = High school graduate or general education development degree (GED); 5 = Some college, no degree; 6 = AA degree, technical or vocational or AA degree, academic program or Bachelor's degree (BA, BS, AB, BBA); 9 = Master's, professional, or doctoral degree)
- RAT\_CAT Ratio of family income to poverty level (1 = Under 0.50; 2 = 0.50 to 0.74; 3 = 0.75 to 0.99; 4 = 1.00 to 1.24; 5 = 1.25 to 1.49; 6 = 1.50 to 1.74; 7 = 1.75 to 1.99; 8 = 2.00 to 2.49; 9 = 2.50 to 2.99; 10 = 3.00 to 3.49; 11 = 3.50 to 3.99; 12 = 4.00 to 4.49; 13 = 4.50 to 4.99; 14 = 5.00 and over; 99 = Unknown)

#### **Details**

The National Health Interview Survey (NHIS) is used to monitor health conditions in the U.S. Data are collected through personal household interviews. Socioeconomic variables are included in this subset along with household and person codes. The nhispart data set contains observations on 3,924 persons extracted from the 2003 U.S. survey. The file contains only persons 18 years and older.

## **Source**

National Health Interview Survey of 2003 conducted by the U.S. National Center for Health Statistics. http://www.cdc.gov/nchs/nhis.htm

# **Examples**

```
data(nhispart)
str(nhispart)
table(nhispart$STRATUM, nhispart$PSU)
table(nhispart$RACERPI2, nhispart$RACRECI2, useNA="always")
```

nLogOdds

nLog0dds

Calculate simple random sample size for estimating a proportion

## **Description**

Calculate the simple random sample size for estimating a proportion using the log-odds transformation.

# Usage

```
nLogOdds(moe.sw, e, alpha=0.05, pU, N=Inf)
```

# Arguments

moe.sw	switch for setting desired margin of error $(1 = CI \text{ half-width on the proportion})$ ;
	2 = CI half-width on a proportion divided by pU)
e	desired margin of error
alpha	1 - (confidence level)
рU	population proportion
N	number of units in finite population

#### **Details**

The function accepts five parameters, which are the same ones as accepted by nPropMoe. The desired margin of error can be specified as the CI half-width on the proportion (moe.sw=1) or as the CI half-width as a proportion of the population value pU (moe.sw=2).

# Value

numeric sample size

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

## References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

# See Also

```
nProp, nPropMoe, nWilson, nCont
```

# **Examples**

```
\label{eq:nlogOdds(moe.sw=1, e=0.05, alpha=0.05, pU=0.2, N=Inf)} $$ nlogOdds(moe.sw=2, e=0.05, alpha=0.05, pU=0.2, N=Inf) $$
```

nProp 51

nProp

Compute simple random sample size for estimating a proportion

# Description

Compute the simple random sample size for estimating a proportion based on different precision requirements.

# Usage

```
nProp(CV0 = NULL, V0 = NULL, pU = NULL, N = Inf)
```

# **Arguments**

CV0	target value of coefficient of variation of the estimated proportion
VØ	target value of variance of the estimated proportion
pU	population proportion
N	number of units in finite population

# **Details**

The precision requirement of  $p_s$  can be set based on either a target coefficient of variation,  $CV_0$ , or a target variance,  $V_0$ . In either case, a value of  $p_U$  must be supplied.

# Value

numeric sample size

## Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

# References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

## See Also

nCont, nLogOdds, nPropMoe, nWilson

52 nProp2sam

## **Examples**

```
# srs sample size so that CV of estimated proportion is 0.05
# assuming the population is large and pU=0.01
# Both examples below are equivalent
nProp(V0=0.0005^2, N=Inf, pU=0.01) #or
nProp(CV0=0.05, N=Inf, pU=0.01)
# srswor sample size so that half-width of 2-sided 95% CI is 0.005
nProp(V0=(0.005/1.96)^2, N=Inf, pU=0.01)
```

nProp2sam

Simple random sample size for difference in proportions

# Description

Compute a simple random sample size for estimating the difference in proportions when samples overlap

# Usage

```
nProp2sam(px, py, pxy, g, r, alt, sig.level=0.05, pow=0.80)
```

# **Arguments**

рх	proportion in group 1
ру	proportion in group 2
pxy	proportion in the overlap has the characteristic in both samples
g	proportion of sample 1 that is in the overlap with sample 2
r	ratio of the size of sample 1 to that of sample 2
alt	should the test be 1-sided or 2-sided; allowable values are alt="one.sided" or alt="two.sided".
sig.level	significance level of the hypothesis test
pow	desired power of the test

# **Details**

nProp2sam computes sample sizes in two groups that are required for testing whether the difference in group proportions is significant. The power of the test is one of the input parameters. The samples have a specified proportion of units in common.

nPropMoe 53

# Value

#### List with values:

n1 sample size in group 1n2 sample size in group 2

px.py.pxy input values of the px, py, pxy parameters

gamma proportion of sample 1 that is in the overlap with sample 2

r ratio of the size of sample 1 to that of sample 2

alt type of test: one-sided or two-sided

sig.level significance level of test

power of the test

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

## References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 4). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

Woodward, M. (1992). Formulas for Sample Size, Power, and Minimum Detectable Relative Risk in Medical Studies. *The Statistician*, 41, 185-196.

## See Also

nDep2sam

## **Examples**

```
nProp2sam(px=0.5, py=0.55, pxy=0.45, g=0.5, r=1, alt="two.sided")
```

nPropMoe

Simple random sample size for a proportion based on margin of error

# Description

Calculates a simple random sample size based on a specified margin of error.

## Usage

```
nPropMoe(moe.sw, e, alpha = 0.05, pU, N = Inf)
```

54 nPropMoe

## Arguments

moe.sw	switch for setting desired margin of error (1 = CI half-width on the proportion; $2 = CI$ half-width on a proportion divided by $p_U$ )
е	desired margin of error; either $e=z_{1-\alpha/2}\sqrt{V(p_s)}$ or $e=z_{1-\alpha/2}CV(p_s)$
alpha	1 - (confidence level)
pU	population proportion
N	number of units in finite population

#### **Details**

The margin of error can be set as the half-width of a normal approximation confidence interval,  $e=z_{1-\alpha/2}\sqrt{V(p_s)}$ , or as the half-width of a normal approximation confidence interval divided by the population proportion,  $e=z_{1-\alpha/2}CV(p_s)$ . The type of margin of error is selected by the parameter moe . sw where moe . sw=1 sets  $e=z_{1-\alpha/2}\sqrt{V(p_s)}$  and moe . sw=2 sets i.e.,  $e=\frac{z_{1-\alpha/2}\sqrt{V(p_s)}}{p_U}$ .

#### Value

numeric sample size

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

```
nCont, nLogOdds, nProp, nWilson
```

# **Examples**

```
# srs sample size so that half-width of a 95% CI is 0.01
# population is large and population proportion is 0.04
nPropMoe(moe.sw=1, e=0.01, alpha=0.05, pU=0.04, N=Inf)

# srswor sample size for a range of margins of error defined as
# half-width of a 95% CI
nPropMoe(moe.sw=1, e=seq(0.01,0.08,0.01), alpha=0.05, pU=0.5)

# srswor sample size for a range of margins of error defined as
# the proportion that the half-width of a 95% CI is of pU
nPropMoe(moe.sw=2, e=seq(0.05,0.1,0.2), alpha=0.05, pU=0.5)
```

NRadjClass 55

NRadjClass Class-based nonresponse adjustments
--

# **Description**

Compute separate nonresponse adjustments in a set of classes.

# Usage

```
NRadjClass(ID, NRclass, resp, preds=NULL, wts=NULL, type)
```

# **Arguments**

ID	identification value for a unit
NRclass	vector of classes to use for nonresponse adjustment. Length is number of respondents plus nonrespondents
resp	indicator for whether unit is a nonrespondent (must be coded $0$ ) or respondent (must be coded $1$ )
preds	response probabilities, typically estimated from a binary regression model as in pclass
wts	vector of survey weights, typically base weights or base weights adjusted for unknown eligibility
type	type of adjustment computed within each value of NRclass. Allowable codes are 1, 2, 3, 4, or 5. (1 = unweighted average of response propensities, i.e., preds; 2 = weighted average response propensity; 3 = unweighted response rate; 4 = weighted response rate; 5 = median response propensity)

# **Details**

The input vectors should include both respondents and nonrespondents in a sample. A single value between 0 and 1 is computed in each nonresponse adjustment class to be used as a nonresponse adjustment. Five alternatives are available for computing the adjustment based on the value of type. The value of the adjustment is merged with individual unit data and stored in the RR field of the output data frame.

# Value

A data frame of respondents only with four columns:

NRcl.no	number of the nonresponse adjustment class for each unit
ID	identification value for a unit
resp	value of the resp variable (always 1)
RR	nonresponse adjustment for each unit

NRFUopt

## Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 13). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

## See Also

pclass

# **Examples**

NRFUopt

Sample sizes for a nonresponse follow-up study

# **Description**

Compute optimal values of the first-phase sample size and the second-phase sampling fraction in a two-phase sample.

# Usage

```
NRFUopt(Ctot=NULL, c1, c2, theta, CV0=NULL, CVpop=NULL, N=Inf, type.sw)
```

# Arguments

Ctot	total variable cost
c1	cost per unit in phase-1
c2	cost per unit in phase-2
theta	probability of response for each unit
CV0	target coefficient of variation for the estimated total or mean
CVpop	Unit coefficient of variation
N	Population size; default is Inf
type.sw	type of allocation; "cost" = target total variable cost, "cv" = target coefficient of variation

NRFUopt 57

#### **Details**

NRFUopt computes the optimal values of the first-phase sample size and the second-phase sampling fraction in a two-phase sample. Both stages are assumed to be selected using simple random sampling without replacement. If type.sw="cost", the optima are computed for a target total, expected cost across both phases. If type.sw="cv", the optima are computed for a target coefficient of variation for an estimated mean.

#### Value

List object with values:

allocation type of allocation: either "fixed cost" or "fixed CV"

"Total variable cost"

expected total cost: fixed cost if type.sw="cost" or computed cost if type.sw="cv";

unrounded sample sizes are used in calculation

"Response rate"

first-phase response rate

CV anticipated coefficient of variation (CV) if type.sw="cost" or target CV if

type.sw="cv"

v.opt optimal fraction of first-phase nonrespondents to select for second-phase follow-

up

n1.opt optimal number of units to sample at first-phase

"Expected n2" expected number of respondents obtained at second-phase

"Expected total cases (2-phase)"

expected number of respondents across both phases

"srs sample for same cv"

size of single-phase simple random sample (srs) needed to obtain same CV as

the two-phase sample

"Cost Ratio: Two phase to srs"

ratio of expected cost for two-phase sample to cost of single-phase srs

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Saerndal, C.E., Swensson, B., and Wretman, J. (1992, examples 15.4.4 and 15.4.5). *Model Assisted Survey Sampling*. New York: Springer.

Valliant, R., Dever, J., Kreuter, F. (2013, chap.17). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

58 nWilson

## **Examples**

```
# optima for fixed target CV
NRFUopt(Ctot=NULL, c1=50, c2=200, theta=0.5, CV0=0.05, CVpop=1, type.sw = "cv")
# optima for fixed total cost
NRFUopt(Ctot=100000, c1=50, c2=200, theta=0.5, CV0=NULL, CVpop=1, type.sw = "cost")
```

nWilson

Calculate a simple random sample size for estimating a proportion

# **Description**

Calculate a simple random sample size for estimating a proportion using the Wilson method.

## Usage

```
nWilson(moe.sw, alpha = 0.05, pU, e)
```

# **Arguments**

moe.sw	switch for setting desired margin of error $(1 = CI \text{ half-width on the proportion}; 2 = CI \text{ half-width on a proportion divided by pU})$
alpha	1 - (confidence level)
pU	population proportion
е	desired margin of error; either the value of CI half-width or the value of the half-width divided by pU

## **Details**

Calculate a simple random sample size using the Wilson (1927) method. A margin of error can be set as the CI half-width on the proportion (moe.sw=1) or as the CI half-width as a proportion of the population value  $p_U$  (moe.sw=2).

#### Value

```
n.sam numeric sample size

"CI lower limit"
lower limit of Wilson confidence interval with computed sample size

"CI upper limit"
upper limit of Wilson confidence interval with computed sample size

"length of CI" length of Wilson confidence interval with computed sample size
```

## Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

pclass 59

## References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

Wilson, E.B. (1927). Probable inference, the law of succession, and statistical inference. *Journal of the American Statistical Association*, 22, 209-212.

#### See Also

```
nCont, nLogOdds, nProp, nPropMoe
```

## **Examples**

```
# srs sample size using Wilson method so that half-width of a 95% CI # is 0.01. Population proportion is 0.04 nWilson(moe.sw = 1, pU = 0.04, e = 0.01)
```

pclass

Form nonresponse adjustment classes based on propensity scores

# **Description**

Fit a binary regression model for response probabilities and divide units into a specified number of classes.

# Usage

```
pclass(formula, data, link="logit", numcl=5, type, design=NULL)
```

# Arguments

formula	symbolic description of the binary regression model to be fitted as used in glm
data	an optional data frame; must be specified if type="unwtd"
link	a specification for the model link function; allowable values are "logit", "probit", or "cloglog"
numcl	number of classes into which units are split based on estimated propensities
type	whether an unweighted or weighted binary regression should be fit; allowable values are "unwtd" or "wtd"
design	sample design object; required if type="wtd"

## **Details**

A typical formula has the form response ~ terms where response is a two-level variable coded as 0 or 1, or is a factor where the first level denotes nonresponse and the second level is response. If type="unwtd", glm is used to fit an unweighted regression. If type="wtd", svyglm in the survey package is used to fit a survey-weighted regression.

60 smho.N874

#### Value

A list with components:

```
p.class propensity class for each unit
propensities estimated response probability for each unit
```

# Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 13). *Practical Tools for Designing and Weighting Survey Samples*. New York: Springer.

#### See Also

NRadjClass

## **Examples**

smho.N874

Survey of Mental Health Organizations Data

## **Description**

Data from the 1998 Survey of Mental Health Organizations (SMHO)

# Usage

```
data(smho.N874)
```

smho.N874 61

#### **Format**

A data frame with 874 observations on the following 6 variables.

**EXPTOTAL** Total expenditures in 1998

BEDS Total inpatient beds

SEENCNT Unduplicated client/patient count seen during year

EOYCNT End of year count of patients on the role

FINDIRCT Hospital receives money from the state mental health agency (1=Yes; 2=No)

hosp. type Hospital type (1 = Psychiatric; 2 = Residential or veterans; 3 = General; 4 = Outpatient, partial care; 5 = Multi-service, substance abuse)

## **Details**

The 1998 SMHO was conducted by the U.S. Substance Abuse and Mental Health Services Administration. It collected data on mental health care organizations and general hospitals that provide mental health care services, with an objective to develop national and state level estimates for total expenditure, full time equivalent staff, bed count, and total caseload by type of organization. The population omits one extreme observation in the smho98 population and has fewer variables than smho98. smho.N874 contains observations on 874 facilities.

# Source

Substance Abuse and Mental Health Services Administration

#### References

Manderscheid, R.W. and Henderson, M.J. (2002). Mental Health, United States, 2002. DHHS Publication No. SMA04-3938. Rockville MD USA: Substance Abuse and Mental Health Services Administration.

#### See Also

smho98

# **Examples**

```
data(smho.N874)
str(smho.N874)
```

62 smho98

smho98

Survey of Mental Health Organizations Data

#### **Description**

Data from the 1998 Survey of Mental Health Organizations (SMHO)

## Usage

data(smho98)

#### **Format**

A data frame with 875 observations on the following variables.

STRATUM Sample design stratum (1 = Psychiatric Hospital, private; 2 = Psychiatric Hospital, public; 3 = Residential, children; 4 = Residential, adults; 5 = General Hospital, public, inpatient or residential care; 6 = General Hospital, public, outpatient care only; 7 = General Hospital, private, inpatient or residential care; 8 = General Hospital, private, outpatient care only; 9 = Military Veterans, inpatient or residential care; 10 = Military Veterans, outpatient care only; 11 = Partial Care 12 = Outpatient care, private; 13 = Outpatient care, public; 14 = Multi-service, private; 15 = Multi-service, public; 16 = Substance Abuse)

BEDS Total inpatient beds

**EXPTOTAL** Total expenditures in 1998

SEENCNT Unduplicated client/patient count seen during year

EOYCNT End of year count of patients on the role

Y\_IP Number of inpatient visits during year

OPCSFRST Number of outpatients on the rolls on the first day of the reporting year

OPCSADDS Number of outpatients admitted, readmitted, or transferred to the organization during the reporting year for less than a 24 hour period and not overnight

OPCSVIST Number of outpatient visits during the reporting year for less than a 24 hour period and not overnight

EMGWALK Number of emergency walk-ins during the reporting year

PSYREHAB Number of visits for psychiatric rehabilitation services

IPCSADDS Number of residential patients added during the reporting year or patients admitted for more than a 24 hour period

#### **Details**

The 1998 SMHO was conducted by the U.S. Substance Abuse and Mental Health Services Administration. It collected data on mental health care organizations and general hospitals that provide mental health care services, with an objective to develop national and state level estimates for total expenditure, full time equivalent staff, bed count, and total caseload by type of organization.

strAlloc 63

## **Source**

Substance Abuse and Mental Health Services Administration

## References

Manderscheid, R.W. and Henderson, M.J. (2002). Mental Health, United States, 2002. DHHS Publication No. SMA04-3938. Rockville MD USA: Substance Abuse and Mental Health Services Administration.

## See Also

```
smho.N874
```

# **Examples**

```
data(smho98)
str(smho98)
summary(smho98)
```

strAlloc

Allocate a sample to strata

# Description

Compute the proportional, Neyman, cost-constrained, and variance-constrained allocations in a stratified simple random sample.

# Usage

# **Arguments**

n.tot	fixed total sample size
Nh	vector of population stratum sizes $(N_h)$ or pop stratum proportions $(W_h)$
Sh	stratum unit standard deviations $(S_h)$ , required unless alloc = "prop"
cost	total variable cost
ch	vector of costs per unit in stratum $h\left(c_{h}\right)$
VØ	fixed variance target for estimated mean
CV0	fixed CV target for estimated mean
ybarU	population mean of $y(\bar{y}_U)$
alloc	type of allocation; must be one of "prop", "neyman", "totcost", "totvar"

64 strAlloc

#### **Details**

alloc="prop" computes the proportional allocation of a fixed total sample size, n.tot, to the strata. alloc="neyman" computes the allocation of a fixed total sample size, n.tot, to the strata that minimizes the variance of an estimated mean. alloc="totcost" computes the allocation of a fixed total sample size, n.tot, to the strata that minimizes the variance of an estimated mean subject to the fixed total cost. alloc="totvar" computes the allocation that minimizes total cost subject to the target coefficient of variation, CV0, or the target variance, V0, of the estimated mean.

#### Value

For proportional allocation, a list with values:

```
alloc type of allocation: "prop", "neyman", "totcost", "totvar"  \text{Nh} \qquad \qquad \text{vector of population sizes } (N_h) \text{ or pop stratum proportions } (W_h)
```

nh vector of stratum sample sizes

"nh/n" proportion of sample allocated to each stratum

For other allocations, the three components above plus:

```
Sh stratum unit standard deviations (S_h)
```

"anticipated SE of estimated mean"

Anticipated SE of the estimated mean for the computed allocation

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

#### References

Cochran, W.G. (1977). Sampling Techniques. John Wiley & Sons.

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 3). *Practical Tools for Designing and Weighting Survey Samples*. Springer.

#### See Also

```
nCont, nLogOdds, nProp, nPropMoe, nWilson
```

#### **Examples**

```
# Neyman allocation
Nh <- c(215, 65, 252, 50, 149, 144)
Sh <- c(26787207, 10645109, 6909676, 11085034, 9817762, 44553355)
strAlloc(n.tot = 100, Nh = Nh, Sh = Sh, alloc = "neyman")

# cost constrained allocation
ch <- c(1400, 200, 300, 600, 450, 1000)
strAlloc(Nh = Nh, Sh = Sh, cost = 100000, ch = ch, alloc = "totcost")

# allocation with CV target of 0.05
strAlloc(Nh = Nh, Sh = Sh, CV0 = 0.05, ch = ch, ybarU = 11664181, alloc = "totvar")</pre>
```

ThirdGrade 65

ThirdGrade

Third grade population

#### **Description**

The ThirdGrade data file is a population of students who participated in the Third International Mathematics and Science Study (TIMSS).

# Usage

```
data(ThirdGrade)
```

#### **Format**

```
A data frame with 2,427 students on the following variables:
```

```
region Geographic region of the U.S. (1 = Northeast; 2 = South; 3 = Central; 4 = West) school.id School identifier (1 - 135) student.id Student identifier (1 - 2427) sex Sex of student (1 = female; 2 = male) language Is language of test spoken at home? (1 = always; 2 = sometimes; 3 = never) math Mathematics test score ethnicity Ethnicity of student (1 = White, non-Hispanic; 2 = Black; 3 = Hispanic; 4 = Asian; 5 = Native American; 6 = Other) science Science test score community Type of location of school (2 = village or rural area; 3 = outskirts of a town or city; 4 = close to center of a town or city) enrollment Number of students in entire school
```

#### **Details**

The Third Grade population consists of 2,427 students in the U.S. who participated in the Third International Mathematics and Science Study (Caslyn, Gonzales, Frase 1999). The methods used in conducting the original study are given in TIMSS International Study Center (1996). Clusters are schools while units within clusters are the students.

#### Source

TIMSS International Study Center 1996.

#### References

Caslyn, C., Gonzales, P., Frase, M. (1999). *Highlights from TIMSS*. National Center for Education Statistics, Washington DC.

TIMSS International Study Center (1996). *Third International Mathematics and Science Study: Technical Report, Volume 1 Design and Development*. Boston College: Chestnut Hill MA.

66 wtdvar

## **Examples**

```
data(ThirdGrade)
str(ThirdGrade)
```

wtdvar

Compute weighted variance

# **Description**

Compute an estimate of a population unit variance from a complex sample with survey weights.

# Usage

```
wtdvar(x,w,na.rm=TRUE)
```

# Arguments

x data vector

w vector of survey weights; must be same length as x

na.rm remove missing values (TRUE or FALSE)

# **Details**

wtdvar is also used by BW3stagePPSe in estimating relvariance components.

# Value

numeric estimate of population unit variance

#### Author(s)

Richard Valliant, Jill A. Dever, Frauke Kreuter

## References

Valliant, R., Dever, J., Kreuter, F. (2013, chap. 9). *Practical Tools for Designing and Weighting Survey Samples*. Springer.

# **Examples**

```
x <- c(1:3)
wts <- c(4, 6, 8)
wtdvar(x=x, w=wts)</pre>
```

# **Index**

* datasets	NRadjClass, 55
Domainy1y2, 29	NRFUopt, 56
hospital, 35	nWilson, 58
labor, 36	pclass, 59
MDarea.pop, 37	strAlloc,63
mibrfss, 38	wtdvar, 66
nhis, 45	* models
nhis.large,46	gamEst, 32
nhispart,48	gammaFit, 33
smho.N874, 60	* survey
smho98, 62	BW2stagePPS, 3
ThirdGrade, 65	BW2stagePPSe,4
* methods	BW2stageSRS, 6
BW2stagePPS, 3	BW3stagePPS, 8
BW2stagePPSe, 4	BW3stagePPSe,9
BW2stageSRS, 6	clus0pt2, 12
BW3stagePPS, 8	clusOpt2fixedPSU, 14
BW3stagePPSe, 9	clusOpt3, 15
clusOpt2, <mark>12</mark>	clusOpt3fixedPSU, 17
clusOpt2fixedPSU, 14	CVcalc2, 19
clusOpt3, 15	CVcalc3, 20
clusOpt3fixedPSU, 17	deff, 21
CVcalc2, 19	deffCR, 24
CVcalc3, 20	deffH, 26
deff, 21	deffK, 27
deffCR, 24	deffS, 28
deffH, 26	dub, 30
deffK, 27	HMT, 34
deffS, 28	nCont, 40
dub, 30	nContMoe, 41
HMT, 34	nDep2sam, 42
nCont, 40	nDomain, 44
nContMoe, 41	nLogOdds, 50
nDep2sam, 42	nProp, 51
nDomain, 44	nProp2sam, 52
nLogOdds, 50	nPropMoe, 53
nProp, 51	NRadjClass, 55
nProp2sam, 52	NRFUopt, 56
nPropMoe, 53	nWilson, 58

68 INDEX

pclass, 59	pclass, <i>56</i> , <i>59</i>
strAlloc, 63 wtdvar, 66	smho.N874, 60, <i>63</i> smho98, <i>61</i> , 62
BW2stagePPS, 3, 6, 7, 9, 11 BW2stagePPSe, 3, 4, 4, 7–9, 11	strAlloc,63
BW2stageSRS, 4, 6, 6, 9, 11 BW3stagePPS, 4, 6, 7, 8, 11	ThirdGrade, 65
BW3stagePPSe, 4, 6, 7, 9, 9, 66	wtdvar, 66
clusOpt2, 12, 15, 16, 18 clusOpt2fixedPSU, 13, 14, 16, 18	
clusOpt3, <i>13</i> , <i>15</i> , 15, <i>18</i> clusOpt3fixedPSU, <i>13</i> , <i>15</i> , <i>16</i> , 17	
CVcalc2, 19 CVcalc3, 20, 20, 21	
deff, 21, 25, 27–29	
deffCR, 23, 24, 27–29 deffH, 23, 25, 26, 28, 29	
deffK, 23, 25, 27, 27, 29 deffS, 23, 25, 27, 28, 28	
Domainy1y2, 29 dub, 30	
gamEst, 32, 33, 34 gammaFit, 32, 33	
HMT, 34 hospital, 35	
labor, 36	
MDarea.pop, 37 mibrfss, 38	
nCont, 40, 42, 45, 50, 51, 54, 59, 64 nContMoe, 41	
nDep2sam, 42, 53	
nDomain, 44 nhis, 39, 45, 48	
nhis.large, <i>39</i> , <i>46</i> , 46 nhispart, 48	
nLogOdds, 41, 42, 45, 50, 51, 54, 59, 64 nProp, 41, 42, 45, 50, 51, 54, 59, 64	
nProp2sam, <i>43</i> , 52	
nPropMoe, <i>41</i> , <i>42</i> , <i>45</i> , <i>50</i> , <i>51</i> , <i>53</i> , <i>59</i> , <i>64</i> NRadjClass, <i>55</i> , <i>60</i>	
NRFUopt, 56 nWilson, 41, 42, 45, 50, 51, 54, 58, 64	