

# Package ‘robsurvey’

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**Type** Package

**Title** Robust Survey Statistics Estimation

**Version** 0.7-3

**Description** Robust (outlier-resistant) estimators of finite population characteristics like of means, totals, ratios, regression, etc. Available methods are M- and GM-estimators of regression, weight reduction, trimming, and winsorization. The package extends the 'survey' <https://CRAN.R-project.org/package=survey> package.

**License** GPL (>= 2)

**Classification/MSC-2010** 62D05, 62F35

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**BugReports** <https://github.com/tobiasschoch/robsurvey/issues>

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**VignetteBuilder** knitr, rmarkdown

**Author** Beat Hulliger [aut] (ORCID: <https://orcid.org/0000-0001-5252-8606>),  
Tobias Schoch [aut, cre] (ORCID:  
<https://orcid.org/0000-0002-1640-3395>),  
Martin Sterchi [ctr, com],  
R-core [ctb, cph] (for zeroIn2.c)

**Maintainer** Tobias Schoch <tobias.schoch@fhnw.ch>

**Repository** CRAN

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## Description

A key *design pattern* of the package is that the majority of the estimating methods is available in two "flavors":

- bare-bone methods
- survey methods

Bare-bone methods are stripped-down versions of the survey methods in terms of functionality and informativeness. These functions may serve users and package developers as building blocks. In particular, bare-bone functions *cannot compute* variances.

The survey methods are much more capable and depend, for variance estimation, on the **survey** package.

## Basic Robust Estimators

### Trimming:

- Bare-bone methods: `weighted_mean_trimmed` and `weighted_total_trimmed`
- Survey methods: `svymean_trimmed` and `svytotat_trimmed`

### Winsorization:

- Bare-bone methods:
  - `weighted_mean_winsorized` and `weighted_total_winsorized`
  - `weighted_mean_k_winsorized` and `weighted_total_k_winsorized`
- Survey methods:
  - `svymean_winsorized` and `svytotat_winsorized`
  - `svymean_k_winsorized` and `svytotat_k_winsorized`

### Dalen's estimators (weight reduction methods):

- Bare-bone methods: `weighted_mean_dalen` and `weighted_total_dalen`
- Survey methods: `svymean_dalen` and `svytotat_dalen`

### M-estimators:

- Bare-bone methods:
  - `weighted_mean_huber` and `weighted_total_huber`
  - `weighted_mean_tukey` and `weighted_total_tukey`
  - `huber2` (weighted Huber Proposal 2 estimator)
- Survey methods:
  - `svymean_huber` and `svytotat_huber`
  - `svymean_tukey` and `svytotat_tukey`
  - `mer` (minimum estimated risk estimator)

## Survey Regression (weighted least squares)

[svyreg](#)

## Robust Regression and Ratio Estimation (weighted)

- Regression M-estimators: [svyreg\\_huberM](#) and [svyreg\\_tukeyM](#)
- Regression GM-estimators (Mallows and Schweppe): [svyreg\\_huberGM](#) and [svyreg\\_tukeyGM](#)
- Ratio M-estimators: [svyratio\\_huber](#) and [svyratio\\_tukey](#)

Note: The functions [svyreg\\_huber](#) and [svyreg\\_tukey](#) are deprecated, use instead [svyreg\\_huberM](#) and [svyreg\\_tukeyM](#), respectively; see also [robsurvey-deprecated](#).

## Robust Generalized Regression (GREG) and Ratio Prediction of the Population Mean and Total

- Regression predictors: [svymean\\_reg](#) and [svytotat\\_reg](#)
- Ratio predictors: [svymean\\_ratio](#) and [svytotat\\_ratio](#)

## Utility functions

- [weighted\\_quantile](#) and [weighted\\_median](#)
- [weighted\\_mad](#) and [weighted\\_IQR](#)
- [weighted\\_mean](#) and [weighted\\_total](#)
- [weighted\\_line](#), [weighted\\_median\\_line](#), and [weighted\\_median\\_ratio](#)

---

class\_svyreg\_rob

*Utility Functions for Objects of Class svyreg\_rob*

---

## Description

Methods and utility functions for objects of class `svyreg_rob`.

## Usage

```
## S3 method for class 'svyreg_rob'
print(x, digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'svyreg_rob'
summary(object, mode = c("design", "model", "compound"),
        digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'svyreg_rob'
coef(object, ...)

## S3 method for class 'svyreg_rob'
vcov(object, mode = c("design", "model", "compound"), ...)
```

```
## S3 method for class 'svyreg_rob'
SE(object, mode = c("design", "model", "compound"), ...)

## S3 method for class 'svyreg_rob'
residuals(object, ...)

## S3 method for class 'svyreg_rob'
fitted(object, ...)

## S3 method for class 'svyreg_rob'
robweights(object)

## S3 method for class 'svyreg_rob'
plot(x, which = 1L:4L,
      hex = FALSE, caption = c("Standardized residuals vs. Fitted Values",
                              "Normal Q-Q", "Response vs. Fitted values",
                              "Sqrt of abs(Residuals) vs. Fitted Values"),
      panel = if (add.smooth) function(x, y, ...) panel.smooth(x, y,
                                                                iter = iter.smooth, ...) else points, sub.caption = NULL, main = "",
      ask = prod(par("mfcol")) < length(which) && dev.interactive(), ...,
      id.n = 3, labels.id = names(residuals(x)), cex.id = 0.75, qqline = TRUE,
      add.smooth = getOption("add.smooth"), iter.smooth = 3,
      label.pos = c(4, 2), cex.caption = 1, cex.oma.main = 1.25)
```

## Arguments

<code>x</code>	object of class <code>svyreg_rob</code> .
<code>digits</code>	[integer] minimal number of significant digits.
<code>...</code>	additional arguments passed to the method.
<code>object</code>	object of class <code>svyreg_rob</code> .
<code>mode</code>	[character] mode of variance estimator: "design", "model" or "compound" (default: "design").
<code>which</code>	[integer] indicating which plots to be drawn; if a subset of the plots is required, you can specify a subset of the numbers 1:4.
<code>hex</code>	[logical] if TRUE, a hexagonally binned plot is shown in place of a scatterplot.
<code>caption</code>	[character] captions to appear above the plots; vector of valid graphics annotations. It can be set to "" or NA to suppress all captions.
<code>panel</code>	panel function. The useful alternative to <a href="#">points</a> , <a href="#">panel.smooth</a> can be chosen by <code>add.smooth = TRUE</code> .
<code>sub.caption</code>	[character] common title—above the figures if there are more than one; used as <code>sub(s.title)</code> otherwise. If NULL, as by default, a possible abbreviated version of <code>deparse(x\$call)</code> is used.
<code>main</code>	[character] title to each plot—in addition to caption.
<code>ask</code>	[logical]; if TRUE, the user is <i>asked</i> before each plot, see <a href="#">par</a> ( <code>ask=.</code> ).
<code>id.n</code>	[integer] number of points to be labelled in each plot, starting with the most extreme.

labels.id	[character] vector of labels from which the labels for extreme points will be chosen. NULL uses observation numbers.
cex.id	[numeric] magnification of point labels.
qqline	[logical] indicating if a <a href="#">qqline</a> should be added to the normal Q-Q plot.
add.smooth	[logical] indicating if a smoother should be added to most plots; see also panel above.
iter.smooth	[integer] the number of robustness iterations, the argument iter in <a href="#">panel.smooth</a> .
label.pos	[numeric] positioning of labels, for the left half and right half of the graph respectively.
cex.caption	[numeric] controls the size of caption.
cex.oma.main	[numeric] controls the size of the sub.caption only if that is <i>above</i> the figures when there is more than one.

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

**Variance** For variance estimation (summary, vcov, and SE) three modes are available:

- "design": design-based variance estimator using linearization; see Binder (1983)
- "model": model-based weighted variance estimator (the sampling design is ignored)
- "compound": design-model-based variance estimator; see Rubin-Bleuer and Schiopu-Kratina (2005) and Binder and Roberts (2009)

**Utility functions** The following utility functions are available:

- summary gives a summary of the estimation properties
- plot shows diagnostic plots for the estimated regression model
- robweights extracts the robustness weights (if available)
- coef extracts the estimated regression coefficients
- vcov extracts the (estimated) covariance matrix
- residuals extracts the residuals
- fitted extracts the fitted values

## References

- Binder, D. A. (1983). On the Variances of Asymptotically Normal Estimators from Complex Surveys. *International Statistical Review* **51**, 279–292. doi:[10.2307/1402588](#)
- Binder, D. A. and Roberts, G. (2009). Design- and Model-Based Inference for Model Parameters. In: *Sample Surveys: Inference and Analysis* ed. by Pfeffermann, D. and Rao, C. R. Volume 29B of *Handbook of Statistics*, Amsterdam: Elsevier, Chap. 24, 33–54 doi:[10.1016/S01697161\(09\)002247](#)
- Rubin-Bleuer, S. and Schiopu-Kratina, I. (2005). On the Two-phase framework for joint model and design-based inference. *The Annals of Statistics* **33**, 2789–2810. doi:[10.1214/009053605000000651](#)

## See Also

Weighted least squares: [svyreg](#); robust weighted regression [svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#)

**Examples**

```

head(workplace)

library(survey)
# Survey design for simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Compute regression M-estimate with Huber psi-function
m <- svyreg_huberM(payroll ~ employment, dn, k = 14)

# Diagnostic plots (e.g., standardized residuals against fitted values)
plot(m, which = 1L)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(m), robweights(m))

# Utility functions
summary(m)
coef(m)
SE(m)
vcov(m)
residuals(m)
fitted(m)
robweights(m)

```

---

class\_svystat\_rob

*Utility Functions for Objects of Class svystat\_rob*


---

**Description**

Methods and utility functions for objects of class svystat\_rob.

**Usage**

```

mse(object, ...)
## S3 method for class 'svystat_rob'
mse(object, ...)
## S3 method for class 'svystat'
mse(object, ...)
## S3 method for class 'svystat_rob'
summary(object, digits = max(3L,

```

```

        getOption("digits") - 3L), ...)
## S3 method for class 'svystat_rob'
coef(object, ...)
## S3 method for class 'svystat_rob'
SE(object, ...)
## S3 method for class 'svystat_rob'
vcov(object, ...)
## S3 method for class 'svystat_rob'
scale(x, ...)
## S3 method for class 'svystat_rob'
residuals(object, ...)
## S3 method for class 'svystat_rob'
fitted(object, ...)
robweights(object)
## S3 method for class 'svystat_rob'
robweights(object)
## S3 method for class 'svystat_rob'
print(x, digits = max(3L, getOption("digits") - 3L), ...)
## S3 method for class 'svystat_rob'
confint(object, parm, level = 0.95, df = Inf, ...)

```

## Arguments

df	[integer] degrees of freedom for t-distribution in confidence interval, use 'degf(design)' for number of PSUs minus number of strata
digits	[integer] minimal number of significant digits.
level	[numeric] the confidence level required.
object	object of class svystat_rob.
parm	a specification of which parameters are to be given confidence intervals, either a vector of numbers or a vector of names. If missing, all parameters are considered.
x	object of class svystat_rob.
...	additional arguments passed to the method.

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

Utility functions:

- mse computes the estimated risk (mean square error) in presence of representative outliers; see also [mer](#)
- summary gives a summary of the estimation properties
- robweights extracts the robustness weights
- coef extracts the estimate of location
- SE extracts the (estimated) standard error



- `vcov` extracts the (estimated) covariance matrix
- `residuals` extracts the residuals
- `fitted` extracts the fitted values

### See Also

[svymean\\_dalen](#), [svymean\\_huber](#), [svymean\\_ratio](#), [svymean\\_reg](#), [svymean\\_tukey](#), [svymean\\_trimmed](#), [svymean\\_winsorized](#)

[svytotal\\_dalen](#), [svytotal\\_huber](#), [svytotal\\_ratio](#), [svytotal\\_reg](#), [svytotal\\_tukey](#), [svytotal\\_trimmed](#), [svytotal\\_winsorized](#)

### Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Estimated one-sided k winsorized population total (i.e., k = 2 observations
# are winsorized at the top of the distribution)
wtot <- svytotal_k_winsorized(~employment, dn, k = 2)

# Show summary statistic of the estimated total
summary(wtot)

# Estimated mean square error (MSE)
mse(wtot)

# Estimate, std. err., variance, and the residuals
coef(wtot)
SE(wtot)
vcov(wtot)
residuals(wtot)

# M-estimate of the total (Huber psi-function; tuning constant k = 3)
mtot <- svytotal_huber(~employment, dn, k = 45)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(mtot), robweights(mtot))
```

---

counties

*Data on a Simple Random Sample of 100 Counties in the U.S.*


---

**Description**

Data from a simple random sample (without replacement) of 100 of the 3141 counties in the United States (U.S. Bureau of the Census, 1994).

**Usage**

```
data(counties)
```

**Format**

A data.frame with 100 observations on the following variables:

```
state state, [character].
county county, [character].
landarea land area, 1990 (square miles), [double].
totpop population total, 1992, [double].
unemp number of unemployed persons, 1991, [double].
farmpop farm population, 1990, [double].
numfarm number of farms, 1987, [double].
farmacre acreage in farms, 1987, [double].
weights sampling weight, [double].
fpc finite population correction, [double].
```

**Details**

The data (and 10 additional variables) are published in Lohr (1999, Appendix C).

**Source**

Lohr, S. L. (1999). *Sampling: Design and Analysis*, Pacific Grove (CA): Duxbury Press.

**Examples**

```
head(counties)

library(survey)
# Survey design for simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~1, fpc = ~fpc, weights = ~weights, data = counties,
    calibrate.formula = ~1)
} else {
```

```
# legacy mode
svydesign(ids = ~1, fpc = ~fpc, weights = ~weights, data = counties)
}
```

---

flour

*Measurement of Copper Content in Wholemeal Flour*

---

### Description

Measurement of copper content in wholemeal flour (measured in parts per million).

### Usage

```
data(flour)
```

### Format

A data.frame with 24 observations (sorted in ascending order) on the following variables:

copper copper content [double].

weight weight [double].

### Details

The data are published in Maronna et al. (2019, p. 2).

### Source

Maronna, R. A., Martin, R. D., Yohai, V. J. and Salibián-Barrera, M. (2019). *Robust Statistics: Theory and Methods (with R)*, Hoboken (NJ): John Wiley and Sons, 2nd edition. doi:[10.1002/9781119214656](https://doi.org/10.1002/9781119214656)

### Examples

```
head(flour)
```

huber2

*Weighted Huber Proposal 2 Estimator***Description**

Weighted Huber Proposal 2 estimator of location and scatter.

**Usage**

```
huber2(x, w, k = 1.5, na.rm = FALSE, maxit = 50, tol = 1e-04, info = FALSE,
       k_Inf = 1e6, df_cor = TRUE)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
maxit	[integer] maximum number of iterations to use (default: 50).
tol	[double] numerical tolerance criterion to stop the iterations (default: 1e-04).
info	[logical] indicating whether additional information should be returned (default: FALSE).
k_Inf	[integer] numerical value that represents Inf (default: 1e+06).
df_cor	[logical] if TRUE, the degrees of freedom of the estimate of scale is adjusted (default: TRUE).

**Details**

Function huber2 computes the weighted Huber (1964) Proposal 2 estimates of location and scale.

The method is initialized by the weighted median (location) and the weighted interquartile range (scale).

**Value**

The return value depends on info:

info = FALSE: estimate of mean or total [double]

info = TRUE: a [list] with items:

- characteristic [character],
- estimator [character],
- estimate [double],
- variance (default: NA),
- robust [list],

- residuals [numeric vector],
- model [list],
- design (default: NA),
- [call]

## Comparison

The `huber2` estimator is initialized by the weighted median and the weighted (scaled) interquartile range. For unweighted data, this estimator *differs* from `hubers` in **MASS**, which is initialized by `mad`.

The difference between the estimators is usually negligible (for sufficiently small values of `tol`). See examples.

## References

Huber, P. J. (1964). Robust Estimation of a Location Parameter. *Annals of Mathematical Statistics* **35**, 73–101. doi:[10.1214/aoms/1177703732](https://doi.org/10.1214/aoms/1177703732)

## Examples

```
head(workplace)

# Weighted "Proposal 2" estimator of the mean
huber2(workplace$employment, workplace$weight, k = 8)

# More information on the estimate, i.e., info = TRUE
m <- huber2(workplace$employment, workplace$weight, k = 8, info = TRUE)

# Estimate of scale
m$scale

# Comparison with MASS::hubers (without weights). We make a copy of MASS::hubers
library(MASS)
hubers_mod <- hubers

# Then we replace mad by the (scaled) IQR as initial scale estimator
body(hubers_mod)[[7]][[3]][[2]] <- substitute(s0 <- IQR(y, type = 2) * 0.7413)

# Define the numerical tolerance
TOLERANCE <- 1e-8

# Comparison
m1 <- huber2(workplace$payroll, rep(1, 142), tol = TOLERANCE)
m2 <- hubers_mod(workplace$payroll, tol = TOLERANCE)$mu
m1 / m2 - 1

# The absolute relative difference is < 4.0-09 (smaller than TOLERANCE)
```

---

losdata	<i>Length-of-Stay (LOS) Hospital Data</i>
---------	---

---

## Description

A simple random sample of 70 patients in inpatient hospital treatment.

## Usage

```
data(losdata)
```

## Format

A data.frame with data on the following variables:

los length of stay (days) [integer].

weight sampling weight [double].

fpc finite population correction [double].

## Details

The losdata are a simple random sample without replacement (SRSWOR) of size  $n = 70$  patients from the (fictive) population of  $N = 2479$  patients in inpatient hospital treatment. We have constructed the losdata as a showcase; though, the LOS measurements are real data that we have taken from the 201 observations in Ruffieux et al. (2000). The original LOS data of Ruffieux et al. (2000) are available in the R package **robustbase**; see `robustbase::data(los)`. Our losdata are a SRSWOR of size  $n = 70$  from the 201 original observations.

Ruffieux et al. (2000) and data.frame los in the R package **robustbase**.

## Source

Ruffieux, C., Paccaud, F. and Marazzi, A. (2000). Comparing rules for truncating hospital length of stay. *Casemix Quarterly* **2**.

## Examples

```
head(losdata)

library(survey)
# Survey design for simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~1, fpc = ~fpc, weights = ~weight, data = losdata,
            calibrate.formula = ~1)
} else {
  # legacy mode
  svydesign(ids = ~1, fpc = ~fpc, weights = ~weight, data = losdata)
}
```

mer

*Minimum Estimated Risk (MER) M-Estimator***Description**

mer is an adaptive M-estimator of the weighted mean or total. It is defined as the estimator that minimizes the estimated mean square error, [mse](#), of the estimator under consideration.

**Usage**

```
mer(object, verbose = TRUE, max_k = 10, init = 1, method = "Brent",
     optim_args = list())
```

**Arguments**

object	an object of class <code>svystat_rob</code> .
verbose	[logical] indicating whether additional information is printed to the console (default: TRUE).
init	[numeric] determines the left boundary value of the search interval and the initial value of the search; we must have <code>init &lt; max_k</code> .
method	[character] the method of <a href="#">optim</a> to be used.
max_k	[numeric vector] defines the right boundary value of the search interval (default: <code>max_k = 1000</code> )
optim_args	[list]: arguments passed on to <a href="#">optim</a> .

**Details**

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

MER-estimators are available for the methods [svymean\\_huber](#), [svytotal\\_huber](#), [svymean\\_tukey](#) and [svytotal\\_tukey](#).

**Value**

Object of class `svystat_rob`

**References**

Hulliger, B. (1995). Outlier Robust Horvitz-Thompson Estimators. *Survey Methodology* **21**, 79–87.

**See Also**

[Overview](#) (of all implemented functions)

## Examples

```
head(losdata)

library(survey)
# Survey design for simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~1, fpc = ~fpc, weights = ~weight, data = losdata,
            calibrate.formula = ~1)
} else {
  # legacy mode
  svydesign(ids = ~1, fpc = ~fpc, weights = ~weight, data = losdata)
}

# M-estimator of the total with tuning constant k = 8
m <- svymean_huber(~los, dn, type = "rhj", k = 8)

# MER estimator
mer(m)
```

---

 MU284pps

*PPS Sample From the MU284 Population*


---

## Description

Probability-proportional-to-size sample (PPS) without replacement of municipalities from the MU284 population in Särndal et al. (1992). The sample inclusion probabilities are proportional to the population size in 1975 (variable P75).

## Usage

```
data(MU284pps)
```

## Format

A data.frame with 60 observations on the following variables:

LABEL identifier variable, [integer].

P85 1985 population size (in thousands), [double].

P75 1975 population size (in thousands), [double].

RMT85 Revenues from the 1985 municipal taxation (in millions of kronor), [double].

CS82 number of Conservative seats in municipal council, [double].

SS82 number of Social-Democrat seats in municipal council (1982), [double].

S82 total number of seats in municipal council (1982), [double].

ME84 number of municipal employees in 1984, [double].

REV84 real estate values according to 1984 assessment (in millions of kronor), [double].



REG geographic region indicator, [integer].

CL cluster indicator (a cluster consists of a set of neighbouring municipalities), [integer].

weights sampling weights, [double].

pi sample inclusion probability, [double].

## Details

The MU284 population of Särndal et al. (1992, Appendix B) is a dataset with observations on the 284 municipalities in Sweden in the late 1970s and early 1980s. The MU284 *population* data are available in the **sampling** package of Tillé and Matei (2021).

The data frame MU284pps is a probability-proportional-to-size sample (PPS) without replacement from the MU284 population. The sample inclusion probabilities are proportional to the population size in 1975 (variable P75). The sample has been selected by Brewer's method; see Tillé (2006, Chap. 7). The sampling weight (inclusion probabilities) are calibrated to the population size and the population total of P75.

## Source

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

Tillé, Y. and Matei, A. (2021). *sampling: Survey Sampling*. R package version 2.9. <https://CRAN.R-project.org/package=sampling>

Tillé, Y. (2006). *Sampling Algorithms*. New York: Springer-Verlag.

## See Also

[MU284strat](#)

## Examples

```
head(MU284pps)

library(survey)
# Survey design with inclusion probabilities proportional to size
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~LABEL, fpc = ~pi, data = MU284pps, pps = "brewer",
    calibrate.formula = ~1)
} else {
  # legacy mode
  svydesign(ids = ~LABEL, fpc = ~pi, data = MU284pps, pps = "brewer")
}
```

---

 MU284strat

*Stratified Sample from the MU284 Population*


---

## Description

Stratified simple random sample (without replacement) of municipalities from the MU284 population in Särndal et al. (1992). Stratification is by geographic region and a take-all stratum (by 1975 population size), which includes the big cities Stockholm, Göteborg, and Malmö.

## Usage

```
data(MU284strat)
```

## Format

A data.frame with 60 observations on the following variables:

LABEL identifier variable, [integer].

P85 1985 population size (in thousands), [double].

P75 1975 population size (in thousands), [double].

RMT85 Revenues from the 1985 municipal taxation (in millions of kronor), [double].

CS82 number of Conservative seats in municipal council, [double].

SS82 number of Social-Democrat seats in municipal council (1982), [double].

S82 total number of seats in municipal council (1982), [double].

ME84 number of municipal employees in 1984, [double].

REV84 real estate values according to 1984 assessment (in millions of kronor), [double].

CL cluster indicator (a cluster consists of a set of neighbouring municipalities), [integer].

REG geographic region indicator, [integer].

Stratum stratum indicator, [integer].

weights sampling weights, [double].

fpc finite population correction, [double].

## Details

The MU284 population of Särndal et al. (1992, Appendix B) is a dataset with observations on the 284 municipalities in Sweden in the late 1970s and early 1980s. The MU284 *population* data are available in the **sampling** package of Tillé and Matei (2021).

The population is divided into two parts based on 1975 population size (P75):

- the MU281 population, which consists of the 281 smallest municipalities;
- the MU3 population of the three biggest municipalities/ cities in Sweden (Stockholm, Göteborg, and Malmö).

The three biggest cities take exceedingly large values (representative outliers) on almost all of the variables. To account for this, a stratified sample has been drawn from the MU284 population using a take-all stratum. The sample data, MU284strat, (of size  $n = 60$ ) consists of

- a stratified simple random sample (without replacement) from the MU281 population, where stratification is by geographic region (REG) with proportional sample size allocation;
- a take-all stratum that includes the three biggest cities/ municipalities (population M3).

### Source

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

Tillé, Y. and Matei, A. (2021). *sampling: Survey Sampling*. R package version 2.9. <https://CRAN.R-project.org/package=sampling>

### See Also

[MU284pps](#)

### Examples

```
head(MU284strat)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~LABEL, strata = ~Stratum, fpc = ~fpc,
            weights = ~weights, data = MU284strat,
            calibrate.formula = ~-1 + Stratum)
} else {
  # legacy mode
  svydesign(ids = ~LABEL, strata = ~Stratum, fpc = ~fpc,
            weights = ~weights, data = MU284strat)
}
```

---

pps

*Sampling with probability proportional to size (pps without replacement)*

---

### Description

Methods to compute the first-order sample inclusion probabilities (given a measure of size) and sampling mechanisms to draw samples with probabilities proportional to size (pps).

**Usage**

```
pps_probabilities(size, n)
pps_draw(x, method = "brewer", sort = TRUE)

## S3 method for class 'prob_pps'
print(x, ...)
```

**Arguments**

size	[numeric vector] measure of size.
n	[integer] sample size.
x	object of class prob_pps.
method	[character] currently only method "brewer" is implemented.
sort	[logical] indicating whether the sampled indices are sorted in ascending order (default: TRUE).
...	additional arguments.

**Details**

Function `pps_probabilities` computes the first-order sample inclusion probabilities for a given sample size  $n$ ; see e.g., Särndal et al., 1992 (p. 90). The probabilities (and additional attributes) are returned as a vector, more precisely as an object of class `prob_pps`. To get the probabilities, use the function `as.numeric()` on the object.

For an object of class `prob_pps` (inclusion probabilities and additional attributes), function `pps_draw` draws a pps sample without replacement and returns the indexes of the population elements. Only the method of Brewer (1963, 1975) is currently implemented.

**Value**

Function `pps_probabilities` returns the probabilities (an object of class `prob_pps`).

Function `pps_draw` returns a pps sample of indexes from the population elements.

**References**

- Brewer, K. W. R. (1963). A Model of Systematic Sampling with Unequal Probabilities. *Australian Journal of Statistics* **5**, 93–105. doi:10.1111/j.1467842X.1963.tb00132.x
- Brewer, K. W. R. (1975). A simple procedure for  $\pi$ pswor, *Australian Journal of Statistics* **17**, 166–172. doi:10.1111/j.1467842X.1975.tb00954.x
- Särndal, C.-E., Swensson, B., Wretman, J. (1992). *Model Assisted Survey Sampling*, New York: Springer-Verlag.

**Examples**

```
# We are going to pretend that the workplace sample is our population.
head(workplace)
```

```
# The population size is N = 142. We want to draw a pps sample (without
# replacement) of size n = 10, where the variable employment is the measure of
# size. The first-order sample inclusion probabilities are calculated as
# follows
p <- pps_probabilities(workplace$employment, n = 10)

# Extract the probabilities
as.numeric(p)

# Now, we draw a pps sample using Brewer's method.
pps_draw(p, method = "brewer")
```

---

robsurvey-deprecated     *Deprecated Functions in Package 'robsurvey'*

---

### Description

These functions are provided for compatibility with older versions of the package only, and may be defunct as soon as the next release.

- [svyreg\\_huber](#)
- [svyreg\\_tukey](#)

Use instead:

- [svyreg\\_huberM](#)
- [svyreg\\_tukeyM](#)

### See Also

[robsurvey-package](#)

---

robsvyreg     *Internal Function for the Regression GM-Estimator*

---

### Description

**Internal** function to call the robust survey regression *GM*-estimator; this function is **only** intended for internal use. The function does **not** check or validate the arguments. In particular, missing values in the data may make the function crash.

### Usage

```
robsvyreg(x, y, w, k, psi, type, xwgt, var = NULL, verbose = TRUE, ...)
svyreg_control(tol = 1e-5, maxit = 100, k_Inf = 1e6, init = NULL,
               mad_center = TRUE, ...)
```

**Arguments**

<code>x</code>	[numeric matrix] design matrix (NA values not allowed).
<code>y</code>	[numeric vector] dependent variable (NA values not allowed).
<code>w</code>	[numeric vector] weights (no NA's allowed).
<code>k</code>	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
<code>psi</code>	[integer] psi-functions: 0: Huber, 1: asymmetric Huber, and 2: Tukey bi-weight.
<code>type</code>	[integer] type of estimator; 0: M-estimator; 1: Mallows and 2: Schweppe type GM-estimator.
<code>xwgt</code>	[numeric vector] weights for design space used in GM-estimators (default: NULL, NA values not allowed).
<code>var</code>	[numeric vector] heteroscedastic variance (default: NULL).
<code>verbose</code>	[logical] indicating whether additional information is printed to the console (default: TRUE).
<code>tol</code>	[double] numerical tolerance criterion to stop the iterations (default: 1e-05).
<code>maxit</code>	[integer] maximum number of iterations to use (default: 100).
<code>k_Inf</code>	[integer] numerical value that represents Inf (default: 1e+06).
<code>init</code>	either NULL or [numeric vector], if <code>init = NULL</code> the regression estimator is initialized by weighted least squares; otherwise, <code>init</code> can be specified as the estimate (i.e., $p$ -vector) to initialize the iteratively re-weighted least squares method (default: NULL).
<code>mad_center</code>	[logical] if TRUE, the weighted MAD is centered about the (weighted) median, otherwise the weighted MAD is centered about zero (default: TRUE).
<code>...</code>	additional arguments passed to the method (see <code>svyreg_control</code> ).

**Details**

Not documented

**Value**

[list]

---

svymean-m-estimator	<i>Weighted Huber and Tukey Mean and Total (M-Estimator) – Robust Horvitz-Thompson Estimator</i>
---------------------	--

---

**Description**

Weighted Huber and Tukey  $M$ -estimator of the population mean and total (robust Horvitz-Thompson estimator)

**Usage**

```
svymean_huber(x, design, k, type = "rwm", asym = FALSE, na.rm = FALSE,
              verbose = TRUE, ...)
svyttotal_huber(x, design, k, type = "rwm", asym = FALSE, na.rm = FALSE,
               verbose = TRUE, ...)
svymean_tukey(x, design, k, type = "rwm", na.rm = FALSE, verbose = TRUE, ...)
svyttotal_tukey(x, design, k, type = "rwm", na.rm = FALSE, verbose = TRUE, ...)
```

**Arguments**

<code>x</code>	a one-sided [formula], e.g., <code>~myVariable</code> .
<code>design</code>	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
<code>k</code>	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
<code>type</code>	[character] type of method: "rwm" or "rht".
<code>asym</code>	[logical] if TRUE, an asymmetric Huber psi-function is used (default: FALSE).
<code>na.rm</code>	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
<code>verbose</code>	[logical] indicating whether additional information is printed to the console (default: TRUE).
<code>...</code>	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.; see <a href="#">svyreg_control</a> ).

**Details**

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

**Methods/ types** `type = "rht"` or `type = "rwm"`; see [weighted\\_mean\\_huber](#) or [weighted\\_mean\\_tukey](#) for more details.

**Variance estimation.** Taylor linearization (residual variance estimator).

**Utility functions** [summary](#), [coef](#), [SE](#), [vcov](#), [residuals](#), [fitted](#), [robweights](#).

**Bare-bone functions** See [weighted\\_mean\\_huber](#) [weighted\\_mean\\_tukey](#), [weighted\\_total\\_huber](#), and [weighted\\_total\\_tukey](#).

**Value**

Object of class [svystat\\_rob](#)

**Failure of convergence**

By default, the method assumes a maximum number of `maxit = 100` iterations and a numerical tolerance criterion to stop the iterations of `tol = 1e-05`. If the algorithm fails to converge, you may consider changing the default values; see [svyreg\\_control](#).

**References**

Hulliger, B. (1995). Outlier Robust Horvitz-Thompson Estimators. *Survey Methodology* **21**, 79–87.

**See Also**

[Overview](#) (of all implemented functions)

**Examples**

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Robust Horvitz-Thompson M-estimator of the population total
svytotal_huber(~employment, dn, k = 9, type = "rht")

# Robust weighted M-estimator of the population mean
m <- svymean_huber(~employment, dn, k = 12, type = "rwm")

# Summary statistic
summary(m)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(m), robweights(m))

# Extract estimate
coef(m)

# Extract estimate of scale
scale(m)

# Extract estimated standard error
SE(m)
```

---

svymean\_dalen

---

*Dalen's Estimators of the Population Mean and Total*


---

**Description**

Dalen's estimators Z2 and Z3 of the population mean and total; see [weighted\\_mean\\_dalen](#) for further details.



**Usage**

```
svymean_dalen(x, design, censoring, type = "Z2", na.rm = FALSE,
              verbose = TRUE, ...)
svyttotal_dalen(x, design, censoring, type = "Z2", na.rm = FALSE,
               verbose = TRUE, ...)
```

**Arguments**

<code>x</code>	a one-sided [formula], e.g., <code>~myVariable</code> .
<code>design</code>	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
<code>censoring</code>	[double] cutoff threshold above which the observations are censored.
<code>type</code>	[character] type of estimator; either "Z2" or "Z3" (default: "Z2").
<code>na.rm</code>	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
<code>verbose</code>	[logical] indicating whether additional information is printed to the console (default: TRUE).
<code>...</code>	additional arguments (currently not used).

**Details**

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

**Methods/ types** `type = "Z2"` or `type = "Z3"`; see [weighted\\_mean\\_dalen](#) for more details.

**Utility functions** [summary](#), [coef](#), [SE](#), [vcov](#), [residuals](#), [fitted](#), [robweights](#).

**Bare-bone functions** See [weighted\\_mean\\_dalen](#) and [weighted\\_total\\_dalen](#).

**Value**

Object of class [svystat\\_rob](#)

**References**

Dalén, J. (1987). Practical Estimators of a Population Total Which Reduce the Impact of Large Observations. R & D Report U/STM 1987:32, Statistics Sweden, Stockholm.

**See Also**

[Overview](#) (of all implemented functions)

[svymean\\_trimmed](#), [svyttotal\\_trimmed](#), [svymean\\_winsorized](#), [svyttotal\\_winsorized](#), [svymean\\_huber](#) and [svyttotal\\_huber](#)

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Dalen's estimator Z3 of the population total
svytotal_dalen(~employment, dn, censoring = 20000, type = "Z3")

# Dalen's estimator Z3 of the population mean
m <- svymean_dalen(~employment, dn, censoring = 20000, type = "Z3")

# Summarize
summary(m)

# Extract estimate
coef(m)

# Extract estimated standard error
SE(m)
```

---

svymean\_ratio

*Robust Ratio Predictor of the Mean and Total*


---

## Description

Robust ratio predictor ( $M$ -estimator) of the population mean and total with Huber and Tukey bi-weight (bisquare) psi-function.

## Usage

```
svytotal_ratio(object, total, variance = "wu", keep_object = TRUE, ...)
svymean_ratio(object, total, N = NULL, variance = "wu",
  keep_object = TRUE, N_unknown = FALSE, ...)
```

## Arguments

object	an object of class [ratio], e.g., result of the Huber ratio $M$ -estimator svyratio_huber.
total	[numeric] vector of population totals of the auxiliary variables.

<code>N</code>	[numeric] population size (see also <code>N_unknown</code> ).
<code>variance</code>	[character] type of variance estimator (default: "wu"); see Details Section.
<code>keep_object</code>	[logical] if TRUE, object is returned as an additional slot of the return value (default: TRUE).
<code>N_unknown</code>	[logical] if TRUE, it is assumed that the population size is unknown; thus, it is estimated (default: FALSE).
<code>...</code>	additional arguments (currently not used).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

The (robust) ratio predictor of the population total or mean is computed in two steps.

- Step 1: Fit the ratio model associated with the predictor by one of the functions [svyratio\\_huber](#) or [svyratio\\_tukey](#). The fitted model is called object.
- Step 2: Based on the fitted model obtained in the first step, we predict the population total and mean, respectively, by the predictors `svytotal_ratio` and `svymean_ratio`, where object is the fitted ratio model.

**Auxiliary data** Two types of auxiliary variables are distinguished: (1) population size  $N$  and (2) the population total of the auxiliary variable (denominator) used in the ratio model.

The option `N_unknown = TRUE` can be used in the predictor of the population mean if  $N$  is unknown.

**Variance estimation** Three variance estimators are implemented (argument `variance`): "base", "wu", and "hajek". These estimators correspond to the estimators  $v_0$ ,  $v_1$ , and  $v_2$  in Wu (1982).

**Utility functions** The return value is an object of class [svystat\\_rob](#). Thus, the utility functions [summary](#), [coef](#), [SE](#), [vcov](#), [residuals](#), [fitted](#), and [robweights](#) are available.

## Value

Object of class [svystat\\_rob](#)

## References

Wu, C.-F. (1982). Estimation of Variance of the Ratio Estimator. *Biometrika* **69**, 183–189.

## See Also

[Overview](#) (of all implemented functions)

[svymean\\_reg](#) and [svytotal\\_reg](#) for (robust) GREG regression predictors

[svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#) for robust regression  $M$ - and  $GM$ -estimators

[svymean\\_huber](#), [svytotal\\_huber](#), [svymean\\_tukey](#) and [svytotal\\_tukey](#) for  $M$ -estimators

**Examples**

```

head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Robust ratio M-estimator with Huber psi-function
rat <- svyratio_huber(~payroll, ~ employment, dn, k = 5)

# Summary of the ratio estimate
summary(rat)

# Diagnostic plots of the ration/regression M-estimate (e.g.,
# standardized residuals against fitted values)
plot(rat, which = 1L)

# Plot of the robustness weights of the ratio/regression M-estimate
# against its residuals
plot(residuals(rat), robweights(rat))

# Robust ratio predictor of the population mean
m <- svymean_ratio(rat, total = 1001233, N = 90840)
m

# Summary of the ratio estimate of the population mean
summary(m)

# Extract estimate
coef(m)

# Extract estimate of scale
scale(m)

# Extract estimated standard error
SE(m)

```

## Description

Generalized regression estimator (GREG) predictor of the mean and total, and robust GREG  $M$ -estimator predictor

## Usage

```
svytotal_reg(object, totals, N = NULL, type, k = NULL, check.names = TRUE,
             keep_object = TRUE, ...)
svymean_reg(object, totals, N = NULL, type, k = NULL, check.names = TRUE,
            keep_object = TRUE, N_unknown = FALSE, ...)
```

## Arguments

object	an object of class [svyreg_rob], e.g., result of the Huber regression $M$ -estimator svyreg_huberM.
totals	[numeric] vector of population totals of the auxiliary variables.
N	[numeric] population size (see also N_unknown).
type	[character] type of predictor; see Details Section.
k	[numeric] robustness tuning constant of the psi-function used in the bias-correction term of the GREG. The definition of k depends on the type of predictor and is discussed in the Details Section.
check.names	[logical] if TRUE, the names of auxiliary are checked against the names of the independent variables of the fitted model object (default: TRUE).
keep_object	[logical] if TRUE, object is returned as an additional slot of the return value (default: TRUE).
N_unknown	[logical] if TRUE, it is assumed that the population size is unknown; thus, it is estimated (default: FALSE).
...	additional arguments (currently not used).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

The (robust) GREG predictor of the population total or mean is computed in two steps.

- Step 1: Fit the regression model associated with the GREG predictor by one of the functions [svyreg](#), [svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) or [svyreg\\_tukeyGM](#). The fitted model is called object.
- Step 2: Based on the fitted model obtained in the first step, we predict the population total and mean, respectively, by the predictors svytotal\_reg and svymean\_reg, where object is the fitted regression model.

The following GREG predictors are available:

**GREG (not robust, k = NULL)** The following *non-robust* GREG predictors are available:

- type = "projective" ignores the bias correction term of the GREG predictor; see Särndal and Wright (1984).

- `type = "ADU"` is the "standard" GREG, which is an asymptotically design unbiased (ADU) predictor; see Särndal et al.(1992, Chapter 6).

If the fitted regression model (object) does include a regression intercept, the predictor types "projective" and "ADU" are identical because the bias correction of the GREG is zero by design.

**Robust GREG** The following *robust* GREG predictors are available:

- `type = "huber"` and `type = "tukey"` are, respectively, the robust GREG predictors with Huber and Tukey bisquare (biweight) psi-function. The tuning constant must satisfy  $0 < k \leq \text{Inf}$ . We can use the Huber-type GREG predictor although the model has been fitted by the regression estimator with Tukey psi-function (and vice versa).
- `type = "BR"` is the bias-corrected robust GREG predictor of Beaumont and Rivest (2009), which is inspired by the bias-corrected robust predictor of Chambers (1986). The tuning constant must satisfy  $0 < k \leq \text{Inf}$ .
- `type = "lee"` is the bias-corrected predictor of Lee (1991; 1992). The tuning constant  $k$  must satisfy  $0 \leq k \leq 1$ .
- `type = "duchesne"` is the bias-corrected, calibration-type estimator/ predictor of Duchesne (1999). The tuning constant  $k$  must be specified as a vector  $k = c(a, b)$ , where  $a$  and  $b$  are the tuning constants of Duchesne's modified Huber psi-function (default values:  $a = 9$  and  $b = 0.25$ ).

**Auxiliary data** Two types of auxiliary variables are distinguished: (1) population size  $N$  and (2) population totals of the auxiliary variables used in the regression model (i.e., non-constant explanatory variables).

The option `N_unknown = TRUE` can be used in the predictor of the population mean if  $N$  is unknown.

The names of the entries of `totals` are checked against the names of the regression fit (object), unless we specify `check.names = FALSE`.

**Utility functions** The return value is an object of class `svystat_rob`. Thus, the utility functions `summary`, `coef`, `SE`, `vcov`, `residuals`, `fitted`, and `robweights` are available.

## Value

Object of class `svystat_rob`

## References

- Beaumont, J.-F. and Rivest, L.-P. (2009). Dealing with outliers in survey data. In: *Sample Surveys: Theory, Methods and Inference* ed. by Pfeffermann, D. and Rao, C. R. Volume 29A of *Handbook of Statistics*, Amsterdam: Elsevier, Chap. 11, 247–280. doi:10.1016/S01697161(08)000114
- Chambers, R. (1986). Outlier Robust Finite Population Estimation. *Journal of the American Statistical Association* **81**, 1063–1069. doi:10.1080/01621459.1986.10478374
- Duchesne, P. (1999). Robust calibration estimators, *Survey Methodology* **25**, 43–56.
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## See Also

[Overview](#) (of all implemented functions)

[svymean\\_ratio](#) and [svytotal\\_ratio](#) for (robust) ratio predictors

[svymean\\_huber](#), [svytotal\\_huber](#), [svymean\\_tukey](#) and [svytotal\\_tukey](#) for  $M$ -estimators

[svyreg](#), [svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#) for robust regression  $M$ - and  $GM$ -estimators

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Robust regression M-estimator with Huber psi-function
reg <- svyreg_huberM(payload ~ employment, dn, k = 3)

# Summary of the regression M-estimate
summary(reg)

# Diagnostic plots of the regression M-estimate (e.g., standardized
# residuals against fitted values)
plot(reg, which = 1L)

# Plot of the robustness weights of the regression M-estimate against
# its residuals
plot(residuals(reg), robweights(reg))

# ADU (asymptotically design unbiased) estimator
m <- svytotal_reg(reg, totals = 1001233, 90840, type = "ADU")
m
```

```

# Robust GREG estimator of the mean; the population means of the auxiliary
# variables are from a register
m <- svymean_reg(reg, totals = 1001233, 90840, type = "huber", k = 20)
m

# Summary of the robust GREG estimate
summary(m)

# Extract estimate
coef(m)

# Extract estimated standard error
SE(m)

# Approximation of the estimated mean square error
mse(m)

```

svymean\_trimmed

*Weighted Trimmed Mean and Total*

## Description

Weighted trimmed population mean and total.

## Usage

```

svymean_trimmed(x, design, LB = 0.05, UB = 1 - LB, na.rm = FALSE, ...)
svytotal_trimmed(x, design, LB = 0.05, UB = 1 - LB, na.rm = FALSE, ...)

```

## Arguments

x	a one-sided [formula], e.g., ~myVariable.
design	an object of class survey.design; see <a href="#">svydesign</a> .
LB	[double] lower bound of trimming such that $0 \leq LB < UB \leq 1$ .
UB	[double] upper bound of trimming such that $0 \leq LB < UB \leq 1$ .
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
...	additional arguments (currently not used).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

**Characteristic.** Population mean or total. Let  $\mu$  denote the estimated trimmed population mean; then, the estimated trimmed total is given by  $\hat{N}\mu$  with  $\hat{N} = \sum w_i$ , where summation is over all observations in the sample.



**Trimming.** The methods trims the  $LB \cdot 100\%$  of the smallest observations and the  $(1 - UB) \cdot 100\%$  of the largest observations from the data.

**Variance estimation.** Large-sample approximation based on the influence function; see Huber and Ronchetti (2009, Chap. 3.3) and Shao (1994).

**Utility functions.** [summary](#), [coef](#), [SE](#), [vcov](#), [residuals](#), [fitted](#), [robweights](#).

**Bare-bone functions.** See [weighted\\_mean\\_trimmed](#) and [weighted\\_total\\_trimmed](#).

## Value

Object of class [svystat\\_rob](#)

## References

Huber, P. J. and Ronchetti, E. (2009). *Robust Statistics*, New York: John Wiley and Sons, 2nd edition. [doi:10.1002/9780470434697](#)

Shao, J. (1994). L-Statistics in Complex Survey Problems. *The Annals of Statistics* **22**, 976–967. [doi:10.1214/aos/1176325505](#)

## See Also

[Overview](#) (of all implemented functions)

[weighted\\_mean\\_trimmed](#) and [weighted\\_total\\_trimmed](#)

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
            data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
            data = workplace)
}

# Estimated trimmed population total (5% symmetric trimming)
svytotal_trimmed(~employment, dn, LB = 0.05, UB = 0.95)

# Estimated trimmed population mean (5% trimming at the top of the distr.)
svymean_trimmed(~employment, dn, UB = 0.95)
```

---

svymean\_winsorized      *Weighted Winsorized Mean and Total*


---

## Description

Weighted winsorized mean and total

## Usage

```
svymean_winsorized(x, design, LB = 0.05, UB = 1 - LB, na.rm = FALSE,
                  trim_var = FALSE, ...)
svymean_k_winsorized(x, design, k, na.rm = FALSE, trim_var = FALSE, ...)
svytotal_winsorized(x, design, LB = 0.05, UB = 1 - LB, na.rm = FALSE,
                  trim_var = FALSE, ...)
svytotal_k_winsorized(x, design, k, na.rm = FALSE, trim_var = FALSE, ...)
```

## Arguments

x	a one-sided [formula], e.g., ~myVariable.
design	an object of class survey.design; see <a href="#">svydesign</a> .
LB	[double] lower bound of winsorization such that $0 \leq LB < UB \leq 1$ .
UB	[double] upper bound of winsorization such that $0 \leq LB < UB \leq 1$ .
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
trim_var	[logical] indicating whether the variance should be approximated by the variance estimator of the trimmed mean/ total (default: FALSE).
k	[integer] number of observations to be winsorized at the top of the distribution.
...	additional arguments (currently not used).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

**Characteristic.** Population mean or total. Let  $\mu$  denote the estimated winsorized population mean; then, the estimated winsorized total is given by  $\hat{N}\mu$  with  $\hat{N} = \sum w_i$ , where summation is over all observations in the sample.

**Modes of winsorization.** The amount of winsorization can be specified in relative or absolute terms:

- *Relative:* By specifying LB and UB, the method winsorizes the LB · 100% of the smallest observations and the (1 - UB) · 100% of the largest observations from the data.

- *Absolute*: By specifying argument  $k$  in the functions with the "infix" `_k_` in their name (e.g., `svymean_k_winsorized`), the largest  $k$  observations are winsorized,  $0 < k < n$ , where  $n$  denotes the sample size. E.g.,  $k = 2$  implies that the largest and the second largest observation are winsorized.

**Variance estimation.** Large-sample approximation based on the influence function; see Huber and Ronchetti (2009, Chap. 3.3) and Shao (1994). Two estimators are available:

`simple_var = FALSE` Variance estimator of the winsorized mean/ total. The estimator depends on the estimated probability density function evaluated at the winsorization thresholds, which can be – depending on the context – numerically unstable. As a remedy, a simplified variance estimator is available by setting `simple_var = TRUE`.

`simple_var = TRUE` Variance is approximated using the variance estimator of the trimmed mean/ total.

**Utility functions.** `summary`, `coef`, `SE`, `vcov`, `residuals`, `fitted` and `robweights`.

**Bare-bone functions.** See:

- `weighted_mean_winsorized`,
- `weighted_mean_k_winsorized`,
- `weighted_total_winsorized`,
- `weighted_total_k_winsorized`.

## Value

Object of class `svyestat_rob`

## References

Huber, P. J. and Ronchetti, E. (2009). *Robust Statistics*, New York: John Wiley and Sons, 2nd edition. doi:[10.1002/9780470434697](https://doi.org/10.1002/9780470434697)

Shao, J. (1994). L-Statistics in Complex Survey Problems. *The Annals of Statistics* **22**, 976–967. doi:[10.1214/aos/1176325505](https://doi.org/10.1214/aos/1176325505)

## See Also

[Overview](#) (of all implemented functions)

[weighted\\_mean\\_winsorized](#), [weighted\\_mean\\_k\\_winsorized](#), [weighted\\_total\\_winsorized](#) and [weighted\\_total\\_k\\_winsorized](#)

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
```

```

# legacy mode
svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
          data = workplace)
}

# Estimated winsorized population mean (5% symmetric winsorization)
svymean_winsorized(~employment, dn, LB = 0.05)

# Estimated one-sided k winsorized population total (2 observations are
# winsorized at the top of the distribution)
svytotal_k_winsorized(~employment, dn, k = 2)

```

---

svyratio\_huber

*Robust Survey Ratio M-Estimator*


---

## Description

svyratio\_huber and svyratio\_tukey compute the robust  $M$ -estimator of the ratio of two variables with, respectively, Huber and Tukey biweight (bisquare) psi-function.

## Usage

```

svyratio_huber(numerator, denominator, design, k, var = denominator,
               na.rm = FALSE, asym = FALSE, verbose = TRUE, ...)
svyratio_tukey(numerator, denominator, design, k, var = denominator,
               na.rm = FALSE, verbose = TRUE, ...)

```

## Arguments

numerator	a one-sided [formula] object (i.e., symbolic description, e.g., ~payroll).
denominator	a one-sided [formula] object (i.e., symbolic description, e.g., ~employment).
design	an object of class survey.design; see <a href="#">svydesign</a> .
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
var	a [formula] object that defines the heteroscedastic variance (default: numerator).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
asym	[logical] toggle for asymmetric Huber psi-function (default: FALSE).
verbose	[logical] indicating whether additional information is printed to the console (default: TRUE).
...	additional arguments passed to the method (e.g., maxit: maxit number of iterations, etc.).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

The functions `svyratio_huber` and `svyratio_tukey` are implemented as wrapper functions of the regression estimators `svyreg_huberM` and `svyreg_tukeyM`. See the help files of these functions (e.g., on how additional parameters can be passed via `...` or on the usage of the `var` argument).

## Value

Object of class `svyreg.rob` and `ratio`

## See Also

[Overview](#) (of all implemented functions)

[summary](#), [coef](#), [residuals](#), [fitted](#), [SE](#) and [vcov](#)

[plot](#) for regression diagnostic plot methods

[svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#) for robust regression estimators

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Compute regression M-estimate with Huber psi-function
m <- svyratio_huber(~payroll, ~employment, dn, k = 8)

# Regression inference
summary(m)

# Extract the coefficients
coef(m)

# Extract estimated standard error
SE(m)

# Extract variance/ covariance matrix
vcov(m)
```

```
# Diagnostic plots (e.g., standardized residuals against fitted values)
plot(m, which = 1L)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(m), robweights(m))
```

svyreg

*Survey Regression Estimator – Weighted Least Squares*

## Description

Weighted least squares estimator of regression

## Usage

```
svyreg(formula, design, var = NULL, na.rm = FALSE, ...)
```

## Arguments

formula	a [formula] object (i.e., symbolic description of the model)
design	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
var	a one-sided [formula] object or variable name ([character]) that defines the heteroscedastic variance or [NULL] indicating homoscedastic variance (default: NULL).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
...	additional arguments (currently not used).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

svyreg computes the regression coefficients by weighted least squares.

Models for `svyreg_rob` are specified symbolically. A typical model has the form `response ~ terms` where `response` is the (numeric) response vector and `terms` is a series of terms which specifies a linear predictor for response; see [formula](#) and [lm](#).

A formula has an implied intercept term. To remove this use either `y ~ x - 1` or `y ~ 0 + x`; see [formula](#) for more details of allowed formulae.

## Value

Object of class [svyreg\\_rob](#).

**See Also**

[Overview](#) (of all implemented functions)

[summary](#), [coef](#), [residuals](#), [fitted](#), [SE](#) and [vcov](#)

[plot](#) for regression diagnostic plot methods

Robust estimating methods [svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#).

**Examples**

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Compute the regression estimate (weighted least squares)
m <- svyreg(payroll ~ employment, dn)

# Regression inference
summary(m)

# Extract the coefficients
coef(m)

# Extract variance/ covariance matrix
vcov(m)

# Diagnostic plots (e.g., Normal Q-Q-plot)
plot(m, which = 2L)
```

---

svyreg\_huber-deprecated

*Deprecated Huber Robust Survey Regression M-Estimator*

---

**Description**

The function `svyreg_huber` is **deprecated**; use instead [svyreg\\_huberM](#).

**Usage**

```
svyreg_huber(formula, design, k, var = NULL, na.rm = FALSE, asym = FALSE,
  verbose = TRUE, ...)
```

**Arguments**

formula	a [formula] object (i.e., symbolic description of the model)
design	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
var	a one-sided [formula] object or variable name ([character]) that defines the heteroscedastic variance or [NULL] indicating homoscedastic variance (default: NULL).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
asym	[logical] toggle for asymmetric Huber psi-function (default: FALSE).
verbose	[logical] indicating whether additional information is printed to the console (default: TRUE).
...	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.).

**Details**

See [svyreg\\_huberM](#).

**Value**

Object of class `svyreg.rob`

**See Also**

[robsurvey-deprecated](#)

---

svyreg\_huberM

*Huber Robust Survey Regression M- and GM-Estimator*


---

**Description**

`svyreg_huberM` and `svyreg_huberGM` compute, respectively, a survey weighted *M*- and *GM*-estimator of regression using the Huber psi-function.

**Usage**

```
svyreg_huberM(formula, design, k, var = NULL, na.rm = FALSE, asym = FALSE,
               verbose = TRUE, ...)
svyreg_huberGM(formula, design, k, type = c("Mallows", "Schweppe"),
               xwgt, var = NULL, na.rm = FALSE, asym = FALSE, verbose = TRUE,
               ...)
```



## Arguments

formula	a [formula] object (i.e., symbolic description of the model)
design	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
var	a one-sided [formula] object or variable name ([character]) that defines the heteroscedastic variance or [NULL] indicating homoscedastic variance (default: NULL).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
asym	[logical] toggle for asymmetric Huber psi-function (default: FALSE).
verbose	[logical] indicating whether additional information is printed to the console (default: TRUE).
type	[character] "Mallows" or "Schweppe".
xwgt	[numerical vector] or [NULL] of weights in the design space (default: NULL); <code>xwgt</code> is only relevant for <code>type = "Mallows"</code> or <code>type = "Schweppe"</code> .
...	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.).

## Details

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

`svyreg_huberM` and `svyreg_huberGM` compute, respectively, *M*- and *GM*-estimates of regression by iteratively re-weighted least squares (IRWLS). The estimate of regression scale is (by default) computed as the (normalized) weighted median of absolute deviations from the weighted median (MAD; see [weighted\\_mad](#)) for each IRWLS iteration. If the weighted MAD is zero (or nearly so), the scale is computed as the (normalized) weighted interquartile range (IQR).

**M-estimator** The regression M-estimator `svyreg_huberM` is robust against residual outliers (granted that the tuning constant `k` is chosen appropriately).

**GM-estimator** Function `svyreg_huberGM` implements the Mallows and Schweppe regression GM-estimator (see argument `type`). The regression GM-estimators are robust against residual outliers *and* outliers in the model's design space (leverage observations; see argument `xwgt`).

**Numerical optimization** See [svyreg\\_control](#).

**Models** Models for `svyreg_rob` are specified symbolically. A typical model has the form `response ~ terms`, where `response` is the (numeric) response vector and `terms` is a series of terms which specifies a linear predictor for response; see [formula](#) and [lm](#).

A formula has an implied intercept term. To remove this use either `y ~ x - 1` or `y ~ 0 + x`; see [formula](#) for more details of allowed formulae.

## Value

Object of class `svyreg.rob`

### Failure of convergence

By default, the method assumes a maximum number of `maxit = 100` iterations and a numerical tolerance criterion to stop the iterations of `tol = 1e-05`. If the algorithm fails to converge, you may consider changing the default values; see [svyreg\\_control](#).

### See Also

[Overview](#) (of all implemented functions)

[Overview](#) (of all implemented functions)

[summary](#), [coef](#), [residuals](#), [fitted](#), [SE](#) and [vcov](#)

[plot](#) for regression diagnostic plot methods

Other robust estimating methods [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#)

### Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Compute regression M-estimate with Huber psi-function
m <- svyreg_huberM(payload ~ employment, dn, k = 8)

# Regression inference
summary(m)

# Extract the coefficients
coef(m)

# Extract variance/ covariance matrix
vcov(m)

# Diagnostic plots (e.g., standardized residuals against fitted values)
plot(m, which = 1L)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(m), robweights(m))
```

---

`svyreg_tukey-deprecated`*Deprecated Tukey Biweight Robust Survey Regression M-Estimator*

---

## Description

The function `svyreg_tukey` is **deprecated**; use instead [svyreg\\_tukeyM](#).

## Usage

```
svyreg_tukey(formula, design, k, var = NULL, na.rm = FALSE, verbose = TRUE,
             ...)
```

## Arguments

<code>formula</code>	a [formula] object (i.e., symbolic description of the model)
<code>design</code>	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
<code>k</code>	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
<code>var</code>	a one-sided [formula] object or variable name ([character]) that defines the heteroscedastic variance or [NULL] indicating homoscedastic variance (default: NULL).
<code>na.rm</code>	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
<code>verbose</code>	[logical] indicating whether additional information is printed to the console (default: TRUE).
<code>...</code>	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.).

## Details

See [svyreg\\_tukeyM](#).

## Value

Object of class `svyreg.rob`

## See Also

[robsurvey-deprecated](#)

svyreg\_tukeyM

*Tukey Biweight Robust Survey Regression M- and GM-Estimator***Description**

svyreg\_tukeyM and svyreg\_tukeyGM compute, respectively, a survey weighted *M*- and *GM*-estimator of regression using the biweight Tukey psi-function.

**Usage**

```
svyreg_tukeyM(formula, design, k, var = NULL, na.rm = FALSE, verbose = TRUE,
              ...)
svyreg_tukeyGM(formula, design, k, type = c("Mallows", "Schweppe"),
               xwgt, var = NULL, na.rm = FALSE, verbose = TRUE, ...)
```

**Arguments**

formula	a [formula] object (i.e., symbolic description of the model)
design	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
var	a one-sided [formula] object or variable name ([character]) that defines the heteroscedastic variance or [NULL] indicating homoscedastic variance (default: NULL).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
verbose	[logical] indicating whether additional information is printed to the console (default: TRUE).
type	[character] "Mallows" or "Schweppe".
xwgt	[numerical vector] or [NULL] of weights in the design space (default: NULL); xwgt is only relevant for type = "Mallows" or type = "Schweppe".
...	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.).

**Details**

Package **survey** must be attached to the search path in order to use the functions (see [library](#) or [require](#)).

svyreg\_tukeyM and svyreg\_tukeyGM compute, respectively, M- and GM-estimates of regression by iteratively re-weighted least squares (IRWLS). The estimate of regression scale is (by default) computed as the (normalized) weighted median of absolute deviations from the weighted median (MAD; see [weighted\\_mad](#)) for each IRWLS iteration. If the weighted MAD is zero (or nearly so), the scale is computed as the (normalized) weighted interquartile range (IQR).

**M-estimator** The regression M-estimator svyreg\_tukeyM is robust against residual outliers (granted that the tuning constant *k* is chosen appropriately).

**GM-estimator** Function `svyreg_huberGM` implements the Mallows and Schweppe regression GM-estimator (see argument `type`). The regression GM-estimators are robust against residual outliers *and* outliers in the model's design space (leverage observations; see argument `xwgt`).

**Numerical optimization** See [svyreg\\_control](#).

**Models** Models for `svyreg_rob` are specified symbolically. A typical model has the form `response ~ terms`, where `response` is the (numeric) response vector and `terms` is a series of terms which specifies a linear predictor for response; see [formula](#) and [lm](#).

A formula has an implied intercept term. To remove this use either `y ~ x - 1` or `y ~ 0 + x`; see [formula](#) for more details of allowed formulae.

## Value

Object of class `svyreg.rob`

## Failure of convergence

By default, the method assumes a maximum number of `maxit = 100` iterations and a numerical tolerance criterion to stop the iterations of `tol = 1e-05`. If the algorithm fails to converge, you may consider changing the default values; see [svyreg\\_control](#).

## See Also

[Overview](#) (of all implemented functions)

[summary](#), [coef](#), [residuals](#), [fitted](#), [SE](#) and [vcov](#)

[plot](#) for regression diagnostic plot methods.

Other robust estimating methods [svyreg\\_huberM](#) and [svyreg\\_huberGM](#)

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}

# Compute regression M-estimate with Tukey bisquare psi-function
m <- svyreg_tukeyM(payroll ~ employment, dn, k = 8)

# Regression inference
summary(m)
```

```
# Extract the coefficients
coef(m)

# Extract variance/ covariance matrix
vcov(m)

# Diagnostic plots (e.g., standardized residuals against fitted values)
plot(m, which = 1L)

# Plot of the robustness weights of the M-estimate against its residuals
plot(residuals(m), robweights(m))
```

---

svysummary

*Weighted Five-Number Summary of a Variable*


---

## Description

Weighted five-number summary used for `survey.design` and `survey.design2` objects (similar to `base::summary` for [numeric vectors]).

## Usage

```
svysummary(object, design, na.rm = FALSE, ...)
```

## Arguments

<code>object</code>	one-sided [formula] for which a summary is desired, e.g., <code>~payroll</code> .
<code>design</code>	an object of class <code>survey.design</code> ; see <a href="#">svydesign</a> .
<code>na.rm</code>	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
<code>...</code>	additional arguments.

## Value

A weighted five-number summary (numeric variable) or a frequency table (factor variable).

## Examples

```
head(workplace)

library(survey)
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
```

```

        svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
                  data = workplace)
    }

    svysummary(~payroll, dn)

```

---

weighted-m-estimator    *Weighted Huber and Tukey Mean and Total (bare-bone functions)*

---

## Description

Weighted Huber and Tukey  $M$ -estimator of the mean and total (bare-bone function with limited functionality; see [svymean\\_huber](#), [svymean\\_tukey](#), [svytotal\\_huber](#), and [svytotal\\_tukey](#) for more capable methods)

## Usage

```

weighted_mean_huber(x, w, k, type = "rwm", asym = FALSE, info = FALSE,
                    na.rm = FALSE, verbose = TRUE, ...)
weighted_total_huber(x, w, k, type = "rwm", asym = FALSE, info = FALSE,
                     na.rm = FALSE, verbose = TRUE, ...)
weighted_mean_tukey(x, w, k, type = "rwm", info = FALSE, na.rm = FALSE,
                    verbose = TRUE, ...)
weighted_total_tukey(x, w, k, type = "rwm", info = FALSE, na.rm = FALSE,
                     verbose = TRUE, ...)

```

## Arguments

<code>x</code>	[numeric vector] data.
<code>w</code>	[numeric vector] weights (same length as <code>x</code> ).
<code>k</code>	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
<code>type</code>	[character] type of method: "rwm" or "rht"; see below (default: "rwm").
<code>asym</code>	[logical] toggle for asymmetric Huber psi-function (default: FALSE).
<code>info</code>	[logical] indicating whether additional information should be returned (default: FALSE).
<code>na.rm</code>	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
<code>verbose</code>	[logical] indicating whether additional information is printed to the console (default: TRUE).
<code>...</code>	additional arguments passed to the method (e.g., <code>maxit</code> : maxit number of iterations, etc.).

## Details

**Characteristic.** Population mean or total. Let  $\mu$  denote the estimated population mean; then, the estimated total is given by  $\hat{N}\mu$  with  $\hat{N} = \sum w_i$ , where summation is over all observations in the sample.

**Type.** Two methods/types are available for estimating the location  $\mu$ :

type = "rwm" (default): robust weighted  $M$ -estimator of the population mean and total, respectively. This estimator is recommended for sampling designs whose inclusion probabilities are *not* proportional to some measure of size. [Legacy note: In an earlier version, the method type = "rwm" was called "rhj"; the type "rhj" is now silently converted to "rwm"]

type = "rht": robust Horvitz-Thompson  $M$ -estimator of the population mean and total, respectively. This estimator is recommended for proportional-to-size sampling designs.

**Variance estimation.** See the related but more capable functions:

- [svymean\\_huber](#) and [svymean\\_tukey](#),
- [svytotal\\_huber](#) and [svytotal\\_tukey](#).

**Psi-function.** By default, the Huber or Tukey psi-function are used in the specification of the  $M$ -estimators. For the Huber estimator, an asymmetric version of the Huber psi-function can be used by setting the argument `asym = TRUE` in the function call.

## Value

The return value depends on `info`:

`info = FALSE`: estimate of mean or total [double]

`info = TRUE`: a [list] with items:

- `characteristic` [character],
- `estimator` [character],
- `estimate` [double],
- `variance` (default: NA),
- `robust` [list],
- `residuals` [numeric vector],
- `model` [list],
- `design` (default: NA),
- [call]

## Failure of convergence

By default, the method assumes a maximum number of `maxit = 100` iterations and a numerical tolerance criterion to stop the iterations of `tol = 1e-05`. If the algorithm fails to converge, you may consider changing the default values; see [svyreg\\_control](#).

## References

Hulliger, B. (1995). Outlier Robust Horvitz-Thompson Estimators. *Survey Methodology* **21**, 79–87.



**See Also**

[Overview](#) (of all implemented functions)

**Examples**

```
head(workplace)

# Robust Horvitz-Thompson M-estimator of the population total
weighted_total_huber(workplace$employment, workplace$weight, k = 9,
  type = "rht")

# Robust weighted M-estimator of the population mean
weighted_mean_huber(workplace$employment, workplace$weight, k = 12,
  type = "rwm")
```

---

weighted_IQR	<i>Weighted Interquartile Range (IQR)</i>
--------------	---

---

**Description**

Weighted (normalized) interquartile range

**Usage**

```
weighted_IQR(x, w, na.rm = FALSE, constant = 0.7413)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
constant	[double] constant scaling factor to make the weighted IQR a consistent estimator of the scale (default: 0.7413).

**Details**

By default, the weighted IQR is normalized to be an unbiased estimate of scale at the Gaussian core model. If normalization is not wanted, put `constant = 1`.

**Value**

Weighted IQR

**See Also**

[Overview](#) (of all implemented functions)

**Examples**

```
head(workplace)

# normalized weighted IQR (default)
weighted_IQR(workplace$employment, workplace$weight)

# weighted IQR (without normalization)
weighted_IQR(workplace$employment, workplace$weight, constant = 1)
```

weighted\_line

*Weighted Robust Line Fitting***Description**

weighted\_line fits a robust line and allows weights.

**Usage**

```
weighted_line(x, y = NULL, w, na.rm = FALSE, iter = 1)

## S3 method for class 'medline'
print(x, ...)
## S3 method for class 'medline'
coef(object, ...)
## S3 method for class 'medline'
residuals(object, ...)
## S3 method for class 'medline'
fitted(object, ...)
```

**Arguments**

x	[numeric vector] explanatory variable.
y	[numeric vector] response variable (default: NULL).
w	[numeric vector] weights (same length as x).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
iter	[integer] number of iterations to use (default: 1).
object	object of class medline.
...	additional arguments passed to the method.

**Details**

weighted\_line uses different quantiles for splitting the sample than stats::line().

**Value**

intercept and slope of the fitted line

**See Also**

[Overview](#) (of all implemented functions)  
[line](#)

**Examples**

```
head(cars)

# compute weighted line
weighted_line(cars$speed, cars$dist, w = rep(1, length(cars$speed)))
m <- weighted_line(cars$speed, cars$dist, w = rep(1:10, each = 5))
m
coef(m)
residuals(m)
fitted(m)
```

weighted\_mad

*Weighted Median Absolute Deviation from the Median (MAD)***Description**

Weighted median of the absolute deviations from the weighted median

**Usage**

```
weighted_mad(x, w, na.rm = FALSE, constant = 1.482602)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
constant	[double] constant scaling factor to make the MAD a consistent estimator of the scale (default: 1.4826).

**Details**

The weighted MAD is computed as the (normalized) weighted median of the absolute deviation from the weighted median; see [weighted\\_median](#). The weighted MAD is normalized to be an unbiased estimate of scale at the Gaussian core model. If normalization is not wanted, put constant = 1.

**Value**

Weighted median absolute deviation from the (weighted) median

**See Also**

[Overview](#) (of all implemented functions)

**Examples**

```
head(workplace)

# normalized weighted MAD (default)
weighted_mad(workplace$employment, workplace$weight)

# weighted MAD (without normalization)
weighted_mad(workplace$employment, workplace$weight, constant = 1)
```

---

weighted_mean	<i>Weighted Total and Mean (Horvitz-Thompson and Hajek Estimators)</i>
---------------	--

---

**Description**

Weighted total and mean (Horvitz-Thompson and Hajek estimators)

**Usage**

```
weighted_mean(x, w, na.rm = FALSE)
weighted_total(x, w, na.rm = FALSE)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

**Details**

weighted\_total and weighted\_mean compute, respectively, the Horvitz-Thompson estimator of the population total and the Hajek estimator of the population mean.

**Value**

Estimated population mean or total

**See Also**

[Overview](#) (of all implemented functions)

## Examples

```
head(workplace)

# Horvitz-Thompson estimator of the total
weighted_total(workplace$employment, workplace$weight)

# Hajek estimator of the mean
weighted_mean(workplace$employment, workplace$weight)
```

---

weighted\_mean\_dalen      *Dalen Estimators of the Mean and Total*

---

## Description

Dalén's estimators of the population mean and the population total (bare-bone functions with limited functionality).

## Usage

```
weighted_mean_dalen(x, w, censoring, type = "Z2", info = FALSE,
                    na.rm = FALSE, verbose = TRUE)
weighted_total_dalen(x, w, censoring, type = "Z2", info = FALSE,
                     na.rm = FALSE, verbose = TRUE)
```

## Arguments

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
censoring	[double] cutoff threshold above which the observations are censored.
type	[character] type of estimator; either "Z2" or "Z3" (default: "Z2").
info	[logical] indicating whether additional information should be returned (default: FALSE).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
verbose	[logical] indicating whether additional information should be printed to the console (default: FALSE).

## Details

Let  $\sum_{i \in s} w_i x_i$  denote the expansion estimator of the  $x$ -total (summation is over all elements  $i$  in sample  $s$ ). The estimators Z2 and Z3 of Dalén (1987) are defined as follows.

**Estimator Z2** The estimator Z2 of the population total sums over  $\min(c, w_i x_i)$ ; hence, it censors the products  $w_i x_i$  to the censoring constant  $c$  (censoring). The estimator of the population  $x$ -mean is defined as the total divided by the population size.

**Estimator Z3** The estimator Z3 of the population total is defined as the sum over the elements  $z_i$ , which is equal to  $z_i = w_i x_i$  if  $w_i y_i \leq c$  and  $z_i = c + (y_i - c/w_i)$  otherwise.

**Value**

The return value depends on info:

info = FALSE: estimate of mean or total [double]

info = TRUE: a [list] with items:

- characteristic [character],
- estimator [character],
- estimate [double],
- variance (default: NA),
- robust [list],
- residuals [numeric vector],
- model [list],
- design (default: NA),
- [call]

**References**

Dalén, J. (1987). Practical Estimators of a Population Total Which Reduce the Impact of Large Observations. R & D Report U/STM 1987:32, Statistics Sweden, Stockholm.

**See Also**

[Overview](#) (of all implemented functions)

**Examples**

```
head(workplace)

# Dalen's estimator of the total (with censoring threshold: 100000)
weighted_total_dalen(workplace$employment, workplace$weight, 100000)
```

---

weighted\_mean\_trimmed    *Weighted Trimmed Mean and Total (bare-bone functions)*

---

**Description**

Weighted trimmed mean and total (bare-bone functions with limited functionality; see [svymean\\_trimmed](#) and [svytotal\\_trimmed](#) for more capable methods)

**Usage**

```
weighted_mean_trimmed(x, w, LB = 0.05, UB = 1 - LB, info = FALSE,
                      na.rm = FALSE)
weighted_total_trimmed(x, w, LB = 0.05, UB = 1 - LB, info = FALSE,
                      na.rm = FALSE)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
LB	[double] lower bound of trimming such that $0 \leq LB < UB \leq 1$ .
UB	[double] upper bound of trimming such that $0 \leq LB < UB \leq 1$ .
info	[logical] indicating whether additional information should be returned (default: FALSE).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

**Details**

**Characteristic.** Population mean or total. Let  $\mu$  denote the estimated trimmed population mean; then, the estimated trimmed population total is given by  $\hat{N}\mu$  with  $\hat{N} = \sum w_i$ , where summation is over all observations in the sample.

**Trimming.** The methods trims the  $LB \cdot 100\%$  of the smallest observations and the  $(1 - UB) \cdot 100\%$  of the largest observations from the data.

**Variance estimation.** See survey methods:

- [svymean\\_trimmed](#),
- [svytotal\\_trimmed](#).

**Value**

The return value depends on info:

info = FALSE: estimate of mean or total [double]

info = TRUE: a [list] with items:

- characteristic [character],
- estimator [character],
- estimate [double],
- variance (default: NA),
- robust [list],
- residuals [numeric vector],
- model [list],
- design (default: NA),
- [call]

**See Also**

[Overview](#) (of all implemented functions)

[svymean\\_trimmed](#) and [svytotal\\_trimmed](#)

**Examples**

```
head(workplace)

# Estimated trimmed population total (5% symmetric trimming)
weighted_total_trimmed(workplace$employment, workplace$weight, LB = 0.05,
  UB = 0.95)

# Estimated trimmed population mean (5% trimming at the top of the distr.)
weighted_mean_trimmed(workplace$employment, workplace$weight, UB = 0.95)
```

---

```
weighted_mean_winsorized
```

*Weighted Winsorized Mean and Total (bare-bone functions)*

---

**Description**

Weighted winsorized mean and total (bare-bone functions with limited functionality; see [svymean\\_winsorized](#) and [svytotal\\_winsorized](#) for more capable methods)

**Usage**

```
weighted_mean_winsorized(x, w, LB = 0.05, UB = 1 - LB, info = FALSE,
  na.rm = FALSE)
weighted_mean_k_winsorized(x, w, k, info = FALSE, na.rm = FALSE)
weighted_total_winsorized(x, w, LB = 0.05, UB = 1 - LB, info = FALSE,
  na.rm = FALSE)
weighted_total_k_winsorized(x, w, k, info = FALSE, na.rm = FALSE)
```

**Arguments**

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
LB	[double] lower bound of winsorization such that $0 \leq LB < UB \leq 1$ .
UB	[double] upper bound of winsorization such that $0 \leq LB < UB \leq 1$ .
info	[logical] indicating whether additional information should be returned (default: FALSE).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).
k	[integer] number of observations to be winsorized at the top of the distribution.



## Details

**Characteristic.** Population mean or total. Let  $\mu$  denote the estimated winsorized population mean; then, the estimated population total is given by  $\hat{N}\mu$  with  $\hat{N} = \sum w_i$ , where summation is over all observations in the sample.

**Modes of winsorization.** The amount of winsorization can be specified in relative or absolute terms:

- *Relative:* By specifying LB and UB, the methods winsorizes the LB · 100% of the smallest observations and the (1 - UB) · 100% of the largest observations from the data.
- *Absolute:* By specifying argument k in the functions with the "infix" \_k\_ in their name, the largest k observations are winsorized,  $0 < k < n$ , where n denotes the sample size. E.g., k = 2 implies that the largest and the second largest observation are winsorized.

**Variance estimation.** See survey methods:

- [svymean\\_winsorized](#),
- [svytotal\\_winsorized](#),
- [svymean\\_k\\_winsorized](#),
- [svytotal\\_k\\_winsorized](#).

## Value

The return value depends on info:

info = FALSE: estimate of mean or total [double]

info = TRUE: a [list] with items:

- characteristic [character],
- estimator [character],
- estimate [double],
- variance (default: NA),
- robust [list],
- residuals [numeric vector],
- model [list],
- design (default: NA),
- [call]

## See Also

[Overview](#) (of all implemented functions)

[svymean\\_winsorized](#), [svymean\\_k\\_winsorized](#), [svytotal\\_winsorized](#) and [svytotal\\_k\\_winsorized](#)

## Examples

```
head(workplace)
```

```
# Estimated winsorized population mean (5% symmetric winsorization)
weighted_mean_winsorized(workplace$employment, workplace$weight, LB = 0.05)
```

```
# Estimated one-sided k winsorized population total (2 observations are
# winsorized at the top of the distribution)
weighted_total_k_winsorized(workplace$employment, workplace$weight, k = 2)
```

---

weighted_median	<i>Weighted Median</i>
-----------------	------------------------

---

## Description

Weighted population median.

## Usage

```
weighted_median(x, w, na.rm = FALSE)
```

## Arguments

x	[numeric vector] data.
w	[numeric vector] weights (same length as x).
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

## Details

Weighted sample median; see [weighted\\_quantile](#) for more information.

## Value

Weighted estimate of the population median

## See Also

[Overview](#) (of all implemented functions)  
[weighted\\_quantile](#)

## Examples

```
head(workplace)

weighted_median(workplace$employment, workplace$weight)
```

---

weighted\_median\_line     *Robust Simple Linear Regression Based on Medians*

---

## Description

Robust simple linear regression based on medians: two methods are available: "slopes" and "product".

## Usage

```
weighted_median_line(x, y = NULL, w, type = "slopes", na.rm = FALSE)
```

## Arguments

x	[numeric vector] explanatory variable.
y	[numeric vector] response variable (default: NULL).
w	[numeric vector] weights (same length as x).
type	[character] "slopes" or "products" (default: "slopes").
na.rm	[logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

## Details

**Overview.** Robust simple linear regression based on medians

**Type.** Two methods/ types are available. Let  $m(x, w)$  denote the weighted median of variable x with weights w:

type = "slopes": The slope is computed as

$$b1 = m\left(\frac{y - m(y, w)}{x - m(x, w)}, w\right).$$

type = "products": The slope is computed as

$$b1 = \frac{m([y - m(y, w)][x - m(x, w)], w)}{m([x - m(x, w)]^2, w)}.$$

## Value

A vector with two components: intercept and slope

## See Also

[Overview](#) (of all implemented functions)

[line](#), [weighted\\_line](#) and [weighted\\_median\\_ratio](#)

**Examples**

```

x <- c(1, 2, 4, 5)
y <- c(3, 2, 7, 4)
weighted_line(y ~ x, w = rep(1, length(x)))
weighted_median_line(y ~ x, w = rep(1, length(x)))
m <- weighted_median_line(y ~ x, w = rep(1, length(x)), type = "prod")
m
coef(m)
fitted(m)
residuals(m)

# cars data
head(cars)
with(cars, weighted_median_line(dist ~ speed, w = rep(1, length(dist))))
with(cars, weighted_median_line(dist ~ speed, w = rep(1, length(dist)),
type = "prod"))

# weighted
w <- c(rep(1,20), rep(2,20), rep(5, 10))
with(cars, weighted_median_line(dist ~ speed, w = w))
with(cars, weighted_median_line(dist ~ speed, w = w, type = "prod"))

# outlier in y
cars$dist[49] <- 360
with(cars, weighted_median_line(dist ~ speed, w = w))
with(cars, weighted_median_line(dist ~ speed, w = w, type = "prod"))

# outlier in x
data(cars)
cars$speed[49] <- 72
with(cars, weighted_median_line(dist ~ speed, w = w))
with(cars, weighted_median_line(dist ~ speed, w = w, type = "prod"))

```

---

weighted\_median\_ratio *Weighted Robust Ratio Estimator Based on Median*

---

**Description**

A weighted median of the ratios  $y/x$  determines the slope of a regression through the origin.

**Usage**

```
weighted_median_ratio(x, y = NULL, w, na.rm = FALSE)
```

**Arguments**

x	[numeric vector] explanatory variable.
y	[numeric vector] response variable (default: NULL).
w	[numeric vector] weights (same length as x).

na.rm                    [logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

**Value**

A vector with two components: intercept and slope

**See Also**

[Overview](#) (of all implemented functions)  
[line](#), [weighted\\_line](#) and [weighted\\_median\\_line](#)

**Examples**

```
x <- c(1,2,4,5)
y <- c(1,0,5,2)
m <- weighted_median_ratio(y ~ x, w = rep(1, length(y)))
m
coef(m)
fitted(m)
residuals(m)
```

---

weighted_quantile	<i>Weighted Quantile</i>
-------------------	--------------------------

---

**Description**

Weighted population quantile.

**Usage**

```
weighted_quantile(x, w, probs, na.rm = FALSE)
```

**Arguments**

x                    [numeric vector] data.  
w                    [numeric vector] weights (same length as x).  
probs                [numeric vector] vector of probabilities with values in [0, 1].  
na.rm                [logical] indicating whether NA values should be removed before the computation proceeds (default: FALSE).

## Details

**Overview.** `weighted_quantile` computes the weighted sample quantiles; argument `probs` allows vector inputs.

**Implementation.** The function is based on a weighted version of the quickselect/Find algorithm with the Bentley and McIlroy (1993) 3-way partitioning scheme. For very small arrays, we use insertion sort.

**Compatibility.** For equal weighting, i.e., when all elements in `w` are equal, `weighted_quantile` is identical with `type = 2` of `stats::quantile`; see also Hyndman and Fan (1996).

## Value

Weighted estimate of the population quantiles

## References

Bentley, J. L. and McIlroy, D. M. (1993). Engineering a Sort Function, *Software - Practice and Experience* **23**, 1249–1265. doi:10.1002/spe.4380231105

Hyndman, R.J. and Fan, Y. (1996). Sample Quantiles in Statistical Packages, *The American Statistician* **50**, 361–365. doi:10.1080/00031305.1996.10473566

## See Also

[Overview](#) (of all implemented functions)

[weighted\\_median](#)

## Examples

```
head(workplace)

# Weighted 25% quantile (1st quartile)
weighted_quantile(workplace$employment, workplace$weight, 0.25)
```

---

wgt\_functions

---

*Weight Functions (for the M- and GM-Estimators)*


---

## Description

Weight functions associated with the Huber and the Tukey biweight psi-functions; and the weight function of Simpson et al. (1992) for GM-estimators.

## Usage

```
huberWgt(x, k = 1.345)
tukeyWgt(x, k = 4.685)
simpsonWgt(x, a, b)
```

## Arguments

x	[numeric vector] data.
k	[double] robustness tuning constant ( $0 < k \leq \infty$ ).
a	[double] robustness tuning constant ( $0 \leq a \leq \infty$ ); see details below.
b	[double] robustness tuning constant ( $0 < b \leq \infty$ ; see details below.

## Details

The functions `huberWgt` and `tukeyWgt` return the weights associated with the respective psi-function.

The function `simpsonWgt` is used (in regression GM-estimators) to downweight leverage observations (i.e., outliers in the model's design space). Let  $d_i$  denote the (robust) squared Mahalanobis distance of the  $i$ -th observation. The Simpson et al. (1992) type of weight is defined as  $\min\{1, (b/d_i)^{a/2}\}$ , where  $a$  and  $b$  are tuning constants.

- By default,  $a = 1$ ; this choice implies that the weights are computed on the basis of the robust Mahalanobis distances. Alternative:  $a = \text{Inf}$  implies a weight of zero for all observations whose (robust) squared Mahalanobis is larger than  $b$ .
- The tuning constants  $b$  is a threshold on the distances.

## Value

Numerical vector of weights

## References

Simpson, D. G., Ruppert, D. and Carroll, R.J. (1992). On One-Step GM Estimates and Stability of Inferences in Linear Regression. *Journal of the American Statistical Association* **87**, 439–450.  
doi:[10.2307/2290275](https://doi.org/10.2307/2290275)

## See Also

[Overview](#) (of all implemented functions)

[svyreg\\_huberM](#), [svyreg\\_huberGM](#), [svyreg\\_tukeyM](#) and [svyreg\\_tukeyGM](#)

## Examples

```
head(flour)

# standardized distance from median (copper content in wholemeal flour)
x <- flour$copper
z <- abs(x - median(x)) / mad(x)

# plot of weight functions vs. distance
plot(z, huberWgt(z, k = 3), ylim = c(0, 1), xlab = "distance",
      ylab = "weight")
points(z, tukeyWgt(z, k = 6), pch = 2, col = 2)
points(z, simpsonWgt(z, a = Inf, b = 3), pch = 3, col = 4)
legend("topright", c("huberWgt(k = 3)", "tukeyWgt(k = 6)",
                    "simpsonWgt(a = Inf, b = 3)"), pch = 1:3, col = c(1, 2, 4))
```

---

within_tolerance	<i>Tolerance Interval</i>
------------------	---------------------------

---

### Description

The function flags observations that fall within the tolerance interval. Observations that fall outside the interval are regarded as (potential) outliers.

### Usage

```
within_tolerance(x, w, method = c("quartile", "modified", "boxplot"),
               constants, lambda = 0.05, info = FALSE,
               boxplot_coef = 1.5)
```

### Arguments

x	[numeric vector] data vector.
w	[numeric vector] design weights (same length as x).
method	[character] one of the methods: "quartile", "modified" (quartile method), or "boxplot".
constants	[numeric vector] a vector of size two with nonnegative tuning constants; it is only used by the methods "quartile" and "modified".
lambda	[numeric] a tuning constant that takes values in the closed unit interval; it is only used by method "modified", default: lambda = 0.05.
info	[logical] if TRUE, the tolerance interval is printed out.
boxplot_coef	[numeric] determines how far the whiskers of the boxplot extend out from the box; the default is 1.5.

### Details

Three methods are available.

**Quartile method** ("quartile") For the quartile method, the tolerance interval is given by

$$[m - c_l \cdot L_l, m + c_u \cdot L_u]$$

with

$$L_l = m - q_1 \quad \text{and} \quad L_u = q_3 - m,$$

where  $m$  denotes the (weighted) median;  $q_1$  and  $q_3$  are, respectively, the first and third (weighted) quartiles. The tuning constants  $c_l$  and  $c_u$  are combined into the vector  $(c_l, c_u)$ , which is available as argument constants; both constants must be nonnegative numbers.

The quartiles are calculated using design weights.



**Modified quartile method** ("modified") For the modified quartile method (Lee, 1995), the tolerance interval is given by replacing  $L_l$  and  $L_u$  with, respectively,

$$L_l = \max(m - q_1, |\lambda \cdot m|),$$

and

$$L_u = \max(q_3 - m, |\lambda \cdot m|)$$

The tuning constant  $\lambda$  can only take values in the closed unit interval and is available as argument `lambda`.

The quartiles are calculated using design weights.

**Boxplot (box-and-whisker plot) method** ("boxplot") The tolerance interval for the boxplot method extends from the lower whisker to the upper whisker. By default, the length of the whiskers is set to 1.5 times the interquartile range; see argument `boxplot_coef`. For more details, see [boxplot](#).

The quartiles, and therefore the interquartile range, are calculated using design weights.

## Value

A vector of logicals, where TRUE indicates that an observation is within the tolerance limits and FALSE indicates a (potential) outlier.

If `info = TRUE`, the function prints the tolerance interval. The endpoints of the interval can be numbers or the symbols 'min.' and 'max.', which denote the minimum and maximum values in the data, respectively.

## References

Lee, H. (1995). Outliers in Business Surveys, in: Cox, B. G. et al. (eds.), *Business Survey Methods*, p. 503–526. New York: John Wiley and Sons.

## See Also

[Overview](#) (of all implemented functions)

## Examples

```
head(workplace)
attach(workplace)

# Show the tolerance limits
within_tolerance payroll, weight, method = "boxplot", info = TRUE)

# Observations that fall outside the tolerance limits are (potential) outliers
outlier <- !within_tolerance payroll, weight, method = "boxplot")
outlier[1:10]
```

workplace

*(Modified) Canadian Workplace and Employee Survey***Description**

The workplace data are from Fuller (2009, pp. 366–367).

**Usage**

```
data(workplace)
```

**Format**

A data.frame with a sample of 142 workplaces on the following variables

ID identifier variable [integer].

weight sampling weight [double].

employment employment total [double].

payroll payroll total (1000 dollars)[double].

strat stratum identifier[integer].

fpc finite population correction [integer].

**Details**

The workplace data represent a sample of workplaces in the retail sector in a Canadian province. The data are *not* those collected by Statistics Canada, but have been generated by Fuller (2009, Example 3.1.1) to display similar characteristics to the original 1999 Canadian Workplace and Employee Survey (WES).

**Sampling design of the 1999 WES:** The WES target population is defined as all workplaces operating in Canada with paid employees. The sampling frame is stratified by industry, geographic region, and size (size is defined using estimated employment). A sample of workplaces has been drawn independently in each stratum using simple random sample without replacement (the stratum-specific sample sizes are determined by Neyman allocation). Several strata containing very large workplaces were sampled exhaustively; see Patak et al (1998). The original sampling weights were adjusted for nonresponse.

**Remarks by Fuller (2009, p. 365):** The original weights of WES were about 2200 for the stratum of small workplaces, about 750 for medium-sized, and about 35 for large workspaces.

**Source**

The data workplace is from Table 6.3 in Fuller (2009, pp. 366–367).

## References

- Fuller, W. A. (2009). *Sampling Statistics*, Hoboken (NJ): John Wiley and Sons. doi:10.1002/9780470523551
- Patak, Z., Hidioglou, M. and Lavallée, P. (1998). The methodology of the Workplace and Employee Survey. *Proceedings of the Survey Research Methods Section, American Statistical Association*, 83–91.

## Examples

```
head(workplace)

library("survey")
# Survey design for stratified simple random sampling without replacement
dn <- if (packageVersion("survey") >= "4.2") {
  # survey design with pre-calibrated weights
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace, calibrate.formula = ~-1 + strat)
} else {
  # legacy mode
  svydesign(ids = ~ID, strata = ~strat, fpc = ~fpc, weights = ~weight,
    data = workplace)
}
```

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