

# Package ‘PF’

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**Type** Package

**Title** Prevented fraction

**Version** 9.6.10

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**Description** Functions related to PF (prevented fraction). Calculate incidence density ratio, confidence interval, and Rao-Scott weights of PF by multiple methods. See <http://goo.gl/eJ6Rxi> for definition of PF.  
No endorsement, claim, or warranty is implied for this package. It is made available for investigational or pedagogical use only. See [https://www.aphis.usda.gov/animal\\_health/vet\\_biologicals/publications/STATWI0007.pdf](https://www.aphis.usda.gov/animal_health/vet_biologicals/publications/STATWI0007.pdf) for futher details.

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**URL** <https://github.com/ABS-dev/PF/>, <https://abs-dev.github.io/PF/>

**BugReports** <https://github.com/ABS-dev/PF/issues/>

**LazyLoad** true

**LazyData** true

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**Imports** methods,  
plyr,  
dplyr,  
tidyR,  
data.table

**Suggests** testthat,  
knitr,  
rmarkdown,  
R.rsp

**Collate** 'aaa.r'  
'aab.r'  
'class.r'  
'generics.r'  
'PF.r'  
'PF-package.r'  
'IDRlisi.r'  
'IDRsc.r'  
'phiWt.r'

'RRlsi.r'  
 'RRmh.r'  
 'RRmpWald.r'  
 'RRor.r'  
 'RRotsst.r'  
 'RRsc.r'  
 'RRStr.r'  
 'RRtosst.r'  
 'rsbWt.r'  
 'tauWt.r'

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**VignetteBuilder** R.rsp

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---

.rr.score.asymp      *Internal function.*

---

## Description

Internal function.

## Usage

```
.rr.score.asymp(y, alpha = 0.05, iter.max = 18, converge = 1e-04, mn = FALSE)
```

## Arguments

|          |   |
|----------|---|
| y        | data  |
| alpha    | alpha   |
| iter.max | maximum number of iterations                                |
| converge | convergence criterion                                       |
| mn       | boolean whether to calculate MN or use default value of 1.0 |

## Examples

```
.rr.score.asymp(c(0, 18, 16, 19), mn = FALSE)
.rr.score.asymp(c(0, 18, 16, 19), mn = TRUE)
```

---

bird      *bird dataset*

---

## Description

bird dataset

## Format

a data.frame with 6 observations of the following 4 variables, no NAs

- y: number positive
- n: total number in group tx x all
- tx: treatment "vac" or "con"
- all: all?

## References

we need some references

**birdm***birdm dataset***Description**

birdm dataset

**Format**

a data.frame with 6 observations of the following 4 variables, no NAs

- **y**: number positive
- **n**: total number in group tx x all
- **tx**: treatment "vac" or "con"
- **all**: all?

**References**

we need some references

**IDRlsi***IDR likelihood support interval.***Description**

Estimates likelihood support interval for the incidence density ratio or prevented fraction based on it.

**Usage**

```
IDRlsi(
  y = NULL,
  formula = NULL,
  data = NULL,
  alpha = 0.05,
  k = 8,
  use.alpha = FALSE,
  pf = TRUE,
  converge = 1e-08,
  rnd = 3,
  start = NULL,
  trace.it = FALSE,
  iter.max = 24,
  compare = c("con", "vac")
)
```

### Arguments

|                        |  |
|------------------------|--|
| <code>y</code>         | Data vector <code>c(y1, n1, y2, n2)</code> where <code>y</code> are the positives, <code>n</code> are the total, and group 1 is compared to group 2 (control or reference).                |
| <code>formula</code>   | Formula of the form <code>cbind(y, n) ~ x</code> , where <code>y</code> is the number positive, <code>n</code> is the group size, <code>x</code> is a factor with two levels of treatment. |
| <code>data</code>      | <code>data.frame</code> containing variables of the formula.   |
| <code>alpha</code>     | Complement of the confidence level.  |
| <code>k</code>         | Likelihood ratio criterion.  |
| <code>use.alpha</code> | Base choice of <code>k</code> on its relationship to <code>alpha</code> ?  |
| <code>pf</code>        | Estimate <i>IDR</i> or its complement <i>PF</i> ?  |
| <code>converge</code>  | Convergence criterion  |
| <code>rnd</code>       | Number of digits for rounding. Affects display only, not estimates.  |
| <code>start</code>     | describe here.   |
| <code>trace.it</code>  | Verbose tracking of the iterations?  |
| <code>iter.max</code>  | Maximum number of iterations   |
| <code>compare</code>   | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.                                   |

### Details

Estimates likelihood support interval for the incidence density ratio based on orthogonal factoring of reparameterized likelihood. The incidence density is the number of cases per subject-time; its distribution is assumed Poisson.

Likelihood support intervals are usually formed based on the desired likelihood ratio, often  $1/8$  or  $1/32$ . Under some conditions the log likelihood ratio may follow the chi square distribution. If so, then  $\alpha = 1 - F(2\log(k), 1)$ , where  $F$  is a chi-square CDF. if `use.alpha = TRUE`~RRsc()` will make the conversion from  $\alpha$  to  $k$ .

The data may also be a matrix, in which case `y` would be entered as `matrix(c(y1, n1 - y1, y2, n2 - y2), 2, 2, byrow = TRUE)`.

### Value

A `rssi` object with the following elements.

- `estimate`: vector with point and interval estimate
- `estimator`: either *PF* or *IDR*
- `y`: `data.frame` with "y1", "n1", "y2", "n2" values.
- `k`: Likelihood ratio criterion
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level

### Author(s)

[PF-package](#)

### References

Royall R. *Statistical Evidence: A Likelihood Paradigm*. Chapman & Hall, Boca Raton, 1997. Section 7.2.

**See Also**[IDRsc](#)**Examples**

```

# Both examples represent the same observation, with data entry by vector
# and matrix notation.

y_vector <- c(26, 204, 10, 205)
IDRlsi(y_vector, pf = FALSE)

# 1 / 8 likelihood support interval for IDR

# corresponds to 95.858% confidence
# (under certain assumptions)

# IDR
# IDR  LL  UL
# 2.61 1.26 5.88

y_matrix <- matrix(c(26, 178, 10, 195), 2, 2, byrow = TRUE)
y_matrix
#      [,1] [,2]
# [1,]    26   178
# [2,]    10   195

IDRlsi(y_matrix, pf = FALSE)

# 1 / 8 likelihood support interval for IDR

# corresponds to 95.858% confidence
# (under certain assumptions)

# IDR
# IDR  LL  UL
# 2.61 1.26 5.88

data1 <- data.frame(group = rep(c("treated", "control"), each = 5),
                     n = c(rep(41, 4), 40, rep(41, 5)),
                     y = c(4, 5, 7, 6, 4, 1, 3, 3, 2, 1),
                     cage = rep(paste("cage", 1:5), 2))
IDRlsi(data = data1, formula = cbind(y, n) ~ group,
       compare = c("treated", "control"), pf = FALSE)

# 1 / 8 likelihood support interval for IDR

# corresponds to 95.858% confidence
# (under certain assumptions)

# IDR
# IDR  LL  UL
# 2.61 1.26 5.88

require(dplyr)
data2 <- data1 |>
  group_by(group) |>

```

```

  summarize(sum_y = sum(y),
            sum_n = sum(n))

  IDRlsi(data = data2, formula = cbind(sum_y, sum_n) ~ group,
          compare = c("treated", "control"), pf = FALSE)

# 1 / 8 likelihood support interval for IDR

# corresponds to 95.858% confidence
#   (under certain assumptions)

# IDR
# IDR   LL   UL
# 2.61 1.26 5.88

```

IDRsc

*IDR confidence interval.*

## Description

Estimates confidence interval for the incidence density ratio or prevented fraction based on it.

## Usage

```

IDRsc(
  y = NULL,
  data = NULL,
  formula = NULL,
  compare = c("con", "vac"),
  alpha = 0.05,
  pf = TRUE,
  rnd = 3
)

```

## Arguments

|                      |   |
|----------------------|---|
| <code>y</code>       | Data vector <code>c(y1, n1, y2, n2)</code> where <code>y</code> are the positives, <code>n</code> are the total, and <code>group 1</code> is compared to <code>group 2</code> (control or reference). |
| <code>data</code>    | <code>data.frame</code> containing variables of <code>formula</code> .  |
| <code>formula</code> | Formula of the form <code>cbind(y, n) ~ x</code> , where <code>y</code> is the number positive, <code>n</code> is the group size, <code>x</code> is a factor with two levels of treatment.            |
| <code>compare</code> | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.  |
| <code>alpha</code>   | Complement of the confidence level.   |
| <code>pf</code>      | Estimate <i>IDR</i> , or its complement <i>PF</i> ?   |
| <code>rnd</code>     | Number of digits for rounding. Affects display only, not estimates.   |

## Details

The incidence density is the number of cases per subject-time; its distribution is assumed Poisson. IDRsc estimates a confidence interval for the incidence density ratio using Siev's formula based on the Poisson score statistic.  $IDR = \widehat{IDR} \left\{ 1 + \left( \frac{1}{y_1} + \frac{1}{y_2} \right) \frac{z_{\alpha/2}^2}{2} \pm \frac{z_{\alpha/2}^2}{2y_1 y_2} \sqrt{y_\bullet \left( y_\bullet z_{\alpha/2}^2 + 4y_1 y_2 \right)} \right\}$

The data may also be a matrix. In that case y would be entered as  
`matrix(c(y1, n1 - y1, y2, n2 - y2), 2, 2, byrow = TRUE).`

## Value

A `rr1` object with the following elements.

- estimate: vector with point and interval estimate
- estimator: either *PF* or *IDR*
- y: data vector
- rnd: how many digits to round the display
- alpha: complement of confidence level

## Author(s)

[PF-package](#)

## References

- Siev D, 1994. Estimating vaccine efficacy in prospective studies. *Preventive Veterinary Medicine* 20:279-296, Appendix 1.
- Graham PL, Mengersen K, Morton AP, 2003. Confidence limits for the ratio of two rates based on likelihood scores:non-iterative method *Statistics in Medicine* 22:2071-2083.
- Siev D, 2004. Letter to the editor. *Statistics in Medicine* 23:693. (Typographical error in formula: replace the two final minus signs with subscript dots.)

## See Also

[IDRlsc](#)

## Examples

```
# All examples represent the same observation, with data entry by vector,
# matrix, and formula+data notation.

y_vector <- c(26, 204, 10, 205)
IDRsc(y_vector, pf = FALSE)

# IDR    LL    UL
# 2.61  1.28  5.34

# IDR    LL    UL
# 2.61  1.28  5.34

y_matrix <- matrix(c(26, 178, 10, 195), 2, 2, byrow = TRUE)
y_matrix
#      [, 1] [, 2]
```

```

# [1, ]   26  178
# [2, ]   10  195

IDRsc(y_matrix, pf = FALSE)

# IDR
# 95% interval estimates

# IDR   LL   UL
# 2.61 1.28 5.34

require(dplyr)
data1 <- data.frame(group = rep(c("treated", "control"), each = 5),
                     n = c(rep(41, 4), 40, rep(41, 5)),
                     y = c(4, 5, 7, 6, 4, 1, 3, 3, 2, 1),
                     cage = rep(paste("cage", 1:5), 2))
data2 <- data1 |>
  group_by(group) |>
  summarize(sum_y = sum(y),
            sum_n = sum(n))
IDRsc(data = data2, formula = cbind(sum_y, sum_n) ~ group,
      compare = c("treated", "control"), pf = FALSE)

# IDR
# 95% interval estimates

# IDR   LL   UL
# 2.61 1.28 5.34

```

New

*New dataset*

## Description

New dataset

## Format

a data frame with 52 observations of the following 3 variables, no NAs

- cage: cage ID. 1 - 26
- tx: treatment. one of "con" or "vac"
- pos: numeric indicator of positive response. 0 = FALSE or 1 = TRUE

## References

We need some references

|                 |                      |
|-----------------|----------------------|
| <b>pf-class</b> | <i>Data class pf</i> |
|-----------------|----------------------|

**Description**

data class pf

**Fields**

- **estimator**: either "PF" or "IDR"
- **rnd**: how many digits to round display
- **alpha**: complement of c.i.

**Author(s)**

PF-package

**See Also**

[rr1](#), [rrsi](#), [rrsc](#), [rrstr](#)

|              |                                       |
|--------------|---------------------------------------|
| <b>phiWt</b> | <i>Binomial dispersion parameter.</i> |
|--------------|---------------------------------------|

**Description**

MME estimate of dispersion parameter phi.

**Usage**

```
phiWt(fit, subset.factor = NULL, fit.only = TRUE, show.warns = FALSE)
```

**Arguments**

- |                      |   |
|----------------------|---|
| <b>fit</b>           | A <a href="#">glm</a> object.   |
| <b>subset.factor</b> | Factor for estimating phi by subset. Will be converted to a factor if it is not a factor. |
| <b>fit.only</b>      | Return only the new fit? If FALSE, also returns the weights and phi estimates.            |
| <b>show.warns</b>    | Show warnings   |

**Details**

Estimates binomial dispersion parameter  $\phi$  by the method of moments. Refits the model, weighting the observations by  $1/\phi$ . Uses quasibinomial family in [glm\(\)](#).

**Value**

A list with the following elements. **fit**: the new model fit, updated by the estimated weights  
**weights**: vector of weights **phi**: vector of phi estimates

**Author(s)**

[PF-package](#)

**References**

Wedderburn RWM, 1974. Quasi-likelihood functions, generalized linear models, and the Gauss-Newton method. *Biometrika* 61:439-447.

**See Also**

[tauWt](#), [RRor](#).

**Examples**

```
birdm.fit <- glm(cbind(y, n - y) ~ tx-1, binomial, birdm)
RRor(phiWt(birdm.fit))
#
# 95% t intervals on 4 df
#
# PF
#      PF      LL      UL
#  0.479 -0.537  0.823
#
#      mu.hat   LL      UL
# txcon  0.768  0.95  0.367
# txvac  0.400  0.78  0.111
#
```

**print.rr1**

*Print values for PF data objects.*

**Description**

Print values for PF data objects.

**Usage**

```
## S3 method for class 'rr1'
print(x, ...)

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

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

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

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

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

**Arguments**

- x object of class rr1, rror, rrsi, rrmp, rrstr, rrsc
- ... other arguments not used by this method

|     |                    |
|-----|--------------------|
| rat | <i>rat dataset</i> |
|-----|--------------------|

**Description**

rat dataset

**Format**

a data.frame with 32 observations of the following 3 variables, no NAs

- y: number positive
- n: total number
- group: treatment group: "control" or "treated"

**References**

Weil's rat data (Table 1 of Rao and Scott)

|           |                       |
|-----------|-----------------------|
| rr1-class | <i>Data class rr1</i> |
|-----------|-----------------------|

**Description**

Data class rr1

**Fields**

- estimate: vector with point and interval estimate
- estimator: either "PF" or "IDR"
- Y: data.frame with restructured input
- rnd: how many digits to round display
- alpha: complement of c.i.

**Author(s)**

PF-package

**See Also**

[IDRsc](#), [RRotsst](#), [RRtosst](#)

---

|       |  |
|-------|--|
| RRlsi | <i>RR likelihood support interval.</i> |
|-------|--|

---

## Description

likelihood support interval for the risk ratio or prevented fraction by the likelihood profile.

## Usage

```
RRlsi(
  y = NULL,
  formula = NULL,
  data = NULL,
  compare = c("vac", "con"),
  alpha = 0.05,
  k = 8,
  use.alpha = FALSE,
  pf = TRUE,
  iter.max = 50,
  converge = 1e-06,
  rnd = 3,
  start = NULL,
  track = FALSE,
  full.track = FALSE
)
```

## Arguments

|                         |  |
|-------------------------|--|
| <code>y</code>          | Data vector <code>c(y1, n1, y2, n2)</code> where <code>y</code> are the positives, <code>n</code> are the total, and group 1 is compared to group 2 (control or reference group).          |
| <code>formula</code>    | Formula of the form <code>cbind(y, n) ~ x</code> , where <code>y</code> is the number positive, <code>n</code> is the group size, <code>x</code> is a factor with two levels of treatment. |
| <code>data</code>       | <code>data.frame</code> containing variables of <code>formula</code> .   |
| <code>compare</code>    | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.                                   |
| <code>alpha</code>      | Complement of the confidence level (see details).  |
| <code>k</code>          | Likelihood ratio criterion.  |
| <code>use.alpha</code>  | Base choice of <code>k</code> on its relationship to <code>alpha</code> ?  |
| <code>pf</code>         | Estimate <i>RR</i> or its complement <i>PF</i> ?   |
| <code>iter.max</code>   | Maximum number of iterations   |
| <code>converge</code>   | Convergence criterion  |
| <code>rnd</code>        | Number of digits for rounding. Affects display only <code>RR</code> , not estimates.   |
| <code>start</code>      | Optional starting value.   |
| <code>track</code>      | Verbose tracking of the iterations?  |
| <code>full.track</code> | Verbose tracking of the iterations?  |

## Details

Estimates a likelihood support interval for *RR* or *PF* by the profile likelihood method using the DUD algorithm.

Likelihood support intervals are usually formed based on the desired likelihood ratio, often 1 / 8 or 1 / 32. Under some conditions the log likelihood ratio may follow the chi square distribution. If so, then  $\alpha = 1 - F(2\log(k), 1)$ , where  $F$  is a chi-square CDF. if `use.alpha = TRUE`, `RRlsi()` will make the conversion from  $\alpha$  to  $k$

The data may also be a matrix. In that case *Y* would be entered as

```
matrix(c(y1, n1-y1, y2, n2-y2), 2, 2, byrow = TRUE).
```

## Value

An object of class `rssi` with the following fields: `estimate`: matrix of point and interval estimates - see details `estimator`: either "PF" or "RR" `y`: data.frame with "y1", "n1", "y2", "n2" values. `rnd`: how many digits to round the display `k`: likelihood ratio criterion `alpha`: complement of confidence level

## Author(s)

[PF-package](#)

## References

Royall R. *Statistical Evidence: A Likelihood Paradigm*. Chapman & Hall, Boca Raton, 1997. Section 7.6

Ralston ML, Jennrich RI, 1978. DUD, A Derivative-Free Algorithm for Nonlinear Least Squares. *Technometrics* 20:7-14.

## Examples

```
# All examples represent the same observation, with data entry by vector,
# matrix, and formula+data notation.

y_vector <- c(4, 24, 12, 28)
RRlsi(y_vector)

# 1 / 8 likelihood support interval for PF

# corresponds to 95.858% confidence
#   (under certain assumptions)

# PF
#      PF      LL      UL
# 0.6111 0.0168 0.8859

y_matrix <- matrix(c(4, 20, 12, 16), 2, 2, byrow = TRUE)
y_matrix
#      [, 1] [, 2]
# [1, ]    4    20
# [2, ]   12    16

RRlsi(y_matrix)
```

```

# 1 / 8 likelihood support interval for PF

# corresponds to 95.858% confidence
# (under certain assumptions)

# PF
#   PF      LL      UL
# 0.6111 0.0168 0.8859

require(dplyr)
data1 <- data.frame(group = rep(c("treated", "control"), each = 2),
y = c(1, 3, 7, 5),
n = c(12, 12, 14, 14),
cage = rep(paste("cage", 1:2), 2))

data2 <- data1 |>
group_by(group) |>
summarize(sum_y = sum(y),
sum_n = sum(n))
RRIlsi(data = data2, formula = cbind(sum_y, sum_n) ~ group,
compare = c("treated", "control"))

# 1 / 8 likelihood support interval for PF
#
# corresponds to 95.858% confidence
# (under certain assumptions)
#
# PF
#   PF      LL      UL
# 0.6111 0.0168 0.8859

```

RRmh

*Mantel-Haenszel method, CI for common RR over strata or clusters with sparse data.*

## Description

Estimates confidence intervals for the risk ratio or prevented fraction from clustered or stratified data, using a Mantel-Haenszel estimator for sparse data.

## Usage

```

RRmh(
  formula = NULL,
  data = NULL,
  compare = c("vac", "con"),
  Y,
  alpha = 0.05,
  pf = TRUE,
  rnd = 3
)

```

## Arguments

|         |   |
|---------|---|
| formula | Formula of the form <code>cbind(y, n) ~ x + cluster(w)</code> , where Y is the number positive, n is the group size, x is a factor with two levels of treatment, and w is a factor indicating the clusters.   |
| data    | <code>data.frame</code> containing variables for formula  |
| compare | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.  |
| Y       | Matrix of data, $K \times 4$ . Each row is a stratum or cluster. The columns are $y_1, n_1, y_2, n_2$ , where the $y$ 's are the number of positive in each group, and the $n$ is the total in each group. Group 1 corresponds to vaccines and group 2 are controls or reference. If data entered by formula and dataframe, Y is generated automatically. |
| alpha   | Complement of the confidence level.   |
| pf      | Estimate $RR$ or its complement $PF$ ?  |
| rnd     | Number of digits for rounding. Affects display only, not estimates.   |

## Details

Based on the Mantel-Haenszel (1959) procedure for sparse data developed by Greenland and Robins (1985). The confidence limits are based on asymptotic normality of the log(risk ratio). Agresti and Hartzel (2000) favor this procedure for small, sparse data sets, but they warn that it is less efficient than maximum likelihood for large data sets.

## Value

An object of class `rr1` with the following fields.

- `estimate`: vector of point and interval estimates: point estimate, lower confidence limit, upper confidence limit
- `estimator`: either "PF" or "RR"
- `y`: `data.frame` of restructured input
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level

## Note

If either all  $y_1$ 's or all  $y_2$ 's are zero, a division by zero may occur, and a NaN returned for some values.

Vignette *Examples for Stratified Designs* forthcoming with more examples.

Call to this function may be one of two formats: (1) specify data and formula or (2) as a matrix Y

```
RRmh(formula, data, compare = c("b", "a"), pf = TRUE, alpha = 0.05, rnd = 3)
```

```
RRmh(Y, pf = TRUE, alpha = 0.05, rnd = 3)
```

## Author(s)

[PF-package](#)

## References

- Mantel N, Haenszel W, 1959. Statistical aspects of the analysis of data from retrospective studies of disease. *Journal of the National Cancer Institute* 22:719-748.
- Greenland S, Robins JM, 1985. Estimation of a common effect parameter from sparse follow-up data. *Biometrics* 41: 55-68. Errata, 45: 1323-1324.
- Agresti A, Hartzel J, 2000. Strategies for comparing treatments on a binary response with multi-centre data. *Statistics in Medicine* 19: 1115-1139.
- Lachin JM, 2000. *Biostatistical Methods: The Assessment of Relative Risks* (Wiley, New York), Sec. 4.3.1.

## See Also

[rr1](#)

## Examples

```
## Table 1 from Gart (1985)
## as data frame

# tx group "b" is control
RRmh(cbind(y, n) ~ tx + cluster(clus),
      Table6,
      compare = c("a", "b"), pf = FALSE)

# RR
# 95% interval estimates
#
#   RR   LL   UL
# 2.67 1.37 5.23
#
## or as matrix
RRmh(Y = table6, pf = FALSE)

# RR
# 95% interval estimates
#
#   RR   LL   UL
# 2.67 1.37 5.23
```

## Description

data class rrmp

**Fields**

- **estimate**: vector with point and interval estimate
- **estimator**: either "PF" or "IDR"
- **Y**: data vector
- **rnd**: how many digits to round display
- **alpha**: complement of c.i.
- **compare**: text vector, same as input
- **multvec**: data.frame showing the multinomial representation of the data

**Author(s)**

[PF-package](#)

**See Also**

[RRmpWald](#)

**RRmpWald**

*Wald confidence intervals for RR from matched pairs*

**Description**

Estimates confidence intervals for the risk ratio or prevented fraction from matched pairs.

**Usage**

```
RRmpWald(
  formula = NULL,
  data = NULL,
  compare = c("vac", "con"),
  affected = 1,
  x,
  alpha = 0.05,
  pf = TRUE,
  tdist = TRUE,
  df = NULL,
  rnd = 3
)
```

**Arguments**

|                 |  |
|-----------------|--|
| <b>formula</b>  | Formula of the form $y \sim x + \text{cluster}(w)$ , where $y$ is the indicator for an individual's positive response, $x$ is a factor with two levels of treatment, and $w$ identifies the pairs. |
| <b>data</b>     | <code>data.frame</code> containing variables in <code>formula</code>   |
| <b>compare</b>  | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.   |
| <b>affected</b> | Indicator for positive response  |

|       |   |
|-------|---|
| x     | Alternative data input. Instead of formula and data frame, data may be input as frequency vector. See example for how to order this vector. |
| alpha | Complement of the confidence level  |
| pf    | Estimate <i>RR</i> or its complement <i>PF</i> ?  |
| tdist | Use t distribution?   |
| df    | Degrees of freedom. When NULL, the function will default to ‘df = N<br>• 2’, where N is the total number of pairs.                          |
| rnd   | Number of digits for rounding. Affects display only, not estimates.   |

## Details

Estimates confidence intervals for the risk ratio or prevented fraction from matched pairs. The response is the tetranomial vector c(11, 12, 21, 22), where the first index is the row and the second index is the column when displayed as a 2x2 table. Wald type confidence intervals are found by applying the delta method to the multinomial variance. This method fails when there are no responders in one of the treatment groups.

Alternative forms of data entry are illustrated by the output, say Y, where c(Y\$xtable) = Y\$freqvec = Y\$multvec\$Freq.

If RR = 0 (PF = 1), the function will return degenerate interval.

## Value

A `rrmp` object with the following fields:

- `estimate`: vector of point and interval estimates - see details
- `estimator`: either "PF" or "RR"
- `compare`: text vector, same as input
- `alpha`: complement of confidence level
- `rnd`: how many digits to round the display
- `multvec`: data frame showing the multinomial representation of the data

## Note

Experimental functions for estimating profile likelihood intervals are in the CVBmisc package.

Call to this function may be one of two formats: (1) specify `data` and `formula` or (2) as a vector `x`

```
RRmpWald(formula, data, compare = c("vac", "con"), affected = 1, alpha = 0.05, pf = TRUE,
          tdist = TRUE, df = NULL, rnd = 3)
```

```
RRmpWald(x, compare = c("vac", "con"), affected = 1, alpha = 0, 05, pf = TRUE, tdist = TRUE,
          df = NULL, rnd = 3)
```

## Author(s)

PF-package

## Examples

```
RRmpWald(pos ~ tx + cluster(cage), New, compare = c("vac", "con"))

thistable <- New |>
  tidyR::spread(tx, pos) |>
  tidyR::drop_na() |>
  dplyr::mutate(vac = factor(vac, levels = 1:0),
    con = factor(con, levels = 1:0)) |>
  with(table(vac, con))
thistable
as.vector(thistable)

RRmpWald(x = as.vector(thistable))
```

RRor

*RR estimate from logistic regression.*

## Description

Model based interval estimate of the risk ratio or prevented fraction from a logistic regression model.

## Usage

```
RRor(
  fit = NULL,
  beta.hat = NULL,
  var.beta.hat = NULL,
  degf = NULL,
  which = c(1, 2),
  pf = TRUE,
  norm = FALSE,
  alpha = 0.05,
  rnd = 3
)
```

## Arguments

|                           |   |
|---------------------------|---|
| <code>fit</code>          | A <a href="#">glm</a> object.   |
| <code>beta.hat</code>     | Parameters estimates from a logistic regression with no intercept.  |
| <code>var.beta.hat</code> | Variance-covariance matrix from a logistic regression with no intercept.                                    |
| <code>degf</code>         | Degrees of freedom.   |
| <code>which</code>        | Numeric vector indicating which parameters to compare, so that $RR = \text{compare}[2] / \text{compare}[1]$ |
| <code>pf</code>           | Estimate $RR$ or its complement $PF$ ?  |
| <code>norm</code>         | Estimate confidence interval using quantiles of Guassian rather than t distribution quantiles?              |
| <code>alpha</code>        | Complement of the confidence level.   |
| <code>rnd</code>          | Number of digits for rounding. Affects display only, not estimates.   |

## Details

Estimates confidence intervals using the delta method on parameters from a generalized linear model with logit link.

$RR = \mu_2/\mu_1$ , where  $\mu_i$  are the estimated probabilities from the model.

## Value

A `rror` object with the following fields.

- `estimate`: vector with point and interval estimate
- `estimator`: either `PF` or `RR`
- `mu`: matrix with rows giving probability estimates for each of the groups
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level
- `norm`: logical indicating Gaussian or t-interval
- `degf`: degrees of freedom

## Note

Call to this function may be one of two formats: (1) specify `fit` or (2) `beta.hat`, `var.beta.hat`, `degf`

`RRor(fit, degf = NULL, pf = TRUE, alpha = 0.05, which = c(1, 2), norm = TRUE, rnd = 3)`

`RRor(beta.hat, var.beta.hat, degf, pf = TRUE, alpha = 0.05, which = c(1, 2), norm = TRUE, rnd = 3)`

## Author(s)

`PF`-package

## See Also

`rror`, `phiWt`, `tauWt` [StatWI007](#) for more examples

## Examples

```
bird.fit <- glm(cbind(y, n - y) ~ tx - 1, binomial, bird)
RRor(tauWt(bird.fit))

# 95% t intervals on 4 df
#
# PF
#      PF      LL      UL
#  0.500 -0.583  0.842
#
#      mu.hat      LL      UL
# txcon  0.733  0.943  0.3121
# txvac  0.367  0.752  0.0997

RRor(phiWt(bird.fit))
# 95% t intervals on 4 df
#
# PF
```

```
#      PF      LL      UL
#  0.500 -0.583  0.842
#
#      mu.hat    LL      UL
# txcon  0.733 0.943 0.3121
# txvac  0.367 0.752 0.0997
```

**rror-class***Data class rror***Description**

data class rror

**Fields**

- estimate: vector with point and interval estimate
- estimator: either "PF" or "IDR"
- Y: data vector
- rnd: how many digits to round display
- alpha: complement of c.i.
- norm: logical indicating Gaussian or t interval
- degf: degrees of freedom
- mu: matrix with rows giving probability estimates for each of the groups

**Author(s)**[PF-package](#)**See Also**[RRor](#)**RRotsst***RR exact CI, OTSSST method.***Description**

Estimates confidence interval for the risk ratio or prevented fraction, exact method based on the score statistic (inverts one two-sided test).

## Usage

```
RRotsst(
  y = NULL,
  data = NULL,
  formula = NULL,
  compare = c("vac", "con"),
  alpha = 0.05,
  pf = TRUE,
  stepstart = 0.1,
  iter.max = 36,
  converge = 1e-06,
  rnd = 3,
  trace.it = FALSE,
  nuisance.points = 120,
  gamma = 1e-06
)
```

## Arguments

|                 |   |
|-----------------|---|
| y               | Data vector $c(y_1, n_1, y_2, n_2)$ where $y$ are the positives, $n$ are the total, and group 1 is compared to group 2 (control or reference).    |
| data            | data.frame containing variables of the formula.   |
| formula         | Formula of the form $cbind(y, n) \sim x$ , where $y$ is the number positive, $n$ is the group size, $x$ is a factor with two levels of treatment. |
| compare         | Text vector stating the factor levels: $compare[1]$ is the vaccine group to which $compare[2]$ (control or reference) is compared.                |
| alpha           | Complement of the confidence level.   |
| pf              | Estimate $RR$ or its complement $PF$ ?  |
| stepstart       | starting interval for step search   |
| iter.max        | Maximum number of iterations  |
| converge        | Convergence criterion   |
| rnd             | Number of digits for rounding. Affects display only, not estimates.   |
| trace.it        | Verbose tracking of the iterations?   |
| nuisance.points | number of points over which to evaluate nuisance parameter  |
| gamma           | parameter for Berger-Boos correction (restricts range of nuisance parameter evaluation)   |

## Details

Estimates confidence intervals based on the score statistic that are 'exact' in the sense of accounting for discreteness. The score statistic is used to select tail area tables, and the binomial probability is estimated over the tail area by taking the maximum over the nuisance parameter. Algorithm is a simple step search.

The data may also be a matrix. In that case  $Y$  would be entered as

```
matrix(c(y1, n1 - y1, y2, n2 - y2), 2, 2, byrow = TRUE).
```

### Value

An object of class `rr1` with the following fields:

- `estimate`: vector with point and interval estimate
- `estimator`: either "PF" or "RR"
- `y`: `data.frame` with "y1", "n1", "y2", "n2" values.
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level

### Author(s)

[PF-package](#)

### References

- Koopman PAR, 1984. Confidence intervals for the ratio of two binomial proportions. *Biometrics* 40:513-517.
- Agresti A, Min Y, 2001. On small-sample confidence intervals for parameters in discrete distribution. *Biometrics* 57: 963-971.
- Berger RL, Boos DD, 1994. P values maximized over a confidence set for the nuisance parameter. *Journal of the American Statistical Association* 89:214-220.

### See Also

[RRotsst](#), `rr1`.

### Examples

```
# All examples represent the same observation, with data entry by multiple
# options.

y_vector <- c(4, 24, 12, 28)
RRotsst(y_vector, rnd = 3)

# PF
# 95% interval estimates

#      PF      LL      UL
# 0.6111 0.0148 0.8519

y_matrix <- matrix(c(4, 20, 12, 16), 2, 2, byrow = TRUE)
RRotsst(y_matrix, rnd = 3)

# PF
# 95% interval estimates

#      PF      LL      UL
# 0.6111 0.0148 0.8519

require(dplyr)
data1 <- data.frame(group = rep(c("treated", "control"), each = 2),
y = c(1, 3, 7, 5),
n = c(12, 12, 14, 14),
```

```

cage = rep(paste("cage", 1:2), 2))

data2 <- data1 |>
  group_by(group) |>
  summarize(sum_y = sum(y),
            sum_n = sum(n))
RRotsst(data = data2, formula = cbind(sum_y, sum_n) ~ group,
        compare = c("treated", "control"))

# PF
# 95% interval estimates
#
# PF      LL      UL
# 0.6111 0.0148 0.8519

```

RRsc

*RR score based asymptotic CI.*

## Description

Estimates confidence intervals for the risk ratio or prevented fraction based on the score statistic.

## Usage

```

RRsc(
  y = NULL,
  data = NULL,
  formula = NULL,
  compare = c("vac", "con"),
  alpha = 0.05,
  pf = TRUE,
  trace.it = FALSE,
  iter.max = 18,
  converge = 1e-06,
  rnd = 3
)

```

## Arguments

|                       |  |
|-----------------------|--|
| <code>y</code>        | Data vector $c(y_1, n_1, y_2, n_2)$ where $y$ are the positives, $n$ are the total, and group 1 is compared to group 2 (control or reference group).     |
| <code>data</code>     | data.frame containing variables of <code>formula</code> .  |
| <code>formula</code>  | Formula of the form $cbind(y, n) \sim x$ , where $y$ is the number positive, $n$ is the group size, $x$ is a factor with two levels of treatment.        |
| <code>compare</code>  | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared. |
| <code>alpha</code>    | Complement of the confidence level.  |
| <code>pf</code>       | Estimate <i>RR</i> or its complement <i>PF</i> ?   |
| <code>trace.it</code> | Verbose tracking of the iterations?  |
| <code>iter.max</code> | Maximum number of iterations   |
| <code>converge</code> | Convergence criterion  |
| <code>rnd</code>      | Number of digits for rounding. Affects display only, not estimates.  |

## Details

Estimates are returned for three estimators based on the score statistic. The score method was introduced by Koopman (1984). Gart and Nam's modification (1988) includes a skewness correction. The method of Miettinen and Nurminen (1985) is a version made slightly more conservative than Koopman's by including a factor of  $(N - 1)/N$ . The starting estimate for the DUD algorithm is obtained by the modified Katz method (log method with 0.5 added to each cell). Both forms of the Katz estimate may be retrieved from the returned object using `RRsc()$estimate`.

The data may also be a matrix. In that case Y would be entered as

```
matrix(c(y1, n1-y1, y2, n2-y2), 2, 2, byrow = TRUE).
```

## Value

A `rrsc` object with the following fields.

- `estimate`: matrix of point and interval estimates - see details
- `estimator`: either "PF" or "RR"
- `y`: data.frame with "y1", "n1", "y2", "n2" values.
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level

## Author(s)

[PF-package](#)

## References

- Gart JJ, Nam J, 1988. Approximate interval estimation of the ratio of binomial parameters: a review and corrections for skewness. *Biometrics* 44:323-338.
- Koopman PAR, 1984. Confidence intervals for the ratio of two binomial proportions. *Biometrics* 40:513-517.
- Miettinen O, Nurminen M, 1985. Comparative analysis of two rates. *Statistics in Medicine* 4:213-226.
- Ralston ML, Jennrich RI, 1978. DUD, A Derivative-Free Algorithm for Nonlinear Least Squares. *Technometrics* 20:7-14.

## See Also

[rrsc](#)

## Examples

```
# All examples represent the same observation, with data entry by using
# multiple notation options.

y_vector <- c(4, 24, 12, 28)
RRsc(y_vector)

# PF
# 95% interval estimates

#      PF      LL      UL
```

```

# MN method    0.611 0.0251 0.857
# score method 0.611 0.0328 0.855
# skew corr    0.611 0.0380 0.876

y_matrix <- matrix(c(4, 20, 12, 16), 2, 2, byrow = TRUE)
#      [, 1] [, 2]
# [1, ]    4    20
# [2, ]   12    16

RRsc(y_matrix)

# PF
# 95% interval estimates

#      PF      LL      UL
# MN method    0.611 0.0251 0.857
# score method 0.611 0.0328 0.855
# skew corr    0.611 0.0380 0.876
require(dplyr)
data1 <- data.frame(group = rep(c("treated", "control"), each = 2),
y = c(1, 3, 7, 5),
n = c(12, 12, 14, 14),
cage = rep(paste("cage", 1:2), 2))

data2 <- data1 |>
group_by(group) |>
summarize(sum_y = sum(y),
sum_n = sum(n))
RRsc(data = data2, formula = cbind(sum_y, sum_n) ~ group,
compare = c("treated", "control"))

# PF
# 95% interval estimates

#      PF      LL      UL
# MN method    0.611 0.0251 0.857
# score method 0.611 0.0328 0.855
# skew corr    0.611 0.0380 0.876

```

**Description**

data class rrsc

**Fields**

- estimate: vector with point and interval estimate
- rnd: how many digits to round display
- alpha: complement of c.i.
- estimator: either "PF" or "RR"
- Y: data.frame with restructured input

**Author(s)**

[PF-package](#)

**See Also**

[rrsc](#)

**rrsi-class**

*Data class rrsi*

**Description**

data class rrsi

**Fields**

- Y: data.frame with restructured input
- k: likelihood ratio criterion
- rnd: digits to round display
- alpha: complement of c.i.
- estimate: vector with point and interval estimate
- estimator: either "PF" or "IDR"

**Author(s)**

[PF-package](#)

**See Also**

[IDRlsi](#), [RRlsi](#)

**RRstr**

*Gart-Nam method, CI for common RR over strata or clusters.*

**Description**

Estimates confidence intervals for the risk ratio or prevented fraction from clustered or stratified data.

**Usage**

```
RRstr(
  formula = NULL,
  data = NULL,
  compare = c("vac", "con"),
  Y,
  alpha = 0.05,
  pf = TRUE,
  trace.it = FALSE,
  iter.max = 24,
  converge = 1e-06,
  rnd = 3,
  multiplier = 0.7,
  divider = 1.1
)
```

**Arguments**

|                         |   |
|-------------------------|---|
| <code>formula</code>    | Formula of the form <code>cbind(y, n) ~ x + cluster(w)</code> , where <code>y</code> is the number positive, <code>n</code> is the group size, <code>x</code> is a factor with two levels of treatment, and <code>w</code> is a factor indicating the clusters. |
| <code>data</code>       | <code>data.frame</code> containing variables of <code>formula</code>  |
| <code>compare</code>    | Text vector stating the factor levels: <code>compare[1]</code> is the control or reference group to which <code>compare[2]</code> is compared   |
| <code>Y</code>          | Matrix of data. Each row is a stratum or cluster. The columns are <code>y2, n2, y1, n1</code> . If data entered by <code>formula</code> and <code>dataframe</code> , <code>Y</code> is generated automatically.   |
| <code>alpha</code>      | Size of the homogeneity test and complement of the confidence level.  |
| <code>pf</code>         | Estimate <i>RR</i> or its complement <i>PF</i> ?  |
| <code>trace.it</code>   | verbose tracking of the iterations?   |
| <code>iter.max</code>   | Maximum number of iterations  |
| <code>converge</code>   | Convergence criterion   |
| <code>rnd</code>        | Number of digits for rounding. Affects display only, not estimates.   |
| <code>multiplier</code> | internal control parameter for algorithm  |
| <code>divider</code>    | internal control parameter for algorithm  |

**Details**

Uses the DUD algorithm to estimate confidence intervals by the method of Gart.

**Value**

A `rrstr` object with the following fields:

- `estimate`: matrix of point and interval estimates - starting value, MLE, and skewness corrected
- `hom`: list of homogeneity statistic, p-value, and degrees of freedom, or error message if appropriate.
- `estimator`: either "PF" or "RR"
- `y`: `data.frame` of restructured input

- compare: groups compared
- rnd: how many digits to round the display
- alpha: size of test; complement of confidence level

### Note

Vignette *Examples for Stratified Designs* forthcoming with more examples.

Call to this function may be one of two formats: (1) specify data and formula or (2) as a matrix Y  
 RRstr(formula, data, compare = c("b", "a"), pf = TRUE, alpha = 0.05, trace.it = FALSE, iter.max = 24, converge = 1e-6, rnd = 3, multiplier = 0.7, divider = 1.1)  
 RRstr(Y, compare = c("b", "a"), pf = TRUE, alpha = 0.05, trace.it = FALSE, iter.max = 24, converge = 1e-6, rnd = 3, multiplier = 0.7, divider = 1.1)

### Author(s)

[PF-package](#)

### References

- Gart JJ, 1985. Approximate tests and interval estimation of the common relative risk in the combination of 2x2 tables. *Biometrika* 72:673-677.
- Gart JJ, Nam J, 1988. Approximate interval estimation of the ratio of binomial parameters: a review and corrections for skewness. *Biometrics* 44:323-338.
- Ralston ML, Jennrich RI, 1978. DUD, A Derivative-Free Algorithm for Nonlinear Least Squares. *Technometrics* 20:7-14.

### See Also

[rrstr](#)

### Examples

```
## Table 1 from Gart (1985)
## as data frame
## "b" is control group
RRstr(cbind(y, n) ~ tx + cluster(clus),
      Table6,
      compare = c("a", "b"), pf = FALSE)

# Test of homogeneity across clusters

# stat      0.954
# df        3
# p         0.812

# RR estimates

#          RR   LL   UL
# starting  2.66 1.37 5.18
# mle       2.65 1.39 5.03
# skew corr 2.65 1.31 5.08

## or as matrix
```

```
RRstr(Y = table6, pf = FALSE)

tst <- data.frame(y = c(0, 2, 0, 4, 0, 3, 0, 7),
n = rep(10, 8),
tx = rep(c("a", "b"), 4),
clus = rep(paste("Row", 1:4, sep = ""), each = 2))
```

**rrstr-class***Data class rrstr***Description**

data class rrstr

**Fields**

- estimate: vector with point and interval estimate
- rnd: how many digits to round display
- alpha: complement of c.i.
- estimator: either "PF" or "RR"
- hom: list of homogeneity statistic, p-value, and degrees of freedom. If Phi == 0 | Phi == 1, homogeneity test is not possible and error message displays
- Y: data.frame of restructured input
- compare: groups compared

**Author(s)**

[PF-package](#)

**See Also**

[rrstr](#)

**RRtosst***RR exact CI, TOSSST method.***Description**

Estimates confidence interval for the risk ratio or prevented fraction; exact method based on the score statistic (inverts two one-sided tests).

## Usage

```
RRtosst(
  y = NULL,
  formula = NULL,
  data = NULL,
  compare = c("vac", "con"),
  alpha = 0.05,
  pf = TRUE,
  stepstart = 0.1,
  iter.max = 36,
  converge = 1e-06,
  rnd = 3,
  trace.it = FALSE,
  nuisance.points = 120,
  gamma = 1e-06
)
```

## Arguments

|                              |  |
|------------------------------|--|
| <code>y</code>               | Data vector <code>c(y1, n1, y2, n2)</code> where <code>y</code> are the positives, <code>n</code> are the total, and group 1 is compared to group 2 (control or reference group).          |
| <code>formula</code>         | Formula of the form <code>cbind(y, n) ~ x</code> , where <code>y</code> is the number positive, <code>n</code> is the group size, <code>x</code> is a factor with two levels of treatment. |
| <code>data</code>            | <code>data.frame</code> containing variables of <code>formula</code> .   |
| <code>compare</code>         | Text vector stating the factor levels: <code>compare[1]</code> is the vaccine group to which <code>compare[2]</code> (control or reference) is compared.                                   |
| <code>alpha</code>           | Complement of the confidence level.  |
| <code>pf</code>              | Estimate <i>RR</i> or its complement <i>PF</i> ?   |
| <code>stepstart</code>       | starting interval for step search  |
| <code>iter.max</code>        | Maximum number of iterations   |
| <code>converge</code>        | Convergence criterion  |
| <code>rnd</code>             | Number of digits for rounding. Affects display only, not estimates.  |
| <code>trace.it</code>        | Verbose tracking of the iterations?  |
| <code>nuisance.points</code> | number of points over which to evaluate nuisance parameter   |
| <code>gamma</code>           | parameter for Berger-Boos correction (restricts range of nuisance parameter evaluation)  |

## Details

Estimates confidence intervals based on the score statistic that are 'exact' in the sense of accounting for discreteness. Inverts two one-sided score tests. The score statistic is used to select tail area tables, and the binomial probability is estimated over the tail area by taking the maximum over the nuisance parameter. Algorithm is a simple step search.

The data may also be a matrix. In that case `Y` would be entered as

```
matrix(c(y1, n1-y1, y2, n2-y2), 2, 2, byrow = TRUE).
```

**Value**

A `r1` object with the following fields.

- `estimate`: vector with point and interval estimate
- `estimator`: either "PF" or "RR"
- `y`: data.frame with "y1", "n1", "y2", "n2" values.
- `rnd`: how many digits to round the display
- `alpha`: complement of confidence level

**Author(s)**

[PF-package](#)

**References**

- Koopman PAR, 1984. Confidence intervals for the ratio of two binomial proportions. *Biometrics* 40:513-517.
- Agresti A, Min Y, 2001. On small-sample confidence intervals for parameters in discrete distribution. *Biometrics* 57: 963-971.
- Berger RL, Boos DD, 1994. P values maximized over a confidence set for the nuisance parameter. *Journal of the American Statistical Association* 89:214-220.

**See Also**

[RRotsst](#), [r1](#)

**Examples**

```
# Both examples represent the same observation, with data entry by vector
# and matrix notation.

y_vector <- c(4, 24, 12, 28)
RRtosst(y_vector)

# PF
# 95% interval estimates

#      PF      LL      UL
# 0.611 0.012 0.902

y_matrix <- matrix(c(4, 20, 12, 16), 2, 2, byrow = TRUE)
#      [, 1] [, 2]
# [1, ]    4    20
# [2, ]   12    16

RRtosst(y_matrix)

# PF
# 95% interval estimates

#      PF      LL      UL
# 0.611 0.012 0.902
```

```

require(dplyr)
data1 <- data.frame(group = rep(c("treated", "control"), each = 2),
                     y = c(1, 3, 7, 5),
                     n = c(12, 12, 14, 14),
                     cage = rep(paste("cage", 1:2), 2))
data2 <- data1 |>
  group_by(group) |>
  summarize(sum_y = sum(y),
            sum_n = sum(n))
RRtosst(data = data2, formula = cbind(sum_y, sum_n) ~ group,
        compare = c("treated", "control"))

# PF
# 95% interval estimates

#      PF      LL      UL
# 0.611 0.012 0.902

```

rsb

*Rao-Scott weights.*

## Description

Rao-Scott weights.

## Usage

```
rsb(y = NULL, n = NULL, formula = NULL, data = NULL, id = NULL)
```

## Arguments

|         |   |
|---------|---|
| y       | vector of number positive.  |
| n       | vector of total number.   |
| formula | Formula of the form cbind(y, n) ~ id, where y is the number positive, n is the total number, id is a factor for estimating the weights by subset. |
| data    | data.frame containing variables of formula.   |
| id      | vector of factor for estimating the weights by subset.  |

## Details

Estimates the cluster design effect  $d_i$  as the variance inflation due to clustering by the method of Rao and Scott. rsb estimates the  $d_i$  for use by rsbWt or other functions.

## Value

A list with the following elements.

- w: vector of weights
- d: vector of  $d_i$  estimates

## Author(s)

PF-package

## References

Rao JNK, Scott AJ, 1992. A simple method for the analysis of clustered binary data. *Biometrics* 48:577-585.

## See Also

[rsbWt](#).

## Examples

```
# Weil's rat data (Table 1 of Rao and Scott)
rsb(rat$y, rat$n, id = rat$group)$d
# control treated
# 1.232495 3.952861
rsb(data = rat, formula = cbind(y, n) ~ group)$d
# control treated
# 1.232495 3.952861
```

[rsbWt](#)

*Rao-Scott weighting.*

## Description

Rao-Scott weighting of clustered binomial observations.

## Usage

```
rsbWt(fit = NULL, subset.factor = NULL, fit.only = TRUE)
```

## Arguments

- fit A [stats::glm](#) object.
- subset.factor Factor for estimating phi by subset. Will be converted to a factor if it is not a factor.
- fit.only Return only the new fit? If FALSE, also returns the weights and phi estimates.

## Details

Estimates the cluster design effect  $d_i$  as the variance inflation due to clustering by the method of Rao and Scott. Observations are then weighted by the inverse of the  $d_i$ .

## Value

A list with the following elements.

- fit: the new model fit, updated by the estimated weights
- weights: vector of weights
- d: vector of  $d_i$  estimates

## Author(s)

[PF-package](#)

## References

Rao JNK, Scott AJ, 1992. A simple method for the analysis of clustered binary data. *Biometrics* 48:577-585.

## See Also

[RRor](#), [rsb](#).

## Examples

```
birdm.fit <- glm(cbind(y, n - y) ~ tx-1, binomial, birdm)
RRor(rsbWt(birdm.fit))
#
# 95% t intervals on 4 df
#
# PF
#      PF      LL      UL
#  0.479 -1.061  0.868
#
#      mu.hat      LL      UL
# txcon  0.768  0.968  0.2659
# txvac  0.400  0.848  0.0737
#
```

Set1

*Set1 dataset*

## Description

Set1 dataset

## Format

a data.frame with 6 observation of the following 4 variables, no NAs

- y: number positive
- n: total number in group tx x clus
- tx: treatment "vac" or "con"
- clus: cluster ID

## References

We need some references

---

|      |                     |
|------|---------------------|
| set1 | <i>set1 dataset</i> |
|------|---------------------|

---

**Description**

set1 dataset

**Format**

a 3 x 4 matrix of data in [Set1](#)

**References**

we need some references!

---

|        |                       |
|--------|-----------------------|
| Table6 | <i>Table6 dataset</i> |
|--------|-----------------------|

---

**Description**

Table6 dataset

**Format**

a data.frame with 8 observations of the following 4 variables, no NAs

- y: number positive
- n: total number in group tx x clus
- tx: treatment "a" or "b"
- clus: cluster ID

**References**

Table 1 from Gart (1985)

---

|        |                       |
|--------|-----------------------|
| table6 | <i>table6 dataset</i> |
|--------|-----------------------|

---

**Description**

table6 dataset

**Format**

matrix for of data in [Table6](#)

**tauWt***Binomial dispersion: intra-cluster correlation parameter.***Description**

MME estimates of binomial dispersion parameter tau (intra-cluster correlation).

**Usage**

```
tauWt(
  fit,
  subset.factor = NULL,
  fit.only = TRUE,
  iter.max = 12,
  converge = 1e-06,
  trace.it = FALSE
)
```

**Arguments**

|                      |  |
|----------------------|--|
| <b>fit</b>           | A <a href="#">glm</a> object.  |
| <b>subset.factor</b> | Factor for estimating phi by subset. Will be converted to a factor if it is not a factor.                  |
| <b>fit.only</b>      | Return only the final fit? If FALSE, also returns the weights and tau estimates.                           |
| <b>iter.max</b>      | Maximum number of iterations.  |
| <b>converge</b>      | Convergence criterion: difference between model degrees of freedom and Pearson's chi-square. Default 1e-6. |
| <b>trace.it</b>      | Display print statements indicating progress   |

**Details**

Estimates binomial dispersion parameter  $\tau$  by the method of moments. Iteratively refits the model by the Williams procedure, weighting the observations by  $1/\phi_{ij}$ , where  $\phi_{ij} = 1 + \tau_j(n_{ij} - 1)$ ,  $j$  indexes the subsets, and  $i$  indexes the observations.

**Value**

A list with the following elements. **fit**: the new model fit, updated by the estimated weights  
**weights**: vector of weights **phi**: vector of phi estimates

**Author(s)**

[PF-package](#)

**References**

- Williams DA, 1982. Extra-binomial variation in logistic linear models. *Applied Statistics* 31:144-148.  
 Wedderburn RWM, 1974. Quasi-likelihood functions, generalized linear models, and the Gauss-Newton method. *Biometrika* 61:439-447.

**See Also**

[phiWt](#), [RRor](#).

**Examples**

```
birdm.fit <- glm(cbind(y, n - y) ~ tx - 1, binomial, birdm)
RRor(tauWt(birdm.fit))

# 95% t intervals on 4 df
#
# PF          LL         UL
# 0.489 -0.578  0.835
#
# mu.hat      LL         UL
# txcon  0.737 0.944 0.320
# txvac  0.376 0.758 0.104
#
# binomial family only
# any link
```

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