

# Package ‘RSurveillance’

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**Type** Package

**Title** Design and analysis of disease surveillance activities

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**Author** Evan Sergeant

**Maintainer** Evan Sergeant <evan@ausvet.com.au>

**Description** This package provides a range of functions for the design and analysis of disease surveillance activities. These functions were originally developed for animal health surveillance activities but can be equally applied to aquatic animal, wildlife, plant and human health surveillance activities. Utilities are included for sample size calculation and analysis of representative surveys for disease freedom, risk-based studies for disease freedom and for prevalence estimation.

**License** GPL-2|GPL-3

**LazyLoad** yes

**Imports** epitools, epiR

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adj.risk	<i>calculate adjusted risk</i>
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## Description

calculate adjusted risk

## Usage

adj.risk(rr, ppr)

**Arguments**

rr                      relative risk values (vector)  
 ppr                    population proportions corresponding to rr values (vector of equal length)

**Value**

vector of adjusted risk values (in order corresponding to rr)

**Examples**

```
## Not run:
# examples for adj.risk
adj.risk(c(5, 1), c(0.1, 0.9))
adj.risk(c(5, 3, 1), c(0.1, 0.1, 0.8))

## End(Not run)
```

---

ap	<i>estimate apparent prevalence</i>
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---

**Description**

estimate apparent prevalence

**Usage**

```
ap(x, n, type = "wilson", conf = 0.95)
```

**Arguments**

x                      number of positives in sample  
 n                      sample size note: either x or n can be a vector, but at least one must be scalar  
 type                  one of c("normal", "exact", "wilson", "jeffreys", "agresti-coull", "all"), default "wilson"  
 conf                  level of confidence required, default = 0.95

**Value**

either 1) if type = "all", a list with 5 elements, each element a matrix with 6 columns, x, n, proportion, lower confidence limit, upper confidence limit, confidence level and CI method; or 2) a matrix of results for the chosen method

**Examples**

```
## Not run:
# examples for ap function
n<- 200
x<- 25
conf<- 0.95
ap(x, n)
ap(seq(10, 100, 10), 200, type = "agresti")
```

```
ap(seq(10, 100, 10), 200, type = "all")

## End(Not run)
```

---

binom.agresti	<i>calculate agresti-coull confidence limits</i>
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---

### Description

calculate agresti-coull confidence limits

### Usage

```
binom.agresti(x, n, conf = 0.95)
```

### Arguments

x	number of positives in sample
n	sample size note: either x or n can be a vector, but at least one must be scalar
conf	level of confidence required, default 0.95

### Value

a dataframe with 6 columns, x, n, proportion, lower confidence limit, upper confidence limit, confidence level and CI method

### Examples

```
## Not run:
# test binom.agresti
binom.agresti(25, 200)
binom.agresti(seq(10, 100, 10), 200)
binom.agresti(50, seq(100, 1000, 100))

## End(Not run)
```

---

binom.cp	<i>calculate clopper pearson exact confidence limits</i>
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---

### Description

calculate clopper pearson exact confidence limits

### Usage

```
binom.cp(x, n, conf = 0.95)
```

**Arguments**

x	number of positives in sample
n	sample size note: either x or n can be a vector, but at least one must be scalar
conf	level of confidence required, default = 0.95

**Value**

a dataframe with 6 columns, x, n, proportion, lower confidence limit, upper confidence limit, confidence level and CI method

**Examples**

```
## Not run:
# test binom.cp
binom.cp(25, 200)
binom.cp(seq(10, 100, 10), 200)
binom.cp(50, seq(100, 1000, 100))

## End(Not run)
```

---

binom.jeffreys	<i>calculate jeffreys confidence limits</i>
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---

**Description**

calculate jeffreys confidence limits

**Usage**

```
binom.jeffreys(x, n, conf = 0.95)
```

**Arguments**

x	number of positives in sample
n	sample size note: either x or n can be a vector, but at least one must be scalar
conf	level of confidence required, default = 0.95

**Value**

a dataframe with 6 columns, x, n, proportion, lower confidence limit, upper confidence limit, confidence level and CI method

**Examples**

```
## Not run:
# test binom.jeffreys
binom.jeffreys(25, 200)
binom.jeffreys(seq(10, 100, 10), 200)
binom.jeffreys(50, seq(100, 1000, 100))

## End(Not run)
```

---

disc.prior	<i>calculate discounted prior</i>
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---

**Description**

calculate discounted prior

**Usage**

```
disc.prior(prior, p.intro)
```

**Arguments**

prior	prior probability of freedom before surveillance
p.intro	probability of introduction for time period (scalar or vector equal length to sep)

**Value**

vector of discounted prior probability of freedom

**Examples**

```
## Not run:
# examples for disc.prior
disc.prior(0.5, 0.01)
disc.prior(0.95, c(0.001, 0.005, 0.01, 0.02, 0.05))
disc.prior(c(0.5, 0.6, 0.7, 0.8, 0.9, 0.95), 0.01)

## End(Not run)
```

---

epi.calc	<i>calculate effective probability of infection</i>
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---

**Description**

calculate effective probability of infection

**Usage**

```
epi.calc(pstar, rr, ppr)
```

**Arguments**

pstar	design prevalence
rr	relative risk values (vector)
ppr	population proportions corresponding to rr values (vector of equal length)

**Value**

list of 2 elements, a vector of EPI values and a vector of corresponding adjusted risks (in corresponding order to rr)

**Examples**

```
## Not run:
# examples for epi.calc
epi.calc(0.1, c(5, 1), c(0.1, 0.9))
epi.calc(0.02, c(5, 3, 1), c(0.1, 0.1, 0.8))

## End(Not run)
```

---

n.2stage	<i>calculate 2-stage sample size</i>
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---

**Description**

calculate 2-stage sample size

**Usage**

```
n.2stage(H = NA, N = NA, sep.sys = 0.95, sep.c, pstar.c, pstar.u,
  se = 1)
```

**Arguments**

H	population size = number of clusters or NA if not known
N	populaton size if known, scalar or vector
sep.sys	desired population sensitivity
sep.c	desired cluster-level sensitivity
pstar.c	specified cluster-level design prevalence (proportion or integer)
pstar.u	specified population-level design prevalence (proportion or integer)
se	unit sensitivity

**Value**

list of number of clusters to sample and sample size per cluster

**Examples**

```
## Not run:
# examples of n.2stage - checked
n.2stage(NA, NA, 0.95, 0.5, 0.01, 0.1, 0.95)
n.2stage(500, NA, 0.95, 0.5, 10, 0.1, 0.95)
n.2stage(1000, c(50, 100, 200, 500, 1000, 5000, NA), 0.95, 0.5, 0.01, 0.05, 0.8)
n.2stage(1000, c(50, 100, 200, 500, 1000, 5000, NA), 0.95, 0.5, 0.01, 1, 0.8)
n.2stage(1000, c(50, 100, 200, 500, 1000, 5000, NA), 0.9, 0.95, 1, 0.1, 0.8)

## End(Not run)
```

---

n.ap	<i>sample size for apparent prevalence (simple proportion)</i>
------	----------------------------------------------------------------

---

**Description**

sample size for apparent prevalence (simple proportion)

**Usage**

```
n.ap(p, precision, conf = 0.95)
```

**Arguments**

p	expected proportion, scalar or vector of values
precision	absolute precision +/- proportion (half width of desired confidence interval), scalar or vector of values note: at least one of p and precision must be a scalar
conf	level of confidence required, default = 0.95

**Value**

a vector of sample sizes

**Examples**

```
## Not run:
# examples of n.ap
n.ap(0.5, 0.1)
n.ap(0.5, 0.1, conf=0.99)
n.ap(0.5, 0.1, N=1000)
n.ap(seq(0.1, 0.5, by = 0.1), 0.05)
n.ap(0.2, c(0.01, 0.02, 0.05, 0.1))

## End(Not run)
```

---

n.binom	<i>calculate sample size assuming sampling with replacement (binomial)</i>
---------	----------------------------------------------------------------------------

---

**Description**

calculate sample size assuming sampling with replacement (binomial)

**Usage**

```
n.binom(sep, pstar, se = 1)
```

**Arguments**

sep	desired population sensitivity
pstar	specified design prevalence
se	unit sensitivity



**Value**

vector of sample sizes

**Examples**

```
## Not run:
# examples for n.binom - checked
n.binom(sep=0.95, pstar=c(0.01, 0.02, 0.05, 0.1, 0.2))
n.binom(c(0.5, 0.8, 0.9, 0.95), 0.01)

## End(Not run)
```

---

n.c.freecalc	<i>calculate optimum sample size and cut-point reactors for given population size and other parameters using freecalc algorithm all parameters must be scalars</i>
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---

**Description**

calculate optimum sample size and cut-point reactors for given population size and other parameters using freecalc algorithm all parameters must be scalars

**Usage**

```
n.c.freecalc(N, sep = 0.95, c = 1, se, sp = 1, pstar, minSpH = 0.95)
```

**Arguments**

N	population size
sep	target population sensitivity
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence
minSpH	minimum desired population specificity

**Value**

a list of 3 elements, a dataframe with 1 row and six columns for the recommended sample size and corresponding values for population sensitivity (SeP), population specificity (SpP), N, c and pstar, a vector of SeP values and a vector of SpP values, for n = 1:N

**Examples**

```
## Not run:
# examples for n.c.hp
n.c.freecalc(120,0.95,c=5,se=0.9,sp=0.99,pstar=0.1, minSpH=0.9)[[1]]
n.c.freecalc(65,0.95,c=5,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)

## End(Not run)
```

---

n.c.hp	<i>all paramaters must be scalars</i>
--------	---------------------------------------

---

### Description

all paramaters must be scalars

### Usage

```
n.c.hp(N, sep = 0.95, c = 1, se, sp = 1, pstar, minSpH = 0.95)
```

### Arguments

N	population size
sep	target population sensitivity
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence
minSpH	minimum desired population specificity

### Value

a list of 3 elements, a dataframe with 1 row and six columns for the recommended sample size and corresponding values for population sensitivity (SeP), population specificity (SpP), N, c and pstar, a vector of SeP values and a vector of SpP values, for n = 1:N

### Examples

```
## Not run:
# examples for n.c.hp
n.c.hp(65,0.95,c=5,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)[[1]]
tmp<- n.c.hp(120,0.95,c=5,se=0.9,sp=0.99,pstar=0.1, minSpH=0.9)

## End(Not run)
```

---

n.freecalc	<i>calculate freecalc sample size for given population size, cut-point number of reactors and other parameters all paramaters must be scalars</i>
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---

### Description

calculate freecalc sample size for given population size, cut-point number of reactors and other parameters all paramaters must be scalars

**Usage**

```
n.freecalc(N, sep = 0.95, c = 1, se, sp = 1, pstar, minSpH = 0.95)
```

**Arguments**

N	population size
sep	target population sensitivity
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence
minSpH	minimum desired population specificity

**Value**

a list of 2 elements, a dataframe with 1 row and six columns for the recommended sample size and corresponding values for population sensitivity (SeP), population specificity (SpP), N, c and pstar and a dataframe of n rows with SeP and SpP values for each value of n up to the recommended value

**Examples**

```
## Not run:
# examples for n.freecalc
n.freecalc(65,0.95,c=1,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)[[1]]
n.freecalc(65,0.95,c=2,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)[[1]]
n.freecalc(65,0.95,c=3,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)

## End(Not run)
```

---

n.freedom	<i>calculate sample size usng appropriate method for data provided</i>
-----------	------------------------------------------------------------------------

---

**Description**

calculate sample size usng appropriate method for data provided

**Usage**

```
n.freedom(N = NA, sep = 0.95, pstar, se = 1)
```

**Arguments**

N	populaton size if known, scalar or vector
sep	desired population sensitivity
pstar	specified design prevalence (proportion or integer)
se	unit sensitivity

**Value**

vector of sample sizes, NA if N is specified and n>N

**Examples**

```
## Not run:
# examples for n.freedom - checked
n.freedom(NA, sep=0.95, pstar=0.01, se=1)
n.freedom(500, sep=0.95, pstar=0.01, se=1)
n.freedom(N=c(100, 500, 1000, 5000, 10000, 100000, NA), sep=0.95, pstar=0.01, se=1)
n.freedom(500, sep=0.95, pstar=0.01, se=c(0.5, 0.6, 0.7, 0.8, 0.9, 0.99, 1))

## End(Not run)
```

---

n.hp	<i>calculate sample size for given population size, cut-point number of reactors and other parameters calculates sample size to achieve specified population sensitivity with population specificity &gt;= specified minimum value if not possible to achieve target sep AND SpH&gt;= min-SpH, returns sample size for maximum achievable sep. all paramaters must be scalars</i>
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---

**Description**

calculate sample size for given population size, cut-point number of reactors and other parameters  
calculates sample size to achieve specified population sensitivity with population specificity >=  
specified minimum value if not possible to achieve target sep AND SpH>= minSpH, returns sample  
size for maximum achievable sep. all paramaters must be scalars

**Usage**

```
n.hp(N, sep = 0.95, c = 1, se, sp = 1, pstar, minSpH = 0.95)
```

**Arguments**

N	population size
sep	target population sensitivity
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence
minSpH	minimum desired population specificity

**Value**

a list of 2 elements, a dataframe with 1 row and six columns for the recommended sample size and corresponding values for population sensitivity (SeP), population specificity (SpP), N, c and pstar and a dataframe of n rows with SeP and SpP values for each value of n up to the recommended value

**Examples**

```
## Not run:
# examples for n.hp
n.hp(65,0.95,c=1,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)[[1]]
n.hp(65,0.95,c=2,se=0.95,sp=0.99,pstar=0.05, minSpH=0.9)

## End(Not run)
```

---

n.hypergeo	<i>calculate sample size assuming sampling without replacement (hypergeometric)</i>
------------	-------------------------------------------------------------------------------------

---

**Description**

calculate sample size assuming sampling without replacement (hypergeometric)

**Usage**

```
n.hypergeo(sep, N, d, se = 1)
```

**Arguments**

sep	desired population sensitivity
N	population size
d	expected number of infected units in population (=pstar*N rounded to next integer)
se	unit sensitivity

**Value**

vector of sample sizes, NA if n>N

**Examples**

```
## Not run:
# examples for n.hypergeo - checked
n.hypergeo(0.95, N=100, d=1, se = 0.95)
n.hypergeo(sep=0.95, N=c(100, 200, 500, 1000, 10000), d=ceiling(0.01*c(100, 200, 500, 1000, 10000)))
n.hypergeo(c(0.5, 0.8, 0.9, 0.95), N=100, d=5)
n.hypergeo(0.95, N=80, d=c(1, 2, 5, 10))
n.hypergeo(0.95, N=80, d=c(1, 2, 5, 10), se = 0.8)

## End(Not run)
```

---

n.pfree	<i>calculate sample size rqued to achieve target confidence of freedom</i>
---------	----------------------------------------------------------------------------

---

**Description**

calculate sample size rqued to achieve target confidence of freedom

**Usage**

```
n.pfree(pfree, prior, p.intro, pstar, se, N = NA)
```

**Arguments**

pfree	desired probability of freedom (scalar or vector)
prior	prior probability of freedom before surveillance
p.intro	probability of introduction for time period (scalar or vactor equal length to sep)
pstar	design prevalence
se	unit sensitivity (scalar or vector)
N	population size

**Value**

vector of sample sizes

**Examples**

```
## Not run:
# examples for n.pfree
n.pfree(0.95, 0.5, 0.01, 0.05, 0.9)
n.pfree(0.95, 0.5, 0.01, 0.05, 0.9, N=300)
n.pfree(pfree = c(0.9, 0.95, 0.98, 0.99), prior = 0.7, 0.01, 0.01, 0.8, 1000)
n.pfree(0.95, 0.7, 0.01, 0.1, 0.96)

## End(Not run)
```

---

n.pooled	<i>sample size for pooled testing for freedom</i>
----------	---------------------------------------------------

---

**Description**

sample size for pooled testing for freedom

**Usage**

```
n.pooled(sep, k, pstar, pse, psp = 1)
```

**Arguments**

sep	desired population sensitivity (scalar or vector)
k	pool size (constant) (scalar or vector)
pstar	design prevalence
pse	pool-level sensitivity
psp	pool-level specificity

**Value**

vector of sample sizes

**Examples**

```
## Not run:
# examples for n.pooled
n.pooled(0.95, 5, 0.01, 1, 1)
n.pooled(0.95, 10, 0.1, 0.9, 1)
n.pooled(0.95, c(2, 5, 10, 20), 0.1, c(0.99, 0.98, 0.97, 0.95), 1)

## End(Not run)
```

---

n.rb	<i>calculate sample size for risk-based sampling \- binomial</i>
------	------------------------------------------------------------------

---

**Description**

calculate sample size for risk-based sampling \- binomial

**Usage**

```
n.rb(pstar, rr, ppr, spr, se, sep)
```

**Arguments**

pstar	design prevalence (scalar)
rr	relative risk values (vector)
ppr	population proportions corresponding to rr values (vector of equal length)
spr	surveillance proportion for each risk group (vector equal length to rr, ppr)
se	unit sensitivity (fixed or vector same length as rr, ppr, n)
sep	required population sensitivity

**Value**

list of 2 elements, a vector of sample sizes for each risk group a scalar of total sample size, a vector of EPI values and a vector of adjusted risks

**Examples**

```
## Not run:
# examples for n.rb
n.rb(0.1, c(5, 3, 1), c(0.1, 0.10, 0.80), c(0.5, 0.3, 0.2), 0.9, 0.95)
n.rb(0.01, c(5, 1), c(0.1, 0.9), c(0.8, 0.2), c(0.9, 0.95), 0.95)

## End(Not run)
```

---

n.rb.varse

*sample size for varying sensitivity*


---

**Description**

sample size for varying sensitivity

**Usage**

```
n.rb.varse(pstar, rr, ppr, spr, se, spr.rg, sep)
```

**Arguments**

pstar	design prevalence
rr	vector of relative risk values
ppr	vector of population proportions for each risk group - same length as rr
spr	vector of surveillance proportions for each risk group - same length as rr
se	vector of sensitivity values
spr.rg	matrix of proportions of samples for each sensitivity value in each risk group (rows - risk groups, columns = sensitivity values) row sums must equal 1
sep	required population sensitivity

**Value**

list of 3 elements, a matrix of sample sizes for each risk and sensitivity group, a vector of EPI values and a vector of mean sensitivity for each risk group

**Examples**

```
## Not run:
# examples for n.rb.varse
m<- rbind(c(0.8, 0.2), c(0.5, 0.5), c(0.7, 0.3))
n.rb.varse(0.01, c(5, 3, 1), c(0.1, 0.1, 0.8), c(0.4, 0.4, 0.2), c(0.92, 0.8), m, 0.95)

m<- rbind(c(0.8, 0.2), c(0.6, 0.4))
n.rb.varse(0.05, c(3, 1), c(0.2, 0.8), c(0.7, 0.3), c(0.95, 0.8), m, 0.95)

m<- rbind(c(1), c(1))
n.rb.varse(0.05, c(3, 1), c(0.2, 0.8), c(0.7, 0.3), c(0.95), m, 0.99)

## End(Not run)
```



---

n.tp	<i>sample size for true prevalence</i>
------	----------------------------------------

---

**Description**

sample size for true prevalence

**Usage**

```
n.tp(p, se, sp, precision, conf = 0.95)
```

**Arguments**

p	estimated tp
se	test unit sensitivity
sp	test unit specificity
precision	absolute precision +/- proportion (half width of desired confidence interval)
conf	desired level of confidence for CI, default = 0.95

**Value**

a vector of sample sizes

**Examples**

```
## Not run:
# examples for n.tp
n.tp(0.1, 0.9, 0.99, 0.05)
n.tp(0.1, 0.9, 0.99, 0.05, conf = 0.99)
n.tp(c(0.05, 0.1, 0.2, 0.3, 0.4, 0.5), 0.9, 0.99, 0.05)
n.tp(0.5, 0.9, 0.99, c(0.01, 0.02, 0.05, 0.1, 0.2))

## End(Not run)
```

---

pfree.1	<i>calculate confidence of freedom for a single time period</i>
---------	-----------------------------------------------------------------

---

**Description**

calculate confidence of freedom for a single time period

**Usage**

```
pfree.1(sep, p.intro, prior = 0.5)
```

**Arguments**

sep	population sensitivity for time period (scalar or vector)
p.intro	probability of introduction for time period (scalar or vector equal length to sep)
prior	prior probability of freedom before surveillance

**Value**

data.frame with columns for sep, p.intro, discounted prior, pfree, pfree.equ and prior.equ

**Examples**

```
## Not run:
# examples for pfree.1
pfree.1(0.8, 0.01, 0.5)
pfree.1(0.6, c(0.001, 0.005, 0.01, 0.02, 0.05), 0.5)
pfree.1(runif(10, 0.4, 0.6), 0.01, 0.5)
pfree.1(runif(10, 0.4, 0.6), runif(10, 0.005, 0.015), 0.5)

## End(Not run)
```

---

pfree.calc	<i>calculate probability (confidence) of freedom for given prior, sep and p.intro over 1 or more time periods</i>
------------	-------------------------------------------------------------------------------------------------------------------

---

**Description**

calculate probability (confidence) of freedom for given prior, sep and p.intro over 1 or more time periods

**Usage**

```
pfree.calc(sep, p.intro, prior = 0.5)
```

**Arguments**

sep	population sensitivity for time period (scalar or vector)
p.intro	probability of introduction for time period (scalar or vector equal length to sep)
prior	prior probability of freedom before surveillance

**Value**

data.frame with columns for sep, p.intro, discounted prior, pfree, pfree.equ and prior.equ

**Examples**

```
## Not run:
# examples for pfree.calc
pfree.calc(0.8, 0.01, 0.5)
pfree.calc(rep(0.6,24), 0.01, 0.5)
pfree.calc(runif(10, 0.4, 0.6), 0.01, 0.5)
pfree.calc(runif(10, 0.4, 0.6), runif(10, 0.005, 0.015), 0.5)

## End(Not run)
```

---

pfree.equ	<i>calculate equilibrium pfree and equilibrium prior Pfree</i>
-----------	----------------------------------------------------------------

---

**Description**

calculate equilibrium pfree and equilibrium prior Pfree

**Usage**

```
pfree.equ(sep, p.intro)
```

**Arguments**

sep	population sensitivity for time period (scalar or vector)
p.intro	probability of introduction for time period (scalar or vector equal length to sep)

**Value**

list of 2 vectors, equilibrium posterior probability of freedom and equilibrium prior (discounted) probability of freedom

**Examples**

```
## Not run:
# examples of pfree.equ
pfree.equ(runif(10, 0.4, 0.6), 0.01)
pfree.equ(0.8, 0.05)
pfree.equ(rep(0.9, 6), c(0.0001, 0.0005, 0.001, 0.005, 0.01, 0.05))

## End(Not run)
```

---

pstar.calc	<i>calculate design prevalence for given sample size and desired population-level sensitivity</i>
------------	---------------------------------------------------------------------------------------------------

---

**Description**

calculate design prevalence for given sample size and desired population-level sensitivity

**Usage**

```
pstar.calc(N = NA, n, sep, se)
```

**Arguments**

N	population size if known, scalar or vector
n	sample size
sep	desired population sensitivity
se	unit sensitivity

**Value**

vector of design prevalence values

**Examples**

```
## Not run:
# examples of pstar.calc- checked
pstar.calc(NA, 280, 0.95, 0.98)
pstar.calc(500, 250, sep=0.95, se=1)
pstar.calc(N=c(100, 500, 1000, 5000, 10000, 100000, NA), n=30, sep=0.95, se=1)
pstar.calc(500, n=30, sep=0.95, se=c(0.5, 0.6, 0.7, 0.8, 0.9, 0.99, 1))

## End(Not run)
```

---

sd.tp	<i>function to calculate variance/sd of tp assuming se and sp known exactly</i>
-------	---------------------------------------------------------------------------------

---

**Description**

function to calculate variance/sd of tp assuming se and sp known exactly

**Usage**

```
sd.tp(x, n, se, sp)
```

**Arguments**

x	number of positive results in sample
n	sample size
se	test unit sensitivity
sp	test unit specificity

**Value**

vector of standard deviation values for true prevalence estimates

**Examples**

```
## Not run:
# example of sd.tp
sd.tp(1:10, 20, 0.9, 0.99)

## End(Not run)
```

---

se.parallel	<i>calculate combined sensitivity for multiple tests in parallel (assuming independence)</i>
-------------	----------------------------------------------------------------------------------------------

---

**Description**

calculate combined sensitivity for multiple tests in parallel (assuming independence)

**Usage**

```
se.parallel(se)
```

**Arguments**

se                      vector of unit sensitivity values

**Value**

scalar of combined sensitivity, assuming independence

**Examples**

```
## Not run:
# examples for se.parallel
se.parallel(c(0.99, 0.95, 0.8))

## End(Not run)
```

---

se.series	<i>calculate combined sensitivity for multiple tests in series (assuming independence)</i>
-----------	--------------------------------------------------------------------------------------------

---

**Description**

calculate combined sensitivity for multiple tests in series (assuming independence)

**Usage**

```
se.series(se)
```

**Arguments**

se                      vector of unit sensitivity values

**Value**

scalar of combined sensitivity, assuming independence

**Examples**

```
## Not run:
# examples for se.series
se.series(c(0.99, 0.95, 0.8))

## End(Not run)
```

sep

*Population sensitivity function to calculate seh using most appropriate option depending on whether or not N provided*

**Description**

Population sensitivity function to calculate seh using most appropriate option depending on whether or not N provided

**Usage**

```
sep(N = NA, n, pstar, se = 1)
```

**Arguments**

N	vector of population sizes: NA or vector of same length as n
n	vector of sample sizes
pstar	design prevalence: single value as a proportion or integer
se	unit sensitivity: single value or vector same length as n

**Value**

value vector of population-level sensitivities

**Examples**

```
## Not run:
# examples for sep - checked
sep(n=300, pstar=0.01, se=1)
sep(NA, 300, 0.01, 1)
sep(10000, 150, 0.02, 1)
sep(n=1:100, pstar = 0.05, se=0.95)
N<- seq(30, 100, by = 5)
se<- 0.95
pstar<- 0.1
n<- rep(30, length(N))
sep(N, n, pstar, se = se)
sep(rep(100, 10), seq(10, 100, by = 10), pstar = 1, se=0.99)
N<- c(55, 134, NA, 44, 256)
n<- c(15, 30, 28, 15, 33)
sep(N, n, 0.1, 0.95)

## End(Not run)
```

---

sep.binom	<i>Population sensitivity assuming sampling with replacement (binomial)</i>
-----------	-----------------------------------------------------------------------------

---

**Description**

Population sensitivity assuming sampling with replacement (binomial)

**Usage**

```
sep.binom(n, pstar, se = 1, sp = 1)
```

**Arguments**

n	integer scalar or vector of number tested (sample size)
pstar	scalar or vector of design prevalence as proportion
se	unit sensitivity of test (proportion)
sp	unit specificity of test (proportion)

**Value**

vector of population-level sensitivities

**Examples**

```
## Not run:
# examples for sep.binom - checked
sep.binom(n=300, pstar = 0.02, se = 0.92)
tested<- seq(10,100, by=10)
prev<- 0.05
sens<- 0.9
sep.binom(tested, prev, sens)

## End(Not run)
```

---

sep.binom.imperfect	<i>any one input can be a vector, all others must be either scalars or vectors of the same length</i>
---------------------	-------------------------------------------------------------------------------------------------------

---

**Description**

any one input can be a vector, all others must be either scalars or vectors of the same length

**Usage**

```
sep.binom.imperfect(n, c = 1, se, sp = 1, pstar)
```

**Arguments**

n	sample size
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence

**Value**

a vector of population-level sensitivities

**Examples**

```
## Not run:
# examples for sep.imperfect.binom
sep.binom.imperfect(1:10*5, 2, 0.95, 0.98, 0.1)
sep.binom.imperfect(50, 1:5, 0.95, 0.98, 0.1)
sep.binom.imperfect(30, 2, 0.9, 0.98, 0.1)
sep.binom.imperfect(30, 1, 0.9, 0.98, 0.1)

## End(Not run)
```

---

sep.exact

---

*Population sensitivity assuming census (all units tested)*


---

**Description**

Population sensitivity assuming census (all units tested)

**Usage**

```
sep.exact(d = 1, se = 1)
```

**Arguments**

d	expected number of infected units in population (=pstar*N rounded to next integer)
se	unit sensitivity of test (proportion)

**Value**

vector of population-level sensitivities



**Examples**

```
## Not run:
# examples for sep.exact - checked
sep.exact(d=1, se = 0.92)
inf<- 1:5
sens<- 0.8
sep.exact(d=inf, se=sens)
sep.exact(se=0.8, d = ceiling(0.01*c(10, 50, 100, 250, 500)))

## End(Not run)
```

---

sep.freecalc	<i>all inputs are scalars</i>
--------------	-------------------------------

---

**Description**

all inputs are scalars

**Usage**

```
sep.freecalc(N, n, c = 1, se, sp = 1, pstar)
```

**Arguments**

N	population size
n	sample size
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence

**Value**

a scalar of population-level sensitivity

**Examples**

```
## Not run:
# examples of sep.freecalc
sep.freecalc(150, 30, 2, 0.9, 0.98, 0.1)
sep.freecalc(150, 30, 1, 0.9, 0.98, 0.1)

## End(Not run)
```

---

sep.hp	<i>any one input can be a vector, all others must be either scalars or vectors of the same length</i>
--------	-------------------------------------------------------------------------------------------------------

---

**Description**

any one input can be a vector, all others must be either scalars or vectors of the same length

**Usage**

```
sep.hp(N, n, c = 1, se, sp = 1, pstar)
```

**Arguments**

N	population size
n	sample size
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
se	unit sensitivity
sp	unit specificity, default=1
pstar	design prevalence

**Value**

a vector of population-level sensitivities

**Examples**

```
## Not run:
# examples of sep.hp
sep.hp(150, 1:5*10, 2, 0.9, 0.98, 0.1)
sep.hp(150, 30, 2, 0.9, 0.98, 15)
sep.hp(150, 30, 1, 0.9, 0.98, 15)
sep.hp(150, 30, 1, 0.9, 0.98, 0.1)

## End(Not run)
```

---

sep.hypergeo	<i>Population sensitivity assuming sampling without replacement (hypergeometric approximation)</i>
--------------	----------------------------------------------------------------------------------------------------

---

**Description**

Population sensitivity assuming sampling without replacement (hypergeometric approximation)

**Usage**

```
sep.hypergeo(N, n, d, se = 1)
```

**Arguments**

N	population size
n	sample size (tested)
d	expected number of infected units in population (=pstar*N rounded to next integer)
se	unit sensitivity of test (proportion)

**Value**

value vector of population-level sensitivities

**Examples**

```
## Not run:
# examples for sep.hypergeo - checked
sep.hypergeo(N=100, n=50, d=1, se = 0.92)
inf<- 1:5
sens<- 0.8
sep.hypergeo(N=100, n=50, d=inf, se=sens)
N<- c(10, 50, 100, 250, 500)
sep.hypergeo(se=0.8, N=N, n=c(5, 25, 50, 125, 250), d = ceiling(0.01*N))

## End(Not run)
```

---

sep.pfree

*calculate population sensitivity for given PFree*


---

**Description**

calculate population sensitivity for given PFree

**Usage**

```
sep.pfree(prior, pfree)
```

**Arguments**

prior	prior probability of freedom before surveillance (scalar or vector)
pfree	desired probability of freedom (scalar or vector)

**Value**

vector of population-level sensitivities

**Examples**

```
## Not run:
# examples of sep.pfree
sep.pfree(0.5, 0.95)
sep.pfree(c(0.5, 0.6, 0.7, 0.8, 0.9, 0.95), 0.99)
sep.pfree(0.5, c(0.8, 0.9, 0.95, 0.99))

## End(Not run)
```

---

sep.pooled	<i>population sensitivity with pooled sampling</i>
------------	----------------------------------------------------

---

**Description**

population sensitivity with pooled sampling

**Usage**

```
sep.pooled(r, k, pstar, pse, psp = 1)
```

**Arguments**

r	number of pools sampled (scalar or vector)
k	pool size (constant) (scalar or vector)
pstar	design prevalence
pse	pool-level sensitivity
psp	pool-level specificity

**Value**

list of 2 elements, vector of sep values and vector of spp values

**Examples**

```
## Not run:
# examples for sep.pooled
sep.pooled(60, 5, 0.01, 1, 1)
sep.pooled(4, 10, 0.1, 0.9, 1)
sep.pooled(1:10*5, 5, 0.02, 0.9, 0.99)
sep.pooled(10, 5, 0.05, c(0.8, 0.9, 0.95, 0.99), 1)

## End(Not run)
```

---

sep.prior	<i>calculate population sensitivity for given prior pfree</i>
-----------	---------------------------------------------------------------

---

**Description**

calculate population sensitivity for given prior pfree

**Usage**

```
sep.prior(prior, p.intro)
```

**Arguments**

prior	prior probability of freedom before surveillance (scalar or vector)
p.intro	probability of introduction for time period (scalar or vector equal length to sep)

**Value**

vector of population-level sensitivities

**Examples**

```
## Not run:
# examples of sep.prior
sep.prior(0.95, 0.01)
sep.prior(c(0.9, 0.95, 0.98, 0.99), 0.01)
sep.prior(0.95, c(0.001, 0.005, 0.01, 0.02, 0.05))

## End(Not run)
```

---

sep.rb.bin

---

*calculate popuation sensitivity for single risk factor - binomial*


---

**Description**

calculate popuation sensitivity for single risk factor - binomial

**Usage**

```
sep.rb.bin(pstar, rr, ppr, n, se)
```

**Arguments**

pstar	design prevalence (scalar)
rr	relative risk values (vector)
ppr	population proportions corresponding to rr values (vector of equal length)
n	sample size per risk category (vector same length as rr and ppr)
se	unit sensitivity (fixed or vector same length as rr, ppr, n)

**Value**

list of 3 elements, a scalar of population-level sensitivity a vector of EPI values and a vector of corresponding Adjusted risks

**Examples**

```
## Not run:
# examples for sep.rb.bin
sep.rb.bin(0.1, c(5, 3, 1), c(0.1, 0.1, 0.8), c(5, 5, 5), 0.9)
sep.rb.bin(0.1, c(5, 1), c(0.1, 0.9), c(10, 5), c(0.95, 0.9))
sep.rb.bin(0.1, c(5, 1), c(0.1, 0.9), c(10, 5), c(0.9, 0.9))
sep.rb.bin(0.01, c(5, 1), c(0.1, 0.9), c(90, 50), c(0.9, 0.9))

## End(Not run)
```

---

sep.rb.bin.varse	<i>calculate population sensitivity for varying unit sensitivity - binomial</i>
------------------	---------------------------------------------------------------------------------

---

## Description

calculate population sensitivity for varying unit sensitivity - binomial

## Usage

```
sep.rb.bin.varse(pstar, rr, ppr, df)
```

## Arguments

pstar	design prevalence (scalar)
rr	relative risk values (vector)
ppr	population proportions corresponding to rr values (vector of equal length)
df	dataframe of values for each sensitivity level col 1 = risk group index, col 2 = unit Se, col 3 = n

## Value

list of 3 elements, a scalar of population-level sensitivity a vector of EPI values and a vector of corresponding Adjusted risks

## Examples

```
## Not run:
# examples for sep.rb.bin.varse
rg<- c(1, 1, 2, 2)
se<- c(0.92, 0.85, 0.92, 0.85)
n<- c(80, 30, 20, 30)
df<- data.frame(rg, se, n)
sep.rb.bin.varse(0.01, c(5, 1), c(0.1, 0.9), df)

rg<- c(1, 1, 2, 2)
se<- c(0.95, 0.8, 0.95, 0.8)
n<- c(20, 10, 10, 5)
df<- data.frame(rg, se, n)
sep.rb.bin.varse(0.05, c(3, 1), c(0.2, 0.8), df)

rg<- c(rep(1, 30), rep(2, 15))
se<- c(rep(0.95, 20), rep(0.8, 10), rep(0.95, 10), rep(0.8, 5))
n<- rep(1, 45)
df<- data.frame(rg, se, n)
sep.rb.bin.varse(0.02, c(3, 1), c(0.2, 0.8), df)

rg<- c(1, 2, 3, 1, 2, 3)
se<- c(0.95, 0.95, 0.95, 0.8, 0.8, 0.8)
n<- c(20, 10, 10, 30, 5, 5)
df<- data.frame(rg, se, n)
sep.rb.bin.varse(0.01, c(5, 3, 1), c(0.1, 0.3, 0.6), df)

## End(Not run)
```

---

sep.rb.hypergeo	<i>calculate population sensitivity for single risk factor - hypergeometric</i>
-----------------	---------------------------------------------------------------------------------

---

**Description**

calculate population sensitivity for single risk factor - hypergeometric

**Usage**

```
sep.rb.hypergeo(pstar, rr, N, n, se)
```

**Arguments**

pstar	design prevalence (scalar)
rr	relative risk values (vector)
N	Population size per risk category (vector same length as rr and ppr)
n	sample size per risk category (vector same length as rr and ppr)
se	unit sensitivity (fixed or vector same length as rr, ppr, n)

**Value**

list of 3 elements, a scalar of population-level sensitivity a vector of EPI values and a vector of corresponding Adjusted risks

**Examples**

```
## Not run:
# examples for sep.rb.bin
sep.rb.hypergeo(0.1, c(5, 3, 1), c(10, 10, 80), c(5, 5, 5), 0.9)
sep.rb.hypergeo(0.1, c(5, 1), c(15, 140), c(10, 5), c(0.95, 0.9))
sep.rb.hypergeo(0.1, c(5, 1), c(23, 180), c(10, 5), c(0.9, 0.9))
sep.rb.hypergeo(0.01, c(5, 1), c(100, 900), c(90, 50), c(0.9, 0.9))

## End(Not run)
```

---

sep.rb.hypergeo.varse	<i>calculate population sensitivity for varying unit sensitivity - hypergeometric</i>
-----------------------	---------------------------------------------------------------------------------------

---

**Description**

calculate population sensitivity for varying unit sensitivity - hypergeometric

**Usage**

```
sep.rb.hypergeo.varse(pstar, rr, N, df)
```

**Arguments**

pstar	design prevalence (scalar)
rr	relative risk values (vector)
N	vector of population size corresponding to rr values (vector of equal length)
df	dataframe of values for each sensitivity level col 1 = risk group index, col 2 = unit Se, col 3 = n

**Value**

list of 5 elements, a scalar of population-level sensitivity a vector of EPI values, a vector of corresponding Adjusted risks a vector of sample sizes (n) per risk group and a vector of mean unit sensitivities per risk group

**Examples**

```
## Not run:
# examples for sep.rb.hypergeo.varse
rg<- c(1, 1, 2, 2)
se<- c(0.92, 0.85, 0.92, 0.85)
n<- c(80, 30, 20, 30)
df<- data.frame(rg, se, n)
sep.rb.hypergeo.varse(0.01, c(5, 1), c(200, 1800), df)

rg<- c(1, 1, 2, 2)
se<- c(0.95, 0.8, 0.95, 0.8)
n<- c(20, 10, 10, 5)
df<- data.frame(rg, se, n)
sep.rb.hypergeo.varse(0.05, c(3, 1), c(100, 400), df)

rg<- c(rep(1, 30), rep(2, 15))
se<- c(rep(0.95, 20), rep(0.8, 10), rep(0.95, 10), rep(0.8, 5))
n<- rep(1, 45)
df<- data.frame(rg, se, n)
sep.rb.hypergeo.varse(0.02, c(3, 1), c(100, 400), df)

rg<- c(1, 2, 3, 1, 2, 3)
se<- c(0.95, 0.95, 0.95, 0.8, 0.8, 0.8)
n<- c(20, 10, 10, 30, 5, 5)
df<- data.frame(rg, se, n)
sep.rb.hypergeo.varse(0.01, c(5, 3, 1), c(100, 300, 600), df)

## End(Not run)
```

---

sep.rb2.binom

---

1-stage risk-based sampling with 2 risk factors - binomial

---

**Description**

1-stage risk-based sampling with 2 risk factors - binomial

**Usage**

```
sep.rb2.binom(pstar, rr1, ppr1, rr2, ppr2, n, se)
```



**Arguments**

pstar	design prevalence
rr1	relative risks for first level risk factor
ppr1	population proportions for first level risk factor
rr2	relative risks for second level risk factor, matrix, rows = levels of rr1, cols = levels of rr2
ppr2	population proportions for second level risk factor, matrix, rows = levels of rr1, cols = levels of rr2
n	matrix of number tested for each risk group (stratified
se	test unit sensitivity

**Value**

list of 4 elements, a scalar of population-level sensitivity a matrix of EPI values, a vector of corresponding Adjusted risks for the first risk factor and a matrix of adjusted risks for the second risk factor

**Examples**

```
## Not run:
# examples for sep.rb2.binom
pstar<- 0.01
rr1<- c(3, 1)
ppr1<- c(0.2, 0.8)
rr2<- rbind(c(4,1), c(4,1))
ppr2<- rbind(c(0.1, 0.9), c(0.3, 0.7))
se<- 0.8
n<- rbind(c(50, 20), c(20, 10))
sep.rb2.binom(pstar, rr1, ppr1, rr2, ppr2, n, se)

## End(Not run)
```

---

sep.rb2.hypergeo	<i>1-stage risk-based sampling with 2 risk factors - hypergeometric</i>
------------------	-------------------------------------------------------------------------

---

**Description**

1-stage risk-based sampling with 2 risk factors - hypergeometric

**Usage**

```
sep.rb2.hypergeo(pstar, rr1, rr2, N, n, se)
```

**Arguments**

pstar	design prevalence
rr1	relative risks for first level risk factor
rr2	relative risks for second level risk factor
N	matrix of population size for each risk group (stratified
n	matrix of number tested (sample size) for each risk group (stratified
se	test unit sensitivity

**Value**

list of 6 elements, a scalar of population-level sensitivity a matrix of EPI values, a vector of corresponding Adjusted risks for the first risk factor and a matrix of adjusted risks for the second risk factor, a vector of population proportions for the first risk factor and a matrix of population proportions for the second risk factor

**Examples**

```
## Not run:
# examples for sep.rb2.hypergeo
pstar<- 0.01
rr1<- c(3, 1)
rr2<- rbind(c(4,1), c(4,1))
N<- rbind(c(100, 500), c(300, 1000))
n<- rbind(c(50, 20), c(20, 10))
se<- 0.8
sep.rb2.hypergeo(pstar, rr1, rr2, N, n, se)

## End(Not run)
```

---

sep.sys

---

*population sensitivity from sampling of individual clusters*


---

**Description**

population sensitivity from sampling of individual clusters

**Usage**

```
sep.sys(H = NA, N = NA, n, pstar.c, pstar.u, se = 1)
```

**Arguments**

H	population size = number of clusters or NA if not known
N	population size within clusters NA if not provided, otherwise a vector of same length as n
n	sample size (tested)
pstar.c	cluster (herd) level design prevalence single value either proportion or integer
pstar.u	unit (animal) level design prevalence single value either proportion or integer
se	unit sensitivity of test (proportion)

**Value**

vector of population-level sensitivities

**Note**

if pstar.c is not a proportion N must be entered (and  $N \geq n$ )

**Examples**

```
## Not run:
# examples for sep.sys - checked
H<- 500
N<- rep(1000, 150)
N[5]<- NA
n<- rep(30, 150)
pstar.u<- 0.1
pstar.c<- 0.01
se<- 0.98
sep.sys(H, N, n, pstar.c, pstar.u, se)
sep.sys(NA, N, n, 0.02, 0.05, 0.95)
N<- round(runif(105)*900+100)
n<- round(runif(105)*30+10)
sse<- sep.sys(1000, N, n, 0.02, 0.05, 0.9)
data.frame(N, n, sse[[2]])

## End(Not run)
```

sep.var.se

*population sensitivity for varying unit sensitivity***Description**

population sensitivity for varying unit sensitivity

**Usage**

```
sep.var.se(N = NA, se, pstar)
```

**Arguments**

N	population size (number of units/clusters. N >= length(se)) or NA if unknown
se	vector of unit sensitivity values
pstar	specified design prevalence

**Value**

value vector of population-level sensitivities

**Examples**

```
## Not run:
# examples of sep.var.se - checked
sens<- c(rep(0.9, 50), rep(0.95, 100))
sep.var.se(NA, sens, 0.01)
sep.var.se(se=sens, pstar=0.01)
sep.var.se(N=500, sens, 0.01)
sep.var.se(NA, runif(150, 0.95, 0.99), 0.02)
sep.var.se(500, runif(150, 0.95, 0.99), 0.02)

## End(Not run)
```

---

sph	<i>Population specificity</i>
-----	-------------------------------

---

**Description**

Population specificity

**Usage**

sph(n, sp)

**Arguments**

n	integer scalar or vector of number tested
sp	unit specificity of test (proportion)

**Value**

value vector of population-level specificities

**Examples**

```
## Not run:
# examples for sph - checked
sph(10, 0.9)
sph(c(10, 20, 50, 100), 0.99)
sph(100, c(0.999, 0.99, 0.98, 0.95, 0.9))

## End(Not run)
```

---

sph.binom	<i>any one input can be a vector, all others must be either scalars or vectors of the same length</i>
-----------	-------------------------------------------------------------------------------------------------------

---

**Description**

any one input can be a vector, all others must be either scalars or vectors of the same length

**Usage**

sph.binom(n, c = 1, sp)

**Arguments**

n	sample size
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
sp	unit specificity

**Value**

a vector of population-level specificities

**Examples**

```
## Not run:
# examples for sph.imperfect.sp
sph.binom(30, 2, 0.98)
sph.binom(30, 1, 0.98)
sph.binom(1:5*10, 2, 0.98)
sph.binom(100, 1:5, 0.98)
sph.binom(100, 3, 95:100/100)
sph.binom(c(5, 10, 15, 20, 30, 50, 100, 200), 2, 0.98)

## End(Not run)
```

---

sph.hp

*HerdPlus herd specificity calculation any one input can be a vector, all others must be either scalars or vectors of the same length*

---

**Description**

HerdPlus herd specificity calculation any one input can be a vector, all others must be either scalars or vectors of the same length

**Usage**

```
sph.hp(N, n, c = 1, sp)
```

**Arguments**

N	population size
n	sample size
c	The cut-point number of reactors to classify a herd/flock as positive, default=1. if reactors < c result is negative, >= c is positive
sp	unit specificity

**Value**

a vector of population-level specificities

**Examples**

```
## Not run:
# examples of sph.hp
sph.hp(150, 30, 2, 0.98)
sph.hp(150, 30, 1, 0.98)
sph.hp(150, 1:5*10, 2, 0.98)
sph.hp(500, 30, 2, 95:100/100)

## End(Not run)
```

---

sse.combined	<i>update between components to account for lack of independence</i>
--------------	----------------------------------------------------------------------

---

## Description

update between components to account for lack of independence

## Usage

```
sse.combined(C = NA, pstar.c, rr, ppr, sep)
```

## Arguments

C	NA or vector of population sizes (number of clusters) for each risk group
pstar.c	cluster level design prevalence
rr	cluster level relative risks
ppr	cluster level population proportions (not required if C is specified)
sep	list of sep values for clusters in each component and corresponding risk group. Each element is a dataframe, first column= clusterid, 2nd =rg.c, 3rd col = sep

## Value

list of 2 elements, a matrix (or vector if C not specified) of population-level (surveillance system) sensitivities (binomial and hypergeometric and adjusted vs unadjusted) and a matrix of adjusted and unadjusted component sensitivities for each component

## Examples

```
## Not run:
# example for sse.combined (checked in excel combined components.xlsx)
C<- c(300, 1200)
pstar<- 0.01
rr<- c(3,1)
ppr<- c(0.2, 0.8)
comp1<- data.frame(id=1:100, rg=c(rep(1,50), rep(2,50)), cse=rep(0.5,100))
comp2<- data.frame(id=seq(2, 120, by=2), rg=c(rep(1,25), rep(2,35)), cse=runif(60, 0.5, 0.8))
comp3<- data.frame(id=seq(5, 120, by=5), rg=c(rep(1,10), rep(2,14)), cse=runif(24, 0.7, 1))
sep<- list(comp1, comp2, comp3)
sse.combined(C, pstar, rr, sep = sep)
sse.combined(C=NA, pstar, rr, ppr, sep = sep)

## End(Not run)
```

---

sse.rb.2stage	<i>calculate system sensitivity for 2 stage risk-based sampling</i>
---------------	---------------------------------------------------------------------

---

**Description**

calculate system sensitivity for 2 stage risk-based sampling

**Usage**

```
sse.rb.2stage(C = NA, pstar.c, pstar.u, rr.c, ppr.c, rr.u, ppr.u, N = NA, n,
  rg, se)
```

**Arguments**

C	Population size (number of clusters), NA = unknown
pstar.c	cluster level design prevalence
pstar.u	unit level design prevalence
rr.c	cluster level relative risks (vector), NA if no cluster level risk factor if risk factor does not apply at either level use rr = c(1,1)
ppr.c	cluster level population proportions for risk categories (vector), NA if no cluster level risk factor
rr.u	unit level relative risks (vector), NA if no unit level risk factor if risk factor does not apply at either level use rr = c(1,1)
ppr.u	matrix, 1 row for each cluster, columns = unit level risk groups
N	cluster sizes NA or matrix of N for each risk group for each cluster, N=NA means cluster sizes not provided)
n	matrix, 1 row for each cluster, columns = unit level risk groups
rg	vector of cluster level risk group for each cluster
se	unit sensitivity for each cluster, scalar or vector of values for each cluster

**Value**

list of 2 elements, a scalar of population-level (surveillance system) sensitivity and a vector of cluster-level sensitivities

**Examples**

```
## Not run:
# examples for sse.rb.2stage
pstar.c<- 0.02
pstar.u<- 0.1
rr.c<- c(5, 1)
ppr.c<- c(0.1, 0.9)
rr.u<- c(3, 1)
se<- 0.9
n<- cbind(rep(10, 50), rep(5, 50))
rg<- c(rep(1, 30), rep(2, 20))
ppr.u<- cbind(rep(0.2, 50), rep(0.8, 50))
N<- cbind(rep(30, 50), rep(120, 50))
```

```

C<- 500
sse.rb.2stage(C=NA, pstar.c, pstar.u, rr.c, ppr.c, rr.u, ppr.u, N=NA, n, rg, se)
sse.rb.2stage(C, pstar.c, pstar.u, rr.c, ppr.c, rr.u, ppr.u, N=NA, n, rg, se)
sse.rb.2stage(C=NA, pstar.c, pstar.u, rr.c, ppr.c, rr.u, ppr.u, N, n, rg, se)
sse.rb.2stage(C, pstar.c, pstar.u, rr.c, ppr.c, rr.u, ppr.u, N, n, rg, se)

## End(Not run)

```

---

tp	<i>estimate true prevalence</i>
----	---------------------------------

---

## Description

estimate true prevalence

## Usage

```
tp(x, n, se, sp, type = "blaker", conf = 0.95)
```

## Arguments

x	number of positive units (scalar)
n	sample size (no units sampled)
se	test unit sensitivity
sp	test unit specificity
type	one of c("normal", "c-p", "sterne", "blaker", "wilson", "all")
conf	desired level of confidence for CI, default = 0.95 note: all parameters must be scalars - won't work for vectors

## Value

list with 2 elements, a matrix of apparent prevalence and lower and upper confidence limits and a matrix of true prevalence and lower and upper confidence limits using the chosen method(s)

## Examples

```

## Not run:
# examples for tp
x<- 20
n<- 120
se<- 0.9
sp<- 0.99
conf<- 0.95
tp(x, n, se, sp, "all")
tp(x, n, se, sp, "c-p")
tp(x, n, 0.95, 0.9, "c-p")

## End(Not run)

```



---

tp.normal	<i>normal approximation CI for tp</i>
-----------	---------------------------------------

---

**Description**

normal approximation CI for tp

**Usage**

```
tp.normal(x, n, se, sp, conf = 0.95)
```

**Arguments**

x	number of positive results in sample
n	sample size
se	test unit sensitivity
sp	test unit specificity
conf	desired level of confidence for CI, default = 0.95

**Value**

list with 2 elements, a matrix of apparent prevalence and wilson lower and upper confidence limits and a matrix of true prevalence and normal approximation lower and upper confidence limits

**Examples**

```
## Not run:  
# examples for tp.normal  
tp.normal(25, 120, 0.9, 0.99)  
tp.normal(seq(5, 25, by=5), 120, 0.9, 0.99)  
  
## End(Not run)
```

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