# Package 'ssutil'

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Title Sample Size Calculation Tools

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Description Functions for sample size estimation and simulation in clinical trials. Includes methods for selecting the best group using the Indifference-zone approach, as well as designs for non-inferiority, equivalence, and negative binomial models. For the sample size calculation for non-inferiority of vaccines, the approach is based on Fleming, Powers, and Huang (2021) <doi:10.1177/1740774520988244>. The Indifference-zone approach is based on Sobel and Huyett (1957) <doi:10.1002/j.1538-7305.1957.tb02411.x> and Bechhofer, Santner, and Goldsman (1995, ISBN:978-0-471-57427-9).

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## Contents

empirical_power_result	2
format.power_single_rate	3
is.empirical_power_result	4
multp	4
multz	5
power best binomial	6
power best normal	7
power single rate	8
print empirical power result	8
print power single rate	9
pronbr	9
sim nower best binomial	10
sim_power_best_bin_rank	11
sim_power_best_pormal	12
sim_power_best_norm_rank	13
sim_power_dest_norm_rank	14
	13
sim_power_nbinom	1/
sim_power_n1_normal	18
ss_best_binomial	20
ss_best_normal	21
ss_ni_ve	22
tidy.empirical_power_result	23
wcs_power_best_binomial	24
	-25

## Index

empirical\_power\_result

Create an Empirical Power Result object

## Description

Constructs an S3 object of class empirical\_power\_result, storing the estimated power, its confidence interval, and the number of simulations used to compute it.

## Usage

empirical\_power\_result(x, n, conf.level = 0.95)

## Arguments

х	Number of successes
n	Number of trials.
conf.level	Confidence level for the returned confidence interval power.

## Details

```
It is a wrap to binom. test
```

#### Value

An object of class empirical\_power\_result, a list with components:

- power: Estimated power.
- conf.low: Lower bound of confidence interval.
- conf.high: Upper bound of confidence interval.
- conf.level: Confidence level for the returned confidence interval.
- nsim: Number of simulations.

## Examples

```
result <- empirical_power_result(
    x = 10,
    n = 100,
    conf.level = 0.95
)
print(result)</pre>
```

format.power\_single\_rate

```
Format method for power_single_rate class
```

## Description

Format method for power\_single\_rate class

## Usage

```
## S3 method for class 'power_single_rate'
format(x, digits = 3, ...)
```

#### Arguments

x	an R object of class power_single_rate
digits	a positive integer indicating how many significant digits are to be used for numeric x.
	further arguments passed to or from other methods

#### Value

A character string with a human-readable summary of the detection power or a markdown-style table, depending on the number of rows.

is.empirical\_power\_result

Check if an object is a sim\_power\_result

## Description

Check if an object is a sim\_power\_result

## Usage

is.empirical\_power\_result(x)

## Arguments

х

Any R object.

## Value

Logical. TRUE if x inherits from "sim\_power\_result".

multp

Calculate the Multivariate Normal Probability

## Description

Computes the multivariate normal probabilities with arbitrary correlation matrices It is the inverse of the multz function

## Usage

multp(q, k, rho, seed = NULL)

## Arguments

q	Numeric. Quantile of the distribution.
k	Integer. Number of variables in the multivariate normal distribution. Must be $>= 1$ .
rho	Numeric. Common correlation coefficient between variables (typically between 0 and 1).
seed	Optional. An object specifying if and how the random number generator should be initialized. Passed to pmvnorm.

#### Value

Numeric. The multivariate probability

## multz

#### Examples

```
q <- 1.3
k <- 3
rho <- 0.5
multp(q, k, rho)</pre>
```

Calculate the Upper Equicoordinate Point of a Multivariate Normal Distribution

## Description

Computes the upper equicoordinate quantile for a multivariate standard normal distribution with unit variances and a common correlation coefficient rho. That is, it returns the value z such that the joint probability  $P(X_1 \le z, ..., X_n \le z) = 1 - \alpha$ .

## Usage

multz(alpha, k, rho, seed = NULL)

## Arguments

alpha	Numeric. Significance level (e.g., 0.05 for a 95% confidence level).
k	Integer. Number of variables in the multivariate normal distribution. Must be $>= 1$ .
rho	Numeric. Common correlation coefficient between variables (typically between 0 and 1).
seed	Optional. An object specifying if and how the random number generator should be initialized. Passed to qmvnorm.

## Value

Numeric. The upper equicoordinate point z such that the joint probability of all variables being less than or equal to z is  $1 - \alpha$ .

## Examples

alpha <- 0.1 # Significance level (10%)
k <- 3 # Number of variables
rho <- 0.5 # Common correlation coefficient
multz(alpha, k, rho)</pre>

power\_best\_binomial Power to Correctly Select the Best Group in a Binomial Test

#### Description

Computes the exact probability of correctly identifying the best group when the outcome follows a binomial distribution. It assumes that p1 is the probability of success in the best group, and that the success probability in all other groups is lower by a fixed difference dif.

#### Usage

```
power_best_binomial(p1, dif, ngroups, npergroup)
```

## Arguments

р1	Numeric. Probability of success in the best group (must be in $[0, 1]$ ).
dif	Numeric. Difference in success probability between the best group and the next best (must be $> 0$ ).
ngroups	Integer. Number of groups (must be greater than 1).
npergroup	Integer. Number of subjects per group (must be positive).

## Details

The formula is based on the exact method described by Sobel and Huyett (1957).

## Value

A numeric value representing the probability of correctly identifying the best group.

#### References

Sobel, M., & Huyett, M. J. (1957). Selecting the Best One of Several Binomial Populations. *Bell System Technical Journal*, 36(2), 537–576. doi:10.1002/j.15387305.1957.tb02411.x

## Examples

power\_best\_binomial(p1 = 0.8, dif = 0.2, ngroups = 4, npergroup = 50)

power\_best\_normal

## Description

Estimate the probability of correctly select the best group among ngroups groups if the difference between the best group and the next best is at least dif(the Indifferent-Zone), and the standard deviation is sd

#### Usage

```
power_best_normal(dif, sd, ngroups, npergroup, seed = NULL)
```

## Arguments

dif	Numeric. Indifferent-zone. Minimum difference that is considered meaningful
sd	Numeric. Common standard deviation of the response variable.
ngroups	Integer. Number of groups (treatments) being compared.
npergroup	Integer. Number in each group.
seed	Optional. Integer seed to use in the internal call to multp().

## Value

Integer. Sample size required per group to achieve the specified power.

#### Note

The function uses the quantile function multp(), which computes critical values for the selection procedure. This implementation assumes equal variances and independent samples.

## Examples

```
power_best_normal(dif = 0.5, sd = 1, ngroups = 3, npergroup = 11)
```

power\_single\_rate Detectable Event Rate with Specified Power and Sample Size

#### Description

Estimates the minimum true proportion of events needed to detect at least one event, given a sample size and desired statistical power.

#### Usage

power\_single\_rate(subjects, power)

#### Arguments

subjects	Integer or vector of integers. Sample size(s).
power	Numeric or vector of numerics. Desired power(s), between 0 and 1.

#### Value

A matrix of class power\_single\_rate with columns:

**n** Sample size

power Requested power

proportion Minimum detectable event rate to observe at least one event

## Examples

```
power_single_rate(30, 0.9)
power_single_rate(c(30, 50, 100), 0.9)
```

print.empirical\_power\_result

Print method for empirical\_power\_result

## Description

Nicely formats the output of an object of class empirical\_power\_result, showing the power estimate, confidence interval, and number of simulations.

#### Usage

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

х	An object of class "empirical_power_result".
	Further arguments passed to or from other methods (ignored).

## Value

Invisibly returns the object passed in.

print.power\_single\_rate

Print method for class power\_single\_rate

## Description

Print method for class power\_single\_rate

## Usage

## S3 method for class 'power\_single\_rate'
print(x, ...)

## Arguments

Х	an object of class power_single_rate
	further arguments passed to or from other methods

## Value

Invisibly returns the object passed in.

prophr	Calculate Event Probability in the Experimental Group Given a Haz-
	ard Ratio

## Description

Computes the event probability in the experimental group based on the event probability in the control group and a specified hazard ratio, assuming proportional hazards.

#### Usage

prophr(p0, hr)

p0	Numeric scalar. Probability of an event in the control group (between 0 and 1).
hr	Numeric scalar. Hazard ratio (must be $> 0$ ).

## Details

This is useful for sample size calculations, for example in PASS (TM), which does not automatically adjust the event rate for the experimental group.

## Value

Numeric. The probability of an event in the experimental group.

#### Examples

prophr(0.05, 0.6)

sim\_power\_best\_binomial

Simulate Power to Select the Best Group Using Binomial Outcomes

#### Description

Estimates the empirical power to correctly identify the best group as having the highest outcome, under a binomial distribution. Assumes that the most promising group has a higher success probability than the others by at least dif, and that outcomes are independent.

## Usage

```
sim_power_best_binomial(
   noutcomes,
   p1,
   dif,
   ngroups,
   npergroup,
   nsim,
   conf.level = 0.95
)
```

## Arguments

noutcomes	Integer. Number of outcomes to evaluate.
р1	Numeric. Probability in the most promising group (scalar or vector).
dif	Numeric. Difference between the best group and the rest.
ngroups	Integer. Number of groups.

npergroup	Integer or vector. Number of subjects per group.
nsim	Integer. Number of simulations.
conf.level	Numeric. Confidence level for the empirical power estimate

#### Details

Multiple outcomes can be evaluated simultaneously. The power is estimated as the proportion of simulations where the most promising group is selected best in all outcomes.

## Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

## See Also

empirical\_power\_result

## Examples

```
sim_power_best_binomial(
    noutcomes = 1,
    p1 = 0.7,
    dif = 0.2,
    ngroups = 3,
    npergroup = 30,
    nsim = 1000
)
```

sim\_power\_best\_bin\_rank

Simulate Power to Rank the Best Group Using Binomial Outcomes

## Description

Estimates the empirical power to rank the most promising group as the best, based on binomial outcomes, via simulation.

#### Usage

```
sim_power_best_bin_rank(
    noutcomes,
    p1,
    dif,
    weights,
    ngroups,
    npergroup,
```

```
nsim,
conf.level = 0.95
)
```

noutcomes	Integer. Number of outcomes to evaluate.
p1	Numeric. Event probability in the best group (scalar or vector of length noutcomes).
dif	Numeric. Difference between the best group and the rest (scalar or vector of length noutcomes).
weights	Numeric vector. Weights for each outcome. If scalar, applied equally.
ngroups	Integer. Number of groups.
npergroup	Integer or vector. Sample size per group.
nsim	Integer. Number of simulations.
conf.level	Numeric. Confidence level for the empirical power estimate#'

#### Details

Each outcome is assumed to follow an independent binomial distribution. The best group is defined as having a probability at least dif higher than the other groups. The function sums weighted ranks across multiple outcomes to determine the top group.

If multiple outcomes are defined, weights can be applied to prioritize some outcomes over others. Weights are automatically scaled to sum 1. The group with the lowest total rank is considered the best.

#### Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

## See Also

empirical\_power\_result

#### Examples

```
sim_power_best_bin_rank(
noutcomes = 2,
p1 = 0.80,
dif = 0.15,
weights = 1,
ngroups = 3,
npergroup = 30,
nsim = 1000,
conf.level = 0.95)
```

12

sim\_power\_best\_normal Simulate Power to Select Best Group (Normal Outcomes)

## Description

Estimates the empirical power to identify the most promising group as the best, when outcomes are normally distributed and independent.

## Usage

```
sim_power_best_normal(
   noutcomes,
   sd,
   dif,
   ngroups,
   npergroup,
   nsim,
   conf.level = 0.95
)
```

#### Arguments

noutcomes	Integer. Number of outcomes to evaluate.
sd	Numeric vector. Standard deviations for each outcome. Can be a single value.
dif	Numeric vector. Difference in means between the best and the other groups.
ngroups	Number of groups to compare.
npergroup	Number of subjects per group. Can be scalar or vector of length ngroups.
nsim	Integer. Number of simulations to perform.
conf.level	Numeric. Confidence level for the empirical power estimate

### Details

The best group (group 1) is assumed to have mean 0, and the rest of the groups have mean -dif.

Multiple outcomes can be evaluated simultaneously. The power is estimated as the proportion of simulations where the most promising group is the best in all outcomes.

The number of subjects per group can be the same or specified per group. In either case, the first group is assumed to be the most promising.

#### Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

## See Also

empirical\_power\_result

## Examples

```
sim_power_best_normal(
    noutcomes = 2,
    sd = c(1, 1.2),
    dif = c(0.2, 0.25),
    ngroups = 3,
    npergroup = c(30, 25, 25),
    nsim = 1000
)
```

sim\_power\_best\_norm\_rank

Simulate Power to Select Best Group by Ranks (Normal Outcomes)

## Description

Estimates the empirical power to identify the most promising group as best, using weighted ranks across outcomes, assuming normally distributed outcomes.

#### Usage

```
sim_power_best_norm_rank(
   noutcomes,
   sd,
   dif,
   weights,
   ngroups,
   npergroup,
   nsim,
   conf.level = 0.95
)
```

## Arguments

noutcomes	Integer. Number of outcomes to evaluate.
sd	Numeric vector. Standard deviations for each outcome.
dif	Numeric vector. Difference in means between the best and other groups
weights	Numeric vector. Weights per outcome.
ngroups	Integer. Number of groups.
npergroup	Integer or vector. Number of subjects per group.
nsim	Integer. Number of simulations.
conf.level	Numeric. Confidence level for the empirical power estimate

14

#### Details

Each outcome is independent and normally distributed. The most promising group is assumed to have a mean at least dif higher than the others. Ranks are weighted and summed per group across outcomes.

If weights is specified, it is internally scaled to sum to 1. The most promising group is always considered to be the first group.

#### Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

#### See Also

empirical\_power\_result

## Examples

```
sim_power_best_norm_rank(
   noutcomes = 3,
   sd = c(1, 0.8, 1.5),
   dif = c(0.2, 0.15, 0.3),
   weights = c(0.5, 0.3, 0.2),
   ngroups = 3,
   npergroup = c(30, 25, 25),
   nsim = 1000
)
```

sim\_power\_equivalence\_normal

Empirical Power for Equivalence (Normal Outcomes)

### Description

Estimates the empirical power to detect equivalence among multiple groups assuming no true difference in normally distributed outcomes. Pairwise two-sample t-tests are used, and equivalence is declared if all confidence intervals for differences between group means lie entirely within the interval defined by llimit and ulimit.

## Usage

```
sim_power_equivalence_normal(
  ngroups,
  npergroup,
  sd,
  llimit,
```

```
ulimit,
nsim,
t_level = 0.95,
conf.level = 0.95
)
```

ngroups	Integer. Number of groups to compare
npergroup	Integer. Number of observations per group.
sd	Numeric. Standard deviation of the outcome distribution (common across groups).
llimit	Numeric. Lower equivalence limit.
ulimit	Numeric. Upper equivalence limit.
nsim	Integer. Number of simulations to perform.
t_level	Numeric. Confidence level used for the t-tests (e.g., 0.95 for 95% CI).
conf.level	Numeric. Confidence level for the empirical power estimate

## Details

This function simulates data under the null hypothesis of no difference between groups and calculates the proportion of simulations in which all pairwise comparisons fall within the specified equivalence limits.

## Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

## See Also

empirical\_power\_result

## Examples

```
#Equivalence testing for three groups with log-scale outcome
sim_power_equivalence_normal(
  ngroups = 3,
  npergroup = 172,
  sd = 0.403,
  llimit = log10(2/3),
  ulimit = log10(3/2),
  nsim = 1000,
  t_level = 0.95
)
```

## Description

Estimates empirical power to detect a relative risk either above or below a specified boundary, depending on the direction of the alternative hypothesis. Simulates count data with over dispersion, fits a model with glm.nb, and evaluates the power to reject the null hypothesis using a negative binomial model.

#### Usage

```
sim_power_nbinom(
    n1,
    n2,
    ir1,
    tm,
    rr,
    boundary,
    dispersion,
    alpha,
    nsim,
    conf.level = 0.95
)
```

#### Arguments

n1	Integer. Number of participants in group 1.
n2	Integer. Number of participants in group 2.
ir1	Numeric. Incidence rate in group 1.
tm	Numeric. Average exposure time per subject (assumed equal across subjects).
rr	Numeric. True relative risk between groups (group 2 rate = $rr \times group 1$ rate).
boundary	Numeric. Relative risk boundary under the null hypothesis.
dispersion	Numeric. Dispersion parameter ( $\phi$ ) for the negative binomial distribution.
alpha	Numeric. Type I error rate (two-sided).
nsim	Integer. Number of simulation iterations.
conf.level	Numeric. Confidence level for the empirical power estimate

## Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

Note

Uses the alternative parameterization of the negative binomial: mu is the mean, and size = 1/dispersion. In glm.nb, dispersion is estimated as theta. The 'boundary' parameter defines the relative risk under the null hypothesis. When rr < 1, rejection occurs if the upper limit of the confidence interval is below the boundary. When rr > 1, rejection occurs if the lower limit is above the boundary.

The alpha parameter is two-sided as it is used to estimate two-sided confidence intervals

#### Author(s)

Chris Gast

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#### See Also

empirical\_power\_result

## Examples

```
sim_power_nbinom(
    n1 = 150, n2 = 150,
    ir1 = 0.55, tm = 1.7,
    rr = 0.6, boundary = 1,
    dispersion = 2,
    alpha = 0.05,
    nsim = 1000
)
```

sim\_power\_ni\_normal Empirical Power for Non-Inferiority (Normal Outcomes)

## Description

Estimates empirical power to declare non-inferiority between two groups across multiple outcomes using t-tests. Simulates normally distributed data under the null (no difference) and applies non-inferiority rules based on user-defined required and optional tests.

## Usage

```
sim_power_ni_normal(
    nsim,
    npergroup,
    ntest,
    ni_limit,
    test_req,
    test_opt,
    sd,
```

18

```
corr = 0,
t_level = 0.95,
conf.level = 0.95
)
```

Integer. Number of simulations to perform.
Integer. Number of observations per group.
Integer. Number of tests (outcomes) to compare.
Numeric. Limit to declare non-inferiority. Can be a scalar or vector of length ntest.
Integer. Number of required tests that must show non-inferiority (first test_req tests).
Integer. Number of optional tests that must also show non-inferiority from the remaining tests.
Numeric. Standard deviation(s) of the outcomes. Scalar or vector of length <code>ntest</code> .
Numeric. Correlation between the tests. Scalar (common correlation), or vector of length ntest*(ntest-1)/2.
Numeric. Confidence level used for the t-tests (e.g., $0.95$ for $95\%$ CI).scalar or vector of length <code>ntest</code> .
Numeric. Confidence level for the empirical power estimate

#### Details

A test is considered non-inferior if the lower bound of its confidence interval is greater than the specified non-inferiority limit. Overall non-inferiority is declared if all test\_req and at least test\_opt of the remaining tests are non-inferior.

#### Value

An S3 object of class empirical\_power\_result, which contains the estimated empirical power and its confidence interval. The object can be printed, formatted, or further processed using associated S3 methods. See also empirical\_power\_result.

## Note

If only one test is used, correlation is ignored.

Use correlation 0 for independent outcomes

When using a correlation vector, it must match the number of test pairs: ntest\*(ntest-1)/2, in this order: (1,2), (1,3), ..., (1,ntest), (2,3), ..., (ntest-1,ntest).

The covariance matrix is derived from the correlation matrix and the standard deviations.

For example: with ntest = 3 and corr = c(0.2, 0.3, 0.4), the resulting correlation matrix is:

	[,1]	[,2]	[,3]
[1, ]	1	0.2	0.3
[2, ]	0.2.	1	0.4
[3, ]	0.3.	0.4	1

#### See Also

empirical\_power\_result

#### Examples

```
sim_power_ni_normal(
    nsim = 1000,
    npergroup = 250,
    ntest = 7,
    ni_limit = log10(2/3),
    test_req = 2,
    test_opt = 3,
    sd = 0.4,
    corr = 0,
    t_level = 0.05
)
```

ss\_best\_binomial Sample Size to Select the Best Group in a Binomial Test

## Description

Computes the minimum sample size per group required to achieve a target probability of correctly selecting the best group in a binomial test. The best group is assumed to have success probability p1, and the other groups have p1 - dif.

## Usage

```
ss_best_binomial(power, p1, dif, ngroups, max_n = 1000)
```

#### Arguments

power	Numeric. Desired probability of correctly selecting the best group (in $[0, 1]$ ).
p1	Numeric. Probability of success in the best group (in [0, 1]).
dif	Numeric. Difference in success probability with the next best group $(> 0)$ .
ngroups	Integer. Number of groups (must be $> 1$ ).
max_n	Integer. Maximum sample size to evaluate (default is 1000).

#### ss\_best\_normal

#### Details

The function searches for the smallest npergroup such that the power from power\_best\_binomial is at least the target power.

#### Value

An integer representing the minimum sample size per group required to reach the specified power.

## Examples

```
ss_best_binomial(power = 0.9, p1 = 0.8, dif = 0.2, ngroups = 4)
```

Sample Size for Selecting the Best Treatment in a Normal Response (Indifference-Zone)

## Description

Calculates the minimum common sample size per group needed to achieve a specified probability (power) of correctly selecting the best group using the indifference-zone approach. This method assumes normally distributed responses with a known and common standard deviation.

#### Usage

ss\_best\_normal(power, dif, sd, ngroups, seed = NULL)

#### Arguments

power	Numeric. Desired probability of correctly selecting the best group.
dif	Numeric. Indifferent-zone. Minimum difference that is considered meaningful
sd	Numeric. Common standard deviation of the response variable.
ngroups	Integer. Number of groups (treatments) being compared.
seed	Optional. Integer seed to use in the internal call to multz().

## Details

The indifference-zone approach guarantees that the probability of correct selection is at least power, assuming the best group's mean exceeds the others by at least dif. The calculation is based on Bechhofer's Procedure Nb.

#### Value

Integer. Sample size required per group to achieve the specified power.

22

The function uses the quantile function multz(), which computes critical values for the selection procedure. This implementation assumes equal variances and independent samples.

## References

Bechhofer, R.E., Santner, T.J., & Goldsman, D.M. (1995). *Design and Analysis of Experiments for Statistical Selection, Screening, and Multiple Comparisons*. Wiley Series in Probability and Statistics. ISBN: 0-471-57427-9.

## Examples

```
ss_best_normal( power = 0.8, dif = 0.5, sd = 1, ngroups = 3)
```

ss\_ni\_ve

Sample Size and Non-Inferiority Margin for Vaccine Efficacy Trials

#### Description

Computes the non-inferiority margin, number of events, and maximum hazard ratio (HR) to declare non-inferiority in vaccine efficacy (VE) trials, based on the approach described by Fleming et al. (2021).

#### Usage

ss\_ni\_ve(ve\_lci, alpha = 0.025, power = 0.9, use70 = FALSE, preserve = 0.5)

#### Arguments

ve_lci	Numeric. Lower bound of the current vaccine's efficacy (e.g., 0.95 for 95% VE).
alpha	Numeric. Type I error rate (default = $0.025$ ).
power	Numeric. Desired power for the test (default = $0.90$ ).
use70	Logical. If TRUE, assumes at least 30% VE for the new vaccine (the 90–70 rule); otherwise, preserves a fixed fraction of the reference VE.
preserve	Numeric. Proportion of the current vaccine's efficacy to preserve under $use70 = FALSE$ (default = 0.5).

#### Details

The method applies either the 95–95 rule or 90–70 rule, depending on whether a minimum VE of 30% is assumed (use70 = TRUE) or 50% of the current VE is preserved.

This implementation approximates Table 1 of the paper using exact binomial confidence intervals via binom.test and the nBinomial1Sample function from **gsDesign**.

#### Value

A named list with:

- Upper limit of the HR used to estimate the sample size: Hazard ratio corresponding to ve\_lci.
- Non-inferior margin in HR scale: Non-inferiority margin expressed as a hazard ratio.
- Alpha: The type I error used.
- Power: The power used.
- Total number of events: Total number of events required in the trial.
- Max HR to declare NI: Maximum observed hazard ratio that satisfies the non-inferiority criterion.
- Max number of events in the experimental group: Maximum number of events in the experimental group still compatible with non-inferiority.
- Non-inferior criteria: Description of the applied non-inferiority rule ("At least 30% VE" or "or preserved effect").

## References

Fleming, T.R., Powers, J.H., & Huang, Y. (2021). The use of active controls and non-inferiority studies in evaluating COVID-19 vaccines. *Clinical Trials*, 18(3), 335–342. doi:10.1177/1740774520988244

#### Examples

ss\_ni\_ve(ve\_lci = 0.95)

tidy.empirical\_power\_result Tidy Method for empirical power result

## Description

Creates a one-row tibble with the power estimate and confidence interval.

## Usage

## S3 method for class 'empirical\_power\_result'
tidy(x, ...)

## Arguments

х	A empirical_power_result object.
	Ignored.

#### Value

A tibble with columns: power, conf.low, conf.high, conf.level.nsim.

wcs\_power\_best\_binomial

Worst-Case Scenario Power for the Best Binomial Group

#### Description

Searches for the probability in the best-performing group that yields the lowest statistical power, given an indifference zone specification, a number of groups, and a number of subjects per group.

## Usage

wcs\_power\_best\_binomial(dif, ngroups, npergroup)

#### Arguments

dif	Numeric. Indifference zone specification (difference threshold)
ngroups	Integer. Number of groups to compare.
npergroup	Integer. Number of subjects per group.

#### Details

Defines an internal function fx that wraps power\_best\_binomial with the supplied parameters, then uses optimize over the interval [0,1] to find the probability p1 that minimizes the resulting power.

#### Value

A named list with components:

p1 Numeric. Probability in the best group that yields the minimum power.

minimum\_power Numeric. The minimum power achieved at p1.

#### See Also

power\_best\_binomial, optimize

#### Examples

```
wcs_power_best_binomial(dif = 0.1, ngroups = 3, npergroup = 50)
```

# Index

binom.test, 3 empirical\_power\_result, 2, 11-20 format.power\_single\_rate, 3 is.empirical\_power\_result,4 multp, 4 multz, 5 optimize, 24 pmvnorm, 4 power\_best\_binomial, 6, 21, 24 power\_best\_normal, 7 power\_single\_rate, 8 print.empirical\_power\_result, 8 print.power\_single\_rate,9 prophr, 9 qmvnorm, 5 sim\_power\_best\_bin\_rank, 11 sim\_power\_best\_binomial, 10 sim\_power\_best\_norm\_rank, 14 sim\_power\_best\_normal, 13 sim\_power\_equivalence\_normal, 15 sim\_power\_nbinom, 17 sim\_power\_ni\_normal, 18  $\texttt{ss\_best\_binomial, } 20$ ss\_best\_normal, 21 ss\_ni\_ve, 22 tidy.empirical\_power\_result, 23

wcs\_power\_best\_binomial, 24