

Package ‘BayesRepDesign’

July 21, 2025

Version 0.42

Date 2023-05-03

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Title Bayesian Design of Replication Studies

Description Provides functionality for determining the sample size of replication studies using Bayesian design approaches in the normal-normal hierarchical model (Pawel et al., 2022) <[doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)>.

License GPL-3

Encoding UTF-8

Imports methods, lamW

Suggests roxygen2, tinytest

RoxygenNote 7.2.3

NeedsCompilation no

URL <https://github.com/SamCH93/BayesRepDesign>

BugReports <https://github.com/SamCH93/BayesRepDesign/issues>

Repository CRAN

Date/Publication 2023-05-03 18:55:43 UTC

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density.designPrior *Density method for class "designPrior"*

Description

Density method for class "designPrior"

Usage

```
## S3 method for class 'designPrior'
density(x, ...)
```

Arguments

x Object of class "designPrior"
 ... Other arguments passed to stats::dnorm

Value

Returns the density function of the design prior

Author(s)

Samuel Pawel

Examples

```
dp <- designPrior(to = 2.3123, so = 0.1, mu = 1.1, tau = 0.2)
f <- density(dp)
tseq <- seq(1, 3.5, 0.01)
plot(tseq, f(theta = tseq), type = "l", xlab = "theta", ylab = "Design prior density")
```

designPrior

*Design prior for effect size***Description**

Creates a design prior for the effect size which can then be used for power and sample size calculations of a replication study. The design prior is obtained from updating an initial prior for the effect size by the data from the original study. A normal-normal hierarchical model is assumed, see Pawel et al. (2022) for details.

Usage

```
designPrior(
  to,
  so,
  mu = 0,
  sp = Inf,
  tau = 0,
  g = sp^2/(tau^2 + so^2),
  h = tau^2/so^2,
  type = c(NA, "conditional", "predictive", "EB")
)
```

Arguments

to	Effect estimate from original study
so	Standard error of effect estimate from original study
mu	The initial prior mean. Defaults to θ
sp	The initial prior standard deviation. Defaults to Inf (an improper uniform prior)
tau	The initial prior heterogeneity standard deviation. Defaults to θ (no heterogeneity)
g	The relative initial prior variance $g = sp^2/(tau^2 + so^2)$ (alternative parametrization of prior standard deviation sp)
h	The relative initial prior heterogeneity variance $h = tau^2/so^2$ (alternative parametrization of prior heterogeneity standard deviation tau)
type	Shortcut for special parameter combinations. The available options are NA, "conditional", "predictive", and "EB" (see details). Defaults to NA

Details

The "conditional" design prior corresponds to a point mass at the original effect estimate, i.e., assuming that the true effect size is equal to the original effect estimate. The "predictive" design prior is obtained from updating a uniform initial prior by the likelihood of the original data. The "EB" design prior is obtained by empirical Bayes estimation of the variance of the normal prior and induces adaptive shrinkage that depends on the p-value of the original effect estimate.

Value

Returns an object of class "designPrior" which is a list containing:

dpMean	The computed mean of the design prior
dpVar	The computed variance of the design prior
to	The specified original effect estimate
so	The specified original standard error
mu	The specified mean of the initial prior
sp	The specified standard deviation of the initial prior
tau	The specified heterogeneity variance

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

See Also

[pors](#), [ssd](#)

Examples

```
designPrior(to = 1.1, so = 1)
```

plot.designPrior *Plot method for class "designPrior"*

Description

Plot method for class "designPrior"

Usage

```
## S3 method for class 'designPrior'  
plot(x, ...)
```

Arguments

x Object of class "designPrior"
... Other arguments passed to plot

Value

Plots the density of the design prior

Author(s)

Samuel Pawel

Examples

```
dp <- designPrior(to = 2.3123, so = 0.1, mu = 1.1, tau = 0.2)  
plot(dp)  
plot(dp, xlim = c(0, 5), length.out = 500)
```

pors *Compute probability of replication success*

Description

This function computes the probability of replication success based on a success region for the replication effect estimate, a design prior, and a replication standard error. If the specified number of sites is larger than 1, the supplied success region has to be formulated in terms of the meta-analytic replication effect estimate across sites.

Usage

```
pors(sregion, dprior, sr, nsites = 1)
```

Arguments

sregion	Success region for replication effect estimate
dprior	Design prior object
sr	Standard error of replication effect estimate
nsites	Number of sites, defaults to 1. The sites are assumed to have the same standard error sr

Value

The probability of replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:[10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
dprior <- designPrior(to = 1.1, so = 1)
sregion <- successRegion(intervals = cbind(1.96, Inf))
pors(sregion = sregion, dprior = dprior, sr = 1)
```

porsBF01

Probability of replication success based on Bayes factor

Description

This function computes the probability to achieve replication success based on a Bayes factor. The Bayes factor is oriented so that values above one indicate evidence for the null hypothesis of the effect size being zero, whereas values below one indicate evidence for the hypothesis of the effect size being non-zero (with normal prior assigned to it).

Usage

```
porsBF01(level, dprior, sr, priormean = 0, priorvar = 1)
```

Arguments

level	Bayes factor level below which replication success is achieved
dprior	Design prior object
sr	Replication standard error
priormean	Mean of the normal prior under the alternative. Defaults to 0
priorvar	Variance of the normal prior under the alternative. Defaults to 1

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:[10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
## specify design prior
to1 <- 2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
porsBF01(level = 1/10, dprior = dprior, sr = c(0.05, 0.04))
```

porsBFr

Probability of replication success based on replication Bayes factor

Description

This function computes the probability to achieve replication success based on the replication Bayes factor. The replication Bayes factor is assumed to be oriented so that values below one indicate replication success, whereas values above one indicate evidence for the null hypothesis.

Usage

```
porsBFr(level, dprior, sr, paradox = TRUE)
```

Arguments

level	Bayes factor level below which replication success is achieved
dprior	Design prior object
sr	Replication standard error
paradox	Should the probability of replication success be computed allowing for the replication paradox (replication success when the effect estimates from original and replication study have a different sign)? Defaults to TRUE

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Verhagen, J. and Wagenmakers, E. J. (2014). Bayesian tests to quantify the result of a replication attempt. *Journal of Experimental Psychology: General*, 145:1457-1475. doi:10.1037/a0036731

Ly, A., Etz, A., Marsman, M., & Wagenmakers, E.-J. (2018). Replication Bayes factors from evidence updating. *Behavior Research Methods*, 51(6), 2498-2508. doi:10.3758/s134280181092x

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
porsBFR(level = 1/10, dprior = dprior, sr = c(0.05, 0.04))
```

porsBFs

Probability of replication success based on the sceptical Bayes factor

Description

This function computes the probability to achieve replication success based on the sceptical Bayes factor. The sceptical Bayes factor is assumed to be oriented so that values below one indicate replication success.

Usage

```
porsBFs(level, dprior, sr, paradox = TRUE)
```

Arguments

level	Threshold for the sceptical Bayes factor below which replication success is achieved
dprior	Design prior object
sr	Replication standard error
paradox	Should the probability of replication success be computed allowing for the replication paradox (replication success when the effect estimates from original and replication study have a different sign)? Defaults to TRUE

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Pawel, S. and Held, L. (2020). The sceptical Bayes factor for the assesment of replication success. Journal of the Royal Statistical Society: Series B (Statistical Methodology), 84(3), 879-911. doi:10.1111/rssb.12491

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1)
porsBFs(level = 1/3, dprior = dprior, sr = 0.05)
```

porsEqu

Probability of replication success based on effect size equivalence

Description

This function computes the probability to achieve replication success on equivalence of original and replication effect size. Effect size equivalence is defined by the confidence interval for the difference between the original and replication effect sizes falling within an equivalence region around zero defined by the specified margin.

Usage

```
porsEqu(level, dprior, margin, sr)
```

Arguments

level	1 - confidence level of confidence interval for effect size difference
dprior	Design prior object
margin	The equivalence margin > 0 for the symmetric equivalence region around zero
sr	Replication standard error

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Anderson, S. F. and Maxwell, S. E. (2016). There's more than one way to conduct a replication study: Beyond statistical significance. Psychological Methods, 21(1), 1-12. doi:10.1037/met0000051

Examples

```
## specify design prior
to1 <- 2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.1)
porsEqu(level = 0.1, dprior = dprior, margin = 0.3, sr = c(0.05, 0.03))
```

porsMeta

Probability of replication success based on meta-analytic significance

Description

This function computes the probability to achieve replication success on statistical significance of the fixed-effects meta-analytic effect estimate obtained from combining original and replication effect estimates.

Usage

```
porsMeta(level, dprior, sr)
```

Arguments

level	Significance level for p-value of the meta-analytic effect estimate (one-sided and in the same direction as the original effect estimate)
dprior	Design prior object
sr	Replication standard error

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Examples

```
## specify design prior
to1 <- 2
so1 <- 1
dprior <- designPrior(to = to1, so = so1, tau = 0.1)
porsMeta(level = 0.025^2, dprior = dprior, sr = c(0.2, 0.1))
```

porsPs

Probability of replication success based on the sceptical p-value

Description

This function computes the probability to achieve replication success based on the sceptical p-value.

Usage

```
porsPs(level, dprior, sr)
```

Arguments

level	Threshold for the (one-sided) sceptical p-value below which replication success is achieved
dprior	Design prior object
sr	Replication standard error

Details

The sceptical p-value is assumed to be uncalibrated as in Held (2020). The package `ReplicationSuccess` allows for sample size and power calculations with the recalibrated sceptical p-value (<https://CRAN.R-project.org/package=ReplicationSuccess>).

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Held, L. (2020). A new standard for the analysis and design of replication studies (with discussion). Journal of the Royal Statistical Society: Series A (Statistics in Society), 183(2), 431-448. doi:10.1111/rssa.12493

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1)
porsPs(level = 0.025, dprior = dprior, sr = c(0.05, 0.01))
```

porsSig

Probability of replication success based on significance

Description

This function computes the probability to achieve replication success on statistical significance of the replication effect estimate.

Usage

```
porsSig(level, dprior, sr)
```

Arguments

level	Significance level for p-value of the replication effect estimate (one-sided and in the same direction as the original effect estimate)
dprior	Design prior object
sr	Replication standard error

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Examples

```
## specify design prior
to1 <- 2
so1 <- 1
dprior <- designPrior(to = to1, so = so1, tau = 0.1)
porsSig(level = 0.025, dprior = dprior, sr = c(0.5, 0.3))
```

porsTOST

Probability of replication success based on TOST equivalence

Description

This function computes the probability to achieve replication success based on establishing the absence of a practically relevant effect size with the Two One-Sided Tests (TOST) procedure in the replication study.

Usage

```
porsTOST(level, dprior, margin, sr)
```

Arguments

level	Significance level for the TOST p-value
dprior	Design prior object
margin	The equivalence margin > 0 for the equivalence region around zero that defines a region of practically irrelevant effect sizes
sr	Replication standard error

Value

The probability to achieve replication success

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Anderson, S. F. and Maxwell, S. E. (2016). There's more than one way to conduct a replication study: Beyond statistical significance. Psychological Methods, 21(1), 1-12. doi:10.1037/met0000051

Examples

```
## specify design prior
to1 <- 2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.1)
porSTOST(level = 0.1, dprior = dprior, margin = 0.3, sr = c(0.05, 0.03))
```

`print.designPrior` *Print method for class "designPrior"*

Description

Print method for class "designPrior"

Usage

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

Arguments

`x` Object of class "designPrior"
`...` Other arguments (for consistency with the generic)

Value

Prints text summary in the console and invisibly returns the "designPrior" object

Author(s)

Samuel Pawel

Examples

```
dp <- designPrior(to = 0.5, so = 0.05, sp = 0.2, tau = 0.1)
print(dp)
```

print.ssdRS	<i>Print method for class "ssdRS"</i>
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Description

Print method for class "ssdRS"

Usage

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

Arguments

x	Object of class "ssdRS"
...	Other arguments (for consistency with the generic)

Value

Prints text summary in the console and invisibly returns the "ssdRS" object

Author(s)

Samuel Pawel

Examples

```
## specify design prior  
to1 <- 2  
so1 <- 1  
dprior <- designPrior(to = to1, so = so1)  
  
## compute required standard error for significance at one-sided 2.5%  
sregionfunSig <- function(sr, alpha = 0.025) {  
  successRegion(intervals = cbind(stats::qnorm(p = 1- alpha)*sr, Inf))  
}  
ssd1 <- ssd(sregionfun = sregionfunSig, dprior = dprior, power = 0.8)  
print(ssd1)
```

```
print.successRegion Print method for class "successRegion"
```

Description

Print method for class "successRegion"

Usage

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

Arguments

x	Object of class "successRegion"
...	Other arguments

Value

Prints text summary in the console and invisibly returns the "successRegion" object

Author(s)

Samuel Pawel

Examples

```
## success region for two-sided significance test  
successRegion(intervals = rbind(c(1.96, Inf), c(-Inf, -1.96)))  
## success region for one-sided significance test  
successRegion(intervals = rbind(c(1.96, Inf)))
```

se2n

Sample size related to standard error and unit standard deviation

Description

This function computes the sample size related to a specified standard error σ and unit standard deviation `unitSD`, which is the standard deviation of one effective unit (one measurement, one pair of measurements, one event, etc.). The relationship $\sigma = \text{unitSD}/\sqrt{n}$ is assumed. The unit standard deviation depends on the parameter type and the assumptions underlying the standard error calculation. The default is `unitSD = 2` which is, under some assumptions, a reasonable approximation to the unit standard deviation for standardized mean differences and log odds/hazard/rate ratios, see Section 2.4 in Spiegelhalter et al. (2004).

Usage

```
se2n(se, unitSD = 2)
```

Arguments

se	Standard error
unitSD	Unit standard deviation. Defaults to 2

Value

The sample size corresponding to the specified standard error and unit standard deviation

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:[10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Spiegelhalter, D.J., Abrams, K.R., Myles, J.P. (2004). Bayesian approaches to clinical trials and health care evaluation. Wiley. doi:[10.1002/0470092602](https://doi.org/10.1002/0470092602)

Examples

```
smd1 <- 0.3
so1 <- 0.05
dprior <- designPrior(to = smd1, so = so1)
ssd1 <- ssdSig(level = 0.025, dprior = dprior, power = 0.8)
se2n(se = ssd1$sr, unitSD = 2) # required n
```

ssd

Sample size determination for replication success

Description

This function computes the standard error of the replication effect estimate required to achieve replication success with a certain probability and based on a certain type of success region.

Usage

```

ssd(
  sregionfun,
  dprior,
  power,
  nsites = 1,
  searchInt = c(.Machine$double.eps^0.5, 4),
  ...
)

```

Arguments

sregionfun	Function that returns the success region for replication effect estimate as a function of the replication standard error
dprior	Design prior object
power	Desired probability of replication success
nsites	Number of sites. Defaults to 1. The sites are assumed to have the same sample size
searchInt	Search interval for standard errors
...	Other arguments passed to uniroot

Value

Returns an object of class "ssdRS" which is a list containing:

designPrior	The specified "designPrior" object
power	The specified power
powerRecomputed	The recomputed power
sr	The required replication standard error
c	The required relative sample size $c = nr/no$ (assuming $so = unitSD/no$ and $sr = unitSD/nr$)

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:[10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
## specify design prior
to1 <- 2
so1 <- 1
dprior <- designPrior(to = to1, so = so1)

## compute required standard error for significance at one-sided 2.5%
sregionfunSig <- function(sr, alpha = 0.025) {
  successRegion(intervals = cbind(stats::qnorm(p = 1- alpha)*sr, Inf))
}
ssd(sregionfun = sregionfunSig, dprior = dprior, power = 0.8)
```

ssdBf01

Sample size determination for replication success based on Bayes factor

Description

This function computes the standard error required to achieve replication success with a certain probability and based on the Bayes factor under normality. The Bayes factor is oriented so that values above one indicate evidence for the null hypothesis of the effect size being zero, whereas values below one indicate evidence for the hypothesis of the effect size being non-zero (with normal prior assigned to it).

Usage

```
ssdBf01(
  level,
  dprior,
  power,
  priormean = 0,
  priorvar = 1,
  searchInt = c(.Machine$double.eps^0.5, 2)
)
```

Arguments

level	Bayes factor level below which replication success is achieved
dprior	Design prior object
power	Desired probability of replication success
priormean	Mean of the normal prior under the alternative. Defaults to 0
priorvar	Variance of the normal prior under the alternative. Defaults to 1
searchInt	Interval for numerical search over replication standard errors

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
ssdBFr01(level = 1/10, dprior = dprior, power = 0.8)
```

ssdBFr

Sample size determination for replication success based on replication Bayes factor

Description

This function computes the standard error required to achieve replication success with a certain probability and based on the replication Bayes factor under normality. The replication Bayes factor is assumed to be oriented so that values below one indicate replication success, whereas values above one indicate evidence for the null hypothesis.

Usage

```
ssdBFr(
  level,
  dprior,
  power,
  searchInt = c(.Machine$double.eps^0.5, 2),
  paradox = TRUE
)
```

Arguments

level	Bayes factor level below which replication success is achieved
dprior	Design prior object
power	Desired probability of replication success

searchInt	Interval for numerical search over replication standard errors
paradox	Should the probability of replication success be computed allowing for the replication paradox (replication success when the effect estimates from original and replication study have a different sign)? Defaults to TRUE

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

- Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)
- Verhagen, J. and Wagenmakers, E. J. (2014). Bayesian tests to quantify the result of a replication attempt. *Journal of Experimental Psychology: General*, 145:1457-1475. [doi:10.1037/a0036731](https://doi.org/10.1037/a0036731)
- Ly, A., Etz, A., Marsman, M., and Wagenmakers, E.-J. (2018). Replication Bayes factors from evidence updating. *Behavior Research Methods*, 51(6), 2498-2508. [doi:10.3758/s134280181092x](https://doi.org/10.3758/s134280181092x)

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
ssdBFr(level = 1/10, dprior = dprior, power = 0.8)
```

ssdBFs	<i>Sample size determination for replication success based on the sceptical Bayes factor</i>
--------	--

Description

This function computes the standard error required to achieve replication success with a certain probability and based on the sceptical Bayes factor. The sceptical Bayes factor is assumed to be oriented so that values below one indicate replication success.

Usage

```
ssdBFs(
  level,
  dprior,
  power,
  searchInt = c(.Machine$double.eps^0.5, 2),
  paradox = TRUE
)
```

Arguments

level	Threshold for the sceptical Bayes factor below which replication success is achieved
dprior	Design prior object
power	Desired probability of replication success
searchInt	Interval for numerical search over replication standard errors
paradox	Should the probability of replication success be computed allowing for the replication paradox (replication success when the effect estimates from original and replication study have a different sign)? Defaults to TRUE

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Pawel, S. and Held, L. (2020). The sceptical Bayes factor for the assesment of replication success. *Journal of the Royal Statistical Society: Series B (Statistical Methodology)*, 84(3), 879-911. [doi:10.1111/rssb.12491](https://doi.org/10.1111/rssb.12491)

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
ssdBFs(level = 1/10, dprior = dprior, power = 0.9)
```

ssdEqu	<i>Sample size determination for replication success based on effect size equivalence</i>
--------	---

Description

This function computes the standard error required to achieve replication success with a certain probability and based on effect size equivalence of original and replication effect size. Effect size equivalence is defined by the confidence interval for the difference between the original and replication effect sizes falling within an equivalence region around zero defined by the specified margin.

Usage

```
ssdEqu(level, dprior, power, margin, searchInt = c(0, 2))
```

Arguments

level	1 - confidence level of confidence interval for effect size difference
dprior	Design prior object
power	Desired probability of replication success
margin	The equivalence margin > 0 for the symmetric equivalence region around zero
searchInt	Interval for numerical search over replication standard errors

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Anderson, S. F. and Maxwell, S. E. (2016). There's more than one way to conduct a replication study: Beyond statistical significance. *Psychological Methods*, 21(1), 1-12. [doi:10.1037/met0000051](https://doi.org/10.1037/met0000051)

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.05)
ssdEqu(level = 0.1, dprior = dprior, power = 0.8, margin = 0.2)
```

ssdMeta	<i>Sample size determination for replication success based on meta-analytic significance</i>
---------	--

Description

This function computes the standard error required to achieve replication success with a certain probability and based on statistical significance of the fixed-effects meta-analytic effect estimate obtained from combining original and replication effect estimates.

Usage

```
ssdMeta(level, dprior, power, searchInt = c(0, 10))
```

Arguments

level	Significance level for the replication effect estimate (one-sided and in the same direction as the original effect estimate)
dprior	Design prior object
power	Desired probability of replication success
searchInt	Interval for numerical search over replication standard errors

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
## specify design prior
to1 <- 2
so1 <- 1
dprior <- designPrior(to = to1, so = so1, tau = 0.25, sp = Inf)
ssdMeta(level = 0.025^2, dprior = dprior, power = 0.95)
```

ssdPs	<i>Sample size determination for replication success based on the sceptical p-value</i>
-------	---

Description

This function computes the standard error required to achieve replication success with a certain probability and based on the sceptical p-value.

Usage

```
ssdPs(level, dprior, power)
```

Arguments

level	Threshold for the (one-sided) sceptical p-value below which replication success is achieved
dprior	Design prior object
power	Desired probability of replication success

Details

The sceptical p-value is assumed to be uncalibrated as in Held (2020). The package `ReplicationSuccess` allows for sample size and power calculations with the recalibrated sceptical p-value (<https://CRAN.R-project.org/package=ReplicationSuccess>).

Value

Returns an object of class "ssdRS". See `ssd` for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Held, L. (2020). A new standard for the analysis and design of replication studies (with discussion). *Journal of the Royal Statistical Society: Series A (Statistics in Society)*, 183(2), 431-448. doi:10.1111/rssa.12493

Examples

```
## specify design prior
to1 <- 0.2
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.03)
ssdPs(level = 0.05, dprior = dprior, power = 0.9)
```

ssdSig	<i>Sample size determination for replication success based on significance</i>
--------	--

Description

This function computes the standard error required to achieve replication success with a certain probability and based on statistical significance of the replication effect estimate.

Usage

```
ssdSig(level, dprior, power)
```

Arguments

level	Significance level for the replication effect estimate (one-sided and in the same direction as the original effect estimate)
dprior	Design prior object
power	Desired probability of replication success

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Examples

```
## specify design prior
to1 <- 2
so1 <- 0.5
dprior <- designPrior(to = to1, so = so1, tau = 0.1)
ssdSig(level = 0.025, dprior = dprior, power = 0.9)
```

ssdTOST	<i>Sample size determination for replication success based on TOST equivalence</i>
---------	--

Description

This function computes the standard error required to achieve replication success with a certain probability and based on establishing the absence of a practically relevant effect size with the Two One-Sided Tests (TOST) procedure in the replication study.

Usage

```
ssdTOST(level, dprior, power, margin, searchInt = c(0, 2))
```

Arguments

level	Significance level for the TOST p-value
dprior	Design prior object
power	Desired probability of replication success
margin	The equivalence margin > 0 for the equivalence region around zero that defines a region of practically irrelevant effect sizes
searchInt	Interval for numerical search over replication standard errors

Value

Returns an object of class "ssdRS". See [ssd](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

Anderson, S. F. and Maxwell, S. E. (2016). There's more than one way to conduct a replication study: Beyond statistical significance. *Psychological Methods*, 21(1), 1-12. [doi:10.1037/met0000051](https://doi.org/10.1037/met0000051)

Examples

```
## specify design prior
to1 <- 0.05
so1 <- 0.05
dprior <- designPrior(to = to1, so = so1, tau = 0.05)
ssdTOST(level = 0.05, dprior = dprior, power = 0.9, margin = 0.3)
```

successRegion	<i>Success region for replication effect estimate</i>
---------------	---

Description

Creates a success region object which can then be used for computing the probability of replication success with [pors](#).

Usage

```
successRegion(intervals)
```

Arguments

`intervals` A 2xN matrix containing N disjoint intervals, the first column containing the lower and the second column containing the upper limits

Value

Returns an object of class "successRegion" which is a matrix containing the success intervals sorted in ascending order

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. [doi:10.48550/arXiv.2211.02552](https://doi.org/10.48550/arXiv.2211.02552)

See Also

[pors](#), [ssd](#)

Examples

```
successRegion(intervals = rbind(c(1.96, Inf), c(-Inf, -1.96)))  
successRegion(intervals = cbind(1.96, Inf))
```

successRegionSig	<i>Success region based on significance</i>
------------------	---

Description

This function returns the success region for the (meta-analytic) replication effect estimate to achieve significance

Usage

```
successRegionSig(sr, to, tau = 0, nsites = 1, level)
```

Arguments

sr	Replication standard error
to	Original effect estimate
tau	Heterogeneity standard deviation used in the calculation of the meta-analytic replication effect estimate and its standard error. Defaults to 0 (fixed effects analysis)
nsites	nsites Number of sites, defaults to 1. The effect estimates from all sites are assumed to have the same standard error sr
level	Significance level for p-value of the (average) replication effect estimate (one-sided and in the same direction as the original effect estimate)

Value

An object of class "successRegion". See [successRegion](#) for details.

Author(s)

Samuel Pawel

References

Pawel, S., Consonni, G., and Held, L. (2022). Bayesian approaches to designing replication studies. arXiv preprint. doi:10.48550/arXiv.2211.02552

Examples

```
successRegionSig(sr = 0.05, to = 0.2, tau = 0.01, nsites = 3, level = 0.025)
```

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