

Package ‘MetaUtility’

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Type Package

Title Utility Functions for Conducting and Reporting Meta-Analysis

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Description Contains functions to estimate the proportion of effects stronger than a threshold of scientific importance (per Mathur & VanderWeele, 2018 [<https://onlinelibrary.wiley.com/doi/full/10.1002/sim.8057>]), to make various effect size conversions, to compute and format inference in a meta-analysis, and to scrape results from existing meta-analyses for re-analysis.

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format_CI	<i>Manuscript-friendly confidence interval formatting</i>
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Description

Formats confidence interval lower and upper bounds into a rounded string.

Usage

```
format_CI(lo, hi, digits = 2)
```

Arguments

lo	Confidence interval lower limit (numeric)
hi	Confidence interval upper limit (numeric)
digits	Digits for rounding

Examples

```
format_CI(0.36, 0.72, 3)
```

format_stat	<i>Manuscript-friendly number formatting</i>
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Description

Formats a numeric result (e.g., p-value) as a manuscript-friendly string in which values below a minimum cutoff (e.g., 10^{-5}) are reported for example as " $< 10^{-5}$ ", values between the minimum cutoff and a maximum cutoff (e.g., 0.01) are reported in scientific notation, and p-values above the maximum cutoff are reported simply as, for example, 0.72.

Usage

```
format_stat(x, digits = 2, cutoffs = c(0.01, 10^-5))
```

Arguments

x	Numeric value to format
digits	Digits for rounding
cutoffs	A vector containing the two cutoffs

Examples

```
format_stat(0.735253)
format_stat(0.735253, digits = 4)
format_stat(0.0123)
format_stat(0.0001626)
format_stat(0.0001626, cutoffs = c(0.01, 10^-3))
```

parse_CI_string	<i>Parse a string with point estimate and confidence interval</i>
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Description

Given a vector of strings such as "0.65 (0.6, 0.70)", for example obtained by running optical character recognition (OCR) software on a screenshot of a published forest plot, parses the strings into a dataframe of point estimates and upper confidence interval limits. Assumes that the point estimate occurs before an opening bracket of the form "(" or "[" and that the confidence interval upper limit follows a the character sep (by default a comma, but might be a hyphen, for example). To further parse this dataframe into point estimates and variances, see `MetaUtility::scrape_meta`.

Usage

```
parse_CI_string(string, sep = ",")
```

Arguments

string	A vector of strings to be parsed.
sep	The character (not including whitespaces) separating the lower from the upper limits.

Examples

```
# messy string of confidence intervals
mystring = c("0.65 [0.6, 0.7]", "0.8(0.5, 0.9)", "1.2 [0.3, 1.5]")
parse_CI_string(mystring)

# now with a hyphen separator
mystring = c("0.65 [0.6- 0.7]", "0.8(0.5 - 0.9)", "1.2 [0.3 -1.5]")
parse_CI_string(mystring, sep="-")
```

prop_stronger *Estimate proportion of true effect sizes above or below a threshold*

Description

Estimates the proportion of true (i.e., population parameter) effect sizes in a meta-analysis that are above or below a specified threshold of scientific importance.

Usage

```
prop_stronger(q, M, t2, se.M = NA, se.t2 = NA, CI.level = 0.95,
  tail = NA, dat = NULL, R = 2000, bootstrap = "ifneeded",
  yi.name = "yi", vi.name = "vi")
```

Arguments

q	True effect size that is the threshold for "scientific importance"
M	Pooled point estimate from meta-analysis
t2	Estimated heterogeneity (τ^2) from meta-analysis
se.M	Estimated standard error of pooled point estimate from meta-analysis
se.t2	Estimated standard error of τ^2 from meta-analysis
CI.level	Confidence level as a proportion
tail	above for the proportion of effects above q; below for the proportion of effects below q.
dat	Dataset of point estimates (with names equal to the passed yi.name) and their variances (with names equal to the passed vi.name). Only required when bootstrapping.
R	Number of bootstrap iterates. Only required when bootstrapping.
bootstrap	If equal to ifneeded, bootstraps if estimated proportion is less than 0.15 or more than 0.85. If equal to never, instead does not return inference in the above edge cases. Only required when bootstrapping.
yi.name	Name of the variable in dat containing the study-level point estimates. Only required when bootstrapping.
vi.name	Name of the variable in dat containing the study-level variances. Only required when bootstrapping.

Details

When the estimated proportion is less than 0.15 or more than 0.85, it is best to bootstrap the confidence interval using the bias-corrected and accelerated (BCa) method (Mathur & VanderWeele, 2018); this is the default behavior of prop_stronger. Sometimes BCa confidence interval estimation fails, in which case prop_stronger instead uses the percentile method, issuing a warning if this is the case. We use a modified "safe" version of the boot package code for bootstrapping such that if any bootstrap iterates fail (usually because of model estimation problems), the error message is printed but the bootstrap iterate is simply discarded so that confidence interval estimation can proceed.

Value

Returns a dataframe containing the point estimate for the proportion, its estimated standard error, and confidence interval limits.

References

1. Mathur MB & VanderWeele TJ. New metrics for meta-analyses of heterogeneous effects. *Statistics in Medicine* (2018).
2. Mathur MB & VanderWeele TJ. New metrics for multisite replication projects. Under review.

Examples

```
##### Example 1: BCG Vaccine and Tuberculosis Meta-Analysis #####

# calculate effect sizes for example dataset
d = metafor::escalc(measure="RR", ai=tpos, bi=tneg,
                   ci=cpos, di=cneg, data=metafor::dat.bcg)

# fit random-effects model
# note that metafor package returns on the log scale
m = metafor::rma.uni(yi= d$yi, vi=d$vi, knha=TRUE,
                    measure="RR", method="REML" )

# pooled point estimate (RR scale)
exp(m$b)

# estimate the proportion of effects stronger than RR = 0.80
# no bootstrapping will be needed
prop_stronger( q = log(0.8),
              M = as.numeric(m$b),
              t2 = m$tau2,
              se.M = as.numeric(m$vb),
              se.t2 = m$se.tau2,
              CI.level = 0.95,
              tail = "below",
              bootstrap = "ifneeded")

## Not run:
# now try a more extreme threshold, q, such that the function will use bootstrapping
# now we will need to pass the final 4 arguments as well
prop_stronger( q = log(0.9),
              M = as.numeric(m$b),
              t2 = m$tau2,
              se.M = as.numeric(m$vb),
              se.t2 = m$se.tau2,
              CI.level = 0.95,
              tail = "below",

              # below arguments control bootstrapping
              # only 100 iterates for demo purposes (should be higher in practice)
```

```

        dat = d,
        R = 100,
        bootstrap = "ifneeded",
        yi.name = "yi",
        vi.name = "vi" )

## End(Not run)

##### Example 2: Meta-Analysis of Multisite Replication Studies #####

# replication estimates (Fisher's z scale) and SEs
# from moral credential example in reference #2
r.fis = c(0.303, 0.078, 0.113, -0.055, 0.056, 0.073,
0.263, 0.056, 0.002, -0.106, 0.09, 0.024, 0.069, 0.074,
0.107, 0.01, -0.089, -0.187, 0.265, 0.076, 0.082)

r.SE = c(0.111, 0.092, 0.156, 0.106, 0.105, 0.057,
0.091, 0.089, 0.081, 0.1, 0.093, 0.086, 0.076,
0.094, 0.065, 0.087, 0.108, 0.114, 0.073, 0.105, 0.04)

d = data.frame( yi = r.fis,
                vi = r.SE^2 )

# meta-analyze the replications
m = metafor::rma.uni( yi = r.fis, vi = r.SE^2, measure = "ZCOR" )

# probability of true effect above r = 0.10 = 28%
# convert threshold on r scale to Fisher's z
q = r_to_z(0.10)

# bootstrap reps should be higher in practice (e.g., 1000)
# here using only 100 for speed
prop_stronger( q = q,
               M = m$b,
               se.M = m$se,
               t2 = m$tau2,
               se.t2 = m$se.tau2,
               tail = "above",
               dat = d,
               R = 250 )

# probability of true effect equally strong in opposite direction
q.star = r_to_z(-0.10)
prop_stronger( q = q.star,
               M = m$b,
               se.M = m$se,
               t2 = m$tau2,
               se.t2 = m$se.tau2,
               tail = "below",
               dat = d,
               R = 250 )

```

round2	<i>Round while keeping trailing zeroes</i>
--------	--

Description

Rounds a numeric value and formats it as a string, keeping trailing zeroes.

Usage

```
round2(x, digits = 2)
```

Arguments

x	Numeric value to round
digits	Digits for rounding

Examples

```
round2(0.03000, digits = 4)

# compare to base round, which drops trailing zeroes and returns a numeric
round(0.03000, digits = 4)
```

r_to_d	<i>Convert Pearson's r to Cohen's d</i>
--------	---

Description

Converts Pearson's r (computed with a continuous X and Y) to Cohen's d for use in meta-analysis. The resulting Cohen's d represents the estimated increase in standardized Y that is associated with a delta-unit increase in X .

Usage

```
r_to_d(r, sx, delta, N = NA, Ns = N, sx.known = FALSE)
```

Arguments

r	Pearson's correlation
sx	Sample standard deviation of X
delta	Contrast in X for which to compute Cohen's d , specified in raw units of X (not standard deviations).
N	Sample size used to estimate r
Ns	Sample size used to estimate sx , if different from N
sx.known	Is sx known rather than estimated? (By default, assumes sx is estimated, which will almost always be the case.)

Details

To preserve the sign of the effect size, the code takes the absolute value of delta. The standard error estimate assumes that X is approximately normal and that N is large.

References

1. Mathur MB & VanderWeele TJ. A simple, interpretable conversion from Pearson's correlation to Cohen's d for meta-analysis. Under revision.

Examples

```
# d for a 1-unit vs. a 2-unit increase in X
r_to_d( r = 0.5,
        sx = 2,
        delta = 1,
        N = 100 )
r_to_d( r = 0.5,
        sx = 2,
        delta = 2,
        N = 100 )

# d when sx is estimated in the same vs. a smaller sample
# point estimate will be the same, but inference will be a little
# less precise in second case
r_to_d( r = -0.3,
        sx = 2,
        delta = 2,
        N = 300,
        Ns = 300 )

r_to_d( r = -0.3,
        sx = 2,
        delta = 2,
        N = 300,
        Ns = 30 )
```

r_to_z

Convert Pearson's r to Fisher's z

Description

Converts Pearson's r to Fisher's z for use in meta-analysis.

Usage

```
r_to_z(r)
```

Arguments

r Pearson's correlation

Examples

```
# convert a Pearson correlation of -0.8 to Fisher's z
r_to_z(-0.8)
```

scrape_meta

Convert forest plot or summary table to meta-analytic dataset

Description

Given relative risks (RR) and upper bounds of 95% confidence intervals (CI) from a forest plot or summary table, returns a dataframe ready for meta-analysis (e.g., via the `metafor` package) with the log-RRs and their variances. Optionally, the user may indicate studies for which the point estimate is to be interpreted as an odds ratios of a common outcome rather than a relative risk; for such studies, the function applies VanderWeele (2017)'s square-root transformation to convert the odds ratio to an approximate risk ratio.

Usage

```
scrape_meta(type = "RR", est, hi, sqrt = FALSE)
```

Arguments

<code>type</code>	RR if point estimates are RRs or ORs (to be handled on log scale); raw if point estimates are raw differences, standardized mean differences, etc. (such that they can be handled with no transformations)
<code>est</code>	Vector of study point estimates on RR or OR scale
<code>hi</code>	Vector of upper bounds of 95% CIs on RRs
<code>sqrt</code>	Vector of booleans (TRUE/FALSE) for whether each study measured an odds ratio of a common outcome that should be approximated as a risk ratio via the square-root transformation

References

1. VanderWeele TJ (2017). On a square-root transformation of the odds ratio for a common outcome. *Epidemiology*.

tau_CI	<i>Return confidence interval for tau for a meta-analysis</i>
--------	---

Description

Returns confidence interval lower and upper limits for tau (the estimated standard deviation of the true effects) for a meta-analysis fit in `metafor::rma`.

Usage

```
tau_CI(meta, CI.level = 0.95)
```

Arguments

meta	A meta-analysis object fit in <code>metafor::rma</code> .
CI.level	Confidence interval level as a proportion (e.g., 0.95)

Examples

```
# calculate effect sizes for example dataset
d = metafor::escalc(measure="RR", ai=tpos, bi=tneg,
                   ci=cpos, di=cneg, data=metafor::dat.bcg)

# fit random-effects model
# note that metafor package returns on the log scale
m = metafor::rma.uni(yi= d$yi, vi=d$vi, knha=TRUE,
                    measure="RR", method="REML" )

tau_CI(m)

# for nicer formatting
format_CI( tau_CI(m)[1], tau_CI(m)[2] )
```

z_to_r	<i>Convert Fisher's z to Pearson's r</i>
--------	--

Description

Converts Fisher's z to Pearson's r for use in meta-analysis.

Usage

```
z_to_r(z)
```

Arguments

z	Fisher's z
---	------------

Examples

```
# convert Fisher's z of 1.1 to Pearson's r  
z_to_r(1.1)
```

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