# Package 'metadat'

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Title Meta-Analysis Datasets

**Depends** R (>= 4.0.0)

Imports utils, tools, mathjaxr

Suggests metafor, numDeriv, BiasedUrn, dfoptim, igraph, ape, testthat, digest, lme4, clubSandwich, meta, netmeta, mvtnorm, gridExtra, rms, bayesmeta, ellipse

**Description** A collection of meta-analysis datasets for teaching purposes, illustrating/testing metaanalytic methods, and validating published analyses.

License GPL (>=2)

ByteCompile TRUE

LazyData TRUE

**Encoding** UTF-8

RdMacros mathjaxr

BuildManual TRUE

URL https://github.com/wviechtb/metadat, https://wviechtb.github.io/metadat/

BugReports https://github.com/wviechtb/metadat/issues

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metadat-package Meta-Analysis Datasets for R

# Description

The **metadat** package contains a large collection of meta-analysis datasets. These datasets are useful for teaching purposes, illustrating/testing meta-analytic methods, and validating published analyses.

#### **Browsing and Searching for Datasets**

A listing of all datasets in the package can be obtained with help(package=metadat). Each datasets is also tagged with one or multiple concept terms. These concept terms refer to various aspects of a dataset, such as the field/topic of research, the outcome measure used for the analysis, the model(s) used for analyzing the data, and the methods/concepts that can be illustrated with the dataset. The datsearch function can be used to search among the existing datasets in the package based on their concept terms or based on a full-text search of their corresponding help files.

You can also read the documentation online at <a href="https://wviechtb.github.io/metadat/">https://wviechtb.github.io/metadat/</a> (where the output from the example analyses corresponding to each dataset is provided).

#### **Contributing New Datasets**

We welcome contributions of new datasets to the package. For each dataset, there must be a citable reference, ideally in a peer-reviewed journal or publication. The general workflow for contributing a new dataset is as follows:

- Install the metadat package in R in the usual manner (i.e., install.packages("metadat")).
- If you are familiar with Git/GitHub and making pull requests, fork the package repository. Otherwise, download the source version of the package from GitHub and unzip the file to some directory on your computer.
- Place the raw data (in a non-binary format) in the data-raw directory. The file should be named dat.<author><year>.<ext>, where <author> is the last name of the first author of the publication from which the data come, <year> is the publication year, and <ext> is the file extension (e.g., .txt, .csv).
- Place a corresponding R script in the data-raw directory named dat.<author><year>.r that reads in the data, possibly does some data cleaning/processing, and then saves the dataset to the data directory (using save), with name dat.<author><year>.rda.
- Start R, load the metadat package (i.e., library(metadat)), and then run the prep\_dat function (either set the working directory to the location of the source package beforehand or use the pkgdir argument of the prep\_dat function to specify the source package location).
- For a new dataset, this should create a boilerplate template for a corresponding help file in the man directory, named dat.<author><year>.Rd. Edit the help file, adding the title and a short description of the dataset in general, a description of each variable in the dataset, further details on the dataset (e.g., the field of research, how the data was collected, the purpose of the dataset / what it was used for, the effect size or outcome measure used in the analysis, the types of analyses/models that can be illustrated with the dataset), a reference for the source of the dataset, one or multiple concept terms, the name and email address of the contributor of the dataset, and (optionally) example code to illustrate the analysis of the dataset.
- Either make a pull request (if you are familiar with this workflow) or zip up the dat.<author><year>.<ext>, dat.<author><year>.r, dat.<author><year>.rda, and dat.<author><year>.Rd files and open up a new issue at GitHub, attaching the zip file.
- If the above makes no sense to you, you can also email one of the package authors with a cleaned, raw data file in .txt or .csv format, along with a meta-data file (format doesn't matter) that includes the information described above.

## dat.aloe2013

## **Citing the Package**

If you use these data, please cite both the **metadat** package (see citation("metadat") for the reference) and the original source of the data as given under the help file of a dataset.

## **Bug/Error Reports**

If you think you have found an error in an existing dataset or a bug in the package in general, please go to <a href="https://github.com/wviechtb/metadat/issues">https://github.com/wviechtb/metadat/issues</a> and open up a new issue.

## Author(s)

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dat.aloe2013	Studies on the Association Between Supervision Quality and Various
	Outcomes in Social, Mental Health, and Child Welfare Workers

## Description

Results from 5 studies examining the association between various measures of supervision quality and various work-related outcomes in social, mental health, and child welfare workers.

#### Usage

dat.aloe2013

## Format

The data frame contains the following columns:

study	character	study author(s) and year
n	integer	sample size
tval	numeric	t-statistic for the test of the association/predictor
preds	integer	number of predictors included in the regression model
R2	numeric	the coefficient of determination (i.e., R-squared value) of the regression model

## Details

The dataset is based on studies that used regression models to examine the association between some measure of perceived supervision quality (e.g., the quality of the relationship with one's supervisor) and some work-related outcome (e.g., job satisfaction) in social, mental health, and child welfare workers. The dataset was extracted from Aloe and Thompson (2013), which in turn is a subset of

the studies included in the meta-analysis by Mor Barak et al. (2009).

The dataset can be used to illustrate the meta-analysis of regression models, using measures such as the (semi-)partial correlation coefficient. For this, the t-statistic from the regression model for the association (i.e., predictor) of interest was extracted from each regression model (tval), as well as the sample size (n), the number of predictors included in the regression model (preds), and the coefficient of determination (i.e., R-squared value) of the regression model (R2). Based on this information, the (semi-)partial correlation coefficient can be computed for each study, as well as its corresponding sampling variance. These values can then be meta-analyzed using standard methods.

## Concepts

social work, (semi-)partial correlations, meta-regression

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Aloe, A. M., & Thompson, C. G. (2013). The synthesis of partial effect sizes. *Journal of the Society for Social Work and Research*, **4**(4), 390–405. https://doi.org/10.5243/jsswr.2013.24

#### References

Mor Barak, M. E., Travis, D. J., Pyun, H., & Xie, B. (2009). The impact of supervision on worker outcomes: A meta-analysis. *Social Service Review*, **83**(1), 3–32. https://doi.org/10.1086/599028

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.aloe2013
dat
## Not run:
### load metafor package
suppressPackageStartupMessages(library(metafor))
### compute the partial correlation coefficients and corresponding sampling variances
dat <- escalc(measure="PCOR", ti=tval, ni=n, mi=preds, data=dat)</pre>
dat
### random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### mixed-effects meta-regression model examining the relationship between the partial
### correlation coefficients and the number of predictors included in the models
res <- rma(yi, vi, mods = ~ preds, data=dat)</pre>
res
```

### compute the r-to-z transformed partial correlation coefficients and their variances

```
dat <- escalc(measure="ZPCOR", ti=tval, ni=n, mi=preds, data=dat)
dat

### random-effects model
res <- rma(yi, vi, data=dat)
res

### back-transformation to the partial correlation scale
predict(res, transf=transf.ztor)

### compute the semi-partial correlation coefficients and their variances
dat <- escalc(measure="SPCOR", ti=tval, ni=n, mi=preds, r2i=R2, data=dat)
dat

### random-effects model
res <- rma(yi, vi, data=dat)
res
## End(Not run)</pre>
```

dat.anand1999	Studies on the Effectiveness of Oral Anticoagulants in Patients with
	Coronary Artery Disease

## Description

Results from 34 trials examining the effectiveness of oral anticoagulants in patients with coronary artery disease.

#### Usage

dat.anand1999

## Format

The data frame contains the following columns:

study	character	author(s) or trial name
year	numeric	publication year
intensity	character	intensity of anticoagulation (low, medium, or high)
asp.t	numeric	concomitant use of aspirin in the treatment group $(0 = no, 1 = yes)$
asp.c	numeric	concomitant use of aspirin in the control group $(0 = no, 1 = yes)$
ai	numeric	number of deaths in the treatment group
n1i	numeric	number of patients in the treatment group
ci	numeric	number of deaths in the control group
n2i	numeric	number of patients in the control group

#### **Details**

The dataset includes the results from 34 randomized clinical trials that examined the effectiveness of oral anticoagulants in patients with coronary artery disease. The results given here are focused on the total mortality in the treatment versus control groups.

#### Concepts

medicine, cardiology, odds ratios, Mantel-Haenszel method

## Note

Strictly speaking, there are only 31 trials, since Breddin et al. (1980) and ATACS (1990) are multiarm trials.

According to a correction, dat.anand1999\$ci[29] should be 1. But then dat.anand1999\$ci[21] would also have to be 1 (if these data indeed refer to the same control group). This appears contradictory, so this correction was not made.

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Anand, S. S., & Yusuf, S. (1999). Oral anticoagulant therapy in patients with coronary artery disease: A meta-analysis. *Journal of the American Medical Association*, **282**(21), 2058–2067. https://doi.org/10.1001/jama.282.21.2058

## Examples

### High- or Moderate-Intensity OA and Aspirin vs Aspirin

dat.assink2016 Studies on the Association between Recidivism and Mental Health

## Description

Results from 17 studies on the association between recidivism and mental health in delinquent juveniles.

## Usage

dat.assink2016

#### Format

The data frame contains the following columns:

study	numeric	study id number
esid	numeric	effect size within study id number
id	numeric	row id number
yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance
pubstatus	numeric	published study $(0 = no; 1 = yes)$
year	numeric	publication year of the study (approximately mean centered)
deltype	character	type of delinquent behavior in which juveniles could have recidivated (either general, overt, or cov

## Details

The studies included in this dataset (which is a subset of the data used in Assink et al., 2015) compared the difference in recidivism between delinquent juveniles with a mental health disorder and a comparison group of juveniles without a mental health disorder. Since studies differed in the way recidivism was defined and assessed, results are given in terms of standardized mean differences, with positive values indicating a higher prevalence of recidivism in the group of juveniles with a mental health disorder.

Multiple effect size estimates could be extracted from most studies (e.g., for different delinquent behaviors in which juveniles could have recidivated), necessitating the use of appropriate models/methods for the analysis. Assink and Wibbelink (2016) illustrate the use of multilevel meta-analysis models for this purpose.

#### Concepts

psychology, criminology, standardized mean differences, multilevel models, cluster-robust inference

#### Note

The year variable is not constant within study 3, as this study refers to two different publications using the same data.

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Assink, M., & Wibbelink, C. J. M. (2016). Fitting three-level meta-analytic models in R: A step-bystep tutorial. *The Quantitative Methods for Psychology*, **12**(3), 154–174. https://doi.org/10.20982/tqmp.12.3.p154

## References

Assink, M., van der Put, C. E., Hoeve, M., de Vries, S. L. A., Stams, G. J. J. M., & Oort, F. J. (2015). Risk factors for persistent delinquent behavior among juveniles: A meta-analytic review. *Clinical Psychology Review*, **42**, 47–61. https://doi.org/10.1016/j.cpr.2015.08.002

```
### copy data into 'dat' and examine data
dat <- dat.assink2016</pre>
head(dat, 9)
## Not run:
### load metafor package
library(metafor)
### fit multilevel model
res <- rma.mv(yi, vi, random = ~ 1 | study/esid, data=dat)</pre>
res
### use cluster-robust inference methods
robust(res, cluster=study, clubSandwich=TRUE)
### LRTs for the variance components
res0 <- rma.mv(yi, vi, random = ~ 1 | study/esid, data=dat, sigma2=c(0,NA))</pre>
anova(res0, res)
res0 <- rma.mv(yi, vi, random = ~ 1 | study/esid, data=dat, sigma2=c(NA,0))</pre>
anova(res0, res)
### examine some potential moderators via meta-regression
rma.mv(yi, vi, mods = ~ pubstatus, random = ~ 1 | study/esid, data=dat)
rma.mv(yi, vi, mods = ~ year, random = ~ 1 | study/esid, data=dat)
dat$deltype <- relevel(factor(dat$deltype), ref="general")</pre>
```

```
rma.mv(yi, vi, mods = ~ deltype, random = ~ 1 | study/esid, data=dat)
rma.mv(yi, vi, mods = ~ year + deltype, random = ~ 1 | study/esid, data=dat)
### assume that the effect sizes within studies are correlated with rho=0.6
V <- vcalc(vi, cluster=study, obs=esid, data=dat, rho=0.6)</pre>
round(V[dat$study %in% c(1,2), dat$study %in% c(1,2)], 4)
### fit multilevel model using this approximate V matrix
res <- rma.mv(yi, V, random = ~ 1 | study/esid, data=dat)</pre>
res
### use cluster-robust inference methods
robust(res, cluster=study, clubSandwich=TRUE)
### use a correlation of 0.7 for effect sizes corresponding to the same type of
### delinquent behavior and a correlation of 0.5 for effect sizes corresponding
### to different types of delinquent behavior
V <- vcalc(vi, cluster=study, type=deltype, obs=esid, data=dat, rho=c(0.7, 0.5))
### fit multilevel model using this approximate V matrix
res <- rma.mv(yi, V, random = ~ 1 | study/esid, data=dat)</pre>
res
### use cluster-robust inference methods
robust(res, cluster=study, clubSandwich=TRUE)
## End(Not run)
```

dat.axfors2021	Mortality Outcomes with Hydroxychloroquine and Chloroquine in
	COVID-19 from an International Collaborative Meta-Analysis of Ran-
	domized Trials

#### Description

Results from 33 trials examining the effectiveness of hydroxychloroquine or chloroquine in patients with COVID-19.

#### Usage

```
dat.axfors2021
```

#### Format

The data frame contains the following columns:

id	character	registry number
acronym	character	shortened registry number
patient_setting	character	patient setting
blinding_exact	character	study blinding

high_dose	character	high or low dose of medication
Published	character	publication status
hcq_cq	character	medication type (hcq = hydroxychloroquine or cq = chloroquine)
hcq_arm_event	numeric	number of deaths in the treatment group
hcq_arm_total	numeric	number of patients in the treatment group
control_arm_event	numeric	number of deaths in the control group
control_arm_total	numeric	number of patients in the control group
Control	character	control group type (Standard of Care or Placebo)

## Details

The dataset includes the results from 33 published and unpublished randomized clinical trials that examined the effectiveness of hydroxychloroquine or chloroquine in patients with COVID-19. The results given here are focused on the total mortality in the treatment versus control groups.

## Concepts

medicine, covid-19, odds ratios

## Author(s)

W. Kyle Hamilton <whamilton@ucmerced.edu> https://kylehamilton.com

#### Source

Axfors, C., Schmitt, A., Janiaud, P., van 't Hooft, J., Moher, D., Goodman, S., ... Hemkens, L. G. (2021, March 9). Hydroxychloroquine and chloroquine for survival in COVID-19: An international collaborative meta-analysis of randomized trials. https://doi.org/10.17605/OSF.IO/QESV4

# References

Axfors, C., Schmitt, A. M., Janiaud, P., van't Hooft, J., Abd-Elsalam, S., Abdo, E. F., Abella, B. S., Akram, J., Amaravadi, R. K., Angus, D. C., Arabi, Y. M., Azhar, S., Baden, L. R., Baker, A. W., Belkhir, L., Benfield, T., Berrevoets, M. A. H., Chen, C.-P., Chen, T.-C., ... Hemkens, L. G. (2021). Mortality outcomes with hydroxychloroquine and chloroquine in COVID-19 from an international collaborative meta-analysis of randomized trials. Nature Communications, 12(1), 2349. https://doi.org/10.1038/s41467-021-22446-z

```
# meta-analysis Hydroxychloroquine
res_hcq <- rma(yi, vi, subset=(hcq_cq=="hcq"), slab = id, data=dat)
print(res_hcq, digits=2)
# meta-analysis Chloroquine
res_cq <- rma(yi, vi, subset=(hcq_cq=="cq"), slab = id, data=dat)</pre>
```

```
print(res_cq, digits=2)
```

## End(Not run)

dat.bakdash2021 Dataset on Situation Awareness and Task Performance Associations

#### Description

Results from 77 papers with 678 effects evaluating associations among measures of situation awareness and task performance.

## Usage

dat.bakdash2021

## Format

The data frame contains the following columns:

Author	character	paper author(s)
Year	integer	year of paper publication
Title	character	title of paper
DOI	character	digital object identifier (DOI)
DTIC.link	character	permanent link for Defense Technical Information Collection (DITC) reports; see: https:/
SA.measure.type	character	type of SA measure
Sample.size	integer	reported sample size
Sample.size.stats	integer	reported sample size based on reported statistics (this reflects excluded participants)
es.z	numeric	z-transformed correlation coefficient; includes ghost results (disclosed and undisclosed non
vi.z	numeric	variance for z-transformed correlation (calculated using Sample.size.stats, not Sample.
SampleID	character	unique identifier for each experiment/study
Outcome	integer	unique value for each effect size

## Details

The dataset contains behavioral experiments from 77 papers/79 studies with a total of 678 effects, evaluating associations among measures of situation awareness ("knowing what is going on") and task performance. Examples of situation awareness include knowledge of current vehicle speed in a simulated driving task and location and heading of aircraft in a simulated air traffic control task. Corresponding examples of task performance include "the number of collisions in a simulated driving task" and "subject matter expert rating of conflict management in a simulated air control

task" (Bakdash et al. 2021a, p. 2). This dataset and the 'Examples' are a highly simplified version of the data and code in Bakdash et al. (2021b; 2021c). The journal article by Bakdash et al. (2021a) describes the systematic review and meta-analysis in detail.

This dataset is used to illustrate multilevel multivariate meta-analytic models for the overall pooled effect and pooled effects by situation awareness measure. We also adjust meta-analytic models using cluster-robust variance estimation / cluster-robust inference with the robust function in *metafor*. Results are shown graphically in a customized forest plot with a prediction interval (estimated plausible range of individual effects). Last, we create a table summarizing the estimated meta-analytic heterogeneity parameters.

The meta-analytic results show most pooled effect sizes in the positive medium range or less. There was also substantial meta-analytic heterogeneity (estimated systematic variance in true effects), nearing the magnitude of the overall pooled effect. We interpret the meta-analytic results as situation awareness typically having limited validity for task performance (i.e., good situation awareness does not tend to have strong probabilistic links with good performance and vice-versa). More formally, measures of situation awareness do not generally and meaningfully capture cognitive processes and other relevant factors underlying task performance.

## **Run-Time:**

The code run-time can be greatly sped-up using a linear algebra library with *R* that makes use of multiple CPU cores. See: https://www.metafor-project.org/doku.php/tips:speeding\_up\_model\_fitting. To measure the run-time, uncomment these three lines: start.time <- Sys.time(), end.time <- Sys.time(), and end.time - start.time. Run-times on Windows 10 x64 with the Intel Math Kernel Library are:

CPU	Run-Time (Minutes)
i7-11850H	2.49
i7-4770	5.38

## Concepts

psychology, human factors, engineering, correlation coefficients, multilevel models, multivariate models, cluster-robust inference

## Author(s)

Jonathan Bakdash, <jonathan.z.bakdash.civ@army.mil>, <jbakdash@gmail.com> Laura Marusich, <laura.m.cooper20.civ@army.mil>, <lmarusich@gmail.com>

#### Source

Bakdash, J. Z., Marusich, L. R., Cox, K. R., Geuss, M. N., Zaroukian, E. G., & Morris, K. M. (2021b). The validity of situation awareness for performance: A meta-analysis (Code Ocean Capsule). https://doi.org/10.24433/C0.1682542.v4

Bakdash, J. Z., Marusich, L. R., Cox, K. R., Geuss, M. N., Zaroukian, E. G., & Morris, K. M. (2021c). The validity of situation awareness for performance: A meta-analysis (Systematic Review, Data, and Code). https://doi.org/10.17605/0SF.IO/4K7ZV

## dat.bakdash2021

## References

Bakdash, J. Z., Marusich, L. R., Cox, K. R., Geuss, M. N., Zaroukian, E. G., & Morris, K. M. (2021a). The validity of situation awareness for performance: A meta-analysis. *Theoretical Issues in Ergonomics Science*, 1–24. https://doi.org/10.1080/1463922X.2021.1921310

Supplemental materials: https://www.tandfonline.com/doi/suppl/10.1080/1463922X.2021.1921310/suppl\_file/

```
### copy data into 'dat' and examine data
dat <- dat.bakdash2021
head(dat[c(1,2,6,8:12)])
## Not run:
#start.time <- Sys.time()</pre>
### load metafor
library(metafor)
### multilevel meta-analytic model to get the overall pooled effect
res.overall <- rma.mv(es.z, vi.z, mods = ~ 1,</pre>
                      random = ~ 1 | SampleID / Outcome,
                       data = dat.
                       test = "t")
res.overall
### get prediction interval
predict(res.overall)
### cluster-robust variance estimation (CRVE) / cluster-robust inference
res.overall.crve <- robust(res.overall, cluster = SampleID)</pre>
res.overall.crve
### get prediction interval
res.overall.crve.pred <- predict(res.overall.crve)</pre>
res.overall.crve.pred
### multilevel meta-analytic model for SA measures
res.sa <- rma.mv(es.z, vi.z, mods = ~ 0 + SA.measure.type,</pre>
                  random = ~ 1 | SampleID / Outcome,
                  data = dat,
                  test = "t")
res.sa
### cluster-robust variance estimation (CRVE) / cluster-robust inference
res.sa.crve <- robust(res.sa, cluster = SampleID)
res.sa.crve
### profile likelihood plots
par(mfrow=c(2,1))
profile(res.sa.crve, progbar = FALSE)
```

```
### format and combine output of meta-analytic models for the forest plot
            <- c(res.sa.crve$beta,
all.z
                                            # SA measures
                res.overall.crve$beta,
                                          # pooled effect for confidence interval (CI)
                res.overall.crve$beta)
                                          # pooled effect for prediction interval (PI)
all.ci.lower <- c(res.sa.crve$ci.lb,</pre>
                                               # SA measures
                  res.overall.crve.pred$ci.lb, # pooled effect, lower CI
                  res.overall.crve.pred$pi.lb) # pooled effect, lower PI
all.ci.upper <- c(res.sa.crve$ci.ub,</pre>
                                               # SA measures
                  res.overall.crve.pred$ci.ub, # pooled effect, upper CI
                  res.overall.crve.pred$pi.ub) # pooled effect, upper PI
### note: there is no p-value for the PI
all.pvals <- c(res.sa.crve$pval, res.overall.crve$pval)</pre>
all.labels <- c(sort(unique(dat$SA.measure.type)), "Overall", "95% Prediction Interval")
### function to round p-values for the forest plot
pvals.round <- function(input) {</pre>
 input <- ifelse(input < 0.001, "< 0.001",</pre>
           ifelse(input < 0.01, "< 0.01",
           ifelse(input < 0.05 & input >= 0.045, "< 0.05",
           ifelse(round(input, 2) == 1.00, "0.99",
           sprintf("%.2f", round(input, 2)))))}
all.pvals.rounded <- pvals.round(all.pvals)
### forest plot
plot.vals <- data.frame(all.labels, all.z, all.ci.lower, all.ci.upper)</pre>
par(mfrow=c(1,1), cex = 1.05)
forest(plot.vals$all.z,
       ci.lb = plot.vals$all.ci.lower,
       ci.ub = plot.vals$all.ci.upper,
       slab = plot.vals$all.labels,
       psize = 1,
       efac = 0, xlim = c(-1.8, 2.5), clim = c(-1, 1),
       transf = transf.ztor, # transform z to r
       at = seq(-0.5, 1, by = 0.25),
       xlab = expression("Correlation Coefficient"~"("*italic('r')*")"),
       main = "\n\n Measures",
       ilab = c(all.pvals.rounded, ""), ilab.xpos = 2.45, ilab.pos = 2.5,
       digits = 2, refline = 0, annotate = FALSE, header = FALSE)
### keep trailing zero using sprintf
output <- cbind(sprintf("%.2f", round(transf.ztor(plot.vals$all.z), 2)),</pre>
                sprintf("%.2f", round(transf.ztor(plot.vals$all.ci.lower), 2)),
                sprintf("%.2f", round(transf.ztor(plot.vals$all.ci.upper), 2)))
### alignment kludge
annotext <- apply(output, 1, function(x) {paste0(" ", x[1], " [", x[2],", ", x[3], "]")})</pre>
text( 1.05, 12:1, annotext, pos = 4, cex = 1.05)
text(-1.475, 14.00, "SA Measure", cex = 1.05)
```

```
text( 2.30, 14.00, substitute(paste(italic('p-value'))), cex = 1.05)
text( 1.55, 14.00, "Correlation [95% CI]", cex = 1.05)
abline(h = 2.5)
### black polygon for overall mean CIs
addpoly(all.z[11], ci.lb = all.ci.lower[11], ci.ub = all.ci.upper[11],
        rows = 2, annotate = FALSE, efac = 1.5, transf = transf.ztor)
### white polygon for PI
addpoly(all.z[12], ci.lb = all.ci.lower[12], ci.ub = all.ci.upper[12],
        rows = 1, col = "white", border = "black",
        annotate = FALSE, efac = 1.5, transf = transf.ztor)
par(mfrow=c(1,1), cex = 1) # reset graph parameters to default
### confidence intervals for the variance components
re.CI.variances <- confint(res.overall)</pre>
re.CI.variances
sigma1.z <- data.frame(re.CI.variances[[1]]["random"])</pre>
sigma2.z <- data.frame(re.CI.variances[[2]]["random"])</pre>
### fit model using alternative multivariate parameterization
res.overall.alt <- rma.mv(es.z, vi.z, mods = ~ 1,</pre>
                           random = ~ factor(Outcome) | factor(SampleID),
                           data = dat,
                           test = "t")
### confidence intervals for the total amount of heterogeneity variance component
res.overall.alt.tau <- confint(res.overall.alt, tau2=1)$random</pre>
### I^2: http://www.metafor-project.org/doku.php/tips:i2_multilevel_multivariate
W <- diag(1/dat$vi.z)</pre>
X <- model.matrix(res.overall)</pre>
P <- W - W %*% X %*% solve(t(X) %*% W %*% X) %*% t(X) %*% W
### I^2 (variance due to heterogeneity): 61%
I2 <- 100 * res.overall.alt$tau2 /</pre>
      (res.overall.alt$tau2 + (res.overall$k-res.overall$p)/sum(diag(P)))
Ι2
### 95% CI for I^2 using uncertainty around tau^2
I2.CI.lb <- 100 * res.overall.alt.tau[1,2] /</pre>
            (res.overall.alt.tau[1,2] + (res.overall$k-res.overall$p)/sum(diag(P)))
I2.CI.lb
I2.CI.ub <- 100 * res.overall.alt.tau[1,3] /</pre>
            (res.overall.alt.tau[1,3] + (res.overall$k-res.overall$p)/sum(diag(P)))
I2.CI.ub
### total amount of heterogeneity (tau)
sqrt(res.overall.alt$tau2)
```

```
### heterogeneity table
table.heterogeneity <- data.frame(matrix(ncol = 3, nrow = 4))</pre>
colnames(table.heterogeneity) <- c("Parameter Value",</pre>
                                      "Lower 95% CI",
                                      "Upper 95% CI")
rownames(table.heterogeneity) <- c("Tau (Total)",</pre>
                                      "Tau1 (Between paper)",
                                      "Tau2 (Within paper)",
                                      "I2 (%)")
table.heterogeneity[1,] <- res.overall.alt.tau[2,]</pre>
table.heterogeneity[2,] <- sigma1.z[2,]</pre>
table.heterogeneity[3,] <- sigma2.z[2,]</pre>
table.heterogeneity[4,] <- c(I2, I2.CI.lb, I2.CI.ub)</pre>
round(table.heterogeneity, 2)
#end.time <- Sys.time()</pre>
#end.time - start.time
## End(Not run)
```

dat.baker2009 Studies on Pharmacologic Treatments for Chronic Obstructive Pulmonary Disease

## Description

Results from 39 trials examining pharmacologic treatments for chronic obstructive pulmonary disease (COPD).

## Usage

dat.baker2009

## Format

The data frame contains the following columns:

study	character	study label
year	numeric	year of publication
id	numeric	study ID
treatment	character	treatment
exac	numeric	number of individuals with one or more COPD exacerbations
total	numeric	number of individuals

#### dat.baker2009

#### Details

This dataset comes from a systematic review of randomized controlled trials on pharmacologic treatments for chronic obstructive pulmonary disease (COPD) (Baker et al., 2009).

The primary outcome, occurrence of one or more episodes of COPD exacerbation, is binary (yes / no). For this outcome, five drug treatments (fluticasone, budesonide, salmeterol, formoterol, tiotropium) and two combinations (fluticasone + salmeterol, budesonide + formoterol) were compared to placebo. The authors considered the two combinations as separate treatments instead of evaluating the individual components.

#### Concepts

medicine, odds ratios, network meta-analysis, component network meta-analysis

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

#### Source

Baker, W. L., Baker, E. L., & Coleman, C. I. (2009). Pharmacologic treatments for chronic obstructive pulmonary disease: A mixed-treatment comparison meta-analysis. *Pharmacotherapy: The Journal of Human Pharmacology and Drug Therapy*, **29**(8), 891–905. https://doi.org/10.1592/phco.29.8.891

```
### Show first 6 rows of the dataset
head(dat.baker2009)
```

```
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
```

```
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
```

```
### Transform data from long arm-based format to contrast-based
### format. Argument 'sm' has to be used for odds ratio as summary
### measure; by default the risk ratio is used in the metabin function
### called internally.
pw <- pairwise(treatment, exac, total, studlab = paste(study, year),
    data = dat.baker2009, sm = "OR")
### Conduct random effects network meta-analysis (NMA)
### with placebo as reference
net <- netmeta(pw, fixed = FALSE, ref = "plac")
### Show network graph
netgraph(net, seq = "optimal", start = "prcomp",
    labels = gsub("+", " +\n", trts, fixed = TRUE),
    plastic = TRUE, thickness = "se.fixed", number = TRUE,
```

```
points = TRUE, cex.points = 5, col.points = "red",
```

```
offset = 0.025)
### Print and plot results for network meta-analysis
net
forest(net)
### Conduct component network meta-analysis (CNMA)
cnet <- netcomb(net)
cnet
### Compare results of NMA and additive CNMA
nb <- netbind(net, cnet, name = c("Standard NMA", "Additive CNMA"))
forest(nb)
### Use previous settings
settings.meta(oldset)
## End(Not run)</pre>
```

dat.bangertdrowns2004 Studies on the Effectiveness of Writing-to-Learn Interventions

## Description

Results from 48 studies on the effectiveness of school-based writing-to-learn interventions on academic achievement.

#### Usage

dat.bangertdrowns2004

## Format

The data frame contains the following columns:

id	numeric	study number
author	character	study author(s)
year	numeric	publication year
grade	numeric	grade level (1 = elementary; 2 = middle; 3 = high-school; 4 = college)
length	numeric	treatment length (in weeks)
minutes	numeric	minutes per assignment
wic	numeric	writing tasks were completed in class $(0 = no; 1 = yes)$
feedback	numeric	feedback on writing was provided $(0 = no; 1 = yes)$
info	numeric	writing contained informational components $(0 = no; 1 = yes)$
pers	numeric	writing contained personal components $(0 = no; 1 = yes)$
imag	numeric	writing contained imaginative components $(0 = no; 1 = yes)$
meta	numeric	prompts for metacognitive reflection $(0 = no; 1 = yes)$
subject	character	subject matter
ni	numeric	total sample size of the study

yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance

#### Details

In each of the studies included in this meta-analysis, an experimental group (i.e., a group of students that received instruction with increased emphasis on writing tasks) was compared against a control group (i.e., a group of students that received conventional instruction) with respect to some content-related measure of academic achievement (e.g., final grade, an exam/quiz/test score). The outcome measure for this meta-analysis was the standardized mean difference (with positive values indicating a higher mean level of academic achievement in the intervention group).

The standardized mean differences given here are bias-corrected and therefore differ slightly from the values reported in the article. Also, since only the total sample size is given in the article, the sampling variances were computed under the assumption that  $n_{i1} = n_{i2} = n_i/2$ .

## Concepts

education, standardized mean differences, meta-regression

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Bangert-Drowns, R. L., Hurley, M. M., & Wilkinson, B. (2004). The effects of school-based writing-to-learn interventions on academic achievement: A meta-analysis. *Review of Educational Research*, **74**(1), 29–58. https://doi.org/10.3102/00346543074001029

```
### copy data into 'dat' and examine data
dat <- dat.bangertdrowns2004
dat[1:10,-13]
## Not run:
### load metafor package
library(metafor)
### fit random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### some examples of mixed-effects meta-regression models
    <- rma(yi, vi, mods = ~ factor(grade), data=dat)
res
res
res <- rma(yi, vi, mods = ~ length, data=dat)</pre>
res
res <- rma(yi, vi, mods = ~ info + pers + imag + meta, data=dat)</pre>
res
```

## End(Not run)

dat.bartos2023

#### Description

Results from 350,757 coin flips by 48 people to examine the presence of same-side bias.

#### Usage

dat.bartos2023

## Format

The data frame contains the following columns:

person	character	person identifier
hsame	numeric	number of flips where the coin landed on heads and on the same side as where it started
hdiff	numeric	number of flips where the coin landed on heads and on the different side as where it started
tsame	numeric	number of flips where the coin landed on tails and on the same side as where it started
tdiff	numeric	number of flips where the coin landed on tails and on the different side as where it started
same	numeric	number of flips where the coin landed on the same side as where it started
flips	numeric	total number of flips

## Details

In a landmark study by Bartoš et al. (2023), 48 people flipped a coin (of various currencies and/or denominations) a total of 350,757 times, recording on each flip whether it landed on heads or tails and whether the coin landed on the same side as where it started or on the different side. The goal of this experiment was to examine the model by Diaconis, Holmes, and Montgomery (2007), according to which flipped coins have a slightly higher than 50% chance (of around 51% according to the D-H-M model) of landing on the same side as where they started.

## Concepts

physics, human factors, proportions, multivariate models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Bartoš, F., Sarafoglou, A., Godmann, H. R., Sahrani, A., Leunk, D. K., Gui, P. Y., Voss, D., Ullah, K., Zoubek, M. J., Nippold, F., Aust, F., Vieira, F. F., Islam, C.-G., Zoubek, A. J., Shabani, S., Petter, J., Roos, I. B., Finnemann, A., Lob, A. B., Hoffstadt, M. F., Nak, J., de Ron, J., Derks, K., Huth, K., Terpstra, S., Bastelica, T., Matetovici, M., Ott, V. L., Zetea, A. S., Karnbach, K., Donzallaz, M. C.,

#### dat.bartos2023

John, A., Moore, R. M., Assion, F., van Bork, R., Leidinger, T. E., Zhao, X., Motaghi, A. K., Pan, T., Armstrong, H., Peng, T., Bialas, M., Pang, J. Y.-C., Fu, B., Yang, S., Lin, X., Sleiffer, D., Bognar, M., Aczel, B., & Wagenmakers, E.-J. (2023). Fair coins tend to land on the same side they started: Evidence from 350,757 flips. *arXiv*, 2310.04153, v2. https://arxiv.org/abs/2310.04153

## References

Diaconis, P., Holmes, S., & Montgomery, R. (2007). Dynamical bias in the coin toss. *SIAM Review*, **49**(2), 211–235. https://doi.org/10.1137/s0036144504446436

```
### copy data into 'dat' and examine data
dat <- dat.bartos2023</pre>
dat
## Not run:
### load metafor package
library(metafor)
### compute proportions and the corresponding sampling variances
dat <- escalc(measure="PR", xi=same, ni=flips, data=dat, slab=person)</pre>
dat
### compute confidence intervals for the individual proportions (as in Table 1)
summary(dat, digits=3)[c(1,6:8,13,14)]
### compute a confidence interval based on the column totals
summary(escalc(measure="PR", xi=sum(dat$same), ni=sum(dat$flips)), digits=3)
### this is the same as meta-analyzing the proportions directly using an equal-effects
### model and also computing the sampling variances under the assumption that the true
### proportions are homogeneous
rma(measure="PR", xi=same, ni=flips, vtype="AV", method="EE", data=dat, digits=3)
### fit a random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### profile likelihood confidence interval for tau^2
confint(res, type="PL")
### forest plot
forest(res, refline=0.5, xlim=c(0.38,0.72), digits=c(3,2), efac=c(0,1))
### funnel plot
funnel(res, xlim=c(0.45,0.6), ylim=c(0,0.02))
### fit a random-effects model excluding those with same-side proportions larger than 0.53
res <- rma(yi, vi, data=dat, subset=yi<=0.53)</pre>
res
confint(res, type="PL")
```

```
### fit a binomial-normal model
res <- rma.glmm(measure="PLO", xi=same, ni=flips, data=dat)</pre>
res
predict(res, transf=plogis)
### conduct a meta-analysis for the proportions of heads (to examine heads-tails bias)
dat <- escalc(measure="PR", xi=hdiff+hsame, ni=flips, data=dat)</pre>
res <- rma(yi, vi, data=dat)</pre>
res
confint(res, type="PL")
### restructure the dataset for a bivariate meta-analysis of same-side and heads proportions
dat <- dat.bartos2023
dat <- dat[rep(1:nrow(dat), each=2),]</pre>
rownames(dat) <- NULL
dat$outcome <- c("heads", "same")</pre>
dat <- escalc(measure="PR", xi=hsame+hdiff, ni=flips, data=dat, include=outcome=="heads")</pre>
dat <- escalc(measure="PR", xi=hsame+tsame, ni=flips, data=dat, include=outcome=="same")</pre>
dat
### construct the 2x2 variance-covariance matrix of the proportions within persons
dat$cov <- with(dat, (hsame/flips * (1-hsame/flips) - hsame/flips * tsame/flips -</pre>
                      hsame/flips * hdiff/flips - hdiff/flips * tsame/flips) / flips)
V <- lapply(split(dat, dat$person), \(x) matrix(c(x$vi[1], x$cov, x$vi[2]), nrow=2))</pre>
### fit bivariate meta-analysis model
res <- rma.mv(yi, V, mods = ~ 0 + outcome, random = ~ outcome | person, struct="UN", data=dat)
res
### create plot with confidence ellipses ('ellipse' package must be installed)
library(ellipse)
plot(NA, xlim=c(0.45,0.62), ylim=c(0.45,0.62), bty="1", xlab="Pr(heads)", ylab="Pr(same)")
abline(h=0.5, lty="dotted")
abline(v=0.5, lty="dotted")
# add confidence ellipses for persons
invisible(tapply(dat, dat$person, \(x) {
   xy <- ellipse(matrix(c(x$vi[1],x$cov,x$vi[2]), nrow=2), centre=x$yi, level=0.95)</pre>
   lines(xy[,1],xy[,2], col="gray80")
}))
# add the points
invisible(tapply(dat, dat$person, \(x) points(x$yi[1], x$yi[2], pch=21, bg="gray80", cex=1.5)))
# add the 95% PI ellipsis based on the model
xy <- ellipse(res$G, centre=coef(res), level=0.95)</pre>
lines(xy[,1],xy[,2], col="gray30", lwd=3, lty="dotted")
# add the 95% CI ellipsis based on the model
xy <- ellipse(vcov(res), centre=coef(res), level=0.95)</pre>
lines(xy[,1],xy[,2], col="gray30", lwd=3)
# add the point for the pooled effects
points(coef(res)[1], coef(res)[2], pch=21, bg="gray40", cex=2)
```

## End(Not run)

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dat.baskerville2012 Studies on the Effectiveness of Practice Facilitation Interventions

## Description

Results from 23 studies on the effectiveness of practice facilitation interventions within the primary care practice setting.

#### Usage

dat.baskerville2012

## Format

The data frame contains the following columns:

author year	character numeric	study author(s) publication year
score	numeric	quality score (0 to 12 scale)
design	character	study design (cct = controlled clinical trial, rct = randomized clinical trial, crct = cluster randomize
alloconc	numeric	allocation concealed $(0 = no, 1 = yes)$
blind	numeric	single- or double-blind study $(0 = no, 1 = yes)$
itt	numeric	intention to treat analysis $(0 = no, 1 = yes)$
fumonths	numeric	follow-up months
retention	numeric	retention (in percent)
country	character	country where study was conducted
outcomes	numeric	number of outcomes assessed
duration	numeric	duration of intervention
pperf	numeric	practices per facilitator
meetings	numeric	(average) number of meetings
hours	numeric	(average) hours per meeting
tailor	numeric	intervention tailored to the context and needs of the practice $(0 = no, 1 = yes)$
smd	numeric	standardized mean difference
se	numeric	corresponding standard error

## Details

Baskerville et al. (2012) describe outreach or practice facilitation as a "multifaceted approach that involves skilled individuals who enable others, through a range of intervention components and approaches, to address the challenges in implementing evidence-based care guidelines within the primary care setting". The studies included in this dataset examined the effectiveness of practice facilitation interventions for improving some relevant evidence-based practice behavior. The effect was quantified in terms of a standardized mean difference, comparing the change (from pre- to post-intervention) in the intervention versus the comparison group (or the difference from baseline in prospective cohort studies).

#### Concepts

medicine, primary care, standardized mean differences, publication bias, meta-regression

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Baskerville, N. B., Liddy, C., & Hogg, W. (2012). Systematic review and meta-analysis of practice facilitation within primary care settings. *Annals of Family Medicine*, **10**(1), 63–74. https://doi.org/10.1370/afm.1312

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.baskerville2012</pre>
dat
## Not run:
### load metafor package
library(metafor)
### random-effects model
res <- rma(smd, sei=se, data=dat, method="DL")</pre>
print(res, digits=2)
### funnel plot
funnel(res, xlab="Standardized Mean Difference", ylim=c(0,0.6))
### rank and regression tests for funnel plot asymmetry
ranktest(res)
regtest(res)
### meta-regression analyses examining various potential moderators
rma(smd, sei=se, mods = ~ score, data=dat, method="DL")
rma(smd, sei=se, mods = ~ alloconc, data=dat, method="DL")
rma(smd, sei=se, mods = ~ blind, data=dat, method="DL")
rma(smd, sei=se, mods = ~ itt,
                                  data=dat, method="DL")
rma(smd, sei=se, mods = ~ duration, data=dat, method="DL")
rma(smd, sei=se, mods = ~ tailor, data=dat, method="DL")
rma(smd, sei=se, mods = ~ pperf,
                                    data=dat, method="DL")
rma(smd, sei=se, mods = ~ I(meetings * hours), data=dat, method="DL")
## End(Not run)
```

dat.bassler2004

Studies on Ketotifen Alone or as Additional Medication for Long-Term Control of Asthma and Wheeze in Children

## dat.bassler2004

## Description

Results from 10 trials reporting the physicians' judgement on the overall efficacy of ketotifen for long-term control of asthma and wheeze in children.

## Usage

dat.bassler2004

#### Format

The data frame contains the following columns:

study	character	study label
Ee	integer	number of children with treament success (ketotifen group)
Ne	integer	number of children (ketotifen group)
Ec	integer	number of children with treament success (control group)
Nc	integer	number of children (control group)
blind	character	blinding of clinicians

#### Details

Results from 10 trials reporting the physicians' judgement on the overall efficacy of Ketotifen for long-term control of asthma and wheeze in children. A prespecified subgroup analysis was conducted to evaluate whether the treatment effect is different in trials with adequate blinding compared to trials with inadequate / unclear blinding.

This data set is used as an example in Schwarzer et al. (2015).

#### Concepts

risk ratios, medicine, subgroup analysis

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

#### Source

Bassler D., Mitra A. A. D., Ducharme F. M., Forster J., & Schwarzer, G. (2004). Ketotifen alone or as additional medication for long-term control of asthma and wheeze in children. *Cochrane Database of Systematic Reviews*, **1**, CD001384. https://doi.org/10.1002/14651858.CD001384.pub2

#### References

Schwarzer, G., Carpenter, J. R., & Rücker, G. (2015). *Meta-analysis with R*. Cham, Switzerland: Springer.

## Examples

```
### Show full data set
dat.bassler2004
## Not run:
### Load meta package
suppressPackageStartupMessages(library("meta"))
### Use DerSimonian-Laird estimator (which was the default in meta in the year 2015).
### Furthermore, print meta-analysis results with two digits.
oldset <- settings.meta(method.tau = "DL", digits = 2)</pre>
### Calculate experimental and control event rates
with(dat.bassler2004, summary(Ee / Ne))
with(dat.bassler2004, summary(Ec / Nc))
### Conduct meta-analysis using the inverse variance method
mb3 <- metabin(Ee, Ne, Ec, Nc, method = "I",</pre>
               data = dat.bassler2004, studlab = study)
mb3
### Conduct subgroup analysis comparing trials with adequate blinding
### to trials with inadequate or unclear blinding
mb3s <- update(mb3, subgroup = blind, print.subgroup.name = FALSE)
mb3s
### Conduct subgroup analysis assuming common between-study variance in subgroups
mb3s.c <- update(mb3s, tau.common = TRUE)</pre>
mb3s.c
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.bcg

Studies on the Effectiveness of the BCG Vaccine Against Tuberculosis

#### Description

Results from 13 studies examining the effectiveness of the Bacillus Calmette-Guerin (BCG) vaccine against tuberculosis.

#### Usage

dat.bcg

#### Format

The data frame contains the following columns:

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## dat.bcg

trial	numeric	trial number
author	character	author(s)
year	numeric	publication year
tpos	numeric	number of TB positive cases in the treated (vaccinated) group
tneg	numeric	number of TB negative cases in the treated (vaccinated) group
cpos	numeric	number of TB positive cases in the control (non-vaccinated) group
cneg	numeric	number of TB negative cases in the control (non-vaccinated) group
ablat	numeric	absolute latitude of the study location (in degrees)
alloc	character	method of treatment allocation (random, alternate, or systematic assignment)

## Details

The 13 studies provide data in terms of  $2 \times 2$  tables in the form:

	TB positive	TB negative
vaccinated group	tpos	tneg
control group	cpos	cneg

The goal of the meta-analysis was to examine the overall effectiveness of the BCG vaccine for preventing tuberculosis and to examine moderators that may potentially influence the size of the effect.

The dataset has been used in several publications to illustrate meta-analytic methods (see 'References').

#### Concepts

medicine, risk ratios, meta-regression

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Colditz, G. A., Brewer, T. F., Berkey, C. S., Wilson, M. E., Burdick, E., Fineberg, H. V., & Mosteller, F. (1994). Efficacy of BCG vaccine in the prevention of tuberculosis: Meta-analysis of the published literature. *Journal of the American Medical Association*, **271**(9), 698–702. https://doi.org/10.1001/jama.1994.035103

## References

Berkey, C. S., Hoaglin, D. C., Mosteller, F., & Colditz, G. A. (1995). A random-effects regression model for meta-analysis. *Statistics in Medicine*, **14**(4), 395–411. https://doi.org/10.1002/sim.4780140406

van Houwelingen, H. C., Arends, L. R., & Stijnen, T. (2002). Advanced methods in meta-analysis: Multivariate approach and meta-regression. *Statistics in Medicine*, **21**(4), 589–624. https://doi.org/10.1002/sim.1040

Viechtbauer, W. (2010). Conducting meta-analyses in R with the metafor package. *Journal of Statistical Software*, **36**(3), 1–48. https://doi.org/10.18637/jss.v036.i03

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.bcg</pre>
dat
## Not run:
### load metafor package
library(metafor)
### calculate log risk ratios and corresponding sampling variances
dat <- escalc(measure="RR", ai=tpos, bi=tneg,</pre>
                             ci=cpos, di=cneg, data=dat,
                             slab=paste0(author, ", ", year))
dat
### random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### average risk ratio with 95% CI
predict(res, transf=exp)
### mixed-effects model with absolute latitude and publication year as moderators
res <- rma(yi, vi, mods = ~ ablat + year, data=dat)</pre>
res
### predicted average risk ratios for 10-60 degrees absolute latitude
### holding the publication year constant at 1970
predict(res, newmods=cbind(seq(from=10, to=60, by=10), 1970), transf=exp)
### note: the interpretation of the results is difficult because absolute
### latitude and publication year are strongly correlated (the more recent
### studies were conducted closer to the equator)
plot(ablat ~ year, data=dat, pch=19, xlab="Publication Year", ylab="Absolute Lattitude")
cor(dat$ablat, dat$year)
## End(Not run)
```

dat.begg1989 Studies on Bone-Marrow Transplantation versus Chemotherapy for the Treatment of Leukemia

#### Description

Results from controlled and uncontrolled studies on the effectiveness of allogeneic bone-marrow transplantation (BMT) and conventional chemotherapy (CMO) in the treatment of acute nonlym-phocytic leukemia.

#### Usage

dat.begg1989

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## dat.begg1989

## Format

The data frame contains the following columns:

study	numeric	study number
trt	character	treatment (BMT or CMO)
arms	numeric	number of arms in the study (1 = uncontrolled studies; 2 = controlled studies)
yi	numeric	2-year disease-free survival rates
sei	numeric	corresponding standard errors
vi	numeric	corresponding sampling variances

#### Details

The dataset includes the results from controlled and uncontrolled studies on the 2-year diseasefree survival rate in patients with acute nonlymphocytic leukemia receiving either allogeneic bonemarrow transplantation (BMT) or conventional chemotherapy (CMO). In the controlled (two-arm) studies (studies 1-4), a cohort of patients in complete remission and potentially eligible for BMT was assembled, and those who consented and for whom a donor could be found received BMT, with the remaining patients used as controls (receiving CMO). In the uncontrolled (one-arm) studies (studies 5-16), only a single group was studied, receiving either BMT or CMO.

The data in this dataset were obtained from Table 1 in Begg and Pilote (1991, p. 902).

#### Concepts

medicine, oncology, single-arm studies, multilevel models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Begg, C. B., & Pilote, L. (1991). A model for incorporating historical controls into a meta-analysis. *Biometrics*, **47**(3), 899–906. https://doi.org/10.2307/2532647

## References

Begg, C. B., Pilote, L., & McGlave, P. B. (1989). Bone marrow transplantation versus chemotherapy in acute non-lymphocytic leukemia: A meta-analytic review. *European Journal of Cancer and Clinical Oncology*, **25**(11), 1519–1523. https://doi.org/10.1016/0277-5379(89)90291-5

```
### copy data into 'dat' and examine data
dat <- dat.begg1989
dat
## Not run:
### load metafor package
library(metafor)</pre>
```

```
### turn trt and arms into factors and set reference levels
dat$trt <- relevel(factor(dat$trt), ref="CMO")</pre>
dat$arms <- relevel(factor(dat$arms), ref="2")</pre>
### create data frame with the treatment differences for the controlled studies
dat2 <- data.frame(yi = dat$yi[c(1,3,5,7)] - dat$yi[c(2,4,6,8)],</pre>
                    vi = dat$vi[c(1,3,5,7)] + dat$vi[c(2,4,6,8)])
dat2
### DerSimonian and Laird method using the treatment differences
res <- rma(yi, vi, data=dat2, method="DL", digits=2)</pre>
res
### Begg & Pilote (1991) model incorporating the uncontrolled studies
res <- rma.mv(yi, vi, mods = ~ trt, random = ~ 1 | study,</pre>
              data=dat, method="ML", digits=2)
res
### model involving bias terms for the uncontrolled studies
res <- rma.mv(yi, vi, mods = ~ trt + trt:arms, random = ~ 1 | study,
              data=dat, method="ML", digits=2)
res
### model with a random treatment effect
res <- rma.mv(yi, vi, mods = ~ trt, random = list(~ 1 | study, ~ trt | study),</pre>
              struct="UN", tau2=c(0,NA), rho=0, data=dat, method="ML", digits=2)
res
### model with a random treatment effect, but with equal variances in both arms
res <- rma.mv(yi, vi, mods = ~ trt, random = list(~ 1 | study, ~ trt | study),</pre>
              struct="CS", rho=0, data=dat, method="ML", digits=2)
res
## End(Not run)
```

dat.berkey1998 Studies on Treatments for Periodontal Disease

#### Description

Results from 5 trials comparing surgical and non-surgical treatments for medium-severity periodontal disease one year after treatment.

#### Usage

dat.berkey1998

#### Format

The data frame contains the following columns:

#### dat.berkey1998

trial	numeric	trial number
author	character	study author(s)
year	numeric	publication year
ni	numeric	number of patients
outcome	character	outcome (PD = probing depth; $AL$ = attachment level)
yi	numeric	observed mean difference in outcome (surgical versus non-surgical)
vi	numeric	corresponding sampling variance
v1i	numeric	variances and covariances of the observed effects
v2i	numeric	variances and covariances of the observed effects

## Details

The dataset includes the results from 5 trials that compared surgical and non-surgical methods for the treatment of medium-severity periodontal disease. Reported outcomes include the change in probing depth (PD) and attachment level (AL) one year after the treatment. The outcome measure used for this meta-analysis was the (raw) mean difference, calculated in such a way that positive values indicate that surgery was more effective than non-surgical treatment in decreasing the probing depth and increasing the attachment level (so, the results from the various trials indicate that surgery is preferable for reducing the probing depth, while non-surgical treatment is preferable for increasing the attachment level). Since each trial provides effect size estimates for both outcomes, the estimates are correlated. A multivariate model can be used to meta-analyze the two outcomes simultaneously.

The v1i and v2i values are the variances and covariances of the observed effects. In particular, for each study, variables v1i and v2i form a  $2 \times 2$  variance-covariance matrix of the observed effects, with the diagonal elements corresponding to the sampling variances of the mean differences (the first for probing depth, the second for attachment level) and the off-diagonal value corresponding to the covariance of the two mean differences. Below, the full (block diagonal) variance-covariance for all studies is constructed from these two variables.

## Concepts

medicine, dentistry, raw mean differences, multivariate models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Berkey, C. S., Antczak-Bouckoms, A., Hoaglin, D. C., Mosteller, F., & Pihlstrom, B. L. (1995). Multiple-outcomes meta-analysis of treatments for periodontal disease. *Journal of Dental Research*, **74**(4), 1030–1039. https://doi.org/10.1177/00220345950740040201

Berkey, C. S., Hoaglin, D. C., Antczak-Bouckoms, A., Mosteller, F., & Colditz, G. A. (1998). Metaanalysis of multiple outcomes by regression with random effects. *Statistics in Medicine*, **17**(22), 2537–2550. https://doi.org/10.1002/(sici)1097-0258(19981130)17:22<2537::aid-sim953>3.0.co;2-c

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.berkey1998
dat
## Not run:
### load metafor package</pre>
```

### construct block diagonal var-cov matrix of the observed outcomes based on variables v1i and v2i
V <- vcalc(vi=1, cluster=author, rvars=c(v1i, v2i), data=dat)</pre>

```
### fit multiple outcomes (meta-regression) model (with REML estimation)
res <- rma.mv(yi, V, mods = ~ 0 + outcome, random = ~ outcome | trial, struct="UN", data=dat)
print(res, digits=3)</pre>
```

### test/estimate difference between the two outcomes
anova(res, X=c(1,-1))

## End(Not run)

library(metafor)

dat.besson2016 Dataset on How Maternal Diet Impacts Copying Styles in Rodents

## Description

Results from 46 studies synthesising maternal nutritional effects on coping styles in rodents.

## Usage

dat.besson2016

#### Format

The data frame contains the following columns:

comp_ID	character	effect-size unique identifier
study_ID	character	study unique identifier
dam_ID	character	dam unique identifier (group of dams subjected to the same treatment)
animal_ID	character	offspring unique identifier (group of offspring from the same dam group s
Reference	character	author's names and date
species	character	species [rats or mice]
strain	character	strain
manip_type	character	maternal nutritional manipulation type [protein or calorie]

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manip_direction	character	direction of maternal nutritional manipulation [- = restriction, + = overfee
nom_manip_val	character	degree of maternal nutritional manipulation as described in the original pu
exp	character	percentage of caloric or protein maternal restriction or increase in caloric
control	character	percentage of caloric or protein maternal restriction or increase in caloric
manip_parameter	character	protein content, percentage fat or intake
vitmin_eql	character	were vitamins equalized across maternal diets? [yes or no]
adlib_con	character	were maternal control groups fed ad libitum? [yes or no]
adlib_exp	character	were maternal experimental groups fed ad libitum? [yes or no]
diet_con	character	name of maternal control diet?
diet_exp	character	name of maternal experimental diet?
dam_diet_start_dPC	numeric	start of the dam diet [in days post-conception]
dam_diet_end_dPC	numeric	end of the dam diet [in days post-conception]
diet_label	character	period of maternal diet manipulation [pregestation = pre-gestation, pre = p
age_mating	numeric	dam age at mating if known
n_con_dam	integer	sample size of the control dam groups
n_exp_dam	integer	sample size of the experimental dam groups
multi_use_con	character	were control groups used multiple time? [yes or no]
dam_housing	character	how were dams housed? [pair, group, or single]
temperature	numeric	temperature during the experiment [°C]
photoperiod	integer	photoperiod during the experiment [number of hours of light]
litter_size	integer	size of the litter [number of pups per dam]
litter_size_equalized	character	has litter size been equalized? [yes or no]
crossfostered	character	have pups been cross-fostered? [yes or no]
sex	character	sex of the offspring that were tested $[m = male, f = female, both = mixed s$
housing	character	offspring housing during the test period [dam, pair, single, or group]
bodymass_mean_contr	numeric	mean body mass of control offspring close to or during the testing period
bodymass_SE_contr	numeric	S.E. for body mass of control offspring close to or during the testing perio
bodymass_mean_exp	numeric	mean body mass of experimental offspring close to or during the testing p
bodymass_SE_exp	numeric	S.E. for body mass of experimental offspring close to or during the testing
bm_N_contr	integer	sample size for body mass of control offspring close to or during the testir
bm_N_exp	integer	sample size for body mass of experimental offspring close to or during the
bm_dPP	integer	age of offspring when body mass was measured [in days post-parturition]
offspring_diet	character	offspring diet after weaning [type of control diet]
offspring_con_adlib	character	were control offspring fed ad libitum after weaning? [yes or no]
offspring_diet_level	character	name of offspring diet after weaning
offspring_diet_end_dPP	integer	end of the offspring diet [in days post-parturition]
post_diet_adlib	character	were experimental offspring fed ad libitum after weaning? [yes or no]
response_age_dPP	numeric	offspring age when behavioural testing started [in days post-parturition]
authors_behaviour_classification	character	author's classification of offspring behaviour [anxiety, exploration, or acti
our_behaviour_classification	character	our classification of offspring behaviour [anxiety, exploration, or activity]
response_test	character	type of test used [elevated T-maze (ETM), open field, etc.] to measure offs
time_trial	integer	duration of the testing [min]
measure	character	measures taken during testing [total distance moved, time spent in open ar
unit	character	unit of the behavioural measure taken [min, s, m, number (#), etc.]
high_better	character	for activity and exploration, a higher number is assumed to be better (i.e.,
night.day	character	time of day when behaviours were measured [night or day]
comparison	character	for a given control-treatment group comparison, animal group codes as us
exp_mean	numeric	mean of the offspring behaviour measured for the experimental group

## dat.besson2016

exp_se	numeric	S.E. of the offspring behaviour measured for the experimental group
exp_n	integer	sample size for the offspring experimental group
con_mean	numeric	mean of offspring behaviour measured for the control group
con_se	numeric	S.E. of the offspring behaviour measured for the control group
con_n	integer	sample size for the offspring control group
con_ID	character	identifier for shared control groups within experiment
percentage	character	is the offspring behaviour measure a percentage? [yes or no]
Data_source	character	figure or table number in the original paper from which the data were extr
measure_comments	character	any comments on the offspring behaviour measures
SE_imputed	character	was S.E. imputed for the offspring behaviour measure? [yes or no]
Comments	character	any comments on the data

## Details

Data from experiments where dams were subject to caloric or protein restriction or were overfed around gestation were included. Offspring activity, exploration, or anxiety were measured outcomes variables from maternal experimental treatments. Multilevel meta-analysis and meta-regression models were used to analyze the meta-analytic data.

# Concepts

ecology, evolution, standardized mean differences

## Author(s)

Daniel Noble, <daniel.noble@anu.edu.au>

#### Source

Besson, A. A., Lagisz, M., Senior, A. M., Hector, K. L., & Nakagawa, S. (2016). Effect of maternal diet on offspring coping styles in rodents: A systematic review and meta-analysis. *Biological Reviews*, **91**(4), 1065–1080. https://doi.org/10.1111/brv.12210

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.besson2016
head(dat)</pre>
```

## Not run:
### load metafor
library(metafor)

### compute SD from SE
dat\$sd\_c <- with(dat, con\_se \* sqrt(con\_n))
dat\$sd\_e <- with(dat, exp\_se \* sqrt(exp\_n))</pre>

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dat.bonett2010

```
### fit model
mod1 <- rma.mv(yi ~ 1, V = vi, random = list(~ 1 | study_ID, ~ 1 | comp_ID), data = dat)
mod1
## End(Not run)</pre>
```

dat.bonett2010 Studies on the Reliability of the CES-D Scale

## Description

Results from 9 studies on the reliability of the Center for Epidemiologic Studies Depression (CES-D) Scale administered to children providing care to an elderly parent.

### Usage

dat.bonett2010

## Format

The data frame contains the following columns:

study	numeric	study number
source	character	source of data
ni	numeric	sample size
mi	numeric	number of items in the scale
ai	numeric	observed value of Cronbach's alpha
caregivers	character	gender of the children in the sample

## Details

The Center for Epidemiologic Studies Depression (CES-D) Scale is a 20-item questionnaire assessing various symptoms of depression, with each item scored on a 4-point scale. The scale has been used in several studies to examine depressive symptoms in children providing care to an elderly parent. The dataset includes information on the reliability of the scale as measured with Cronbach's alpha in 9 such studies. Also, the gender composition of the children in each sample is indicated.

## Concepts

psychology, Cronbach's alpha, reliability generalization, meta-regression

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Bonett, D. G. (2010). Varying coefficient meta-analytic methods for alpha reliability. *Psychological Methods*, **15**(4), 368–385. https://doi.org/10.1037/a0020142

## References

Bonett, D. G. (2002). Sample size requirements for testing and estimating coefficient alpha. *Journal* of *Educational and Behavioral Statistics*, **27**(4), 335–340. https://doi.org/10.3102/10769986027004335

Hakstian, A. R., & Whalen, T. E. (1976). A k-sample significance test for independent alpha coefficients. *Psychometrika*, **41**(2), 219–231. https://doi.org/10.1007/BF02291840

```
### copy data into 'dat' and examine data
dat <- dat.bonett2010</pre>
dat
## Not run:
### load metafor package
library(metafor)
### meta-analysis using the raw alpha values
res <- rma(measure="ARAW", ai=ai, mi=mi, ni=ni, data=dat)</pre>
res
### meta-analysis using transformed alpha values (using the
### transformation suggested by Hakstian & Whalen, 1976)
res <- rma(measure="AHW", ai=ai, mi=mi, ni=ni, data=dat)</pre>
res
predict(res, transf=transf.iahw)
### meta-analysis using transformed alpha values (using the
### transformation suggested by Bonett, 2002)
res <- rma(measure="ABT", ai=ai, mi=mi, ni=ni, data=dat)</pre>
res
predict(res, transf=transf.iabt)
### forest plot
forest(res, slab=source, xlim=c(0,4.5),
       atransf=transf.iabt, refline=coef(res))
### examine whether female/mixed samples yield different alphas (with raw alphas)
res <- rma(measure="ARAW", ai=ai, mi=mi, ni=ni, mods = ~ caregivers, data=dat)
res
predict(res, newmods=c(0,1), digits=2)
## End(Not run)
```

dat.bornmann2007

## Description

Results from 21 studies on gender differences in grant and fellowship awards.

### Usage

dat.bornmann2007

# Format

The data frame contains the following columns:

study	character	study reference
obs	numeric	observation within study
doctype	character	document type
gender	character	gender of the study authors
year	numeric	(average) cohort year
org	character	funding organization / program
country	character	country of the funding organization / program
type	character	fellowship or grant application
discipline	character	discipline / field
waward	numeric	number of women who received a grant/fellowship award
wtotal	numeric	number of women who applied for an award
maward	numeric	number of men who received a grant/fellowship award
mtotal	numeric	number of men who applied for an award

### Details

The studies in this dataset examine whether the chances of receiving a grant or fellowship award differs for men and women. Note that many studies provide multiple comparisons (e.g., for different years / cohorts / disciplines). A multilevel meta-analysis model can be used to account for the multilevel structure in these data.

# Concepts

sociology, odds ratios, multilevel models

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Bornmann, L., Mutz, R., & Daniel, H. (2007). Gender differences in grant peer review: A metaanalysis. *Journal of Informetrics*, 1(3), 226–238. https://doi.org/10.1016/j.joi.2007.03.001

### References

Marsh, H. W., Bornmann, L., Mutz, R., Daniel, H.-D., & O'Mara, A. (2009). Gender effects in the peer reviews of grant proposals: A comprehensive meta-analysis comparing traditional and multi-level approaches. *Review of Educational Research*, **79**(3), 1290–1326. https://doi.org/10.3102/0034654309334143

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.bornmann2007
head(dat, 16)
## Not run:
### load metafor package
library(metafor)
### calculate log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=waward, n1i=wtotal, ci=maward, n2i=mtotal, data=dat)
### fit multilevel meta-analysis model
res <- rma.mv(yi, vi, random = ~ 1 | study/obs, data=dat)</pre>
res
### estimated average odds ratio (with 95% CI/PI)
predict(res, transf=exp, digits=2)
### test for a difference between fellowship and grant applications
res <- rma.mv(yi, vi, mods = ~ type, random = ~ 1 | study/obs, data=dat)</pre>
res
predict(res, newmods=0:1, transf=exp, digits=2)
## End(Not run)
```

dat.bourassa1996 Studies on the Association between Handedness and Eye-Dominance

## Description

Results from 47 studies on the association between handedness and eye-dominance.

### Usage

```
dat.bourassa1996
```

#### Format

The data frame contains the following columns:

study	numeric	study number
sample	numeric	sample number

### dat.bourassa1996

author	character	(first) author
year	numeric	publication year
selection	character	selection of subjects on the basis of eyedness or handedness
investigator	character	investigator (psychologist, educationalist, or other)
hand_assess	character	method to assess handedness (questionnaire or performance based)
eye_assess	character	method to assess eyedness (see 'Details')
mage	numeric	mean age of sample
lh.le	numeric	number of left-handed left-eyed individuals
lh.re	numeric	number of left-handed right-eyed individuals
rh.le	numeric	number of right-handed left-eyed individuals
rh.re	numeric	number of right-handed right-eyed individuals
sex	character	sex of the sample (combined, male, or female)

## Details

The 47 studies included in this meta-analysis examined the association between handedness and eye-dominance (ocular dominance or eyedness). Results are given in terms of  $2 \times 2$  tables, indicating the number of left-handed left-eyed, left-handed right-eyed, right-handed left-eyed, and right-handed right-eyed individuals in each sample. Note that some studies included multiple (independent) samples, so that the meta-analysis included 54 samples in total. Also, for some studies, the combined data of the males and females are further broken down into the two subgroups.

In some studies, there was indication that the selection of subjects was not random with respect to handedness and/or eyedness. While this should not influence the size of the association as measured with the odds ratio, this invalidates those studies for assessing the overall percentage of left-eyed and left-handed individuals.

Handedness was assessed in the individual studies either based on a questionnaire or based on task performance. Eyedness was assessed based on various methods: E.1 methods are based on task performance, while E.2.a denotes assessment based on a questionnaire. The performance based methods could be further broken down into: E.1.a.i (monocular procedure with object/instrument held in one hand), E.1.a.ii (monocular procedure with object/instrument held in both hands), E.1.b (binocular procedure), E.1.c (a combination of the previous methods), and E.1.d (some other method).

# Concepts

psychology, odds ratios, multilevel models

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Bourassa, D. C., McManus, I. C., & Bryden, M. P. (1996). Handedness and eye-dominance: A meta-analysis of their relationship. *Laterality*, **1**(1), 5–34. https://doi.org/10.1080/713754206

### Examples

```
### copy data into 'dat' and examine data
dat <- dat.bourassa1996
head(dat, 10)</pre>
```

## Not run:
### load metafor package
library(metafor)

### calculate log(OR) and corresponding sampling variance with 1/2 correction
dat <- escalc(measure="OR", ai=lh.le, bi=lh.re, ci=rh.le, di=rh.re, data=dat, add=1/2, to="all")
head(dat, 10)</pre>

```
### overall association between handedness and eyedness
res <- rma(yi, vi, data=dat, subset=sex=="combined")
res
predict(res, transf=exp, digits=2)</pre>
```

```
### multilevel model to account for heterogeneity at the study and sample levels
res <- rma.mv(yi, vi, random = ~ 1 | study/sample, data=dat, subset=sex=="combined")
res
predict(res_transf=ovp_digits=2)</pre>
```

```
predict(res, transf=exp, digits=2)
```

```
### restructure the dataset to keep only the male/female data when it is reported
### separately and the combined data when this is the only data reported
dat <- lapply(split(dat, dat$sample), function(x) {</pre>
   if (nrow(x) == 3L) {
      x[-which(x$sex == "combined"),]
   } else {
      Х
   }
})
dat <- do.call(rbind, dat)</pre>
rownames(dat) <- NULL</pre>
dat
### multilevel model to account for heterogeneity at the study, sample, and subgroup levels
res <- rma.mv(yi, vi, random = ~ 1 | study/sample/sex, data=dat)</pre>
res
predict(res, transf=exp, digits=2)
## End(Not run)
```

dat.cannon2006

Studies on the Effectiveness of Intensive Versus Moderate Statin Therapy for Preventing Coronary Death or Myocardial Infarction

#### Description

Results from 4 trials examining the effectiveness of intensive (high dose) versus moderate (standard dose) statin therapy for preventing coronary death or myocardial infarction.

### 42

## dat.cannon2006

### Usage

dat.cannon2006

## Format

The data frame contains the following columns:

trial	character	trial name
рор	character	study population (post-ACS: post acute coronary syndrome; stable CAD: stable coronary artery disease)
nt	numeric	number of patients in the high dose group
nc	numeric	number of patients in the standard dose group
ep1t	numeric	number of events in the high dose group for end point 1: coronary death or non-fatal myocardial infarcti
ep1c	numeric	number of events in the standard dose group for end point 1: coronary death or non-fatal myocardial infa
ep2t	numeric	number of events in the high dose group for end point 2: coronary death or any cardiovascular event (M
ep2c	numeric	number of events in the standard dose group for end point 2: coronary death or any cardiovascular event
ep3t	numeric	number of events in the high dose group for end point 3: cardiovascular death
ep3c	numeric	number of events in the standard dose group for end point 3: cardiovascular death
ep4t	numeric	number of events in the high dose group for end point 4: non-cardiovascular death
ep4c	numeric	number of events in the standard dose group for end point 4: non-cardiovascular death
ep5t	numeric	number of events in the high dose group for end point 5: deaths (all-cause mortality)
ep5c	numeric	number of events in the standard dose group for end point 5: deaths (all-cause mortality)
ep6t	numeric	number of events in the high dose group for end point 6: stroke
ep6c	numeric	number of events in the standard dose group for end point 6: stroke

# Details

The data were obtained from Figures 2, 3, 4, and 5 in Cannon et al. (2006). The authors used the Mantel-Haenszel method for combining the results from the 4 trials. This approach is implemented in the rma.mh function.

# Concepts

medicine, cardiology, odds ratios, Mantel-Haenszel method

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Cannon, C. P., Steinberg, B. A., Murphy, S. A., Mega, J. L., & Braunwald, E. (2006). Meta-analysis of cardiovascular outcomes trials comparing intensive versus moderate statin therapy. *Journal of the American College of Cardiology*, **48**(3), 438–445. https://doi.org/10.1016/j.jacc.2006.04.070

```
### copy data into 'dat' and examine data
dat <- dat.cannon2006
dat</pre>
```

dat.cohen1981 Studies on the Relationship between Course Instructor Ratings and Student Achievement

# Description

Results from 20 studies on the correlation between course instructor ratings and student achievement.

## Usage

dat.cohen1981

### Format

The data frame contains the following columns:

study	character	study author(s) and year
sample	character	course type
control	character	ability control
ni	numeric	sample size of the study (number of sections)
ri	numeric	observed correlation

### Details

The studies included in this dataset examined to what extent students' ratings of a course instructor correlated with their achievement in the course. Instead of correlating individual ratings and achievement scores, the studies were carried out in multisection courses, in which the sections had different instructors but all sections used a common achievement measure (e.g., a final exam). The correlation coefficients reflect the correlation between the mean instructor rating and the mean achievement score of each section. Hence, the unit of analysis are the sections, not the individuals.

## dat.cohen1981

Note that this dataset (extracted from Table A.3 in Cooper & Hedges, 1994) only contains studies with at least 10 sections.

### Concepts

education, correlation coefficients

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Cooper, H., & Hedges, L. V. (1994). Appendix A: Data Sets. In H. Cooper & L. V. Hedges (Eds.), *The handbook of research synthesis* (pp. 543–547). New York: Russell Sage Foundation.

## References

Cohen, P. A. (1981). Student ratings of instruction and student achievement: A meta-analysis of multisection validity studies. *Review of Educational Research*, **51**(3), 281–309. https://doi.org/10.3102/003465430510

```
### copy data into 'dat' and examine data
dat <- dat.cohen1981
dat[c(1,4,5)]
## Not run:
### load metafor package
library(metafor)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat[c(1,4,5)])
dat
### meta-analysis of the transformed correlations using a random-effects model
res <- rma(yi, vi, data=dat, digits=2)
res
### predicted average correlation with 95% CI
predict(res, transf=transf.ztor)
## End(Not run)
```

dat.colditz1994

### Description

Results from 13 studies examining the effectiveness of the Bacillus Calmette-Guerin (BCG) vaccine against tuberculosis.

#### Usage

dat.colditz1994

## Format

The data frame contains the following columns:

trial	numeric	trial number
author	character	author(s)
year	numeric	publication year
tpos	numeric	number of TB positive cases in the treated (vaccinated) group
tneg	numeric	number of TB negative cases in the treated (vaccinated) group
cpos	numeric	number of TB positive cases in the control (non-vaccinated) group
cneg	numeric	number of TB negative cases in the control (non-vaccinated) group
ablat	numeric	absolute latitude of the study location (in degrees)
alloc	character	method of treatment allocation (random, alternate, or systematic assignment)

## Details

The 13 studies provide data in terms of  $2 \times 2$  tables in the form:

	TB positive	TB negative
vaccinated group	tpos	tneg
control group	cpos	cneg

The goal of the meta-analysis was to examine the overall effectiveness of the BCG vaccine for preventing tuberculosis and to examine moderators that may potentially influence the size of the effect.

The dataset has been used in several publications to illustrate meta-analytic methods (see 'References').

## Concepts

medicine, risk ratios, meta-regression

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Colditz, G. A., Brewer, T. F., Berkey, C. S., Wilson, M. E., Burdick, E., Fineberg, H. V., & Mosteller, F. (1994). Efficacy of BCG vaccine in the prevention of tuberculosis: Meta-analysis of the published literature. *Journal of the American Medical Association*, **271**(9), 698–702. https://doi.org/10.1001/jama.1994.035103

#### References

Berkey, C. S., Hoaglin, D. C., Mosteller, F., & Colditz, G. A. (1995). A random-effects regression model for meta-analysis. *Statistics in Medicine*, **14**(4), 395–411. https://doi.org/10.1002/sim.4780140406

van Houwelingen, H. C., Arends, L. R., & Stijnen, T. (2002). Advanced methods in meta-analysis: Multivariate approach and meta-regression. *Statistics in Medicine*, **21**(4), 589–624. https://doi.org/10.1002/sim.1040

Viechtbauer, W. (2010). Conducting meta-analyses in R with the metafor package. *Journal of Statistical Software*, **36**(3), 1–48. https://doi.org/10.18637/jss.v036.i03

```
### copy data into 'dat' and examine data
dat <- dat.colditz1994</pre>
dat
## Not run:
### load metafor package
library(metafor)
### calculate log risk ratios and corresponding sampling variances
dat <- escalc(measure="RR", ai=tpos, bi=tneg,</pre>
                             ci=cpos, di=cneg, data=dat,
                             slab=paste0(author, ", ", year))
dat
### random-effects model
res <- rma(yi, vi, data=dat)
res
### average risk ratio with 95% CI
predict(res, transf=exp)
### mixed-effects model with absolute latitude and publication year as moderators
res <- rma(yi, vi, mods = ~ ablat + year, data=dat)</pre>
res
### predicted average risk ratios for 10-60 degrees absolute latitude
### holding the publication year constant at 1970
predict(res, newmods=cbind(seq(from=10, to=60, by=10), 1970), transf=exp)
### note: the interpretation of the results is difficult because absolute
### latitude and publication year are strongly correlated (the more recent
### studies were conducted closer to the equator)
plot(ablat ~ year, data=dat, pch=19, xlab="Publication Year", ylab="Absolute Lattitude")
cor(dat$ablat, dat$year)
```

```
## End(Not run)
```

dat.collins1985a Studies on the Treatment of Upper Gastrointestinal Bleeding by a Histamine H2 Antagonist

### Description

Results from studies examining the effectiveness of histamine H2 antagonists (cimetidine or ranitidine) in treating patients with acute upper gastrointestinal hemorrhage.

## Usage

dat.collins1985a

### Format

The data frame contains the following columns:

id	numeric	study number
trial	character	first author of trial
year	numeric	year of publication
ref	numeric	reference number
trt	character	C = cimetidine, R = ranitidine
ctrl	character	P = placebo, AA = antacids, UT = usual treatment
nti	numeric	number of patients in treatment group
b.xti	numeric	number of patients in treatment group with persistent or recurrent bleedings
o.xti	numeric	number of patients in treatment group in need of operation
d.xti	numeric	number of patients in treatment group that died
nci	numeric	number of patients in control group
b.xci	numeric	number of patients in control group with persistent or recurrent bleedings
o.xci	numeric	number of patients in control group in need of operation
d.xci	numeric	number of patients in control group that died

# Details

The data were obtained from Tables 1 and 2 in Collins and Langman (1985). The authors used Peto's (one-step) method for meta-analyzing the 27 trials. This approach is implemented in the rma.peto function. Using the same dataset, van Houwelingen, Zwinderman, and Stijnen (1993) describe some alternative approaches for analyzing these data, including fixed- and random-effects conditional logistic models. Those are implemented in the rma.glmm function.

# Concepts

medicine, odds ratios, Peto's method, generalized linear models

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Collins, R., & Langman, M. (1985). Treatment with histamine H2 antagonists in acute upper gastrointestinal hemorrhage. *New England Journal of Medicine*, **313**(11), 660–666. https://doi.org/10.1056/NEJM1985091.

### References

van Houwelingen, H. C., Zwinderman, K. H., & Stijnen, T. (1993). A bivariate approach to metaanalysis. *Statistics in Medicine*, **12**(24), 2273–2284. https://doi.org/10.1002/sim.4780122405

```
### copy data into 'dat' and examine data
dat <- dat.collins1985a</pre>
dat
## Not run:
### load metafor package
library(metafor)
### meta-analysis of log ORs using Peto's method (outcome: persistent or recurrent bleedings)
res <- rma.peto(ai=b.xti, n1i=nti, ci=b.xci, n2i=nci, data=dat)</pre>
print(res, digits=2)
### meta-analysis of log ORs using a conditional logistic regression model (FE model)
res <- rma.glmm(measure="OR", ai=b.xti, n1i=nti, ci=b.xci, n2i=nci, data=dat,</pre>
                model="CM.EL", method="FE")
summary(res)
predict(res, transf=exp, digits=2)
### plot the likelihoods of the odds ratios
llplot(measure="OR", ai=b.xti, n1i=nti, ci=b.xci, n2i=nci, data=dat,
       lwd=1, refline=NA, xlim=c(-4,4), drop00=FALSE)
### meta-analysis of log odds ratios using a conditional logistic regression model (RE model)
res <- rma.glmm(measure="OR", ai=b.xti, n1i=nti, ci=b.xci, n2i=nci, data=dat,
                model="CM.EL", method="ML")
summary(res)
predict(res, transf=exp, digits=2)
### meta-analysis of log ORs using Peto's method (outcome: need for surgery)
res <- rma.peto(ai=o.xti, n1i=nti, ci=o.xci, n2i=nci, data=dat)</pre>
print(res, digits=2)
### meta-analysis of log ORs using Peto's method (outcome: death)
res <- rma.peto(ai=d.xti, n1i=nti, ci=d.xci, n2i=nci, data=dat)</pre>
print(res, digits=2)
## End(Not run)
```

dat.collins1985b

# Description

Results from 9 studies examining the effects of diuretics in pregnancy on various outcomes.

# Usage

dat.collins1985b

# Format

The data frame contains the following columns:

id	numeric	study number
author	character	study author(s)
year	numeric	publication year
pre.nti	numeric	number of women in treatment group followed up for pre-eclampsia outcome
pre.nci	numeric	number of women in control/placebo group followed up for pre-eclampsia outcome
pre.xti	numeric	number of women in treatment group with any form of pre-eclampsia
pre.xci	numeric	number of women in control/placebo group with any form of pre-eclampsia
oedema	numeric	dummy variable indicating whether oedema was a diagnostic criterion
fup.nti	numeric	number of women in treatment group followed up for mortality outcomes
fup.nci	numeric	number of women in control/placebo group followed up for mortality outcomes
ped.xti	numeric	number of perinatal deaths in treatment group
ped.xci	numeric	number of perinatal deaths in control/placebo group
stb.xti	numeric	number of stillbirths in treatment group
stb.xci	numeric	number of stillbirths in control/placebo group
ned.xti	numeric	number of neonatal deaths in treatment group
ned.xci	numeric	number of neonatal deaths in control/placebo group

# Details

The 9 studies in this dataset examined the effects of diuretics in pregnancy on various outcomes, including the presence of any form of pre-eclampsia, perinatal death, stillbirth, and neonatal death.

# Concepts

medicine, obstetrics, odds ratios, Peto's method

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## dat.craft2003

## Source

Collins, R., Yusuf, S., & Peto, R. (1985). Overview of randomised trials of diuretics in pregnancy. *British Medical Journal*, **290**(6461), 17–23. https://doi.org/10.1136/bmj.290.6461.17

## Examples

### copy data into 'dat' and examine data
dat <- dat.collins1985b
dat

## Not run:
### load metafor package
library(metafor)

### calculate (log) odds ratio and sampling variance
dat <- escalc(measure="OR", n1i=pre.nti, n2i=pre.nci, ai=pre.xti, ci=pre.xci, data=dat)
summary(dat, digits=2, transf=exp)

### meta-analysis using Peto's method for any form of pre-eclampsia
rma.peto(n1i=pre.nti, n2i=pre.nci, ai=pre.xti, ci=pre.xci, data=dat, digits=2)

### meta-analysis including only studies where oedema was not a diagnostic criterion
rma.peto(n1i=pre.nti, n2i=pre.nci, ai=pre.xti, ci=pre.xci, data=dat, digits=2, subset=(oedema==0))</pre>

### meta-analyses of mortality outcomes (perinatal deaths, stillbirths, and neonatal deaths)
rma.peto(n1i=fup.nti, n2i=fup.nci, ai=ped.xti, ci=ped.xci, data=dat, digits=2)
rma.peto(n1i=fup.nti, n2i=fup.nci, ai=stb.xti, ci=stb.xci, data=dat, digits=2)
rma.peto(n1i=fup.nti, n2i=fup.nci, ai=ned.xti, ci=ned.xci, data=dat, digits=2)

## End(Not run)

dat.craft2003 Studies on the Relationship between the Competitive State Anxiety Inventory-2 and Sport Performance

## Description

Results from 10 studies on the relationship between the Competitive State Anxiety Inventory-2 (CSAI-2) and sport performance.

### Usage

dat.craft2003

## Format

The data frame contains the following columns:

study	numeric	study number
ni	numeric	sample size
sport	character	type of sport (T = team sport, I = individual sport)
ri	numeric	correlation coefficient
var1	character	variable 1 of the correlation coefficient (see 'Details')
var2	character	variable 2 of the correlation coefficient (see 'Details')

## Details

The 10 studies included in this dataset are a subset of the studies included in the meta-analysis by Craft et al. (2003) on the relationship between the Competitive State Anxiety Inventory-2 (CSAI-2) and sport performance.

The CSAI-2 has three subscales: cognitive anxiety (acog), somatic anxiety (asom), and self-confidence (conf). The studies included in this dataset administered the CSAI-2 prior to some sport competition and then measured sport performance based on the competition. Most studies provided all 6 correlations (3 for the correlations among the 3 subscales and 3 for the correlations between the subscales and sport performance), but 2 studies (with study numbers 6 and 17) only provided a subset.

#### Concepts

psychology, correlation coefficients, multivariate models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Becker, B. J., & Aloe, A. M. (2019). Model-based meta-analysis and related approaches. In H. Cooper, L. V. Hedges, & J. C. Valentine (Eds.), *The handbook of research synthesis and meta-analysis* (3nd ed., pp. 339–363). New York: Russell Sage Foundation.

### References

Craft, L. L., Magyar, T. M., Becker, B. J., & Feltz, D. L. (2003). The relationship between the Competitive State Anxiety Inventory-2 and sport performance: A meta-analysis. *Journal of Sport and Exercise Psychology*, **25**(1), 44–65. https://doi.org/10.1123/jsep.25.1.44

### Examples

```
### copy data into 'dat' and examine data
dat <- dat.craft2003
head(dat, 18)
## Not run:
### load matefea methems</pre>
```

### load metafor package
library(metafor)

### construct dataset and var-cov matrix of the correlations

## dat.crede2010

```
tmp <- rcalc(ri ~ var1 + var2 | study, ni=ni, data=dat)</pre>
V <- tmp$V
dat <- tmp$dat
### examine data for study 1
dat[dat$study == 1,]
V[dat$study == 1, dat$study == 1]
### examine data for study 6
dat[dat$study == 6,]
V[dat$study == 6, dat$study == 6]
### examine data for study 17
dat[dat$study == 17,]
V[dat$study == 17, dat$study == 17]
### multivariate random-effects model
res <- rma.mv(yi, V, mods = ~ 0 + var1.var2, random = ~ var1.var2 | study, struct="UN", data=dat)</pre>
res
## End(Not run)
```

dat.crede2010	
---------------	--

Studies on the Relationship between Class Attendance and Grades in College Students

## Description

Results from 68 studies on the relationship between class attendence and class performance and/or grade point average in college students.

### Usage

dat.crede2010

# Format

The data frame contains the following columns:

studyid	numeric	study number
year	numeric	publication year
source	character	study source (journal, dissertation, other)
sampleid	numeric	sample within study number
criterion	character	criterion variable (grade, gpa)
class	character	class type (science, nonscience)
ni	numeric	sample size
ri	numeric	observed correlation

## Details

The 68 studies included in this dataset provide information about the relationship between class attendance of college students and their performance (i.e., grade) in the class and/or their overall grade point average. Some studies included multiple samples and hence the dataset actually contains 97 correlation coefficients.

The dataset was obtained via personal communication. Note that this dataset differs just slightly from the one used by Credé et al. (2010).

## Concepts

education, correlation coefficients, multilevel models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Personal communication.

### References

Credé, M., Roch, S. G., & Kieszczynka, U. M. (2010). Class attendance in college: A meta-analytic review of the relationship of class attendance with grades and student characteristics. *Review of Educational Research*, **80**(2), 272–295. https://doi.org/10.3102/0034654310362998

# Examples

### copy data into 'dat' and examine data
dat <- dat.crede2010
head(dat, 18)</pre>

## Not run: ### load metafor package library(metafor)

### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat)</pre>

```
*****
```

```
### meta-analysis for the relationship between attendance and grades
res <- rma(yi, vi, data=dat, subset=criterion=="grade")
res</pre>
```

### estimated average correlation with 95% CI/PI
predict(res, transf=transf.ztor, digits=2)

```
### examine if relationship between attendance and grades differs for nonscience/science classes
res <- rma(yi, vi, mods = ~ class, data=dat, subset=criterion=="grade")
res</pre>
```

```
### estimated average correlations for nonscience and science classes
predict(res, newmods=c(0,1), transf=transf.ztor, digits=2)
### examine if relationship between attendance and grades has changed over time
res <- rma(yi, vi, mods = ~ year, data=dat, subset=criterion=="grade")</pre>
res
### meta-analysis for the relationship between attendance and GPA
res <- rma(yi, vi, data=dat, subset=criterion=="gpa")</pre>
res
### estimated average correlation with 95% CI/PI
predict(res, transf=transf.ztor, digits=2)
### examine if relationship between attendance and GPA has changed over time
res <- rma(yi, vi, mods = ~ year, data=dat, subset=criterion=="gpa")</pre>
res
### use a multilevel model to examine the relationship between attendance and grades
res <- rma.mv(yi, vi, random = ~ 1 | studyid/sampleid, data=dat, subset=criterion=="grade")
res
predict(res, transf=transf.ztor, digits=2)
### use a multilevel model to examine the relationship between attendance and gpa
res <- rma.mv(yi, vi, random = ~ 1 | studyid/sampleid, data=dat, subset=criterion=="gpa")</pre>
res
predict(res, transf=transf.ztor, digits=2)
## End(Not run)
```

dat.crisafulli2020 Duchenne Muscular Dystrophy (DMD) Prevalence Data

### Description

26 studies reporting estimates of the birth prevalence of Duchenne muscular dystrophy.

## Usage

```
dat.crisafulli2020
```

#### Format

The data frame contains the following columns:

study	character	study label (first author, year)
pubyear	integer	publication year
country	factor	origin of investigated population
from, to	integer	time span of investigation (years)
cases	integer	number of DMD cases
total	integer	corresponding total population

### Details

*Duchenne muscular dystrophy* (DMD) is a rare disease that is caused by a genetic mutation and is characterized by impairment through muscle weakness and a reduced life expectancy.

Crisafulli et al. (2020) reported on a systematic review of data on the epidemiology of DMD, including estimates of the *birth prevalence* (which is of the order of a few per ten thousand). One of the originally reported studies (Koenig, 2019) is omitted here, as it constitutes an obvious outlier, and the reliability of the reported data is doubtful; Crisafulli et al. (2020) pointed out that "Concerning birth prevalence, Koenig et al. were found to be outliers. This study had problems with data collection in the last study year, as due to privacy issues, DMD cases were under-reported."

# Concepts

medicine, epidemiology, proportions, dose-response models

### Author(s)

Christian Roever, <christian.roever@med.uni-goettingen.de>

#### Source

Crisafulli, S., Sultana, J., Fontana, A., Salvo, F., Messina, S., & Trifiro, G. (2020). Global epidemiology of Duchenne muscular dystrophy: an updated systematic review and meta-analysis. *Orphanet Journal of Rare Diseases*, **15**, 141. https://doi.org/10.1186/s13023-020-01430-8

## dat.curtin2002

```
# illustrate in a forest plot
forest(rma01, xlim=c(-12,-5))
## End(Not run)
```

dat.curtin2002

Studies on Potassium Supplementation to Reduce Diastolic Blood Pressure

# Description

Results from 21 cross-over studies evaluating the effect of potassium supplementation to reduce diastolic blood pressure.

## Usage

dat.curtin2002

## Format

The data frame contains the following columns:

author	character	first author
year	character	year of publication
Ν	integer	total sample size
mean	numeric	mean difference in diastolic blood pressure
SE	numeric	standard error
corr	numeric	within-patient correlation

## Details

Results from 21 cross-over studies evaluating the effect of potassium supplementation to reduce diastolic blood pressure (Curtin et al., 2002, Table II).

This data set is used as an example in Schwarzer et al. (2015), Chapter 2.

# Concepts

raw mean differences

# Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Curtin, F., Altman, D. G., & Elbourne, D. (2002). Meta-analysis combining parallel and cross-over clinical trials. I: Continuous outcomes. *Statistics in Medicine*, **21**(15), 2131–2144. https://doi.org/10.1002/sim.1205

# References

Schwarzer, G., Carpenter, J. R., & Rücker, G. (2015). *Meta-analysis with R.* Cham, Switzerland: Springer.

# Examples

dat.curtis1998 Studies on the Effects of Elevated CO2 Levels on Woody Plant Mass

# Description

Results from studies examining the effects of elevated CO2 levels on woody plant mass.

## Usage

```
dat.curtis1998
```

#### Format

The data frame contains the following columns:

id	numeric	observation number
paper	numeric	paper number

## dat.curtis1998

genus	character	genus name
species	character	species name
fungrp	character	plant functional group
co2.ambi	numeric	ambient CO2 level (control group)
co2.elev	numeric	elevated CO2 level (treatment group)
units	character	units for CO2 exposure levels
time	numeric	maximum length of time (days) of CO2 exposure
pot	character	growing method (see 'Details')
method	character	CO2 exposure facility (see 'Details')
stock	character	planting stock code
xtrt	character	interacting treatment code (see 'Details')
level	character	interacting treatment level codes (see 'Details')
m1i	numeric	mean plant mass under elevated CO2 level (treatment group)
sd1i	numeric	standard deviation of plant mass underelevated CO2 level (treatment group)
n1i	numeric	number of observations under elevated CO2 level (treatment group)
m2i	numeric	mean plant mass under ambient CO2 level (control group)
sd2i	numeric	standard deviation of plant mass under ambient CO2 level (control group)
n2i	numeric	number of observations under ambient CO2 level (control group)

# Details

The studies included in this dataset compared the total above- plus below-ground biomass (in grams) for plants that were either exposed to ambient (around 35 Pa) and elevated CO2 levels (around twice the ambient level). The co2.ambi and co2.elev variables indicate the CO2 levels in the control and treatment groups, respectively (with the units variable specifying the units for the CO2 exposure levels). Many of the studies also varied one or more additional environmental variables (defined by the xtrt and level variables):

- NONE = no additional treatment factor
- FERT = soil fertility (either a CONTROL, HIGH, or LOW level)
- LIGHT = light treatment (always a LOW light level)
- FERT+L = soil fertility and light (a LOW light and soil fertility level)
- H2O = well watered vs drought (either a WW or DRT level)
- TEMP = temperature treatment (either a HIGH or LOW level)
- OZONE = ozone exposure (either a HIGH or LOW level)
- UVB = ultraviolet-B radiation exposure (either a HIGH or LOW level)

In addition, the studies differed with respect to various design variables, including CO2 exposure duration (time), growing method (pot: number = pot size in liters; GRND = plants rooted in ground; HYDRO = solution or aeroponic culture), CO2 exposure facility (method: GC = growth chamber; GH = greenhouse; OTC = field-based open-top chamber), and planting stock (stock: SEED = plants started from seeds; SAP = plants started from cuttings). The goal of the meta-analysis was to examine the effects of elevated CO2 levels on plant physiology and growth and the interacting effects of the environmental (and design) variables.

## Concepts

ecology, ratios of means

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Hedges, L. V., Gurevitch, J., & Curtis, P. S. (1999). The meta-analysis of response ratios in experimental ecology. *Ecology*, **80**(4), 1150–1156. https://doi.org/10.1890/0012-9658(1999)080[1150:TMAORR]2.0.C0;2 (data obtained from *Ecological Archives*, E080-008-S1, at: https://doi.org/10.6084/m9.figshare.c.3297278)

## References

Curtis, P. S., & Wang, X. (1998). A meta-analysis of elevated CO2 effects on woody plant mass, form, and physiology. *Oecologia*, **113**(3), 299–313. https://doi.org/10.1007/s004420050381

```
### copy data into 'dat' and examine data
dat <- dat.curtis1998</pre>
head(dat)
## Not run:
### load metafor package
library(metafor)
### calculate (log transformed) ratios of means and corresponding sampling variances
dat <- escalc(measure="ROM", m1i=m1i, sd1i=sd1i, n1i=n1i,</pre>
                              m2i=m2i, sd2i=sd2i, n2i=n2i, data=dat)
head(dat)
### meta-analysis using a random-effects model
res <- rma(yi, vi, method="DL", data=dat)</pre>
res
### average ratio of means with 95% CI
predict(res, transf=exp, digits=2)
### meta-analysis for plants grown under nutrient stress
res <- rma(yi, vi, method="DL", data=dat, subset=(xtrt=="FERT" & level=="LOW"))</pre>
predict(res, transf=exp, digits=2)
### meta-analysis for plants grown under low light conditions
res <- rma(yi, vi, method="DL", data=dat, subset=(xtrt=="LIGHT" & level=="LOW"))</pre>
predict(res, transf=exp, digits=2)
## End(Not run)
```

dat.dagostino1998

Studies on the Effectiveness of Antihistamines in Reducing Symptoms of the Common Cold

## Description

Results from 9 studies on the effectiveness of antihistamines in reducing the severity of runny nose and sneezing in the common cold.

## Usage

dat.dagostino1998

# Format

The data frame contains the following columns:

study	numeric	study id
cold	character	natural or induced cold study
scale.rn	character	scale for measuring runny nose severity
scale.sn	character	scale for measuring sneezing severity
drug	character	type of antihistamine studied
tnt	numeric	total sample size of the treatment group
tnc	numeric	total sample size of the control (placebo) group
outcome	character	outcome variable (see 'Details')
mt	numeric	mean in the treatment group
sdt	numeric	SD in the treatment group
mc	numeric	mean in the control group
sdc	numeric	SD in the control group
xt	numeric	number of patients reaching the therapy goal in the treatment group
xc	numeric	number of patients reaching the therapy goal in the control (placebo) group
nt	numeric	sample size of the treatment group for measuring the outcome
nc	numeric	sample size of the control group for measuring the outcome

# Details

The studies for this meta-analysis were assembled to examine the effectiveness of antihistamines in reducing the severity of runny nose and sneezing in the common cold. Effectiveness was measured after one and two days of treatment in terms of 4 different outcome variables:

- 1. rnic1 and rnic2 (continuous): incremental change (improvement) in runny nose severity at day 1 and day 2,
- 2. rngoal1 and rngoal2 (dichotomous): reaching the goal of therapy (of at least a 50% reduction in runny nose severity) at day 1 and day 2,
- 3. snic1 and snic2 (continuous): incremental change (improvement) in sneezing severity at day 1 and day 2, and

4. rngoal1 and rngoal2 (dichotomous): reaching the goal of therapy (of at least a 50% reduction in sneezing severity) at day 1 and day 2.

For the continuous outcomes, standardized mean differences can be computed to quantify the difference between the treatment and control groups. For the dichotomous outcomes, one can compute (log) odds ratios to quantify the difference between the treatment and control groups.

### Concepts

medicine, standardized mean differences, odds ratios, multivariate models

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

D'Agostino, R. B., Sr., Weintraub, M., Russell, H. K., Stepanians, M., D'Agostino, R. B., Jr., Cantilena, L. R., Jr., Graumlich, J. F., Maldonado, S., Honig, P., & Anello, C. (1998). The effectiveness of antihistamines in reducing the severity of runny nose and sneezing: A meta-analysis. *Clinical Pharmacology & Therapeutics*, **64**(6), 579–596. https://doi.org/10.1016/S0009-9236(98)90049-2

```
### copy data into 'dat' and examine data
dat <- dat.dagostino1998</pre>
head(dat, 16)
## Not run:
### load metafor package
library(metafor)
### compute standardized mean differences and corresponding sampling variances
dat <- escalc(measure="SMD", m1i=mt, m2i=mc, sd1i=sdt, sd2i=sdc, n1i=nt, n2i=nc, data=dat,
              add.measure=TRUE)
### compute log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=xt, ci=xc, n1i=nt, n2i=nc, data=dat,</pre>
              replace=FALSE, add.measure=TRUE, add=1/2, to="all")
### inspect data for the first study
head(dat, 8)
### fit a random-effects model for incremental change in runny nose severity at day 1
res <- rma(yi, vi, data=dat, subset=outcome=="rnic1")</pre>
res
### fit a random-effects model for reaching the goal of therapy for runny nose severity at day 1
res <- rma(yi, vi, data=dat, subset=outcome=="rngoal1")</pre>
res
predict(res, transf=exp)
```

```
### construct approximate V matrix assuming a correlation of 0.7 for sampling errors within studies
dat$esid <- ave(dat$study, dat$study, FUN=seq)
V <- vcalc(vi, cluster=study, obs=esid, rho=0.7, data=dat)
### fit a model for incremental change in runny nose severity at day 1 and at day 2, allowing for
### correlated sampling errors (no random effects added, since there does not appear to be any
### noteworthy heterogeneity in these data)
res <- rma.mv(yi, V, mods = ~ 0 + outcome, data=dat, subset=outcome %in% c("rnic1","rnic2"))
res
### test if there is a difference in effects at day 1 and day 2
anova(res, X=c(1,-1))
## End(Not run)
```

dat.damico2009 Studies on Topical plus Systemic Antibiotics to Prevent Respiratory Tract Infections

## Description

Results from 16 studies examining the effectiveness of topical plus systemic antibiotics to prevent respiratory tract infections (RTIs).

#### Usage

dat.damico2009

# Format

The data frame contains the following columns:

study	character	first author
year	numeric	publication year
xt	numeric	number of RTIs in the treatment group
nt	numeric	number of patients in the treatment group
xc	numeric	number of RTIs in the control group
nc	numeric	number of patients in the control group
conceal	numeric	allocation concealment ( $0 = not$ adequate, $1 = adequate$ )
blind	numeric	blinding ( $0 = open, 1 = double-blind$ )

## Details

The dataset includes the results from 16 studies that examined the effectiveness of topical plus systemic antibiotics versus no prophylaxis to prevent respiratory tract infections (RTIs).

## Concepts

medicine, odds ratios

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

D'Amico, R., Pifferi, S., Torri, V., Brazzi, L., Parmelli, E., & Liberati, A. (2009). Antibiotic prophylaxis to reduce respiratory tract infections and mortality in adults receiving intensive care. *Cochrane Database of Systematic Reviews*, **4**, CD000022. https://doi.org/10.1002/14651858.CD000022.pub3

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.damico2009
dat

## Not run:
### load metafor package
library(metafor)
### meta-analysis of the (log) odds ratios using the Mantel-Haenszel method
rma.mh(measure="OR", ai=xt, n1i=nt, ci=xc, n2i=nc, data=dat, digits=2)
### calculate log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=xt, n1i=nt, ci=xc, n2i=nc, data=dat)
### meta-analysis using a random-effects model
res <- rma(yi, vi, data=dat, method="DL")
res
predict(res, transf=exp, digits=2)
## End(Not run)</pre>
```

dat.debruin2009 Studies on Standard Care Quality and HAART-Adherence

# Description

Results from 13 trials providing information about standard care quality and HAART-adherence in control groups.

#### Usage

```
dat.debruin2009
```

#### Format

The data frame contains the following columns:

## dat.debruin2009

author	character	(first) author of study
year	numeric	publication year
scq	numeric	standard care quality
ni	numeric	number of patients in the standard care group
xi	numeric	number of patients with an undetectable viral load in standard care group
mi	numeric	number of patients with a detectable viral load in standard care group
ethnicity	character	dominant ethnicity of the patients in the standard care group
patients	character	inclusion of patients continuing or starting (a new) treatment
select	character	baseline selection of patients with adherence problems or no selection
sens	character	sensitivity of viral load assessments (<400 vs. >=400 copies/ml)

#### Details

Highly active antiretroviral therapy (HAART) refers to a combination of multiple antiretroviral drugs that can effectively suppress the HIV virus. However, achieving viral suppression (to the point that the virus becomes essentially undetectable in a blood sample) requires high levels of adherence to an often complicated medication regimen. A number of trials have examined various interventions that aim to increase adherence levels. In each trial, patients receiving the intervention are compared to patients in a control group receiving standard care (often referred to as 'care as usual'). However, the quality of standard care can vary substantially between these studies. de Bruin et al. (2009) assessed the quality of standard care provided (based on a quantification of the number of behavior change techniques applied) and examined to what extent the quality of standard care was related to the proportion of patients achieving effective viral suppression in the control groups.

#### Concepts

psychology, medicine, proportions, single-arm studies, meta-regression

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

de Bruin, M., Viechtbauer, W., Hospers, H. J., Schaalma, H. P., & Kok, G. (2009). Standard care quality determines treatment outcomes in control groups of HAART-adherence intervention studies: Implications for the interpretation and comparison of intervention effects. *Health Psychology*, **28**(6), 668–674. https://doi.org/10.1037/a0015989

```
### copy data into 'dat' and examine data
dat <- dat.debruin2009
dat
## Not run:
### load metafor package
library(metafor)</pre>
```

```
### calculate proportions and corresponding sampling variances
dat <- escalc(measure="PR", xi=xi, ni=ni, data=dat)
dat
### random-effects model
res <- rma(yi, vi, data=dat)
print(res, digits=2)
### mixed-effects meta-regression model with all predictors/covariates
res <- rma(yi, vi, mods = ~ scq + ethnicity + patients + select + sens, data=dat)
print(res, digits=3)
### mixed-effects meta-regression model with scq and ethnicity as predictors/covariates
res <- rma(yi, vi, mods = ~ scq + ethnicity, data=dat)
print(res, digits=3)
## End(Not run)
```

dat.dogliotti2014 Studies on Antithrombotic Treatments to Prevent Strokes

### Description

Results from 20 trials examining the effectiveness of antithrombotic treatments to prevent strokes in patients with non-valvular atrial fibrillation.

## Usage

dat.dogliotti2014

### Format

The data frame contains the following columns:

study	character	study label
id	numeric	study ID
treatment	character	treatment
stroke	numeric	number of strokes
total	numeric	number of individuals

## Details

This dataset comes from a systematic review aiming to estimate the effects of eight antithrombotic treatments including placebo in reducing the incidence of major thrombotic events in patients with non-valvular atrial fibrillation (Dogliotti et al., 2014).

The review included 20 studies with 79,808 participants, four studies are three-arm studies. The primary outcome is stroke reduction (yes / no).

```
66
```

## Concepts

medicine, odds ratios, network meta-analysis, Mantel-Haenszel method

### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Dogliotti, A., Paolasso, E., & Giugliano, R. P. (2014). Current and new oral antithrombotics in non-valvular atrial fibrillation: A network meta-analysis of 79808 patients. *Heart*, **100**(5), 396–405. https://doi.org/10.1136/heartjnl-2013-304347

### Examples

```
### Show first 7 rows / 3 studies of the dataset
head(dat.dogliotti2014, 7)
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
### Change appearance of confidence intervals
cilayout("(", "-")
### Transform data from long arm-based format to contrast-based
### format. Argument 'sm' has to be used for odds ratio as summary
### measure; by default the risk ratio is used in the metabin function
### called internally.
pw <- pairwise(treat = treatment, n = total, event = stroke,</pre>
  studlab = study, data = dat.dogliotti2014, sm = "OR")
### Print log odds ratios (TE) and standard errors (seTE)
head(pw, 5)[, 1:5]
### Conduct network meta-analysis (NMA) with placebo as reference
net <- netmeta(pw, ref = "plac")</pre>
### Details on excluded study
selvars <- c("studlab", "event1", "n1", "event2", "n2")</pre>
subset(pw, studlab == "WASPO, 2007")[, selvars]
### Show network graph
netgraph(net, seq = "optimal", number = TRUE)
### Conduct Mantel-Haenszel NMA
net.mh <- netmetabin(pw, ref = "plac")</pre>
```

### Compare results of inverse variance and Mantel-Haenszel NMA

```
nb <- netbind(net, net.mh, random = FALSE,
    name = c("Inverse variance", "Mantel-Haenszel"))
forest(nb, xlim = c(0.15, 2), at = c(0.2, 0.5, 1, 2))
### Print and plot results for inverse variance NMA
net
forest(net)
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.dong2013	Studies on Safety of Inhaled Medications for Chronic Obstructive Pul-
	monary Disease

# Description

Results from 41 trials examining the safety of inhaled medications in patients with chronic obstructive pulmonary disease.

## Usage

dat.dong2013

# Format

The data frame contains the following columns:

. .

id	integer	study ID
treatment	character	treatment
death	integer	mortality
randomized	integer	number of individuals

. ....

## Details

This network meta-analysis compared the safety of inhaled medications in patients with chronic obstructive pulmonary disease (Dong et al., 2013).

Mortality was reported in 41 randomized trials, with a total of 52 462 patients. Mortality was low, with 2 408 deaths (4.6%) reported across all studies. There were nine studies that reported zero events in at least one of the treatment arms and three additional studies had zero events in all treatment arms.

This dataset was used in Efthimiou et al. (2019) to illustrate the Mantel-Haenszel method for network meta-analysis.

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#### dat.dong2013

### Concepts

medicine, odds ratios, network meta-analysis, Mantel-Haenszel method

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Dong, Y.-H., Lin, H.-H., Shau, W.-Y., Wu, Y.-C., Chang, C.-H., & Lai, M.-S. (2013). Comparative safety of inhaled medications in patients with chronic obstructive pulmonary disease: Systematic review and mixed treatment comparison meta-analysis of randomised controlled trials. *Thorax*, **68**(1), 48–56. https://doi.org/10.1136/thoraxjnl-2012-201926

## References

Efthimiou, O., Rücker, G., Schwarzer, G., Higgins, J., Egger, M., & Salanti, G. (2019). A Mantel-Haenszel model for network meta-analysis of rare events. *Statistics in Medicine*, **38**(16), 2992–3012. https://doi.org/10.1002/sim.8158

### Examples

### Show first 6 rows / 3 studies of the dataset head(dat.dong2013)

```
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
```

```
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
```

```
### Change appearance of confidence intervals
cilayout("(", "-")
```

```
### Network graph
netgraph(net, seq = "optimal", col = "black", plastic = FALSE,
points = TRUE, pch = 21, cex.points = 3, col.points = "black",
```

```
bg.points = "gray", thickness = "se.fixed",
number.of.studies = TRUE)
### Show results for Mantel-Haenszel NMA
net
forest(net)
### League table with network estimates in lower triangle and direct
### estimates in upper triangle
netleague(net)
### Assess inconsistency
print(netsplit(net), show = "both", ci = TRUE, overall = FALSE,
nchar.trts = 6)
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.dorn2007

Studies on Complementary and Alternative Medicine for Irritable Bowel Syndrome

### Description

Results from 19 trials examining complementary and alternative medicine (CAM) for irritable bowel syndrome (IBS).

#### Usage

dat.dorn2007

# Format

The data frame contains the following columns:

id	numeric	trial id number
study	character	(first) author
year	numeric	publication year
country	character	country where trial was conducted
ibs.crit	character	IBS diagnostic criteria (Manning, Rome I, Rome II, or Other)
days	numeric	number of treatment days
visits	numeric	number of practitioner visits
jada	numeric	Jadad score
x.a	numeric	number of responders in the active treatment group
n.a	numeric	number of participants in the active treatment group
x.p	numeric	number of responders in the placebo group
n.p	numeric	number of participants in the placebo group

#### dat.dorn2007

## Details

The dataset includes the results from 19 randomized clinical trials that examined the effectiveness of complementary and alternative medicine (CAM) for irritable bowel syndrome (IBS).

## Concepts

medicine, alternative medicine, risk ratios

## Note

The data were extracted from Table I in Dorn et al. (2009). Comparing the funnel plot in Figure 1 with the one obtained below indicates that the data for study 5 (Davis et al., 2006) in the table were not the ones that were used in the actual analyses.

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Dorn, S. D., Kaptchuk, T. J., Park, J. B., Nguyen, L. T., Canenguez, K., Nam, B. H., Woods, K. B., Conboy, L. A., Stason, W. B., & Lembo, A. J. (2007). A meta-analysis of the placebo response in complementary and alternative medicine trials of irritable bowel syndrome. *Neurogastroenterology* & *Motility*, **19**(8), 630–637. https://doi.org/10.1111/j.1365-2982.2007.00937.x

```
### copy data into 'dat' and examine data
dat <- dat.dorn2007
dat
## Not run:
### load metafor package
library(metafor)
### calculate log risk ratios and corresponding sampling variances
dat <- escalc(measure="RR", ai=x.a, n1i=n.a, ci=x.p, n2i=n.p, data=dat)</pre>
### random-effects model
res <- rma(yi, vi, data=dat, digits=2, method="DL")</pre>
res
### estimated average risk ratio
predict(res, transf=exp)
### funnel plot with study 5 highlighted in red
funnel(res, atransf=exp, at=log(c(0.1, 0.2, 0.5, 1, 2, 5, 10)),
       ylim=c(0,1), steps=6, las=1, col=ifelse(id == 5, "red", "black"))
### change log risk ratio for study 5
dat$yi[5] <- -0.44
```

## End(Not run)

dat.dumouchel1994 Nitrogen dioxide data set

### Description

Nine studies investigating the effect of NO2 exposure on respiratory illness in children.

### Usage

dat.dumouchel1994

### Format

The data frame contains the following columns:

study	character	study label
smoke	factor	adjustment for smoking (y/n)
no2	factor	direct measurement of NO2 concentration (y/n)
gender	factor	adjustment for gender (y/n)
or	numeric	odds ratio for childhood respiratory illness
lower	numeric	lower bound of 95 percent CI
upper	numeric	upper bound of 95 percent CI

# Details

Hasselblad et al. (1992) investigated the effects of nitrogen dioxide (NO2) exposure on the occurrence of respiratory illness in children. Their data were picked up by DuMouchel (1994) as an illustrative example in his article on Bayesian meta-analysis, and were also part of his "hblm" S-Plus software package. DuMouchel's dataset differs slightly from the figures quoted by Hasselblad et al. (1992), apparently because he had additional, more detailed data available.

The data set features three study-level covariables reflecting characteristics of the study designs, namely, whether the quoted estimate had been adjusted for parents' smoking status, whether NO2 exposure had been measured directly (or presence of a gas stove in the household had been used as a proxy instead), and whether the quoted effect had been adjusted for gender. Inclusion of the covariables allows to account for the studies' design features, quantify their effects, and adjust for these.
# Concepts

medicine, odds ratios, meta-regression

### Author(s)

Christian Roever, <christian.roever@med.uni-goettingen.de>

# Source

DuMouchel, W. H. (1994). Hierarchical Bayes linear models for meta-analysis. Technical Report 27, National Institute of Statistical Sciences (NISS); Research Triangle Park, NC, USA. https://www.niss.org/research/

# References

Hasselblad, V., Eddy, D. M., & Kotchmar, D. J. (1992). Synthesis of environmental evidence: Nitrogen dioxide epidemiology studies. *Journal of the Air and Waste Management Association*, **42**(5), 662–671. https://doi.org/10.1080/10473289.1992.10467018

```
# show data:
dat.dumouchel1994
## Not run:
# derive effect sizes (log-ORs):
library(metafor)
no2 <- escalc(measure="OR", yi=log(or),</pre>
               sei=(log(upper)-log(lower))/(2*qnorm(0.975)),
               slab=study, data=dat.dumouchel1994)
summary(no2)
# compute overall meta-analysis:
library(bayesmeta)
bm01 <- bayesmeta(no2, tau.prior="DuMouchel")</pre>
# show results:
bm01
forestplot(bm01)
traceplot(bm01)
# perform meta-regression;
# specify regressor matrix:
X <- model.matrix( ~ smoke + no2 + gender, data=no2)</pre>
colnames(X) <- c("intercept", "smoke", "no2", "gender")</pre>
# perform regression:
bm02 <- bmr(no2, X=X, tau.prior="DuMouchel")</pre>
# show results:
bm02
forestplot(bm02)
```

```
#forestplot(bm02, xlab="log-OR",
            X.mean=rbind("none"
                                      = c(1,0,0,0),
#
#
                          "smoke"
                                      = c(1,1,0,0),
#
                          "no2"
                                      = c(1,0,1,0),
                          "gender"
#
                                      = c(1,0,0,1),
                          "all three" = c(1,1,1,1)))
#
traceplot(bm02)
## End(Not run)
```

```
dat.egger2001
```

Studies on the Effectiveness of Intravenous Magnesium in Acute Myocardial Infarction

## Description

Results from 16 trials examining the effectiveness of intravenous magnesium in the prevention of death following acute myocardial infarction.

## Usage

dat.egger2001

### Format

The data frame contains the following columns:

id	numeric	trial id number
study	character	first author or trial name
year	numeric	publication year
ai	numeric	number of deaths in the magnesium group
n1i	numeric	number of patients in the magnesium group
ci	numeric	number of deaths in the control group
n2i	numeric	number of patients in the control group

#### Details

The dataset includes the results from 16 randomized clinical trials that examined the effectiveness of intravenous magnesium in the prevention of death following acute myocardial infarction. Studies 1-7 were included in the meta-analyses by Teo et al. (1991) and Horner (1992) and were combined with the results from the LIMIT-2 trial (Woods et al., 1992) in Yusuf et al. (1993), suggesting that magnesium is an effective treatment for reducing mortality. However, the results from the ISIS-4 mega trial (ISIS-4 Collaborative Group, 1995) indicated no reduction in mortality with magnesium treatment. Publication bias has been suggested as one possible explanation for the conflicting findings (Egger & Davey Smith, 1995).

The present dataset includes some additional trials and are based on Table 18.2 from Egger, Davey Smith, and Altman (2001).

## dat.egger2001

### Concepts

medicine, cardiology, Peto's method, publication bias

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Egger, M., Davey Smith, G., & Altman, D. G. (Eds.) (2001). Systematic reviews in health care: *Meta-analysis in context* (2nd ed.). London: BMJ Books.

### References

Egger, M., & Davey Smith, G. (1995). Misleading meta-analysis: Lessons from "an effective, safe, simple" intervention that wasn't. *British Medical Journal*, **310**(6982), 752–754. https://doi.org/10.1136/bmj.310.6982

Horner, S. M. (1992). Efficacy of intravenous magnesium in acute myocardial infarction in reducing arrhythmias and mortality: Meta-analysis of magnesium in acute myocardial infarction. *Circulation*, **86**(3), 774–779. https://doi.org/10.1161/01.cir.86.3.774

ISIS-4 Collaborative Group (1995). ISIS-4: A randomised factorial trial assessing early oral captopril, oral mononitrate, and intravenous magnesium sulphate in 58,050 patients with suspected acute myocardial infarction. *Lancet*, **345**(8951), 669–685. https://doi.org/10.1016/S0140-6736(95)90865-X

Teo, K. K., Yusuf, S., Collins, R., Held, P. H., & Peto, R. (1991). Effects of intravenous magnesium in suspected acute myocardial infarction: Overview of randomised trials. *British Medical Journal*, **303**(6816), 1499–1503. https://doi.org/10.1136/bmj.303.6816.1499

Woods, K. L., Fletcher, S., Roffe, C., & Haider, Y. (1992). Intravenous magnesium sulphate in suspected acute myocardial infarction: Results of the second Leicester Intravenous Magnesium Intervention Trial (LIMIT-2). *Lancet*, **339**(8809), 1553–1558. https://doi.org/10.1016/0140-6736(92)91828-v

Yusuf, S., Teo, K., & Woods, K. (1993). Intravenous magnesium in acute myocardial infarction: An effective, safe, simple, and inexpensive treatment. *Circulation*, **87**(6), 2043–2046. https://doi.org/10.1161/01.cir.87.

#### See Also

dat.li2007

```
### copy data into 'dat' and examine data
dat <- dat.egger2001
dat
## Not run:
### load metafor package
library(metafor)
### meta-analysis of trials 1-7 using Peto's method (as in Teo et al., 1991)
res <- rma.peto(ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, subset=1:7)
print(res, digits=2)</pre>
```

```
### meta-analysis of trials 1-7 and LIMIT-2 (as in Yusuf et al., 1993)
res <- rma.peto(ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, subset=c(1:7,14))
print(res, digits=2)
### meta-analysis of all trials except ISIS-4
res <- rma.peto(ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, subset=-16)
print(res, digits=2)
### meta-analysis of all trials including ISIS-4
res <- rma.peto(ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat)
print(res, digits=2)
### contour-enhanced funnel plot centered at 0
funnel(res, refline=0, level=c(90, 95, 99), shade=c("white", "gray", "darkgray"))
## End(Not run)</pre>
```

dat.fine1993	Studies on Radiation Therapy with or without Adjuvant Chemotherapy
	in Patients with Malignant Gliomas

#### Description

Results from 17 trials comparing post-operative radiation therapy with and without adjuvant chemotherapy in patients with malignant gliomas.

#### Usage

dat.fine1993

# Format

The data frame contains the following columns:

study	numeric	study number
nei	numeric	sample size in the experimental group receiving radiotherapy plus adjuvant chemotherapy
nci	numeric	sample size in the control group receiving radiotherapy alone
e1i	numeric	number of survivors at 6 months in the experimental group
c1i	numeric	number of survivors at 6 months in the control group
e2i	numeric	number of survivors at 12 months in the experimental group
c2i	numeric	number of survivors at 12 months in the control group
e3i	numeric	number of survivors at 18 months in the experimental group
c3i	numeric	number of survivors at 18 months in the control group
e4i	numeric	number of survivors at 24 months in the experimental group
c4i	numeric	number of survivors at 24 months in the control group

## dat.fine1993

## Details

The 17 trials report the post-operative survival of patients with malignant gliomas receiving either radiation therapy with adjuvant chemotherapy or radiation therapy alone. Survival was assessed at 6, 12, 18, and 24 months in all but one study (which assessed survival only at 12 and at 24 months).

The data were reconstructed by Trikalinos and Olkin (2012) based on Table 2 in Fine et al. (1993) and Table 3 in Dear (1994). The data can be used to illustrate how a meta-analysis can be conducted of effect sizes reported at multiple time points using a multivariate model.

# Concepts

medicine, oncology, odds ratios, longitudinal models

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Dear, K. B. G. (1994). Iterative generalized least squares for meta-analysis of survival data at multiple times. *Biometrics*, **50**(4), 989–1002. https://doi.org/10.2307/2533438

Trikalinos, T. A., & Olkin, I. (2012). Meta-analysis of effect sizes reported at multiple time points: A multivariate approach. *Clinical Trials*, **9**(5), 610–620. https://doi.org/10.1177/1740774512453218

## References

Fine, H. A., Dear, K. B., Loeffler, J. S., Black, P. M., & Canellos, G. P. (1993). Meta-analysis of radiation therapy with and without adjuvant chemotherapy for malignant gliomas in adults. *Cancer*, **71**(8), 2585–2597. https://doi.org/10.1002/1097-0142(19930415)71:8<2585::aid-cncr2820710825>3.0.co;2-s

```
### copy data into 'dat' and examine data
dat <- dat.fine1993</pre>
dat
## Not run:
### load metafor package
library(metafor)
### calculate log(ORs) and sampling variances for each time point
dat <- escalc(measure="OR", ai=e1i, n1i=nei, ci=c1i, n2i=nci, data=dat, var.names=c("y1i","v1i"))</pre>
dat <- escalc(measure="OR", ai=e2i, n1i=nei, ci=c2i, n2i=nci, data=dat, var.names=c("y2i","v2i"))</pre>
dat <- escalc(measure="OR", ai=e3i, n1i=nei, ci=c3i, n2i=nci, data=dat, var.names=c("y3i","v3i"))</pre>
dat <- escalc(measure="OR", ai=e4i, n1i=nei, ci=c4i, n2i=nci, data=dat, var.names=c("y4i","v4i"))
### calculate the covariances (equations in Appendix of Trikalinos & Olkin, 2012)
dat$v12i <- with(dat, nei / (e1i * (nei - e2i)) + nci / (c1i * (nci - c2i)))
dat$v13i <- with(dat, nei / (e1i * (nei - e3i)) + nci / (c1i * (nci - c3i)))
dat$v14i <- with(dat, nei / (e1i * (nei - e4i)) + nci / (c1i * (nci - c4i)))
dat$v23i <- with(dat, nei / (e2i * (nei - e3i)) + nci / (c2i * (nci - c3i)))
```

```
dat$v24i <- with(dat, nei / (e2i * (nei - e4i)) + nci / (c2i * (nci - c4i)))
dat$v34i <- with(dat, nei / (e3i * (nei - e4i)) + nci / (c3i * (nci - c4i)))
### create dataset in long format
dat.long <- data.frame(study=rep(1:nrow(dat), each=4), time=1:4,</pre>
                        yi=c(t(dat[c("y1i","y2i","y3i","y4i")])),
                        vi=c(t(dat[c("v1i","v2i","v3i","v4i")])))
### var-cov matrices of the studies
V <- lapply(split(dat, dat$study),</pre>
            function(x) matrix(c( x$v1i, x$v12i, x$v13i, x$v14i,
                                  x$v12i, x$v2i, x$v23i, x$v24i,
                                  x$v13i, x$v23i, x$v3i, x$v34i,
                             x$v14i, x$v24i, x$v34i, x$v4i), nrow=4, ncol=4, byrow=TRUE))
### remove rows for the missing time points in study 17
dat.long <- na.omit(dat.long)</pre>
### remove corresponding rows/columns from var-cov matrix
V[[17]] <- V[[17]][c(2,4),c(2,4)]</pre>
### make a copy of V
Vc <- V
### replace any (near) singular var-cov matrices with ridge corrected versions
repl.Vi <- function(Vi) {</pre>
   res <- eigen(Vi)</pre>
   if (any(res$values <= 0.08)) {</pre>
      round(res$vectors %*% diag(res$values + 0.08) %*% t(res$vectors), 12)
   } else {
      Vi
   }
}
Vc <- lapply(Vc, repl.Vi)</pre>
### do not correct var-cov matrix of study 17
Vc[[17]] <- V[[17]]
### construct block diagonal matrix
Vc <- bldiag(Vc)</pre>
### multivariate fixed-effects model
res <- rma.mv(yi, Vc, mods = ~ 0 + factor(time), method="FE", data=dat.long)</pre>
print(res, digits=3)
### multivariate random-effects model with heteroscedastic AR(1) structure for the true effects
res <- rma.mv(yi, Vc, mods = ~ 0 + factor(time), random = ~ time | study,</pre>
              struct="HAR", data=dat.long, control=list(optimizer="hjk"))
print(res, digits=3)
### profile the variance components
par(mfrow=c(2,2))
profile(res, tau2=1, xlim=c( 0, 0.2))
```

# dat.franchini2012

```
profile(res, tau2=2, xlim=c( 0, 0.2))
profile(res, tau2=3, xlim=c( 0, 0.2))
profile(res, tau2=4, xlim=c(0.1, 0.3))
### profile the autocorrelation coefficient
par(mfrow=c(1,1))
profile(res, rho=1)
## End(Not run)
```

dat.franchini2012 Studies on Dopamine Agonists to Reduce "Off-Time" in Patients with Advanced Parkinson Disease

# Description

Results from 7 trials examining the effectiveness of four dopamine agonists and placebo to reduce "off-time" in patients with advanced Parkinson disease.

## Usage

dat.franchini2012

### Format

The data frame contains the following columns:

Study	character	study label
Treatment1	character	treatment 1
y1	numeric	treatment effect arm 1
sd1	numeric	standard deviation arm 2
n1	integer	sample size arm 1
Treatment2	character	treatment 2
y2	numeric	treatment effect arm 2
sd2	numeric	standard deviation arm 2
n2	integer	sample size arm 1
Treatment3	character	treatment 3
y3	numeric	treatment effect arm 3
sd3	numeric	standard deviation arm 2
n3	integer	sample size arm 1

#### **Details**

This network meta-analysis compared the effectiveness of four active treatments and placebo in patients with advanced Parkinson disease (Franchini et al., 2012). The outcome is mean lost work-time reduction in patients given dopamine agonists as adjunct therapy. The data are given as sample size, mean, and standard deviation in each trial arm.

This dataset was used as an example in the supplemental material of Dias et al. (2013) where

placebo is coded as 1 and the four active drugs as 2 to 5.

# Concepts

medicine, raw mean differences, network meta-analysis

### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Dias, S., Sutton, A. J., Ades, A. E., & Welton, N. J. (2013). Evidence synthesis for decision making 2: A generalized linear modeling framework for pairwise and network meta-analysis of randomized controlled trials. *Medical Decision Making*, **33**(5), 607–617. https://doi.org/10.1177/0272989X12458724

Franchini, A. J., Dias, S., Ades, A. E., Jansen, J. P., & Welton, N. J. (2012). Accounting for correlation in network meta-analysis with multi-arm trials. *Research Synthesis Methods*, **3**(2), 142–160. https://doi.org/10.1002/jrsm.1049

```
### Show results from first three studies; third study is a three-arm
### study
head(dat.franchini2012, 3)
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print mean differences with two digits
oldset <- settings.meta(digits = 2)</pre>
### Transform data from wide arm-based format to contrast-based
### format. Argument 'sm' must not be provided as the mean difference
### is the default in R function metacont() called internally.
pw <- pairwise(list(Treatment1, Treatment2, Treatment3),</pre>
  n = list(n1, n2, n3),
 mean = list(y1, y2, y3),
  sd = list(sd1, sd2, sd3),
  data = dat.franchini2012, studlab = Study, sm = "MD")
### Show calculated mean differences (TE) for first three studies
pw[1:5, c(3:7, 10, 1)]
### Conduct network meta-analysis
net <- netmeta(pw)</pre>
net
### Draw network graph
netgraph(net, points = TRUE, cex.points = 3, cex = 1.5,
  plastic = TRUE, thickness = "se.fixed",
  iterate = TRUE, start = "eigen")
```

# dat.frank2008

### Use previous settings
settings.meta(oldset)

## End(Not run)

dat.frank2008

Studies on the Association Between the CASP8 -652 6N del Promoter Polymorphism and Breast Cancer Risk

# Description

Results from 4 case-control studies examining the association between the CASP8 -652 6N del promoter polymorphism and breast cancer risk.

## Usage

dat.frank2008

# Format

The data frame contains the following columns:

study	character	study identifier
bc.ins.ins	numeric	number of cases who have a homozygous insertion polymorphism
bc.ins.del	numeric	number of cases who have a heterozygous insertion/deletion polymorphism
bc.del.del	numeric	number of cases who have a homozygous deletion polymorphism
ct.ins.ins	numeric	number of controls who have a homozygous insertion polymorphism
ct.ins.del	numeric	number of controls who are heterozygous insertion/deletion polymorphism
ct.del.del	numeric	number of controls who have a homozygous deletion polymorphism

## Details

The 4 studies included in this dataset are case-control studies that have examined the association between the CASP8 -652 6N del promoter polymorphism and breast cancer risk. Breast cancer cases and controls were genotyped and either had a homozygous insertion, a heterozygous insertion/deletion, or a homozygous deletion polymorphism.

Ziegler et al. (2011) used the same dataset to illustrate the use of meta-analytic methods to examine deviations from Hardy-Weinberg equilibrium across multiple studies. The relative excess heterozy-gosity (REH) is the proposed measure for such a meta-analysis, which can be computed by setting measure="REH".

# Concepts

medicine, oncology, genetics, odds ratios

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Frank, B., Rigas, S. H., Bermejo, J. L., Wiestler, M., Wagner, K., Hemminki, K., Reed, M. W., Sutter, C., Wappenschmidt, B., Balasubramanian, S. P., Meindl, A., Kiechle, M., Bugert, P., Schmutzler, R. K., Bartram, C. R., Justenhoven, C., Ko, Y.-D., Brüning, T., Brauch, H., Hamann, U., Pharoah, P. P. D., Dunning, A. M., Pooley, K. A., Easton, D. F., Cox, A. & Burwinkel, B. (2008). The CASP8 -652 6N del promoter polymorphism and breast cancer risk: A multicenter study. *Breast Cancer Research and Treatment*, **111**(1), 139–144. https://doi.org/10.1007/s10549-007-9752-z

## References

Ziegler, A., Steen, K. V. & Wellek, S. (2011). Investigating Hardy-Weinberg equilibrium in casecontrol or cohort studies or meta-analysis. *Breast Cancer Research and Treatment*, **128**(1), 197– 201. https://doi.org/10.1007/s10549-010-1295-z

```
### copy data into 'dat' and examine data
dat <- dat.frank2008
dat
## Not run:
### load metafor package
library(metafor)
### calculate log odds ratios comparing ins/del versus ins/ins
dat <- escalc(measure="OR", ai=bc.ins.del, bi=bc.ins.ins,</pre>
                            ci=ct.ins.del, di=ct.ins.ins, data=dat)
### fit random-effects model and get the pooled odds ratio (with 95% CI)
res <- rma(yi, vi, data=dat)
res
predict(res, transf=exp, digits=2)
### calculate log odds ratios comparing del/del versus ins/ins
dat <- escalc(measure="OR", ai=bc.del.del, bi=bc.ins.ins,</pre>
                             ci=ct.del.del, di=ct.ins.ins, data=dat)
### fit random-effects model and get the pooled odds ratio (with 95% CI)
res <- rma(yi, vi, data=dat)</pre>
res
predict(res, transf=exp, digits=2)
### calculate log odds ratios comparing ins/del+del/del versus ins/ins
dat <- escalc(measure="OR", ai=bc.ins.del+bc.del.del, bi=bc.ins.ins,</pre>
                             ci=ct.ins.del+ct.del.del, di=ct.ins.ins, data=dat)
### fit random-effects model and get the pooled odds ratio (with 95% CI)
```

```
res <- rma(yi, vi, data=dat)</pre>
res
predict(res, transf=exp, digits=2)
****
### compute the relative excess heterozygosity in the controls
dat <- escalc(measure="REH", ai=ct.ins.ins, bi=ct.ins.del, ci=ct.del.del,</pre>
            slab=study, data=dat)
### fit random-effects model and get the pooled REH value (with 90% CI)
res <- rma(yi, vi, data=dat, level=90)</pre>
res
predict(res, transf=exp, digits=2)
### draw forest plot
forest(res, atransf=exp, xlim=c(-1.4,1.4), at=log(c(0.5,5/7,1,7/5,2)))
segments(log(5/7), -2, log(5/7), res$k+1, lty="dotted")
segments(log(7/5), -2, log(7/5), res$k+1, lty="dotted")
## End(Not run)
```

dat.furukawa2003	Studies on Low Dosage Tricyclic Antidepressants for the Treatment of
	Depression

## Description

Results on depression severity from 17 studies comparing low dosage tricyclic antidepressants (TCA) and placebo for the treatment of depression.

### Usage

dat.furukawa2003

# Format

The data frame contains the following columns:

author	character	First author with information on dosage in parentheses
Ne	integer	number of patients in low TCA group
Me	numeric	depression severity (low TCA)
Se	numeric	standard deviation (low TCA)
Nc	integer	number of patients in placebo group
Mc	numeric	depression severity (placebo)
Sc	numeric	standard deviation (placebo)

# Details

Furukawa et al. (2003) carried out a systematic review comparing low dosage tricyclic antidepressants (TCA) with placebo for the treatment of depression. They reported the effect on presence/absence of depression and on depression severity at various time points. Here we focus on depression severity at four weeks. Most studies used some version of the Hamilton Depression Rating Scale, however, some studies used the Montgomery-Asberg Depression Rating Scale. Accordingly, it is not possible to pool the estimated effects directly.

This data set is used as an example in Schwarzer et al. (2015).

#### Concepts

standardized mean differences

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

# Source

Furukawa, T. A., McGuire, H., & Barbui, C. (2003). Low dosage tricyclic antidepressants for depression. *Cochrane Database of Systematic Reviews*, **3**, CD003197. https://doi.org/10.1002/14651858.CD003197

#### References

Schwarzer, G., Carpenter, J. R., & Rücker, G. (2015). *Meta-analysis with R*. Cham, Switzerland: Springer.

## Examples

dat.gibson2002

Studies on the Effectiveness of Self-Management Education and Regular Medical Review for Adults with Asthma

# Description

Results from 15 trials examining the effectiveness of self-management education and regular medical review for adults with asthma.

# Usage

dat.gibson2002

# Format

The data frame contains the following columns:

author	character	first author of study
year	numeric	publication year
n1i	numeric	number of participants in the intervention group
m1i	numeric	mean number of days off work/school in the intervention group
sd1i	numeric	standard deviation of the number of days off work/school in the intervention group
n2i	numeric	number of participants in the control/comparison group
m2i	numeric	mean number of days off work/school in the control/comparison group
sd2i	numeric	standard deviation of the number of days off work/school in the control/comparison group
ai	numeric	number of participants who had one or more days off work/school in the intervention group
bi	numeric	number of participants who no days off work/school in the intervention group
ci	numeric	number of participants who had one or more days off work/school in the control/comparison group
di	numeric	number of participants who no days off work/school in the control/comparison group
type	numeric	numeric code for the intervention type (see 'Details')

# Details

Asthma management guidelines typically recommend for patients to receive education and regular medical review. While self-management programs have been shown to increase patient knowledge, it is less clear to what extent they actually impact health outcomes. The systematic review by Gibson et al. (2002) examined the effectiveness of self-management education and regular medical review for adults with asthma. In each study, participants receiving a certain management intervention were compared against those in a control/comparison group with respect to a variety of health outcomes. One of the outcomes examined in a number of studies was the number of days off work/school.

The majority of studies reporting this outcome provided means and standard deviations allowing a meta-analysis of standardized mean differences. Seven studies also reported the number of participants who had one or more days off work/school in each group. These studies could be metaanalyzed using, for example, (log) risk ratios. Finally, one could also consider a combined analysis based on standardized mean differences computed from the means and standard deviations where available and using probit transformed risk differences (which also provide estimates of the standardized mean difference) for the remaining studies. Some degree of patient education was provided in all studies. In addition, the type variable indicates what additional intervention components were included in each study:

- 1. optimal self-management (writing action plan, self-monitoring, regular medical review),
- 2. self-monitoring and regular medical review,
- 3. self-monitoring only,
- 4. regular medical review only,

predict(res, transf=exp, digits=2)

5. written action plan only.

#### Concepts

medicine, primary care, risk ratios, standardized mean differences

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Gibson, P. G., Powell, H., Wilson, A., Abramson, M. J., Haywood, P., Bauman, A., Hensley, M. J., Walters, E. H., & Roberts, J. J. L. (2002). Self-management education and regular practitioner review for adults with asthma. *Cochrane Database of Systematic Reviews*, **3**, CD001117. https://doi.org/10.1002/14651858.CD001117

```
### copy data into 'dat' and examine data
dat <- dat.gibson2002
dat
## Not run:
### load metafor package
library(metafor)
### compute standardized mean differences and corresponding sampling variances
dat <- escalc(measure="SMD", m1i=m1i, sd1i=sd1i, n1i=n1i, m2i=m2i, sd2i=sd2i, n2i=n2i, data=dat)</pre>
dat
### fit an equal-effects model to the standardized mean differences (as in Gibson et al., 2002)
res <- rma(yi, vi, data=dat, method="EE")</pre>
print(res, digits=2)
### compute log risk ratios and corresponding sampling variances
dat <- escalc(measure="RR", ai=ai, bi=bi, ci=ci, di=di, data=dat)</pre>
dat
### fit an equal-effects model to the log risk ratios
res <- rma(yi, vi, data=dat, method="EE")</pre>
print(res, digits=2)
```

```
### note: Gibson et al. (2002) used the Mantel-Haenszel method for their analysis
rma.mh(measure="RR", ai=ai, bi=bi, ci=ci, di=di, data=dat, digits=2)
### compute standardized mean differences where possible and otherwise probit transformed
### risk differences (which also provide estimates of the standardized mean differences)
dat <- escalc(measure="SMD", m1i=m1i, sd1i=sd1i, n1i=n1i,</pre>
                              m2i=m2i, sd2i=sd2i, n2i=n2i, data=dat, add.measure=TRUE)
dat <- escalc(measure="PBIT", ai=ai, bi=bi, ci=ci, di=di, data=dat, replace=FALSE, add.measure=TRUE)</pre>
dat
### fit a random-effects model to these estimates
res <- rma(yi, vi, data=dat)</pre>
print(res, digits=2)
### meta-regression model examining if there are systematic differences based on the
### type of measure used (there are only 2 studies where measure="PBIT", so this isn't
### very conclusive here, but shown for illustration purposes)
res <- rma(yi, vi, mods = ~ measure, data=dat)</pre>
print(res, digits=2)
predict(res, newmods=1, digits=2)
## End(Not run)
```

dat.graves2010 Studies on the Effectiveness of Injected Cholera Vaccines

## Description

Results from 17 studies on the effectiveness of injected vaccines against cholera.

### Usage

dat.graves2010

#### Format

The data frame contains the following columns:

study	character	author/study name and publication year
ai	numeric	number of cholera cases in the vaccinated group
n1i	numeric	number of individuals in the vaccinated group
ci	numeric	number of cholera cases in the placebo group
n2i	numeric	number of individuals in the placebo group

### Details

Cholera is an infection caused by certain strains of the bacterium *Vibrio cholerae*. When untreated, mortality rates can be as high as 50-60%. Proper sanitation practices are usually effective in preventing outbreaks, but a number of oral and injectable vaccines have also been developed. The

Cochrane review by Graves et al. (2010) examined the effectiveness of injectable vaccines for preventing cholera cases and death. The present dataset includes results from 17 studies that reported the number of cholera cases in vaccinated and placebo/comparison groups up to 7 months after the treatment.

## Concepts

medicine, risk ratios, Mantel-Haenszel method

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Graves, P. M., Deeks, J. J., Demicheli, V., & Jefferson, T. (2010). Vaccines for preventing cholera: Killed whole cell or other subunit vaccines (injected). *Cochrane Database of Systematic Reviews*, **8**, CD000974. https://doi.org/10.1002/14651858.CD000974.pub2

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.graves2010
dat
## Not run:
### load metafor package
library(metafor)
### analysis using the Mantel-Haenszel method
rma.mh(measure="RR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, digits=2)
```

## End(Not run)

dat.gurusamy2011 Studies on Interventions to Reduce Mortality after Liver Transplantation

#### Description

Results from 14 trials examining the mortality risk of interventions for decreasing blood loss and blood transfusion requirements during liver transplantation.

#### Usage

dat.gurusamy2011

### Format

The data frame contains the following columns:

study	character	study information
treatment	character	treatment
death	integer	mortality at 60 days post-transplantation
n	integer	number of individuals

# Details

This network meta-analysis compared the effectiveness of seven interventions for decreasing blood loss and blood transfusion requirements during liver transplantation (Gurusamy et al., 2011).

Fourteen studies reported mortality at 60 days, in 1,002 patients. Forty-five deaths were reported across all studies (4.5%). Six studies observed deaths in all treatment arms while three studies did not observe any deaths.

This dataset was used in Efthimiou et al. (2019) to introduce the Mantel-Haenszel method for network meta-analysis.

One of the treatments (solvent detergent plasma) was only included in one study with zero events in both treatment arms; this study was excluded from all network meta-analyses. In addition, no death was observed in the antithrombin III arm of the only study evaluating this treatment which was excluded from the Mantel-Haenszel network meta-analysis.

### Concepts

medicine, odds ratios, network meta-analysis, Mantel-Haenszel method

# Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

#### Source

Gurusamy, K. S., Pissanou, T., Pikhart, H., Vaughan, J., Burroughs, A. K., & Davidson, B. R. (2011). Methods to decrease blood loss and transfusion requirements for liver transplantation. *Cochrane Database of Systematic Reviews*, **12**, CD009052. https://doi.org/10.1002/14651858.CD009052.pub2

## References

Efthimiou, O., Rücker, G., Schwarzer, G., Higgins, J., Egger, M., & Salanti, G. (2019). A Mantel-Haenszel model for network meta-analysis of rare events. *Statistics in Medicine*, **38**(16), 2992–3012. https://doi.org/10.1002/sim.8158

```
### Show first 6 rows of the dataset
head(dat.gurusamy2011)
### Only study evaluating solvent detergent plasma
subset(dat.gurusamy2011, study == "Williamson 1999")
### Only study evaluating antithrombin III
subset(dat.gurusamy2011, study == "Baudo 1992")
```

```
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
### Change appearance of confidence intervals
cilayout("(", "-")
### Transform data from long arm-based format to contrast-based
### format. Argument 'sm' has to be used for odds ratio as summary
### measure; by default the risk ratio is used in the metabin function
### called internally.
pw <- pairwise(treatment, death, n, studlab = study,</pre>
  data = dat.gurusamy2011, sm = "OR")
### Conduct Mantel-Haenszel network meta-analysis (NMA)
net.MH <- netmetabin(pw, ref = "cont")</pre>
### Conduct inverse variance (IV) network meta-analysis
net.IV <- netmeta(pw, ref = "cont")</pre>
### Network graph (Mantel-Haenszel NMA)
netgraph(net.MH, seq = "optimal", col = "black", plastic = FALSE,
  points = TRUE, pch = 21, cex.points = 3, col.points = "black",
  bg.points = "gray", thickness = "se.fixed",
  number.of.studies = TRUE)
### Full network graph (based on inverse variance method, including
### study comparing Antithrombin III with Control/Placebo)
netgraph(net.IV,
  seq = "optimal", col = "black", plastic = FALSE,
  points = TRUE, pch = 21, cex.points = 3, col.points = "black",
  bg.points = "gray", thickness = "se.fixed",
  number.of.studies = TRUE)
### Compare results for Mantel-Haenszel and IV NMA
forest(netbind(net.MH, net.IV,
  random = FALSE, name = c("MH NMA", "IV NMA")))
### Show results for Mantel-Haenszel NMA
net.MH
forest(net.MH)
### League table with network estimates in lower triangle and direct
### estimates in upper triangle
netleague(net.MH)
### Assess inconsistency
print(netsplit(net.MH), show = "both", ci = TRUE, overall = FALSE,
  nchar.trts = 6)
```

### Use previous settings
settings.meta(oldset)

## End(Not run)

dat.hackshaw1998

Studies on the Risk of Lung Cancer in Women Exposed to Environmental Tobacco Smoke

# Description

Results from 37 studies on the risk of lung cancer in women exposed to environmental tobacco smoke (ETS) from their smoking spouse.

## Usage

dat.hackshaw1998

### Format

The data frame contains the following columns:

study author	numeric character	study number first author of study
year	numeric	publication year
country	character	country where study was conducted
design	character	study design (either cohort or case-control)
cases	numeric	number of lung cancer cases
or	numeric	odds ratio
or.lb	numeric	lower bound of 95% CI for the odds ratio
or.ub	numeric	upper bound of 95% CI for the odds ratio
yi	numeric	log odds ratio
vi	numeric	corresponding sampling variance

# Details

The dataset includes the results from 37 studies (4 cohort, 33 case-control) examining if women (who are lifelong nonsmokers) have an elevated risk for lung cancer due to exposure to environmental tobacco smoke (ETS) from their smoking spouse. Values of the log odds ratio greater than 0 indicate an increased risk of cancer in exposed women compared to women not exposed to ETS from their spouse.

Note that the log odds ratios and corresponding sampling variances were back-calculated from the reported odds ratios and confidence interval (CI) bounds (see 'Examples'). Since the reported values were rounded to some extent, this introduces some minor inaccuracies into the back-calculations. The overall estimate reported in Hackshaw et al. (1997) and Hackshaw (1998) can be fully reproduced though.

### Concepts

medicine, oncology, epidemiology, smoking, odds ratios

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Hackshaw, A. K., Law, M. R., & Wald, N. J. (1997). The accumulated evidence on lung cancer and environmental tobacco smoke. *British Medical Journal*, **315**(7114), 980–988. https://doi.org/10.1136/bmj.315.7114.

Hackshaw, A. K. (1998). Lung cancer and passive smoking. *Statistical Methods in Medical Research*, **7**(2), 119–136. https://doi.org/10.1177/096228029800700203

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.hackshaw1998</pre>
head(dat, 10)
## Not run:
### load metafor package
library(metafor)
### random-effects model using the log odds ratios
res <- rma(yi, vi, data=dat, method="DL")</pre>
res
### estimated average odds ratio with CI (and prediction interval)
predict(res, transf=exp, digits=2)
### illustrate how the log odds ratios and corresponding sampling variances
### can be back-calculated based on the reported odds ratios and CI bounds
dat$yi <- NULL
dat$vi <- NULL
dat <- data.frame(dat)</pre>
head(dat, 10)
dat <- conv.wald(out=or, ci.lb=or.lb, ci.ub=or.ub, data=dat, transf=log)</pre>
head(dat, 10)
## End(Not run)
```

dat.hahn2001Studies on the Effectiveness of Different Rehydration Solutions for the<br/>Prevention of Unscheduled Intravenous Infusion in Children with Di-<br/>arrhoea

# dat.hahn2001

# Description

Results from 12 trials examining the effectiveness of a reduced versus standard rehydration solution for the prevention of unscheduled intravenous infusion in children with diarrhoea.

# Usage

dat.hahn2001

## Format

The data frame contains the following columns:

tion grou
ution gro
-

# Details

The dataset includes the results from 12 randomized clinical trials that examined the effectiveness of a reduced osmolarity oral rehydration solution (total osmolarity <250 mmol/l with reduced sodium) with a standard WHO oral rehydration solution (sodium 90 mmol/l, glucose 111mmol/l, total osmolarity 311 mmol/l) for the prevention of unscheduled intravenous infusion in children with diarrhoea.

# Concepts

medicine, odds ratios, Mantel-Haenszel method

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Hahn, S., Kim, Y., & Garner, P. (2001). Reduced osmolarity oral rehydration solution for treating dehydration due to diarrhoea in children: Systematic review. *British Medical Journal*, **323**(7304), 81–85. https://doi.org/10.1136/bmj.323.7304.81

```
### copy data into 'dat' and examine data
dat <- dat.hahn2001
dat
## Not run:
### load metafor package
library(metafor)</pre>
```

```
### meta-analysis of (log) odds rations using the Mantel-Haenszel method
res <- rma.mh(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, digits=2, slab=study)
res
### forest plot (also show studies that were excluded from the analysis)
options(na.action="na.pass")
forest(res, atransf=exp, xlim=c(-11,9), at=log(c(0.01, 0.1, 1, 10, 100)))
options(na.action="na.omit")
## End(Not run)</pre>
```

dat.hannum2020	Studies	Comparing	Objective	and	Subjective	Olfactory	Loss	in
	COVID-	19 Patients						

# Description

Results from 35 studies measuring olfactory loss in COVID-19 patients using either objective or subjective measures.

#### Usage

dat.hannum2020

# Format

The data frame contains the following columns:

authorName	character	(first) author of study
DOI	character	article DOI number
ni	numeric	number of Covid-19 positive patients in the study
xi	numeric	number of Covid-19 positive patients in the study with olfactory loss
percentOlfactoryLoss	numeric	percent of the sample with olfactory loss
objectivity	character	objective or subjective measure used
measured	character	outcome measure
testType	character	type of test used
country	character	country where patients were treated
patientType	character	type of patient information and location where being treated

# Details

One of the symptoms of COVID-19 infection is olfactory loss (loss of smell) either recently acquired anosmia (complete loss of smell) or hyposmia (partial loss of smell). One challenge to reaching this symptom is the wide range of reported prevalence for this symptom ranging from 5 percent to 98 percent. In this dataset studies were grouped into one of two groups based on the type of method used to measure smell loss (either subjective measures, such as self-reported smell loss, or objective measures using rated stimuli).

## Concepts

medicine, covid-19, proportions

## Author(s)

W. Kyle Hamilton <whamilton@ucmerced.edu> https://kylehamilton.com

# Source

Ramirez VA, Hannum ME, Lipson SJ, Herriman RD, Toskala AK, Lin C, Joseph PV, Reed DR. 2020. COVID-19 Smell Loss Prevalence Tracker. Available from: https://vicente-ramirez.shinyapps.io/COVID19\_01 and https://github.com/vramirez4/01factoryLoss (accessed August 11, 2021)

# References

Hannum, M. E., Ramirez, V. A., Lipson, S. J., Herriman, R. D., Toskala, A. K., Lin, C., Joseph, P. V., & Reed, D. R. (2020). Objective sensory testing methods reveal a higher prevalence of olfactory loss in COVID-19 positive patients compared to subjective methods: A systematic review and meta-analysis. *Chemical Senses*, **45**(9), 865–874. https://doi.org/10.1093/chemse/bjaa064

# Examples

```
# copy data into 'dat' and examine data
dat <- dat.hannum2020
dat
## Not run:
# load metafor package
library(metafor)
# compute effect size
dat <- escalc(measure="PR", xi=xi, ni=ni, data=dat)</pre>
# split data into objective and subjective datasets
dat_split <- split(dat, dat$objectivity)</pre>
dat_objective <- dat_split[["Objective"]]</pre>
dat_subjective <- dat_split[["Subjective"]]</pre>
# random-effects model all studies
res_all <- rma(yi, vi, data=dat)</pre>
print(res_all, digits=2)
# random-effects model objective
res_objective <- rma(yi, vi, data=dat_objective)</pre>
print(res_objective, digits=2)
# random-effects model subjective
res_subjective <- rma(yi, vi, data=dat_subjective)
print(res_subjective, digits=2)
```

## End(Not run)

dat.hart1999

## Description

Results from 6 clinical trials examining the effectiveness of adjusted-dose warfarin for preventing strokes in patients with atrial fibrillation.

### Usage

dat.hart1999

# Format

The data frame contains the following columns:

trial	numeric	trial number
study	character	study name (abbreviated)
year	numeric	publication year
x1i	numeric	number of strokes in the warfarin group
n1i	numeric	number of patients in the warfarin group
t1i	numeric	total person-time (in years) in the warfarin group
x2i	numeric	number of strokes in the placebo/control group
n2i	numeric	number of patients in the placebo/control group
t2i	numeric	total person-time (in years) in the placebo/control group
compgrp	character	type of comparison group (placebo or control)
prevtype	character	type of prevention (primary or secondary)
trinr	character	target range for the international normalized ratio (INR)

## Details

The 6 studies provide data with respect to the number of strokes in the warfarin and the comparison (placebo or control) group. In addition, the number of patients and the total person-time (in years) is provided for the two groups. The goal of the meta-analysis was to examine the effectiveness of adjusted-dose warfarin for preventing strokes in patients with atrial fibrillation.

# Concepts

medicine, cardiology, incidence rates

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Hart, R. G., Benavente, O., McBride, R., & Pearce, L. A. (1999). Antithrombotic therapy to prevent stroke in patients with atrial fibrillation: A meta-analysis. *Annals of Internal Medicine*, **131**(7),

492-501. https://doi.org/10.7326/0003-4819-131-7-199910050-00003

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.hart1999
dat
## Not run:
### load metafor package
library(metafor)
### calculate log incidence rate ratios and corresponding sampling variances
dat <- escalc(measure="IRR", x1i=x1i, x2i=x2i, t1i=t1i, t2i=t2i, data=dat)</pre>
dat
### meta-analysis of log incidence rate ratios using a random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### average incidence rate ratio with 95% CI
predict(res, transf=exp)
### forest plot with extra annotations
par(mar=c(5,4,1,2))
forest(res, xlim=c(-11, 5), at=log(c(0.05, 0.25, 1, 4)), atransf=exp,
       slab=paste0(study, " (", year, ")"),
       ilab=cbind(paste(x1i, "/", t1i, sep=" "),
                  paste(x2i, "/", t2i, sep=" ")),
       ilab.xpos=c(-6.5,-4), cex=0.85, header="Study (Year)")
op <- par(cex=0.85, font=2)</pre>
text(c(-6.5,-4), 8.5, c("Warfarin", "Control"))
text(c(-6.5,-4), 7.5, c("Strokes / PT", "Strokes / PT"))
segments(x0=-8, y0=8, x1=-2.75, y1=8)
par(op)
### meta-analysis of incidence rate differences using a random-effects model
res <- rma(measure="IRD", x1i=x1i, x2i=x2i, t1i=t1i, t2i=t2i, data=dat)
res
## End(Not run)
```

dat.hartmannboyce2018 Studies on the Effectiveness of Nicotine Replacement Therapy for Smoking Cessation

#### Description

Results from 133 studies examining the effectiveness of nicotine replacement therapy (NRT) for smoking cessation at 6+ months of follow-up.

#### Usage

dat.hartmannboyce2018

#### Format

The data frame contains the following columns:

study	numeric	study identifier
x.nrt	numeric	number of participants in the NRT group who were abstinent at the follow-up
n.nrt	numeric	number of participants in the NRT group
x.ctrl	numeric	number of participants in the control group who were abstinent at the follow-up
n.ctrl	numeric	number of participants in the control group
treatment	character	type of NRT provided in the treatment group

# Details

The dataset includes the results from 133 studies examining the effectiveness of nicotine replacement therapy (NRT) for smoking cessation. The results given in this dataset pertain to abstinence at 6+ months of follow-up. NRT was provided to participants in the treatment groups in various forms as indicated by the treatment variable (e.g., gum, patch, inhalator). Note that the dataset includes 136 rows, since a few studies included multiple treatments.

### Concepts

medicine, smoking, risk ratios, Mantel-Haenszel method

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Hartmann-Boyce, J., Chepkin, S. C., Ye, W., Bullen, C. & Lancaster, T. (2018). Nicotine replacement therapy versus control for smoking cessation. *Cochrane Database of Systematic Reviews*, **5**, CD000146. https://doi.org//10.1002/14651858.CD000146.pub5

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.hartmannboyce2018
head(dat, 10)</pre>
```

```
## Not run:
### load metafor package
library(metafor)
```

### turn treatment into a factor with the desired ordering
dat\$treatment <- factor(dat\$treatment, levels=unique(dat\$treatment))</pre>

### meta-analysis per treatment using the M-H method

# dat.hasselblad1998

## End(Not run)

dat.hasselblad1998 Studies on the Effectiveness of Counseling for Smoking Cessation

# Description

Results from 24 studies on the effectiveness of various counseling types for smoking cessation.

## Usage

dat.hasselblad1998

#### Format

The data frame contains the following columns:

id	numeric	id number for each treatment arm
study	numeric	study id number
authors	character	study author(s)
year	numeric	publication year
trt	character	intervention group
xi	numeric	number of individuals abstinent
ni	numeric	number of individuals in group

# Details

The dataset includes the results from 24 studies on the effectiveness of various counseling types for smoking cessation (i.e., self-help, individual counseling, group counseling, and no contact). The dataset indicates the total number of individuals within each study arm and the number that were abstinent from 6 to 12 months. The majority of the studies compared two interventions types against each other, while 2 studies compared three types against each other simultaneously.

The data can be used for a 'network meta-analysis' (also called a 'mixed treatment comparison'). The code below shows how such an analysis can be conducted using an arm-based and a contrastbased model (see Salanti et al., 2008, for more details).

# Concepts

medicine, psychology, smoking, odds ratios, network meta-analysis

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Hasselblad, V. (1998). Meta-analysis of multitreatment studies. *Medical Decision Making*, **18**(1), 37–43. https://doi.org/10.1177/0272989X9801800110

# References

Gleser, L. J., & Olkin, I. (2009). Stochastically dependent effect sizes. In H. Cooper, L. V. Hedges, & J. C. Valentine (Eds.), *The handbook of research synthesis and meta-analysis* (2nd ed., pp. 357–376). New York: Russell Sage Foundation.

Law, M., Jackson, D., Turner, R., Rhodes, K., & Viechtbauer, W. (2016). Two new methods to fit models for network meta-analysis with random inconsistency effects. *BMC Medical Research Methodology*, **16**, 87. https://doi.org/10.1186/s12874-016-0184-5

Salanti, G., Higgins, J. P. T., Ades, A. E., & Ioannidis, J. P. A. (2008). Evaluation of networks of randomized trials. *Statistical Methods in Medical Research*, **17**(3), 279–301. https://doi.org/10.1177/0962280207080643

```
### copy data into 'dat' and examine data
dat <- dat.hasselblad1998</pre>
dat
## Not run:
### load metafor package
library(metafor)
### create network graph ('igraph' package must be installed)
library(igraph, warn.conflicts=FALSE)
pairs <- data.frame(do.call(rbind,</pre>
  sapply(split(dat$trt, dat$study), function(x) t(combn(x,2)))), stringsAsFactors=FALSE)
lvls <- c("no_contact", "self_help", "ind_counseling", "grp_counseling")</pre>
pairs$X1 <- factor(pairs$X1, levels=lvls)</pre>
pairs$X2 <- factor(pairs$X2, levels=lvls)</pre>
tab <- table(pairs[,1], pairs[,2])</pre>
tab # adjacency matrix
g <- graph_from_adjacency_matrix(tab, mode = "plus", weighted=TRUE, diag=FALSE)
vertex_attr(g, "name") <- c("No Contact", "Self-Help",</pre>
                             "Individual\nCounseling", "Group\nCounseling")
plot(g, edge.curved=FALSE, edge.width=E(g)$weight, layout=layout_on_grid,
   vertex.size=45, vertex.color="lightgray", vertex.label.color="black", vertex.label.font=2)
### calculate log odds for each study arm
dat <- escalc(measure="PLO", xi=xi, ni=ni, add=1/2, to="all", data=dat)</pre>
dat
### convert trt variable to factor with desired ordering of levels
dat$trt <- factor(dat$trt, levels=c("no_contact", "self_help", "ind_counseling", "grp_counseling"))</pre>
```

```
### add a space before each level (this makes the output a bit more legible)
levels(dat$trt) <- paste0(" ", levels(dat$trt))</pre>
### network meta-analysis using an arm-based model with fixed study effects
### by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons
res <- rma.mv(yi, vi, mods = ~ 0 + factor(study) + trt,</pre>
              random = ~ trt | study, rho=1/2, data=dat, btt="trt")
res
### all pairwise odds ratios of interventions versus no contact
predict(res, newmods=cbind(matrix(0, nrow=3, ncol=24), diag(3)),
        intercept=FALSE, transf=exp, digits=2)
### all pairwise odds ratios comparing interventions (ic vs sh, gc vs sh, and gc vs ic)
predict(res, newmods=cbind(matrix(0, nrow=3, ncol=24), rbind(c(-1,1,0), c(-1,0,1), c(0,-1,1))),
        intercept=FALSE, transf=exp, digits=2)
### forest plot of ORs of interventions versus no contact
forest(c(0,res$beta[25:27]), sei=c(0,res$se[25:27]), psize=1, xlim=c(-3,4), digits=c(2,1), efac=2,
       slab=c("No Contact", "Self-Help", "Individual Counseling", "Group Counseling"),
    atransf=exp, at=log(c(0.5, 1, 2, 4, 8)), xlab="Odds Ratio for Intervention vs. No Contact",
       header=c("Intervention", "Odds Ratio [95% CI]"))
*****************
### restructure dataset to a contrast-based format
dat <- to.wide(dat.hasselblad1998, study="study", grp="trt", ref="no_contact", grpvars=6:7)</pre>
### calculate log odds ratios for each treatment comparison
dat <- escalc(measure="OR", ai=xi.1, n1i=ni.1,</pre>
                            ci=xi.2, n2i=ni.2, add=1/2, to="all", data=dat)
dat
### calculate the variance-covariance matrix of the log odds ratios for multitreatment studies
### see Gleser & Olkin (2009), equation (19.11), for the covariance equation
calc.v <- function(x) {</pre>
 v <- matrix(1/(x$xi.2[1] + 1/2) + 1/(x$ni.2[1] - x$xi.2[1] + 1/2), nrow=nrow(x), ncol=nrow(x))</pre>
  diag(v) <- x$vi</pre>
  v
}
V <- bldiag(lapply(split(dat, dat$study), calc.v))</pre>
### add contrast matrix to dataset
dat <- contrmat(dat, grp1="trt.1", grp2="trt.2")</pre>
dat
### network meta-analysis using a contrast-based random-effects model
### by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons
res <- rma.mv(yi, V, mods = ~ 0 + self_help + ind_counseling + grp_counseling,</pre>
              random = ~ comp | study, rho=1/2, data=dat)
res
```

```
### predicted odds ratios of interventions versus no contact
predict(res, newmods=diag(3), transf=exp, digits=2)
### fit random inconsistency effects model (see Law et al., 2016)
res <- rma.mv(yi, V, mods = ~ 0 + self_help + ind_counseling + grp_counseling,
             random = list(~ comp | study, ~ comp | design), rho=1/2, phi=1/2, data=dat)
res
```

## End(Not run)

dat.hine1989 Studies on Prophylactic Use of Lidocaine After a Heart Attack

# Description

Results from 6 studies evaluating mortality from prophylactic use of lidocaine in acute myocardial infarction.

### Usage

dat.hine1989

### Format

The data frame contains the following columns:

study	numeric	study number
source	character	source of data
n1i	numeric	number of patients in lidocaine group
n2i	numeric	number of patients in control group
ai	numeric	number of deaths in lidocaine group
ci	numeric	number of deaths in control group

### Details

Hine et al. (1989) conducted a meta-analysis of death rates in randomized controlled trials in which prophylactic lidocaine was administered to patients with confirmed or suspected acute myocardial infarction. The dataset describes the mortality at the end of the assigned treatment period for control and intravenous lidocaine treatment groups for six studies. The question of interest is whether there is a detrimental effect of lidocaine. Because the studies were conducted to compare rates of arrhythmias following a heart attack, the studies, taken individually, are too small to detect important differences in mortality rates.

The data in this dataset were obtained from Table I in Normand (1999, p. 322).

# Concepts

medicine, cardiology, risk differences

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Normand, S. T. (1999). Meta-analysis: Formulating, evaluating, combining, and reporting. *Statis-tics in Medicine*, **18**(3), 321–359. https://doi.org/10.1002/(sici)1097-0258(19990215)18:3<321::aid-sim28>3.0

#### References

Hine, L. K., Laird, N., Hewitt, P., & Chalmers, T. C. (1989). Meta-analytic evidence against prophylactic use of lidocaine in acute myocardial infarction. *Archives of Internal Medicine*, **149**(12), 2694–2698. https://doi.org/10.1001/archinte.1989.00390120056011

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.hine1989
dat
## Not run:
### load metafor package
library(metafor)
### calculate risk differences and corresponding sampling variances
dat <- escalc(measure="RD", n1i=n1i, n2i=n2i, ai=ai, ci=ci, data=dat)
dat
### meta-analysis of risk differences using a random-effects model
res <- rma(yi, vi, data=dat)
res
## End(Not run)</pre>
```

dat.ishak2007 Studies on Deep-Brain Stimulation in Patients with Parkinson's disease

## Description

Results from 46 studies examining the effects of deep-brain stimulation on motor skills of patients with Parkinson's disease.

#### Usage

dat.ishak2007

## Format

The data frame contains the following columns:

study	character	(first) author and year
y1i	numeric	observed mean difference at 3 months
v1i	numeric	sampling variance of the mean difference at 3 months
y2i	numeric	observed mean difference at 6 months
v2i	numeric	sampling variance of the mean difference at 6 months
y3i	numeric	observed mean difference at 12 months
v3i	numeric	sampling variance of the mean difference at 12 months
y4i	numeric	observed mean difference at the long-term follow-up
v4i	numeric	sampling variance of the mean difference at the long-term follow-up
mdur	numeric	mean disease duration (in years)
mbase	numeric	mean baseline UPDRS score

#### Details

Deep-brain stimulation (DBS), which is delivered through thin surgically implanted wires in specific areas of the brain and controlled by the patient, is meant to provide relief of the debilitating symptoms of Parkinson's disease. The dataset includes the results from 46 studies examining the effects of DBS of the subthalamic nucleus on motor functioning, measured with the Unified Parkinson's Disease Rating Scale (UPDRS). The effect size measure for this meta-analysis was the mean difference of the scores while the stimulator is active and the baseline scores (before implantation of the stimulator). Since lower scores on the UPDRS indicate better functioning, negative numbers indicate improvements in motor skills. Effects were generally measured at 3, 6, and 12 months after implantation of the stimulator, with some studies also including a further long-term follow-up. However, the number of measurements differed between studies - hence the missing data on some of the measurement occasions.

Since the same patients were followed over time within a study, effect size estimates from multiple measurement occasions are likely to be correlated. A multivariate model accounting for the correlation in the effects can be used to meta-analyze these data. A difficulty with this approach is the lack of information about the correlation of the measurements over time in the individual studies. The approach taken by Ishak et al. (2007) was to assume an autoregressive (AR1) structure for the estimates within the individual studies. In addition, the correlation in the true effects was modeled, again using an autoregressive structure.

#### Concepts

medicine, raw mean differences, longitudinal models

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Ishak, K. J., Platt, R. W., Joseph, L., Hanley, J. A., & Caro, J. J. (2007). Meta-analysis of longitudinal studies. *Clinical Trials*, **4**(5), 525–539. https://doi.org/10.1177/1740774507083567

### Examples

### copy data into 'dat' and examine data

```
dat <- dat.ishak2007</pre>
head(dat, 5)
## Not run:
### load metafor package
library(metafor)
### create long format dataset
dat <- reshape(dat, direction="long", idvar="study", v.names=c("yi","vi"),</pre>
               varying=list(c(2,4,6,8), c(3,5,7,9)))
dat <- dat[order(study, time),]</pre>
### remove missing measurement occasions from dat.long
dat <- dat[!is.na(yi),]</pre>
rownames(dat) <- NULL
head(dat, 8)
### construct the full (block diagonal) V matrix with an AR(1) structure
### assuming an autocorrelation of 0.97 as estimated by Ishak et al. (2007)
V <- vcalc(vi, cluster=study, time1=time, phi=0.97, data=dat)</pre>
### plot data
with(dat, interaction.plot(time, study, yi, type="b", pch=19, lty="solid", xaxt="n",
          legend=FALSE, xlab="Time Point", ylab="Mean Difference", bty="l"))
axis(side=1, at=1:4, lab=c("1 (3 months)", "2 (6 months)", "3 (12 months)", "4 (12+ months)"))
### multivariate model with heteroscedastic AR(1) structure for the true effects
res <- rma.mv(yi, V, mods = ~ 0 + factor(time), random = ~ time | study,</pre>
              struct = "HAR", data = dat)
print(res, digits=2)
## End(Not run)
```

dat.kalaian1996 Studies on the Effectiveness of Coaching for the SAT

# Description

Results from studies examining the effectiveness of coaching on the performance on the Scholastic Aptitude Test (SAT).

#### Usage

```
dat.kalaian1996
```

#### Format

The data frame contains the following columns:

id	numeric	row (effect) id
study	character	study identifier
year	numeric	publication year
n1i	numeric	number of participants in the coached group
n2i	numeric	number of participants in the uncoached group
outcome	character	subtest (verbal or math)
yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance
hrs	numeric	hours of coaching
ets	numeric	study conducted by the Educational Testing Service (ETS) $(0 = no, 1 = yes)$
homework	numeric	assignment of homework outside of the coaching course $(0 = no, 1 = yes)$
type	numeric	study type $(1 = randomized study, 2 = matched study, 3 = nonequivalent comparison study)$

#### Details

The effectiveness of coaching for the Scholastic Aptitude Test (SAT) has been examined in numerous studies. This dataset contains standardized mean differences comparing the performance of a coached versus uncoached group on the verbal and/or math subtest of the SAT. Studies may report a standardized mean difference for the verbal subtest, the math subtest, or both. In the latter case, the two standardized mean differences are not independent (since they were measured in the same group of subjects). The number of hours of coaching (variable hrs), whether the study was conducted by the Educational Testing Service (variable ets), whether homework was assigned outside of the coaching course (variable homework), and the study type (variable type) may be potential moderators of the treatment effect.

# Concepts

education, standardized mean differences, multivariate models, meta-regression

#### Note

The dataset was obtained from Table 1 in Kalaian and Raudenbush (1996). However, there appear to be some inconsistencies between the data in the table and those that were actually used for the analyses (see 'Examples').

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Kalaian, H. A., & Raudenbush, S. W. (1996). A multivariate mixed linear model for meta-analysis. *Psychological Methods*, **1**(3), 227–235. https://doi.org/10.1037/1082-989X.1.3.227

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.kalaian1996
head(dat, 12)</pre>
```

```
## Not run:
### load metafor package
library(metafor)
### check ranges
range(dat$yi[dat$outcome == "verbal"]) # -0.35 to 0.74 according to page 230
range(dat$yi[dat$outcome == "math"]) # -0.53 to 0.60 according to page 231
### comparing this with Figure 1 in the paper reveals some discrepancies
par(mfrow=c(1,2), mar=c(5,5,1,3.4))
plot(log(dat$hrs[dat$outcome == "verbal"]), dat$yi[dat$outcome == "verbal"],
    pch=19, col=rgb(0,0,0,0.4), xlab="Log(Coaching Hours)", ylab="Effect Size (verbal)",
     xlim=c(1,6), ylim=c(-0.5,1), xaxs="i", yaxs="i")
abline(h=c(-0.5,0,0.5), lty="dotted")
abline(v=log(c(5,18)), lty="dotted")
plot(log(dat$hrs[dat$outcome == "math"]), dat$yi[dat$outcome == "math"],
     pch=19, col=rgb(0,0,0,0.4), xlab="Log(Coaching Hours)", ylab="Effect Size (math)",
     xlim=c(1,6), ylim=c(-1.0,1), xaxs="i", yaxs="i")
abline(h=c(-0.5,0,0.5), lty="dotted")
abline(v=log(c(5,18)), lty="dotted")
### construct variance-covariance matrix assuming rho = 0.66 for effect sizes
### corresponding to the 'verbal' and 'math' outcome types
V <- vcalc(vi, cluster=study, type=outcome, data=dat, rho=0.66)</pre>
### fit multivariate random-effects model
res <- rma.mv(yi, V, mods = ~ 0 + outcome,</pre>
              random = ~ outcome | study, struct="UN",
              data=dat, digits=3)
res
### test whether the effect differs for the math and verbal subtest
anova(res, X=c(1,-1))
### log-transform and mean center the hours of coaching variable
dat$loghrs <- log(dat$hrs) - mean(log(dat$hrs), na.rm=TRUE)</pre>
### fit multivariate model with log(hrs) as moderator
res <- rma.mv(yi, V, mods = ~ 0 + outcome + outcome:loghrs,</pre>
              random = ~ outcome | study, struct="UN",
              data=dat, digits=3)
res
### fit model with tau2 = 0 for outcome verbal (which also constrains rho = 0)
res <- rma.mv(yi, V, mods = ~ 0 + outcome + outcome:loghrs,</pre>
              random = ~ outcome | study, struct="UN", tau2=c(NA,0),
              data=dat, digits=3)
res
## End(Not run)
```

dat.kearon1998

Studies on the Accuracy of Venous Ultrasonography for the Diagnosis of Deep Venous Thrombosis

## Description

Results from diagnostic accuracy studies examining the accuracy of venous ultrasonography for the diagnosis of deep venous thrombosis.

# Usage

dat.kearon1998

# Format

The data frame contains the following columns:

id	numeric	study id
author	character	study author(s)
year	numeric	publication year
patients	character	patient group (either symptomatic or asymptomatic patients)
tp	numeric	number of true positives
np	numeric	number of positive patients (cases)
tn	numeric	number of true negatives
nn	numeric	number of negative patients (non-cases)

#### Details

The studies included in the dataset examined the accuracy of venous ultrasonography for the diagnossis of a first deep venous thrombosis in symptomatic and asymptomatic patients. Cases and non-cases were determined based on contrast venography. Venous ultrasonography was then used to make a diagnosis, leading to a given number of true positives and negatives.

A subset of this dataset (using only the studies with asymptomatic patients) was used by Deeks et al. (2005) to illustrate methods for detecting publication bias (or small-study effects) in meta-analyses of diagnostic accuracy studies.

#### Concepts

medicine, odds ratios, diagnostic accuracy studies, multivariate models, publication bias

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Kearon, C., Julian, J. A., Math, M., Newman, T. E., & Ginsberg, J. S. (1998). Noninvasive diagnosis of deep venous thrombosis. *Annals of Internal Medicine*, **128**(8), 663–677. https://doi.org/10.7326/0003-4819-128-8-
## dat.kearon1998

#### References

Deeks, J. J., Macaskill, P., & Irwig, L. (2005). The performance of tests of publication bias and other sample size effects in systematic reviews of diagnostic test accuracy was assessed. *Journal of Clinical Epidemiology*, **58**(9), 882–893. https://doi.org/10.1016/j.jclinepi.2005.01.016

```
### copy data into 'dat' and examine data
dat <- dat.kearon1998
head(dat)
## Not run:
### load metafor package
library(metafor)
### calculate diagnostic log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=tp, n1i=np, ci=nn-tn, n2i=nn, data=dat, add=1/2, to="all")
head(dat)
### fit random-effects model for the symptomatic patients
res <- rma(yi, vi, data=dat, subset=patients=="symptomatic")</pre>
res
### fit random-effects model for the asymptomatic patients
res <- rma(yi, vi, data=dat, subset=patients=="asymptomatic")</pre>
res
### estimated average diagnostic odds ratio (with 95% CI)
predict(res, transf=exp, digits=2)
### regression test for funnel plot asymmetry using SE as predictor
reg <- regtest(res, model="lm")</pre>
reg
### corresponding funnel plot
funnel(res, atransf=exp, xlim=c(0,7), at=log(c(1,10,100,1000)), ylim=c(0,1.5), steps=4)
ys <- seq(0, 2, length=100)</pre>
lines(coef(reg$fit)[1] + coef(reg$fit)[2]*ys, ys, lwd=2, lty=3)
### regression test for funnel plot asymmetry using total sample size as predictor
reg <- regtest(res, model="lm", predictor="ni")</pre>
reg
### corresponding funnel plot
funnel(res, yaxis="ni", atransf=exp, xlim=c(0,7), at=log(c(1,10,100,1000)), ylim=c(0,300), steps=4)
ys <- seq(0, 300, length=100)
lines(coef(reg$fit)[1] + coef(reg$fit)[2]*ys, ys, lwd=2, lty=3)
### regression test for funnel plot asymmetry using 1/sqrt(ESS) as predictor (Deeks et al., 2005)
dat$invessi <- 1/(4*dat$np) + 1/(4*dat$nn)</pre>
tmp <- rma(yi, invessi, data=dat, subset=patients=="asymptomatic")</pre>
reg <- regtest(tmp, model="lm")</pre>
```

```
reg
### corresponding funnel plot
funnel(tmp, atransf=exp, xlim=c(0,7), at=log(c(1,10,100,1000)), ylim=c(0,0.15), steps=4,
       refline=coef(res), level=0, ylab="1/root(ess)")
ys <- seq(0, 0.2, length=100)</pre>
lines(coef(reg$fit)[1] + coef(reg$fit)[2]*ys, ys, lwd=2, lty=3)
### convert data to long format
dat <- to.long(measure="OR", ai=tp, n1i=np, ci=tn, n2i=nn,</pre>
               data=dat.kearon1998, subset=patients=="asymptomatic", append=FALSE)
dat$group <- factor(dat$group, levels=c(1,2), labels=c("sensitivity", "specificity"))</pre>
dat
### calculate logit-transformed sensitivities
dat <- escalc(measure="PLO", xi=out1, mi=out2, data=dat, add=1/2, to="all",</pre>
              include=group=="sensitivity")
dat
### calculate logit-transformed specificities
dat <- escalc(measure="PLO", xi=out1, mi=out2, data=dat, add=1/2, to="all",</pre>
              include=group=="specificity")
dat
### bivariate random-effects model for logit sensitivity and specificity
res <- rma.mv(yi, vi, mods = ~ 0 + group, random = ~ group | study, struct="UN", data=dat)
res
### estimated average sensitivity and specificity based on the model
predict(res, newmods = rbind(c(1,0),c(0,1)), transf=transf.ilogit, tau2.levels=c(1,2), digits=2)
### estimated average diagnostic odds ratio based on the model
predict(res, newmods = c(1,1), transf=exp, digits=2)
## End(Not run)
```

dat.knapp2017	Studies on Differences in Planning Performance in Schizophrenia Pa-
	tients versus Healthy Controls

# Description

Results from 31 studies examining differences in planning performance in schizophrenia patients versus healthy controls.

### Usage

dat.knapp2017

### dat.knapp2017

## Format

The data frame contains the following columns:

author	character	study author(s)
year	numeric	publication year
study	numeric	study id number
task	character	type of task
difficulty	numeric	task difficulty
group1	character	identifier for patient group within studies
group2	character	identifier for control group within studies
comp	numeric	identifier for comparisons within studies
yi	numeric	standardized mean difference for planning performance
vi	numeric	corresponding sampling variance
n_sz	numeric	number of schizophrenic patients
n_hc	numeric	number of healthy controls
yi	numeric	standardized mean difference for IQ
vi	numeric	corresponding sampling variance

## Details

The studies included in this dataset examined differences between schizophrenia patients and healthy controls with respect to their performance on the tower of London test (https://en.wikipedia. org/wiki/Tower\_of\_London\_test) or a similar cognitive tasks measuring planning ability. The outcome measure for this meta-analysis was the standardized mean difference (with positive values indicating better performance in the healthy controls compared to the schizophrenia patients).

The dataset has a more complex structure for several reasons:

- 1. Studies 2, 3, 9, and 20 included more than one schizophrenia patient group and the standardized mean differences were computed by comparing these groups against a single healthy control group.
- 2. Studies 6, 12, 14, 15, 18, 19, 22, and 26 had the patients and controls complete different tasks of varying complexity (essentially the average number of moves required to complete a task). Study 6 also included two different task types.
- 3. Study 24 provides two standardized mean differences, one for men and the other for women.
- 4. Study 29 provides three standardized mean differences, corresponding to the three different COMT Val158Met genotypes (val/val, val/met, and met/met).

All 4 issues described above lead to a multilevel structure in the dataset, with multiple standardized mean differences nested within some of the studies. Issues 1. and 2. also lead to correlated sampling errors.

# Concepts

psychology, standardized mean differences, multilevel models, multivariate models, cluster-robust inference, meta-regression

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Knapp, F., Viechtbauer, W., Leonhart, R., Nitschke, K., & Kaller, C. P. (2017). Planning performance in schizophrenia patients: A meta-analysis of the influence of task difficulty and clinical and sociodemographic variables. *Psychological Medicine*, **47**(11), 2002–2016. https://doi.org/10.1017/S00332917170004

```
### copy data into 'dat' and examine data
dat <- dat.knapp2017
dat[-c(1:2)]
## Not run:
### load metafor package
library(metafor)
### fit a standard random-effects model ignoring the issues described above
res <- rma(yi, vi, data=dat)</pre>
res
### fit a multilevel model with random effects for studies and comparisons within studies
### (but this ignored the correlation in the sampling errors)
res <- rma.mv(yi, vi, random = ~ 1 | study/comp, data=dat)</pre>
res
### create variable that indicates the task and difficulty combination as increasing integers
dat$task.diff <- unlist(lapply(split(dat, dat$study), function(x) {</pre>
  task.int <- as.integer(factor(x$task))</pre>
  diff.int <- as.integer(factor(x$difficulty))</pre>
  diff.int[is.na(diff.int)] <- 1</pre>
  paste0(task.int, ".", diff.int)}))
### construct correlation matrix for two tasks with four different difficulties where the
### correlation is 0.4 for different difficulties of the same task, 0.7 for the same
### difficulty of different tasks, and 0.28 for different difficulties of different tasks
R <- matrix(0.4, nrow=8, ncol=8)</pre>
R[5:8,1:4] <- R[1:4,5:8] <- 0.28
diag(R[1:4,5:8]) <- 0.7
diag(R[5:8,1:4]) <- 0.7
diag(R) <- 1
rownames(R) <- colnames(R) <- paste0(rep(1:2, each=4), ".", 1:4)</pre>
R
### construct an approximate V matrix accounting for the use of shared groups and
### for correlations among tasks/difficulties as specified in the R matrix above
V <- vcalc(vi, cluster=study, grp1=group1, grp2=group2, w1=n_sz, w2=n_hc,</pre>
           obs=task.diff, rho=R, data=dat)
### correlation matrix for study 3 with four patient groups and a single control group
round(cov2cor(V[dat$study == 3, dat$study == 3]), 2)
### correlation matrix for study 6 with two tasks with four difficulties
cov2cor(V[dat$study == 6, dat$study == 6])
```

```
### correlation matrix for study 24 with two independent groups
cov2cor(V[dat$study == 24, dat$study == 24])
### correlation matrix for study 29 with three independent groups
cov2cor(V[dat$study == 29, dat$study == 29])
### fit multilevel model as above, but now use this V matrix in the model
res <- rma.mv(yi, V, random = ~ 1 | study/comp, data=dat)</pre>
res
predict(res, digits=2)
### use cluster-robust inference methods based on this model
robust(res, cluster=study)
### use methods from the clubSandwich package
robust(res, cluster=study, clubSandwich=TRUE)
### examine if task difficulty is a potential moderator of the effect
res <- rma.mv(yi, V, mods = ~ difficulty, random = ~ 1 | study/comp, data=dat)
res
sav <- robust(res, cluster=study)</pre>
sav
sav <- robust(res, cluster=study, clubSandwich=TRUE)</pre>
sav
### draw bubble plot
regplot(sav, xlab="Task Difficulty", ylab="Standardized Mean Difference", las=1, digits=1, bty="l")
## End(Not run)
```

dat.konstantopoulos2011

Studies on the Effects of Modified School Calendars on Student Achievement

# Description

Results from 56 studies on the effects of modified school calendars on student achievement.

### Usage

```
dat.konstantopoulos2011
```

#### Format

The data frame contains the following columns:

**district** numeric district id number **school** numeric school id number (within district)

study	numeric	study id number
yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance
year	numeric	year of the study

## Details

Instead of following the more traditional school calendar with a long summer break (in addition to a short winter and spring break), some schools have switched to a modified school calendar comprising more frequent but shorter intermittent breaks (e.g., 9 weeks of school followed by 3 weeks off), while keeping the total number of days at school approximately the same. The effects of using such a modified calendar on student achievement have been examined in a number of studies and were meta-analyzed by Cooper et al. (2003).

The dataset (taken from Konstantopoulos, 2011) contains the results from 56 studies, each comparing the level of academic achievement in a group of students following a modified school calendar with that of a group of students following a more traditional school calendar. The difference between the two groups was quantified in terms of a standardized mean difference (with positive values indicating a higher mean level of achievement in the group following the modified school calendar).

The studies were conducted at various schools that were clustered within districts. The data therefore have a multilevel structure, with schools nested within districts. A multilevel meta-analysis of these data can be used to estimate and account for the amount of heterogeneity between districts and between schools within districts.

#### Concepts

education, standardized mean differences, multilevel models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Konstantopoulos, S. (2011). Fixed effects and variance components estimation in three-level metaanalysis. *Research Synthesis Methods*, **2**(1), 61–76. https://doi.org/10.1002/jrsm.35

# References

Cooper, H., Valentine, J. C., Charlton, K., & Melson, A. (2003). The effects of modified school calendars on student achievement and on school and community attitudes. *Review of Educational Research*, **73**(1), 1–52. https://doi.org/10.3102/00346543073001001

```
### copy data into 'dat' and examine data
dat <- dat.konstantopoulos2011
dat</pre>
```

## dat.landenberger2005

```
## Not run:
### load metafor package
library(metafor)
### fit random-effects model
res <- rma(yi, vi, data=dat)</pre>
print(res, digits=3)
### fit random-effects model using rma.mv()
res <- rma.mv(yi, vi, random = ~ 1 | study, data=dat)</pre>
print(res, digits=3)
### fit multilevel random-effects model
res.ml <- rma.mv(yi, vi, random = ~ 1 | district/school, data=dat)</pre>
print(res.ml, digits=3)
### profile variance components
profile(res.ml, progbar=FALSE)
### fit multivariate parameterization of the model
res.mv <- rma.mv(yi, vi, random = ~ school | district, data=dat)</pre>
print(res.mv, digits=3)
### tau^2 = sum of the two variance components from the multilevel model
round(sum(res.ml$sigma2), digits=3)
### rho = intraclass correlation coefficient based on the multilevel model
round(res.ml$sigma2[1] / sum(res.ml$sigma2), digits=3)
## End(Not run)
```

dat.landenberger2005 Studies on the Effectiveness of CBT for Reducing Recidivism

## Description

Results from 58 studies on the effectiveness of cognitive-behavioral therapy (CBT) for reducing recidivism in juvenile and adult offenders.

### Usage

dat.landenberger2005

## Format

The data frame contains the following columns:

study	character	(first) author and year
pubtype	character	publication type (book chapter, journal article, report, or thesis)
country	character	country where study was carried out (Canada, New Zealand, UK, or USA)

# dat.landenberger2005

design	character	study design (matched groups, nonequivalent groups, or randomized trial)
program	character	purpose of setting up the CBT program (for demonstration, practice, or research purposes)
setting	character	treatment setting (community or prison)
designprob	character	indication of study design problems (no, favors the control group, or favors the treatment g
n.ctrl.rec	numeric	number of recidivists in the control group
n.ctrl.non	numeric	number of non-recidivists in the control group
n.cbt.rec	numeric	number of recidivists in the CBT group
n.cbt.non	numeric	number of non-recidivists in the CBT group
interval	numeric	recidivism interval (in months)
group	numeric	study group (adults or juveniles)
age	numeric	mean age of the study group
male	numeric	percentage of males in the study group
minority	numeric	percentage of minorities in the study group
length	numeric	treatment length (in weeks)
sessions	numeric	number of CBT sessions per week
hrs_week	numeric	treatment hours per week
hrs_total	numeric	total hours of treatment
cbt.cogskills	character	CBT component: cognitive skills (yes, no)
cbt.cogrestruct	character	CBT component: cognitive restructuring (yes, no)
cbt.intpprbsolv	character	CBT component: interpersonal problem solving (yes, no)
cbt.socskills	character	CBT component: social skills (yes, no)
cbt.angerctrl	character	CBT component: anger control (yes, no)
cbt.victimimpact	character	CBT component: victim impact (yes, no)
cbt.subabuse	character	CBT component: substance abuse (yes, no)
cbt.behavmod	character	CBT component: behavior modification (yes, no)
cbt.relapseprev	character	CBT component: relapse prevention (yes, no)
cbt.moralrsng	character	CBT component: moral reasoning (yes, no)
cbt.roletaking	character	CBT component: role taking (yes, no)
cbt.other	character	CBT component: other (yes, no)

## Details

Landenberger and Lipsey (2005) conducted a meta-analysis of 58 experimental and quasi-experimental studies of the effects of cognitive-behavioral therapy (CBT) on the recidivism rates of adult and juvenile offenders (see also Lipsey et al., 2007). The present dataset includes the results of these studies and a range of potential moderator variables to identify factors associated with variation in treatment effects.

## Concepts

psychology, criminology, odds ratios, meta-regression

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Personal communication.

## References

Landenberger, N. A., & Lipsey, M. W. (2005). The positive effects of cognitive-behavioral programs for offenders: A meta-analysis of factors associated with effective treatment. *Journal of Experimental Criminology*, **1**, 451–476. https://doi.org/10.1007/s11292-005-3541-7

Lipsey, M. W., Landenberger, N. A., & Wilson, S. J. (2007). Effects of cognitive-behavioral programs for criminal offenders. *Campbell Systematic Reviews*, **3**(1), 1–27. https://doi.org/10.4073/csr.2007.6

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.landenberger2005
head(dat)
## Not run:
### load metafor package
library(metafor)</pre>
```

### calculate log odds ratios (for non-recidivism in CBT vs. control groups) and sampling variances dat <- escalc(measure="OR", ai=n.cbt.non, bi=n.cbt.rec, ci=n.ctrl.non, di=n.ctrl.rec, data=dat)</pre>

```
### fit random-effects model
res <- rma(yi, vi, data=dat)
res</pre>
```

### estimated average OR and corresponding 95% CI/PI
predict(res, transf=exp, digits=2)

```
### examine if number of treatment sessions per week is a potential moderator
res <- rma(yi, vi, mods = ~ sessions, data=dat)
res</pre>
```

### predicted ORs for 1, 2, 5, or 10 sessions per week
predict(res, newmods=c(1,2,5,10), transf=exp, digits=2)

## End(Not run)

dat.laopaiboon2015 Studies on the Effectiveness of Azithromycin for Treating Lower Respiratory Tract Infections

## Description

Results from 15 studies on the effectiveness of azithromycin versus amoxycillin or amoxycillin/clavulanic acid (amoxyclav) in the treatment of acute lower respiratory tract infections.

### Usage

dat.laopaiboon2015

# Format

The data frame contains the following columns:

author	character	author(s)
year	numeric	publication year
ai	numeric	number of clinical failures in the group treated with azithromycin
n1i	numeric	number of patients in the group treated with azithromycin
ci	numeric	number of clinical failures in the group treated with amoxycillin or amoxyclav
n2i	numeric	number of patients in the group treated with amoxycillin or amoxyclav
age	character	whether the trial included adults or children
diag.ab	numeric	trial included patients with a diagnosis of acute bacterial bronchitis
diag.cb	numeric	trial included patients with a diagnosis of chronic bronchitis with acute exacerbation
diag.pn	numeric	trial included patients with a diagnosis of pneumonia
ctrl	character	antibiotic in control group (amoxycillin or amoxyclav)

## Details

Azithromycin is an antibiotic useful for the treatment of a number of bacterial infections. Laopaiboon et al. (2015) conducted a meta-analysis of trials comparing the effectiveness of azithromycin versus amoxycillin or amoxycillin/clavulanic acid (amoxyclav) in the treatment of acute lower respiratory tract infections, including acute bacterial bronchitis, acute exacerbations of chronic bronchitis, and pneumonia. The results from 15 trials are included in this dataset.

### Concepts

medicine, risk ratios

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Laopaiboon, M., Panpanich, R., & Swa Mya, K. (2015). Azithromycin for acute lower respiratory tract infections. *Cochrane Database of Systematic Reviews*, **3**, CD001954. https://doi.org/10.1002/14651858.CD00195

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.laopaiboon2015
dat
### Not run:
### load metafor package
library(metafor)
#### analysis using the Mantel-Haenszel method
rma.mh(measure="RR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, digits=3)
### calculate log risk ratios and corresponding sampling variances</pre>
```

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## dat.lau1992

```
dat <- escalc(measure="RR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat)
### random-effects model
res <- rma(yi, vi, data=dat)
res
### average risk ratio with 95% CI
predict(res, transf=exp)
## End(Not run)</pre>
```

dat.lau1992 Studies on Intravenous Streptokinase for Acute Myocardial Infarction

## Description

Results from 33 trials comparing intravenous streptokinase versus placebo or no therapy in patients who had been hospitalized for acute myocardial infarction.

## Usage

dat.lau1992

# Format

The data frame contains the following columns:

trial	character	trial name
year	numeric	publication year
ai	numeric	number of deaths in the streptokinase group
n1i	numeric	number of patients in the streptokinase group
ci	numeric	number of deaths in the control group
n2i	numeric	number of patients in the control group

## Details

In the paper by Lau et al. (1992), the data are used to illustrate the idea of a cumulative metaanalysis, where the results are updated as each trial is added to the dataset. See 'Examples' for code that replicates the results and shows corresponding forest plots.

## Concepts

medicine, cardiology, odds ratios, cumulative meta-analysis

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Lau, J., Antman, E. M., Jimenez-Silva, J., Kupelnick, B., Mosteller, F., & Chalmers, T. C. (1992). Cumulative meta-analysis of therapeutic trials for myocardial infarction. *New England Journal of Medicine*, **327**(4), 248–254. https://doi.org/10.1056/NEJM199207233270406

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.lau1992</pre>
dat
## Not run:
### load metafor package
library(metafor)
### meta-analysis of log odds ratios using the MH method
res <- rma.mh(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, slab=trial)
print(res, digits=2)
### forest plot
forest(res, xlim=c(-11,9), atransf=exp, at=log(c(0.01, 0.1, 1, 10, 100)),
       ilab=dat$year, ilab.xpos=-7)
text(-7, 35, "Year", font=2)
### cumulative meta-analysis
sav <- cumul(res)</pre>
### forest plot of the cumulative results
forest(sav, xlim=c(-5,4), atransf=exp, at=log(c(0.1, 0.5, 1, 2, 10)),
       ilab=dat$year, ilab.xpos=-3)
text(-3, 35, "Year", font=2)
id <- c(4, 8, 15, 33) # rows for which the z/p-values should be shown (as in Lau et al., 1992)
text(1.1, (res$k:1)[id], paste0("z = ", fmtx(sav$zval[id], digits=2),
                        fmtp(sav$pval[id], pname=", p", equal=TRUE, sep=TRUE, add0=TRUE)))
```

## End(Not run)

dat.lee2004

Studies on Acupoint P6 Stimulation for Preventing Nausea

#### Description

Results from studies examining the effectiveness of wrist acupuncture point P6 stimulation for preventing postoperative nausea.

### Usage

dat.lee2004

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## dat.lee2004

## Format

The data frame contains the following columns:

id	numeric	trial id number
study	character	first author
year	numeric	study year
ai	numeric	number of patients experiencing nausea in the treatment group
n1i	numeric	total number of patients in treatment group
ci	numeric	number of patients experiencing nausea in the sham group
n2i	numeric	total number of patients in the sham group

## Details

Postoperative nausea and vomiting are common complications following surgery and anaesthesia. As an alternative to drug therapy, acupuncture has been studied as a potential treatment in several trials. The dataset contains the results from 16 clinical trials examining the effectiveness of wrist acupuncture point P6 stimulation for preventing postoperative nausea.

## Concepts

medicine, alternative medicine, risk ratios

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Lee, A., & Done, M. L. (2004). Stimulation of the wrist acupuncture point P6 for preventing postoperative nausea and vomiting. *Cochrane Database of Systematic Reviews*, **3**, CD003281. https://doi.org/10.1002/14651858.CD003281.pub2

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.lee2004
dat</pre>
```

## Not run: ### load metafor package library(metafor)

```
### meta-analysis based on log risk ratios
res <- rma(measure="RR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat)
res
predict(res, transf=exp, digits=2)</pre>
```

## End(Not run)

dat.lehmann2018

# Description

Results from studies in which participants rated the attractiveness of photos that featured red or a control color. See OSF project at https://osf.io/xy47p/.

# Usage

dat.lehmann2018

## Format

The data frame contains the following columns:

Short_Title	character	Shortened citation formatted as: Author name(s), year of publication - Experiment n
Full_Citation	character	Full citation in APA format.
Short_Citation	character	Shortened citation of different format, exactly as it would appear in an in-text citatio
Year	numeric	Year study published (whether in journal or published online).
Study	character	Experiment number. If only one experiment presented in a paper, then 'Exp 1', other
Peer_Reviewed	character	Whether the experiment was published in a peer-reviewed journal or not. 'Yes' = pe
Source_Type	character	Location where experiment is available, including journal articles, conference proce
Preregistered	character	Whether experiment was pre-registered or not.
Moderator_Group	character	In some studies, a moderator was intentionally investigated that was meant to reduce
Gender	character	Gender of rater (male or female). In all cases, gender of stimuli will be opposite.
Color_Contrast	character	The color used as the contrast against red. In some cases, not every contrast color wa
Color_Form	character	Location of color in photo. Background = background or border color manipulated;
Photo_Type	character	Amount of body visible in photo. Head Shot = head only; Bust = head, shoulders, so
DV_Type	character	Scale used for DV. 'Perceived attractiveness' = the perceived attractiveness scale use
DV_Items	numeric	Number of items in DV scale.
DV_Scale	character	Full length of DV scale, if clear.
DV_ScaleBottom	numeric	Lower anchor of DV scale.
DV_ScaleTop	numeric	Upper anchor of DV scale.
Location	character	Country where study took place, if clear. 'Worldwide' in some cases of online partic
Continent	character	Continent where study took place, for the sake of creating larger categories for analy
Participants	character	Basic notes about participants. Students = high school, undergraduate, or graduate st
Participant_Notes	character	A finer grained description of participant characteristics.
Design	character	Whether study was a between- or within-subjects design.
Eth_Majority	character	Basic notes about participant ethnicity for ease of analysis. This represents the ethni
Eth_Majority_Detail	character	A finer grained description of participant characteristics, including in some cases par
Eth_Stim	character	Ethnicity of the people pictured in the stimulus materials.
Eth_Match	character	Whether the ethnic majority of the participant pool matched the ethnicity of stimulus
Red_Age	numeric	Mean age of participants in red group. If not given for specific group, then mean age
Control_Age	numeric	Mean age of participants in control group. If not given for specific group, then mean
Color_Red	character	Specific values of red color, if given.
Color_Control	character	Specific values of control color, if given.

### dat.lehmann2018

Red_Original Color_Match Presentation_Control Stimuli_Presentation Red_N Red_M Red_SD Control_N Control_M Control_SD SD_diff RM_r Control_Attractiveness Notes Total.SampleSize pooled	character character character numeric numeric numeric numeric numeric numeric numeric numeric numeric character numeric numeric numeric	<ul> <li>Whether the red color used in the study is within 5 units of the LCh values for red us</li> <li>Whether the control color used in the study is within 5 units of the red color on the I</li> <li>Whether the color of the stimulus viewed by each participant was consistent, as in pa</li> <li>Method for presenting stimuli. 'Paper' = stimuli printed on paper, shown in-person;</li> <li>Number of participants in red group.</li> <li>Mean rating of DV in red group.</li> <li>Standard deviation of DV in control group.</li> <li>Mean rating of DV in control group.</li> <li>Standard deviation of DV in control group.</li> <li>Calculated for within-subjects studies, standard deviation of difference scores.</li> <li>Calculated for within-subjects studies, correlation between participant ratings of red</li> <li>Attractiveness of stimuli in control condition, calculated as (Control_M - DV_Scale</li> <li>Any additional notes on the study.</li> <li>Pooled standard deviation for within-subjects studies.</li> </ul>
-		
pooled yi vi	numeric numeric numeric	Standardized mean difference. Corresponding sampling variance.

## Details

This is data from a meta-analysis of studies that test the red-romance hypothesis, which is that the color red enhances heterosexual attraction in romantic contexts. Analyzing male participants only, the meta-analysis should show a small, statistically significant effect (d = 0.26 [0.12, 0.40], p = .0004, N = 2,961). Analyzing female participants only should show a very small effect (d = 0.13 [0.01, 0.25], p = .03, N = 2,739). The analyses in the published meta-analysis found clear evidence of upward bias in the estimate for female participants and equivocal evidence for male participants. Moderator analyses suggest effect sizes may have declined over time (both genders), may be largest when an original shade of red is used (men only), and may be smaller in pre-registered studies (women only).

## Concepts

psychology, attraction, standardized mean differences

# Author(s)

Robert Calin-Jageman, <rcalinjageman@dom.edu>, https://calin-jageman.net

# Source

Lehmann, G. K., Elliot, A. J., & Calin-Jageman, R. J. (2018). Meta-analysis of the effect of red on perceived attractiveness. *Evolutionary Psychology*, **16**(4). https://doi.org/10.1177/1474704918802412 https://osf.io/xy47p/

### Examples

### copy data into 'dat' and examine data
dat <- dat.lehmann2018</pre>

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```
head(dat)
## Not run:
### load metafor package
library(metafor)
### meta-analyses for male and female participants
red_romance_malep <- dat[dat$Gender == "Males", ]
red_romance_femalep <- dat[dat$Gender == "Females", ]
res_malep <- rma(yi, vi, data=red_romance_malep, test="knha")
res_femalep
## End(Not run)</pre>
```

```
dat.li2007
```

Studies on the Effectiveness of Intravenous Magnesium in Acute Myocardial Infarction

## Description

Results from 22 trials examining the effectiveness of intravenous magnesium in the prevention of death following acute myocardial infarction.

### Usage

dat.li2007

## Format

The data frame contains the following columns:

id	numeric	trial id number
study	character	first author or trial name
year	numeric	publication year
ai	numeric	number of deaths in the magnesium group
n1i	numeric	number of patients in the magnesium group
ci	numeric	number of deaths in the control group
n2i	numeric	number of patients in the control group

## Details

The dataset includes the results from 22 randomized clinical trials that examined the effectiveness of intravenous magnesium in the prevention of death following acute myocardial infarction. It is similar to the dataset dat.egger2001, with some slight differences in the included trials and data used.

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## dat.lim2014

### Concepts

medicine, cardiology, odds ratios, publication bias

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Li, J., Zhang, Q., Zhang, M., & Egger, M. (2007). Intravenous magnesium for acute myocardial infarction. *Cochrane Database of Systematic Reviews*, **2**, CD002755. https://doi.org/10.1002/14651858.CD002755.pub2

### See Also

dat.egger2001

### Examples

```
### copy data into 'dat' and examine data
dat <- dat.li2007
dat
## Not run:
### load metafor package
library(metafor)
### meta-analysis of all trials except ISIS-4
res <- rma(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, method="EE", subset=-14)
print(res, digits=2)
predict(res, transf=exp, digits=2)
### meta-analysis of all trials including ISIS-4
res <- rma(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, method="EE")
print(res, digits=2)
predict(res, transf=exp, digits=2)
### contour-enhanced funnel plot centered at 0
funnel(res, refline=0, level=c(90, 95, 99), shade=c("white", "gray", "darkgray"))
## End(Not run)
```

dat.lim2014

Studies on the Association Between Maternal Size, Offspring Size, and Number of Offsprings

### Description

Results from studies examining the association between maternal size, offspring size, and number of offsprings.

## Usage

dat.lim2014

## Format

The object is a list containing data frames m\_o\_size, m\_o\_fecundity, o\_o\_unadj, and o\_o\_adj that contain the following columns and the corresponding phylogenetic trees called m\_o\_size\_tree, m\_o\_fecundity\_tree, o\_o\_unadj\_tree, and o\_o\_adj\_tree:

article	numeric	article id
author	character	study author(s)
year	numeric	publication year
species	character	species
amniotes	character	whether the species was amniotic
environment	character	whether the species were wild or captive
reprounit	character	whether the data were based on lifetime reproductive output or a single reproductive event (only
ri	numeric	correlation coefficient
ni	numeric	sample size

# Details

The object dat.lim2014 includes 4 datasets:

m_o_size	on the correlation between maternal size and offspring size
<pre>m_o_fecundity</pre>	on the correlation between maternal size and number of offsprings
o_o_unadj	on the correlation between offspring size and number of offsprings
o_o_adj	on the correlation between offspring size and number of offsprings adjusted for maternal size

Objects m\_o\_size\_tree, m\_o\_fecundity\_tree, o\_o\_unadj\_tree, and o\_o\_adj\_tree are the corresponding phylogenetic trees for the species included in each of these datasets.

## Concepts

ecology, evolution, correlation coefficients, multilevel models, phylogeny

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Lim, J. N., Senior, A. M., & Nakagawa, S. (2014). Heterogeneity in individual quality and reproductive trade-offs within species. *Evolution*, **68**(8), 2306–2318. https://doi.org/10.1111/evo.12446

## References

Cinar, O., Nakagawa, S., & Viechtbauer, W. (in press). Phylogenetic multilevel meta-analysis: A simulation study on the importance of modelling the phylogeny. *Methods in Ecology and Evolution*. https://doi.org/10.1111/2041-210X.13760

# 126

Hadfield, J. D., & Nakagawa, S. (2010). General quantitative genetic methods for comparative biology: Phylogenies, taxonomies and multi-trait models for continuous and categorical characters. *Journal of Evolutionary Biology*, **23**(3), 494–508. https://doi.org/10.1111/j.1420-9101.2009.01915.x

Nakagawa, S., & Santos, E. S. A. (2012). Methodological issues and advances in biological metaanalysis. *Evolutionary Ecology*, **26**(5), 1253–1274. https://doi.org/10.1007/s10682-012-9555-5

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.lim2014$o_o_unadj</pre>
dat[1:14, -c(2:3)]
## Not run:
### load metafor package
library(metafor)
### load ape package
library(ape, warn.conflicts=FALSE)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat)</pre>
### copy tree to 'tree'
tree <- dat.lim2014$o_o_unadj_tree</pre>
### compute branch lengths
tree <- compute.brlen(tree)</pre>
### compute phylogenetic correlation matrix
A <- vcv(tree, corr=TRUE)</pre>
### make copy of the species variable
dat$species.phy <- dat$species</pre>
### create effect size id variable
dat$esid <- 1:nrow(dat)</pre>
### fit multilevel phylogenetic meta-analytic model
res <- rma.mv(yi, vi,</pre>
   random = list(~ 1 | article, ~ 1 | esid, ~ 1 | species, ~ 1 | species.phy),
   R=list(species.phy=A), data=dat)
res
## End(Not run)
```

dat.linde2005

Studies on the Effectiveness of St. John's Wort for Treating Depression

## Description

Results from 26 studies on the effectiveness of Hypericum perforatum extracts (St. John's wort) for treating depression.

# Usage

dat.linde2005

## Format

The data frame contains the following columns:

numeric	study number		
character	study author(s)		
numeric	publication year		
character	study location		
numeric	total sample size		
numeric	sample restricted to patients who met criteria for major depression		
numeric	HRSD baseline score		
numeric	HRSD version (17 or 21 items)		
numeric	study duration (in weeks)		
character	Hypericum extract preparation		
numeric	dosage (in mg)		
numeric	definition of response (see 'Details')		
numeric	number of responses in treatment group		
numeric	number of patients in treatment group		
numeric	number of responses in placebo group		
numeric	number of patients in placebo group		
numeric	stratification variable used by the authors (see 'Details')		
	character numeric character numeric numeric numeric numeric numeric numeric numeric numeric numeric numeric numeric numeric numeric		

# Details

The dataset includes the results from 26 double-blind placebo-controlled trials on the effectiveness of Hypericum perforatum extracts (St. John's wort) for treating depression (note that 2 studies did not provide sufficient response information).

Data were extracted from Table 1 and Figure 3 from Linde et al. (2005). For study duration, the assessment week (instead of the total study duration) was coded for Philipp et al. (1999) and Montgomery et al. (2000). For dosage, the midpoint was coded when a range of values was given.

The definition of what constitutes a response differed across studies and is coded as follows:

- 1. HRSD score reduction of at least 50% or HRSD score after therapy <10,
- 2. HRSD reduction of at least 50%,
- 3. based on HRSD scale but exact definition not reported,
- 4. global patient assessment of efficacy,
- 5. at least 'much improved' on the Clinical Global Impression sub-scale for global improvement.

The group variable corresponds to the variable used by Linde et al. (2005) to stratify their analyses and is coded as follows:

- 1. smaller trials restricted to major depression,
- 2. larger trials restricted to major depression,
- 3. smaller trials not restricted to major depression,
- 4. larger trials not restricted to major depression.

#### Concepts

medicine, psychiatry, risk ratios

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Linde, K., Berner, M., Egger, M., & Mulrow, C. (2005). St John's wort for depression: Metaanalysis of randomised controlled trials. *British Journal of Psychiatry*, **186**(2), 99–107. https://doi.org/10.1192/bjp.18

# References

Viechtbauer, W. (2007). Accounting for heterogeneity via random-effects models and moderator analyses in meta-analysis. *Zeitschrift für Psychologie / Journal of Psychology*, **215**(2), 104–121. https://doi.org/10.1027/0044-3409.215.2.104

### Examples

```
### copy data into 'dat' and examine data
dat <- dat.linde2005
head(dat)</pre>
```

## Not run: ### load metafor package library(metafor)

### remove studies with no response information and study with no responses in either group dat <- dat[-c(5,6,26),]

```
### calculate log risk ratios and corresponding sampling variances
dat <- escalc(measure="RR", ai=ai, ci=ci, n1i=n1i, n2i=n2i, data=dat)
head(dat)
```

```
### meta-analysis of the log risk ratios using a random-effects model
res <- rma(yi, vi, data=dat, method="DL")
res</pre>
```

```
### mixed-effects meta-regression model with stratification variable
res <- rma(yi, vi, mods = ~ 0 + factor(group), data=dat, method="DL")
res</pre>
```

```
### predicted average risk ratio for each level of the stratification variable
predict(res, newmods=diag(4), transf=exp, digits=2)
```

```
## End(Not run)
```

dat.linde2015

# Description

Results from 66 trials examining eight classes of antidepressants and placebo for the primary care setting.

# Usage

dat.linde2015

## Format

The data frame contains the following columns:

id	integer	study ID		
author	character	first author		
year	integer	year of publication		
treatment1	character	treatment 1		
treatment2	character	treatment 2		
treatment3	character	treatment 3		
n1	integer	number of patients (arm 1)		
resp1	integer	number of early responder (arm 1)		
remi1	integer	number of early remissions (arm 1)		
loss1	integer	number of patients loss to follow-up (arm 1)		
loss.ae1	integer	number of patients loss to follow-up due to adverse events (arm 1)		
ae1	integer	number of patients with adverse events (arm 1)		
n2	integer	number of patients (arm 2)		
resp2	integer	number of early responder (arm 2)		
remi2	integer	number of early remissions (arm 2)		
loss2	integer	number of patients loss to follow-up (arm 2)		
loss.ae2	integer	number of patients loss to follow-up due to adverse events (arm 2)		
ae2	integer	number of patients with adverse events (arm 2)		
n3	integer	number of patients (arm 3)		
resp3	integer	number of early responder (arm 3)		
remi3	integer	number of early remissions (arm 3)		
loss3	integer	number of patients loss to follow-up (arm 3)		
loss.ae3	integer	number of patients loss to follow-up due to adverse events (arm 3)		
ae3	integer	number of patients with adverse events (arm 3)		

### Details

This dataset comes from a systematic review of 8 pharmacological treatments of depression and placebo in primary care with 66 studies (8 of which were 3-arm studies) including 14,785 patients.

The primary outcome is early response, defined as at least a 50% score reduction on a depression scale after completion of treatment. Secondary outcomes (also measured as dichotomous) were early remission (defined as having a symptom score below a fixed threshold after completion of treatment), lost to follow-up, lost to follow-up due to adverse events, and any adverse event. The odds ratio was used as effect measure.

This dataset was used as an example in Rücker and Schwarzer (2017) who introduced methods to resolve conflicting rankings of outcomes in network meta-analysis.

### Concepts

medicine, psychiatry, odds ratios, network meta-analysis

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Linde, K., Kriston, L., Rücker, G., Jamil, S., Schumann, I., Meissner, K., Sigterman, K., & Schneider, A. (2015). Efficacy and acceptability of pharmacological treatments for depressive disorders in primary care: Systematic review and network meta-analysis. *Annals of Family Medicine*, **13**(1), 69–79. https://doi.org/10.1370/afm.1687

### References

Rücker, G., & Schwarzer, G. (2017). Resolve conflicting rankings of outcomes in network metaanalysis: Partial ordering of treatments. *Research Synthesis Methods*, **8**(4), 526–536. https://doi.org/10.1002/jrsm.127

```
### Show results from first three studies (including three-arm study
### Lecrubier 1997)
head(dat.linde2015, 3)
```

```
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
```

```
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
```

```
### Change appearance of confidence intervals
cilayout("(", "-")
```

```
### Transform data from wide arm-based format to contrast-based format
### (outcome: early response). Argument 'sm' has to be used for odds
### ratio as summary measure; by default the risk ratio is used in the
### metabin function called internally.
pw1 <- pairwise(list(treatment1, treatment2, treatment3),</pre>
  event = list(resp1, resp2, resp3),
  n = list(n1, n2, n3),
  studlab = id, data = dat.linde2015, sm = "OR")
### Conduct random effects network meta-analysis for primary outcome
### (early response); small number of early responses is bad (argument
### small.values)
net1 <- netmeta(pw1, fixed = FALSE, reference = "Placebo", seq = trts,</pre>
  small.values = "bad")
net1
### Random effects NMA for early remission
pw2 <- pairwise(treat = list(treatment1, treatment2, treatment3),</pre>
  event = list(remi1, remi2, remi3),
  n = list(n1, n2, n3),
  studlab = id, data = dat.linde2015, sm = "OR")
net2 <- netmeta(pw2, fixed = FALSE,</pre>
   seq = trts, ref = "Placebo", small.values = "bad")
net2
### Ranking of treatments
nr1 <- netrank(net1)</pre>
nr2 <- netrank(net2)</pre>
nr1
nr2
### Partial order of treatment rankings (two outcomes)
outcomes <- c("Early response", "Early remission")</pre>
po12 <- netposet(nr1, nr2, outcomes = outcomes)</pre>
plot(po12)
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.linde2016 Studies on Antidepressants for the Primary Care Setting

### Description

Results from 93 trials examining 22 interventions (including placebo and usual care) for the primary care of depression.

### dat.linde2016

### Usage

dat.linde2016

### Format

The data frame contains the following columns:

id	integer	study ID
lnOR	numeric	response after treatment (log odds ratio)
selnOR	numeric	standard error of log odds ratio
treat1	character	first treatment
treat2	character	second treatment

# Details

This dataset comes from a network meta-analysis of 22 treatments of depression in primary care (Linde et al., 2016), based on 93 trials (79 two-arm trials, 13 three-arm trials, and one four-arm trial). The primary outcome was response after treatment (yes/no), defined as a reduction from baseline by at least 50% on a depression scale. The dataset contains log odds ratios with standard errors for all pairwise comparisons.

The interventions comprised both medical and psychological treatments, also in combination, including placebo and usual care (UC) (Linde et al., 2016). Pharmacological interventions were tricyclic antidepressants (TCA), selective serotonin reuptake inhibitors (SSRI), serotonin-noradrenaline reuptake inhibitors (SNRI), noradrenaline reuptake inhibitors (NRI), low- dose serotonin (5-HT2) antagonists and reuptake inhibitors (low-dose SARI), noradrenergic and specific serotonergic agents (NaSSa), reversible inhibitors of monoaminoxidase A (rMAO-A), hypericum extracts, and an individualized drug. Psychological interventions were cognitive behavioral therapy (CBT; four forms: face-to-face CBT, remote therapist-led CBT, guided self-help CBT, and no or minimal contact CBT), face-to-face problem-solving therapy (PST), face-to-face interpersonal psychotherapy, faceto-face CBT + SSRI, face-to-face PST + SSRI, and face-to-face interpersonal psychotherapy + SSRI.

The dataset was used as an example in Rücker et al. (2020) to illustrate component network metaanalysis using frequentist methods.

#### Concepts

medicine, psychiatry, odds ratios, network meta-analysis, component network meta-analysis

## Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

### Source

Linde, K., Rücker, G., Schneider, A., & Kriston, L. (2016). Questionable assumptions hampered interpretation of a network meta-analysis of primary care depression treatments. *Journal of Clinical Epidemiology*, **71**, 86–96. https://doi.org/10.1016/j.jclinepi.2015.10.010

## References

Rücker, G., Petropoulou, M., & Schwarzer, G. (2020). Network meta-analysis of multicomponent interventions. *Biometrical Journal*, **62**(3), 808–821. https://doi.org/10.1002/bimj.201800167

```
### Show results of first three studies (first study has three treatment
### arms)
head(dat.linde2016, 5)
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)</pre>
### Define order of treatments in printouts and forest plots
trts <- c("SSRI",</pre>
  "Face-to-face CBT", "Face-to-face interpsy", "Face-to-face PST",
  "Face-to-face CBT + SSRI", "Face-to-face interpsy + SSRI",
  "Face-to-face PST + SSRI",
  "Face-to-face psychodyn", "Other face-to-face",
  "TCA", "SNRI", "NRI", "Low-dose SARI", "NaSSa", "rMAO-A", "Ind drug",
  "Hypericum",
  "Remote CBT", "Self-help CBT", "No contact CBT",
  "UC", "Placebo")
### Conduct random effects network meta-analysis
net <- netmeta(lnOR, selnOR, treat1, treat2, id,</pre>
  data = dat.linde2016, reference.group = "placebo",
  seq = trts, sm = "OR", fixed = FALSE)
### Network graph
netgraph(net, seq = "o", number = TRUE)
### Show results
net
forest(net, xlim = c(0.2, 50))
### Additive component network meta-analysis with placebo as inactive
### treatment
nc <- netcomb(net, inactive = "placebo")</pre>
nc
forest(nc, xlim = c(0.2, 50))
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.lopez2019

# Description

Results from 76 studies examining the effectiveness of cognitive behavioral therapy (CBT) for depression in adults.

### Usage

dat.lopez2019

## Format

The data frame contains the following columns:

study	character	(first) author and year of study		
treatment	character	treatment provided (see 'Details')		
scale	character	scale used to measure depression symptoms		
n	numeric	group size		
diff	numeric	standardized mean change		
se	numeric	corresponding standard error		
group	numeric	type of therapy $(0 = individual, 1 = group therapy)$		
tailored	numeric	whether the intervention was tailored to each patient $(0 = no, 1 = yes)$		
sessions	numeric	number of sessions		
length	numeric	average session length (in minutes)		
intensity	numeric	product of sessions and length		
multi	numeric	intervention included multimedia elements $(0 = no, 1 = yes)$		
cog	numeric	intervention included cognitive techniques $(0 = no, 1 = yes)$		
ba	numeric	intervention included behavioral activation $(0 = no, 1 = yes)$		
psed	numeric	intervention included psychoeducation $(0 = no, 1 = yes)$		
home	numeric	intervention included homework $(0 = no, 1 = yes)$		
prob	numeric	intervention included problem solving $(0 = no, 1 = yes)$		
soc	numeric	intervention included social skills training $(0 = no, 1 = yes)$		
relax	numeric	intervention included relaxation $(0 = no, 1 = yes)$		
goal	numeric	intervention included goal setting $(0 = no, 1 = yes)$		
final	numeric	intervention included a final session $(0 = no, 1 = yes)$		
mind	numeric	intervention included mindfulness $(0 = no, 1 = yes)$		
act	numeric	intervention included acceptance and commitment therapy $(0 = no, 1 = yes)$		

# Details

The dataset includes the results from 76 studies examining the effectiveness of cognitive behavioral therapy (CBT) for treating depression in adults. Studies included two or more of the following treatments/conditions:

1. treatment as usual (TAU),

- 2. no treatment,
- 3. wait list,
- 4. psychological or attention placebo,
- 5. face-to-face CBT,
- 6. multimedia CBT,
- 7. hybrid CBT (i.e., multimedia CBT with one or more face-to-face sessions).

Multimedia CBT was defined as CBT delivered via self-help books, audio/video recordings, telephone, computer programs, apps, e-mail, or text messages.

Variable diff is the standardized mean change within each group, with negative values indicating a decrease in depression symptoms.

## Concepts

psychiatry, standardized mean changes, network meta-analysis

# Author(s)

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#### Source

Personal communication.

### References

López-López, J. A., Davies, S. R., Caldwell, D. M., Churchill, R., Peters, T. J., Tallon, D., Dawson, S., Wu, Q., Li, J., Taylor, A., Lewis, G., Kessler, D. S., Wiles, N., & Welton, N. J. (2019). The process and delivery of CBT for depression in adults: A systematic review and network meta-analysis. *Psychological Medicine*, **49**(12), 1937–1947. https://doi.org/10.1017/S003329171900120X

```
### copy data into 'dat' and examine data
dat <- dat.lopez2019
dat[1:10,1:6]
## Not run:
### load metafor package
library(metafor)
### create network graph ('igraph' package must be installed)
library(igraph, warn.conflicts=FALSE)
pairs <- data.frame(do.call(rbind,
    sapply(split(dat$treatment, dat$study), function(x) t(combn(x,2)))), stringsAsFactors=FALSE)
pairs$X1 <- factor(pairs$X1, levels=sort(unique(dat$treatment)))
pairs$X2 <- factor(pairs$X2, levels=sort(unique(dat$treatment)))
tab <- table(pairs[,1], pairs[,2])
tab # adjacency matrix</pre>
```

```
g <- graph_from_adjacency_matrix(tab, mode = "plus", weighted=TRUE, diag=FALSE)
plot(g, edge.curved=FALSE, edge.width=E(g)$weight/2,
   layout=layout_in_circle(g, order=c("Wait list", "No treatment", "TAU", "Multimedia CBT",
                                          "Hybrid CBT", "F2F CBT", "Placebo")),
   vertex.size=45, vertex.color="lightgray", vertex.label.color="black", vertex.label.font=2)
### restructure data into wide format
dat <- to.wide(dat, study="study", grp="treatment", ref="TAU",</pre>
               grpvars=c("diff","se","n"), postfix=c("1","2"))
### compute contrasts between treatment pairs and corresponding sampling variances
dat$yi <- with(dat, diff1 - diff2)</pre>
dat$vi <- with(dat, se1^2 + se2^2)</pre>
### calculate the variance-covariance matrix for multitreatment studies
calc.v <- function(x) {</pre>
   v <- matrix(x$se2[1]^2, nrow=nrow(x), ncol=nrow(x))</pre>
   diag(v) <- x$vi</pre>
   V
}
V <- bldiag(lapply(split(dat, dat$study), calc.v))</pre>
### add contrast matrix to the dataset
dat <- contrmat(dat, grp1="treatment1", grp2="treatment2")</pre>
### network meta-analysis using a contrast-based random-effects model
### by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons
### the treatment left out (TAU) becomes the reference level for the treatment comparisons
res <- rma.mv(yi, V, data=dat,</pre>
      mods = ~ 0 + No.treatment + Wait.list + Placebo + F2F.CBT + Hybrid.CBT + Multimedia.CBT,
         random = \sim comp | study, rho=1/2)
res
### forest plot of the contrast estimates (treatments versus TAU)
forest(coef(res), diag(vcov(res)), slab=sub(".", " ", names(coef(res)), fixed=TRUE),
       xlim=c(-5,5), alim=c(-3,3), psize=1, header="Treatment",
       xlab="Difference in Standardized Mean Change (compared to TAU)")
### fit random inconsistency effects model (might have to switch optimizer to get convergence)
res <- rma.mv(yi, V, data=dat,</pre>
      mods = ~ 0 + No.treatment + Wait.list + Placebo + F2F.CBT + Hybrid.CBT + Multimedia.CBT,
         random = list(~ comp | study, ~ comp | design), rho=1/2, phi=1/2,
         control=list(optimizer="BFGS"))
res
## End(Not run)
```

dat.maire2019

Studies on Temporal Trends in Fish Community Structures in French Rivers

## Description

Results from studies examining changes in the abundance of fish species in French rivers.

### Usage

dat.maire2019

## Format

.

The object is a list containing a data frame called dat that contains the following columns and distance matrix called dmat:

site	character	study site	
station	character	sampling station at site	
site_station	character	site and station combined	
s1	numeric	Mann-Kendal trend statistic for relative abundance of non-local species	
vars1	numeric	corresponding sampling variance (corrected for temporal autocorrelation)	
s2	numeric	Mann-Kendal trend statistic for relative abundance of northern species	
vars2	numeric	corresponding sampling variance (corrected for temporal autocorrelation)	
s3	numeric	Mann-Kendal trend statistic for relative abundance of non-native species	
vars3	numeric	corresponding sampling variance (corrected for temporal autocorrelation)	
const	numeric	constant value of 1	

## Details

The dataset includes the results from 35 sampling stations (at 11 sites along various French rivers) examining the abundance of various fish species over time (i.e., over 19-37 years, all until 2015). The temporal trend in these abundance data was quantified in terms of Mann-Kendal trend statistics, with positive values indicating monotonically increasing trends. The corresponding sampling variances were corrected for the temporal autocorrelation in the data (Hamed & Rao, 1998).

The distance matrix dmat indicates the distance of the sampling stations (1-423 river-km). For stations not connected through the river network, a high distance value of 10,000 river-km was set (effectively forcing the spatial correlation to be 0 for such stations).

The dataset can be used to illustrate a meta-analysis allowing for spatial correlation in the outcomes.

## Concepts

ecology, climate change, spatial correlation

### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Maire, A., Thierry, E., Viechtbauer, W., & Daufresne, M. (2019). Poleward shift in large-river fish communities detected with a novel meta-analysis framework. *Freshwater Biology*, **64**(6), 1143–1156. https://doi.org/10.1111/fwb.13291

## 138

### dat.mccurdy2020

## References

Hamed, K. H., & Rao, A. R. (1998). A modified Mann-Kendall trend test for autocorrelated data. *Journal of Hydrology*, **204**(1-4), 182–196. https://doi.org/10.1016/S0022-1694(97)00125-X

### Examples

```
### copy data into 'dat' and examine data
dat <- dat.maire2019$dat</pre>
dat[-10]
### copy distance matrix into 'dmat' and examine first 5 rows/columns
dmat <- dat.maire2019$dmat
dmat[1:5,1:5]
## Not run:
### load metafor package
library(metafor)
### fit a standard random-effects model ignoring spatial correlation
res1 <- rma.mv(s1, vars1, random = ~ 1 | site_station, data=dat)</pre>
res1
### fit model allowing for spatial correlation
res2 <- rma.mv(s1, vars1, random = ~ site_station | const, struct="SPGAU",</pre>
               data=dat, dist=list(dmat), control=list(rho.init=10))
res2
### add random effects for sites and stations within sites
res3 <- rma.mv(s1, vars1, random = list(~ 1 | site/station, ~ site_station | const), struct="SPGAU",
               data=dat, dist=list(dmat), control=list(rho.init=10))
res3
### likelihood ratio tests comparing the models
anova(res1, res2)
anova(res2, res3)
### profile likelihood plots for model res2
profile(res2, cline=TRUE)
### effective range (river-km for which the spatial correlation is \geq 0.05)
sqrt(3) * res2$rho
### note: it was necessary to adjust the starting value for rho in models
### res2 and res3 so that the optimizer does not get stuck in a local maximum
profile(res2, rho=1, xlim=c(0,200), steps=100)
## End(Not run)
```

dat.mccurdy2020 Studies on the Generation Effect

## Description

Results from 126 articles that examined the so-called 'generation effect'.

## Usage

dat.mccurdy2020

## Format

The data frame contains the following columns:

article	numeric	article identifier		
experiment	character	experiment (within article) identifier		
sample	numeric	sample (within experiment) identifier		
id	numeric	row identifier		
pairing	numeric	identifier to indicate paired conditions within experiments		
yi	numeric	mean recall rate for the condition		
vi	numeric	corresponding sampling variance		
ni	numeric	number of participants for the condition		
stimuli	numeric	number of stimuli for the condition		
condition	factor	condition ('read' or 'generate')		
gen_difficulty	factor	generation difficulty ('low' or 'high')		
manip_type	factor	manipulation type of the generate versus read condition (using a 'within' or 'between' subje		
present_style	factor	presentation style ('mixed' or 'pure' list presentation)		
word_status	factor	word status ('words', 'non-words', or 'numbers')		
memory_test	factor	memory test ('recognition', 'cued recall', or 'free recall')		
memory_type	factor	memory type ('item', 'source', 'font color', 'font type', 'order', 'cue word', 'background co		
gen_constraint	factor	generation constraint ('low', 'medium', or 'high')		
learning_type	factor	learning type ('incidental' or 'intentional')		
stimuli_relation	factor	stimuli relation ('semantic', 'category', 'antonym', 'synonym', 'rhyme', 'compound words'		
gen_mode	factor	generation mode ('verbal/speaking', 'covert/thinking', or 'writing/typing')		
gen_task	factor	generation task ('anagram', 'letter transposition', 'word fragment', 'sentence completion', '		
attention	factor	attention ('divided' or 'full')		
pacing	factor	pacing ('self-paced' or 'timed')		
filler_task	factor	filler task ('yes' or 'no')		
age_grp	factor	age group ('younger' or 'older' adults)		
retention_delay	factor	retention delay ('immediate', 'short', or 'long')		

# Details

The generation effect is the memory benefit for self-generated compared with read or experimenterprovided information (Jacoby, 1978; Slamecka & Graf, 1978). In a typical study, participants are presented with a list of stimuli (usually words or word pairs). For half of the stimuli, participants self-generate a target word (e.g., open-cl\_\_\_\_), while for the other half, participants simply read an intact target word (e.g., above-below). On a later memory test for the target words, the common finding is that self-generated words are better remembered than read words (i.e., the generation effect).

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Although several theories have been proposed to explain the generation effect, there is still some debate on the underlying memory mechanism(s) contributing to this phenomenon. The meta-analysis by McCurdy et al. (2020) translated various theories on the generation effect into hypotheses that could then be tested in moderator analyses based on a dataset containing 126 articles, 310 experiments, and 1653 mean recall estimates collected under various conditions.

Detailed explanations of the various variables coded (and how these can be used to test various hypotheses regarding the generation effect) can be found in the article. The most important variable is condition, which denotes whether a particular row of the dataset corresponds to the results of a 'read' or a 'generate' condition.

The data structure is quite complex. Articles may have reported the findings from multiple experiments involving one or multiple samples that were examined under various conditions. The pairing variable indicates which rows of the dataset represent a pairing of a read condition with one or multiple corresponding generate conditions within an experiment. A pairing may involve the same sample of subjects (when using a within-subjects design for comparing the conditions) or different samples (when using a between-subjects design).

#### Concepts

psychology, memory, proportions, raw means, multilevel models, cluster-robust inference

# Author(s)

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## Source

McCurdy, M. P., Viechtbauer, W., Sklenar, A. M., Frankenstein, A. N., & Leshikar, E. D. (2020). Theories of the generation effect and the impact of generation constraint: A meta-analytic review. *Psychonomic Bulletin & Review*, **27**(6), 1139–1165. https://doi.org/10.3758/s13423-020-01762-3

## References

Slamecka, N. J., & Graf, P. (1978). The generation effect: Delineation of a phenomenon. *Journal of Experimental Psychology: Human Learning and Memory*, **4**(6), 592–604. https://doi.org/10.1037/0278-7393.4.6.59

Jacoby, L. L. (1978). On interpreting the effects of repetition: Solving a problem versus remembering a solution. *Journal of Verbal Learning and Verbal Behavior*, **17**(6), 649–668. https://doi.org/10.1016/S0022-5371(

```
data=dat, sparse=TRUE, digits=3)
res
### proportion of total amount of heterogeneity due to each component
data.frame(source=res$s.names, sigma2=round(res$sigma2, 3),
    prop=round(res$sigma2 / sum(res$sigma2), 2))
### apply cluster-robust inference methods
sav <- robust(res, cluster=article, clubSandwich=TRUE)
sav
### estimated average recall rate in read and generate conditions
predict(sav, newmods = c(0,1), digits=3)
## End(Not run)</pre>
```

dat.mcdaniel1994 Studies on the Validity of Employment Interviews

## Description

Results from 160 studies on the correlation between employment interview assessments and job performance.

### Usage

dat.mcdaniel1994

### Format

The data frame contains the following columns:

study	numeric	study number
ni	numeric	sample size of the study
ri	numeric	observed correlation
type	character	interview type ( $j = job$ -related, $s = situational$ , $p = psychological$ )
struct	character	interview structure (u = unstructured, s = structured)

### Details

The 160 studies provide data in terms of the correlation between employment interview performance and actual job performance. In addition, the interview type and the interview structure are indicated.

McDaniel et al. (1994) describe the interview type and structure variables as follows. "Questions in situational interviews [...] focus on the individual's ability to project what his or her behavior would be in a given situation. [...] Job-related interviews are those in which the interviewer is a personnel officer or hiring authority and the questions attempt to assess past behaviors and job-related information, but most questions are not considered situational. Psychological interviews

are conducted by a psychologist, and the questions are intended to assess personal traits, such as dependability." In structured interviews, "the questions and acceptable responses were specified in advance and the responses were rated for appropriateness of content. [...] Unstructured interviews gather applicant information in a less systematic manner than do structured interviews. Although the questions may be specified in advance, they usually are not, and there is seldom a formalized scoring guide. Also, all persons being interviewed are not typically asked the same questions."

The goal of the meta-analysis was to examine the overall criterion-related validity of employment interviews and to examine whether the validity depends on the type and structure of the interview.

The data in this dataset were obtained from Table A.2 in Rothstein, Sutton, and Borenstein (2005, p. 325-329). Note that the type and struct variables contain some NAs.

### Concepts

psychology, correlation coefficients, meta-regression

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Rothstein, H. R., Sutton, A. J., & Borenstein, M. (Eds.). (2005). *Publication bias in meta-analysis: Prevention, assessment, and adjustments*. Chichester, England: Wiley.

## References

McDaniel, M. A., Whetzel, D. L., Schmidt, F. L., & Maurer, S. D. (1994). The validity of employment interviews: A comprehensive review and meta-analysis. *Journal of Applied Psychology*, **79**(4), 599–616. https://doi.org/10.1037/0021-9010.79.4.599

```
### copy data into 'dat' and examine data
dat <- dat.mcdaniel1994
head(dat)
## Not run:
### load metafor package
library(metafor)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat)
head(dat)
### meta-analysis of the transformed correlations using a random-effects model
res <- rma(yi, vi, data=dat)
res
### average correlation with 95% CI
predict(res, transf=transf.ztor, digits=2)</pre>
```

```
### mixed-effects model with interview type as factor
### note: job-related interviews is the reference level
res <- rma(yi, vi, mods = ~ factor(type), data=dat)</pre>
res
### estimated average correlation for each level of interview type
res <- rma(yi, vi, mods = ~ 0 + factor(type), data=dat)</pre>
predict(res, newmods=diag(3), transf=transf.ztor, digits=2)
### mixed-effects model with interview structure as factor
### note: structured interviews is the reference level
res <- rma(yi, vi, mods = ~ factor(struct), data=dat)</pre>
res
### estimated average correlation for each level of interview structure
res <- rma(yi, vi, mods = ~ 0 + factor(struct), data=dat)</pre>
predict(res, newmods=diag(2), transf=transf.ztor, digits=2)
### note: the interpretation of the results is difficult since all
### situational interviews were structured, almost all psychological
### interviews were unstructured, and actually for the majority of
### the psychological interviews it was unknown whether the interview
### was structured or unstructured
table(dat$type, dat$struct, useNA="always")
### meta-analysis of raw correlations using a random-effects model
res <- rma(measure="COR", ri=ri, ni=ni, data=dat.mcdaniel1994)</pre>
res
## End(Not run)
```

### Description

dat.michael2013

Results from studies exploring how a superfluous fMRI brain image influences the persuasiveness of a scientific claim.

The Non-Persuasive Power of a Brain Image

## Usage

```
dat.michael2013
```

#### Format

The data frame contains the following columns:

Study	character	name of the study: Citation - Experiment - Subgroup
No_brain_n	numeric	sample size for no-brain-image condition
No_brain_m	numeric	mean agreement rating for no-brain-image condition
-------------------	-----------	--
No_brain_s	numeric	standard deviation for no-brain-image condition
Brain_n	numeric	sample size for brain-image condition
Brain_m	numeric	mean agreement rating for brain-image condition
Brain_s	numeric	standard deviation for brain-image condition
Included_Critique	character	'Critique' if article included critical commentary on conclusions, otherwise 'No_critique'
Medium	character	'Paper' if conducted in person; 'Online' if conducted online
Compensation	character	notes on compensation provided to participants
Participant_Pool	character	notes on where participants were recruited
yi	numeric	raw mean difference, calculated as Brain_m - No_brain_m
vi	numeric	corresponding sampling variance

## Details

The dataset contains the data from the meta-analysis by Michael et al. (2013) of experiments on the persuasive power of a brain image. The meta-analysis analyzed an original study by McCabe and Castel (2008) as well as 10 replication attempts conducted by the authors of the meta-analysis.

In each study, participants read an article about using brain imaging as a lie detector. The article either included a superfluous fMRI image of a brain (brain) or not (no\_brain). After reading the article, all participants responded to the statement "Do you agree or disagree with the conclusion that brain imaging can be used as a lie detector?" on a scale from 1 (strongly disagree) to 4 (strongly agree).

The original study by McCabe and Castel (2008) reported a relatively large increase in agreement due to the presence of brain images. Meta-analysis of the original study with the 10 replications suggests, however, a small, possibly null effect: an estimated average raw mean difference of 0.07 points, 95% CI [-0.00, 0.14], under a random-effects model.

In some studies, the article included a passage critiquing the primary claims made in the article; this is coded in the Included\_Critique column for analysis as a possible moderator. Note that Experiment 3 by McCabe and Castel (2008) was a 2x2 between subjects design: brain image presence was manipulated as well as the inclusion of a critique. The two different critique conditions are recorded as separate rows in this dataset. Analysis of this dataset with metafor yields the same results (given rounding) reported in the manuscript.

#### Concepts

psychology, persuasion, raw mean differences

## Author(s)

Robert Calin-Jageman, <rcalinjageman@dom.edu>, https://calin-jageman.net

### Source

Michael, R. B., Newman, E. J., Vuorre, M., Cumming, G., & Garry, M. (2013). On the (non)persuasive power of a brain image. *Psychonomic Bulletin & Review*, **20**(4), 720—725. https://doi.org/10.3758/s13423-013-0391-

# References

McCabe, D. P., & Castel, A. D. (2008). Seeing is believing: The effect of brain images on judgments of scientific reasoning. *Cognition*, **107**(1), 343–352. https://doi.org/10.1016/j.cognition.2007.07.017

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.michael2013</pre>
dat
## Not run:
### load metafor package
library(metafor)
### Data prep
# yi and vi are already provided, but here's how you would use escalc() to obtain
# a raw-mean difference and its variance.
# Note the measure parameter is "MD" for 'raw mean difference'
dat <- metafor::escalc(</pre>
  measure = "MD",
  m1i = Brain_m,
 m2i = No_brain_m,
  sd1i = Brain_s,
  sd2i = No_brain_s,
  n1i = Brain_n,
 n2i = No_brain_n,
  data = dat
)
### meta-analysis using a random-effects model of the raw mean differences
res <- rma(yi, vi, data=dat)</pre>
print(res, digits=2)
### examine if Included_Critique is a potential moderator
res <- rma(yi, vi, mods = ~ Included_Critique, data=dat)</pre>
print(res, digits=2)
## End(Not run)
```

dat.molloy2014	Studies on the Relationship between Conscientiousness and Medica-
	tion Adherence

# Description

Results from 16 studies on the correlation between conscientiousness and medication adherence.

#### Usage

dat.molloy2014

#### dat.molloy2014

## Format

The data frame contains the following columns:

authors	character	study authors
year	numeric	publication year
ni	numeric	sample size of the study
ri	numeric	observed correlation
controls	character	number of variables controlled for
design	character	whether a cross-sectional or prospective design was used
a_measure	character	type of adherence measure (self-report or other)
c_measure	character	type of conscientiousness measure (NEO or other)
meanage	numeric	mean age of the sample
quality	numeric	methodological quality

# Details

Conscientiousness, one of the big-5 personality traits, can be defined as "socially prescribed impulse control that facilitates task- and goal-directed behaviour, such as thinking before acting, delaying gratification, following norms and rules and planning, organising and prioritising tasks" (John & Srivastava, 1999). Conscientiousness has been shown to be related to a number of health-related behaviors (e.g., tobacco/alcohol/drug use, diet and activity patterns, risky behaviors). A recent meta-analysis by Molloy et al. (2014) examined to what extent conscientiousness is related to medication adherence, that is, the extent to which (typically chronically ill) patients follow a prescribed medication regimen (e.g., taking a daily dose of a cholesterol lowering drug in patients with high LDL serum cholesterol levels). The results from the 16 studies included in this meta-analysis are provided in this dataset.

Variable a\_measure indicates whether adherence was measured based on self-reports or a more 'objective' measure (e.g., electronic monitoring of pill bottle openings, pill counts). Variable c\_measure indicates whether conscientiousness was measured with some version of the NEO personality inventory or some other scale. Methodological quality was scored by the authors on a 1 to 4 scale with higher scores indicating higher quality (see article for details on how this score was derived).

### Concepts

psychology, medicine, correlation coefficients

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Molloy, G. J., O'Carroll, R. E., & Ferguson, E. (2014). Conscientiousness and medication adherence: A meta-analysis. *Annals of Behavioral Medicine*, **47**(1), 92–101. https://doi.org/10.1007/s12160-013-9524-4

## References

John, O. P., & Srivastava, S. (1999). The Big Five Trait taxonomy: History, measurement, and theoretical perspectives. In L. A. Pervin & O. P. John (Eds.), *Handbook of personality: Theory and research* (2nd ed., pp. 102-138). New York: Guilford Press.

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.molloy2014</pre>
dat[-c(5:6)]
## Not run:
### load metafor package
library(metafor)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat, slab=paste(authors, year, sep=", "))</pre>
dat[-c(5:6)]
### meta-analysis of the transformed correlations using a random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### average correlation with 95% CI
predict(res, digits=3, transf=transf.ztor)
### forest plot
forest(res, addpred=TRUE, xlim=c(-1.6,1.6), atransf=transf.ztor,
       at=transf.rtoz(seq(-0.4, 0.6, by=0.2)), digits=c(2,1), cex=0.9,
       header="Author(s), Year")
### funnel plot
funnel(res)
## End(Not run)
```

dat.moura2021 Studies on Assortative Mating

# Description

Results from 457 studies on assortative mating in various species.

#### Usage

dat.moura2021

# dat.moura2021

# Format

The object is a list containing a data frame called dat that contains the following columns and a phylogenetic tree called tree:

study.id	character	study id
effect.size.id	numeric	effect size id
species	character	species
species.id	character	species id (as in the Open Tree of Life reference taxonomy)
subphylum	character	the subphyla of the species
phylum	character	the phyla of the species
assortment.trait	character	the measure of body size
trait.dimensions	character	dimensionality of the measure
field.collection	character	whether data were collected in the field
publication.year	numeric	publication year of the study
pooled.data	character	whether data were pooled either spatially and/or temporally
spatially.pooled	character	whether data were pooled spatially
temporally.pooled	character	whether data were pooled temporally
ri	numeric	correlation coefficient
ni	numeric	sample size

### Details

The 457 studies included in this dataset provide 1828 correlation coefficients describing the similarity in some measure of body size in mating couples in 341 different species.

## Concepts

ecology, evolution, correlation coefficients, multivariate models, phylogeny, meta-regression

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Rios Moura, R., Oliveira Gonzaga, M., Silva Pinto, N., Vasconcellos-Neto, J., & Requena, G. S. (2021). Assortative mating in space and time: Patterns and biases. *Ecology Letters*, **24**(5), 1089–1102. https://doi.org/10.1111/ele.13690

## References

Cinar, O., Nakagawa, S., & Viechtbauer, W. (in press). Phylogenetic multilevel meta-analysis: A simulation study on the importance of modelling the phylogeny. *Methods in Ecology and Evolution*. https://doi.org/10.1111/2041-210X.13760

Hadfield, J. D., & Nakagawa, S. (2010). General quantitative genetic methods for comparative biology: Phylogenies, taxonomies and multi-trait models for continuous and categorical characters. *Journal of Evolutionary Biology*, **23**(3), 494–508. https://doi.org/10.1111/j.1420-9101.2009.01915.x

Nakagawa, S., & Santos, E. S. A. (2012). Methodological issues and advances in biological metaanalysis. *Evolutionary Ecology*, **26**(5), 1253–1274. https://doi.org/10.1007/s10682-012-9555-5

#### dat.nakagawa2007

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.moura2021$dat</pre>
head(dat)
## Not run:
### load metafor package
library(metafor)
### load ape package
library(ape, warn.conflicts=FALSE)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=ri, ni=ni, data=dat)</pre>
### copy tree to 'tree'
tree <- dat.moura2021$tree</pre>
### turn tree into an ultrametric one
tree <- compute.brlen(tree)</pre>
### compute phylogenetic correlation matrix
A <- vcv(tree, corr=TRUE)</pre>
### make copy of the species.id variable
dat$species.id.phy <- dat$species.id</pre>
### fit multilevel phylogenetic meta-analytic model
res <- rma.mv(yi, vi,</pre>
  random = list(~ 1 | study.id, ~ 1 | effect.size.id, ~ 1 | species.id, ~ 1 | species.id.phy),
   R=list(species.id.phy=A), data=dat)
res
### examine if spatial and/or temporal pooling of data tends to yield larger correlations
res <- rma.mv(yi, vi,</pre>
   mods = ~ spatially.pooled * temporally.pooled,
  random = list(~ 1 | study.id, ~ 1 | effect.size.id, ~ 1 | species.id, ~ 1 | species.id.phy),
   R=list(species.id.phy=A), data=dat)
res
### estimated average correlation without pooling, when pooling spatially,
### when pooling temporally, and when pooling spatially and temporally
predict(res, newmods = rbind(c(0,0,0),c(1,0,0),c(0,1,0),c(1,1,1)), transf=transf.ztor, digits=2)
## End(Not run)
```

dat.nakagawa2007

Assessing the Function of House Sparrows' Bib Size Using a Flexible Meta-Analysis Method

## Description

A meta-analysis on the association between the size of a male's bib and their social status in house sparrows (*Passer domesticus*).

## Usage

dat.nakagawa2007

#### Format

The data frame contains the following columns:

StudyID	character	identity of primary study
Place	character	location of study population
Correlation	numeric	correlation coefficient
SampleSize	integer	sample size of population

# Details

Each study measures the association between a sparrows bib size and its social status. Effects are quantified as correlation coefficients.

# Concepts

ecology, correlation coefficients

#### Author(s)

Daniel Noble, <daniel.noble@anu.edu.au>

# Source

Nakagawa, S., Ockendon, N., Gillespie, D. O. S, Hatchwell, B. J., & Burke, T. (2007). Assessing the function of house sparrows' bib size using a flexible meta-analysis method. *Behavioral Ecology*, **18**(5), 831–840. https://doi.org/10.1093/beheco/arm050

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.nakagawa2007
dat
### Not run:
### load metafor package
library(metafor)
### calculate Zr
dat <- escalc(measure="ZCOR", ri=Correlation, ni=SampleSize, data=dat)</pre>
```

## dat.nielweise2007

```
### fit meta-analytic model
res <- rma.mv(yi, vi, random = ~ 1 | StudyID, data=dat)
res
## End(Not run)</pre>
```

dat.nielweise2007	Studies on Anti-Infective-Treated Central Venous Catheters for Pre-
	vention of Catheter-Related Bloodstream Infections

# Description

Results from 18 studies comparing the risk of catheter-related bloodstream infection when using anti-infective-treated versus standard catheters in the acute care setting.

## Usage

dat.nielweise2007

# Format

The data frame contains the following columns:

study	numeric	study number
author	character	(first) author
year	numeric	publication year
ai	numeric	number of CRBSIs in patients receiving an anti-infective catheter
n1i	numeric	number of patients receiving an anti-infective catheter
ci	numeric	number of CRBSIs in patients receiving a standard catheter
n2i	numeric	number of patients receiving a standard catheter

# Details

The use of a central venous catheter may lead to a catheter-related bloodstream infection (CRBSI), which in turn increases the risk of morbidity and mortality. Anti-infective-treated catheters have been developed that are meant to reduce the risk of CRBSIs. Niel-Weise et al. (2007) conducted a meta-analysis of studies comparing infection risk when using anti-infective-treated versus standard catheters in the acute care setting. The results from 18 such studies are included in this dataset.

The dataset was used in the article by Stijnen et al. (2010) to illustrate various generalized linear mixed-effects models for the meta-analysis of proportions and odds ratios (see 'References').

## Concepts

medicine, odds ratios, generalized linear models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Niel-Weise, B. S., Stijnen, T., & van den Broek, P. J. (2007). Anti-infective-treated central venous catheters: A systematic review of randomized controlled trials. *Intensive Care Medicine*, **33**(12), 2058–2068. https://doi.org/10.1007/s00134-007-0897-3

#### References

Stijnen, T., Hamza, T. H., & Ozdemir, P. (2010). Random effects meta-analysis of event outcome in the framework of the generalized linear mixed model with applications in sparse data. *Statistics in Medicine*, **29**(29), 3046–3067. https://doi.org/10.1002/sim.4040

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.nielweise2007
dat
## Not run:
### load metafor package
library(metafor)
### standard (inverse-variance) random-effects model
res <- rma(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, drop00=TRUE)
print(res, digits=3)
predict(res, transf=exp, digits=2)
### random-effects conditional logistic model
res <- rma.glmm(measure="OR", ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, model="CM.EL")
print(res, digits=3)
predict(res, transf=exp, digits=2)
## End(Not run)</pre>
```

dat.nielweise2008 Studies on Anti-Infective-Treated Central Venous Catheters for Prevention of Catheter-Related Bloodstream Infections

### Description

Results from 18 studies comparing the risk of catheter-related bloodstream infection when using anti-infective-treated versus standard catheters for total parenteral nutrition or chemotherapy.

#### Usage

dat.nielweise2008

## Format

The data frame contains the following columns:

## dat.nielweise2008

study	numeric	study number
authors	character	study authors
year	numeric	publication year
x1i	numeric	number of CRBSIs in patients receiving an anti-infective catheter
t1i	numeric	total number of catheter days for patients receiving an anti-infective catheter
x2i	numeric	number of CRBSIs in patients receiving a standard catheter
t2i	numeric	total number of catheter days for patients receiving a standard catheter

## Details

The use of a central venous catheter may lead to a catheter-related bloodstream infection (CRBSI), which in turn increases the risk of morbidity and mortality. Anti-infective-treated catheters have been developed that are meant to reduce the risk of CRBSIs. Niel-Weise et al. (2008) conducted a meta-analysis of studies comparing infection risk when using anti-infective-treated versus standard catheters for total parenteral nutrition or chemotherapy. The results from 9 such studies are included in this dataset.

The dataset was used in the article by Stijnen et al. (2010) to illustrate various generalized linear mixed-effects models for the meta-analysis of incidence rates and incidence rate ratios (see 'References').

# Concepts

medicine, incidence rates, generalized linear models

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Niel-Weise, B. S., Stijnen, T., & van den Broek, P. J. (2008). Anti-infective-treated central venous catheters for total parenteral nutrition or chemotherapy: A systematic review. *Journal of Hospital Infection*, **69**(2), 114–123. https://doi.org/10.1016/j.jhin.2008.02.020

## References

Stijnen, T., Hamza, T. H., & Ozdemir, P. (2010). Random effects meta-analysis of event outcome in the framework of the generalized linear mixed model with applications in sparse data. *Statistics in Medicine*, **29**(29), 3046–3067. https://doi.org/10.1002/sim.4040

```
### copy data into 'dat' and examine data
dat <- dat.nielweise2008
dat
## Not run:
### load metafor package
library(metafor)</pre>
```

```
### standard (inverse-variance) random-effects model
res <- rma(measure="IRR", x1i=x1i, t1i=t1i, x2i=x2i, t2i=t2i, data=dat)
print(res, digits=3)
predict(res, transf=exp, digits=2)
### random-effects conditional Poisson model
res <- rma.glmm(measure="IRR", x1i=x1i, t1i=t1i, x2i=x2i, t2i=t2i, data=dat, model="CM.EL")
print(res, digits=3)
predict(res, transf=exp, digits=2)
## End(Not run)
## End(Not run)</pre>
```

dat.normand1999 Studies on the Length of Hospital Stay of Stroke Patients

### Description

Results from 9 studies on the length of the hospital stay of stroke patients under specialized care and under conventional/routine (non-specialist) care.

#### Usage

dat.normand1999

#### Format

The data frame contains the following columns:

study	numeric	study number
source	character	source of data
n1i	numeric	number of patients under specialized care
m1i	numeric	mean length of stay (in days) under specialized care
sd1i	numeric	standard deviation of the length of stay under specialized care
n2i	numeric	number of patients under routine care
m2i	numeric	mean length of stay (in days) under routine care
sd2i	numeric	standard deviation of the length of stay under routine care

# Details

The 9 studies provide data in terms of the mean length of the hospital stay (in days) of stroke patients under specialized care and under conventional/routine (non-specialist) care. The goal of the meta-analysis was to examine the hypothesis whether specialist stroke unit care will result in a shorter length of hospitalization compared to routine management.

## Concepts

medicine, raw mean differences, standardized mean differences

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Normand, S. T. (1999). Meta-analysis: Formulating, evaluating, combining, and reporting. *Statistics in Medicine*, **18**(3), 321–359. https://doi.org/10.1002/(sici)1097–0258(19990215)18:3<321::aid-sim28>3.0

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.normand1999
dat
## Not run:
### load metafor package
library(metafor)
### calculate mean differences and corresponding sampling variances
dat <- escalc(measure="MD", m1i=m1i, sd1i=sd1i, n1i=n1i, m2i=m2i, sd2i=sd2i, n2i=n2i, data=dat)
dat
### meta-analysis of mean differences using a random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
### meta-analysis of standardized mean differences using a random-effects model
res <- rma(measure="SMD", m1i=m1i, sd1i=sd1i, n1i=n1i, m2i=m2i, sd2i=sd2i, n2i=n2i,
           data=dat, slab=source)
res
### draw forest plot
forest(res, xlim=c(-7,5), alim=c(-3,1), header="Study/Source")
### calculate (log transformed) ratios of means and corresponding sampling variances
dat <- escalc(measure="ROM", m1i=m1i, sd1i=sd1i, n1i=n1i, m2i=m2i, sd2i=sd2i, n2i=n2i, data=dat)</pre>
dat
### meta-analysis of the (log transformed) ratios of means using a random-effects model
res <- rma(yi, vi, data=dat)</pre>
res
predict(res, transf=exp, digits=2)
## End(Not run)
```

dat.obrien2003 Studies on the Relationship Between BMI and Risk of Preeclampsia

## Description

Results from 13 studies on the relationship between maternal body mass index (BMI) and the risk of preeclampsia.

# Usage

dat.obrien2003

#### Format

The data frame contains the following columns:

study	numeric	study id
author	character	(first) author of the study
year	numeric	publication year
ref	numeric	reference number
ch	character	exclusion due to chronic hypertension (yes/no)
dm	character	exclusion due to diabetes mellitus (yes/no)
mg	character	exclusion due to multiple gestation (yes/no)
bmi.lb	numeric	lower bound of the BMI interval
bmi.ub	numeric	upper bound of the BMI interval
bmi	numeric	midpoint of the BMI interval
cases	numeric	number of preeclampsia cases in the BMI group
total	numeric	number of individuals in the BMI group

# Details

The dataset includes the results from 13 studies examining the relationship between maternal body mass index (BMI) and the risk of preeclampsia. For each study, results are given in terms of the number of preeclampsia cases within two or more groups defined by the lower and upper BMI bounds as shown in the dataset (NA means that the interval is either open to the left or right). The bmi variable is the interval midpoint as defined by O'Brien et al. (2003).

## Concepts

medicine, obstetrics, risk ratios, proportions, multilevel models, dose-response models

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

O'Brien, T. E., Ray, J. G., & Chan, W.-S. (2003). Maternal body mass index and the risk of preeclampsia: A systematic overview. *Epidemiology*, **14**(3), 368–374. https://doi.org/10.1097/00001648-200305000-

## Examples

### copy data into 'dat' and examine data
dat <- dat.obrien2003</pre>

```
dat
## Not run:
### load metafor package
library(metafor)
### restructure the data into a wide format
dat2 <- to.wide(dat, study="study", grp="grp", ref=1, grpvars=c("bmi","cases","total"),</pre>
                addid=FALSE, adddesign=FALSE, postfix=c(1,2))
dat2[1:10, -c(2:3)]
### calculate log risk ratios and corresponding sampling variances
dat2 <- escalc(measure="RR", ai=cases1, n1i=total1, ci=cases2, n2i=total2, data=dat2)</pre>
dat2[1:10, -c(2:7)]
### forest plot of the risk ratios
dd <- c(0,diff(dat2$study))</pre>
dd[dd > 0] <- 1
rows <- (1:nrow(dat2)) + cumsum(dd)</pre>
rows <- 1 + max(rows) - rows</pre>
slabs <- mapply(function(x,y,z) as.expression(bquote(.(x)^.(y)~.(z))),</pre>
                dat2$author, dat2$ref, dat2$year)
with(dat2, forest(yi, vi, slab=slabs, xlim=c(-7,5.5), cex=0.8,
  psize=1, pch=19, efac=0, rows=rows, ylim=c(0,max(rows)+3), yaxs="i",
 atransf=exp, at=log(c(0.05,0.1,0.2,0.5,1,2,5,10,20)), ilab=comp, ilab.xpos=-4, ilab.pos=4))
text(-4.4, max(rows)+2, "Comparison", font=2, cex=0.8, pos=4)
### within-study mean center the BMI variable
dat$bmicent <- with(dat, bmi - ave(bmi, study))</pre>
### compute the proportion of preeclampsia cases and corresponding sampling variances
dat <- escalc(measure="PR", xi=cases, ni=total, data=dat)</pre>
### convert the proportions to percentages (and convert the variances accordingly)
dat$yi <- dat$yi*100</pre>
dat$vi <- dat$vi*100^2
dat[1:10, -c(2:3)]
### fit multilevel meta-regression model to examine the relationship between the
### (centered) BMI variable and the risk of preeclampsia
res <- rma.mv(yi, vi, mods = ~ bmicent, random = ~ 1 | study/grp, data=dat)
res
### draw scatterplot with regression line
res$slab <- dat$ref
regplot(res, xlab=expression("Within-Study Mean Centered BMI"~(kg/m^2)),
        ylab="Preeclampsia Prevalence (%)", las=1, bty="l",
        at=seq(0,18,by=2), olim=c(0,100), psize=2, bg="gray90",
        label=TRUE, offset=0, labsize=0.6)
### fit model using a random slope for bmicent
res <- rma.mv(yi, vi, mods = ~ bmicent, random = ~ bmicent | study, struct="GEN", data=dat)
res
```

```
### load rms package
library(rms)
### fit restricted cubic spline model
res <- rma.mv(yi, vi, mods = ~ rcs(bmicent, 4), random = ~ 1 | study/grp, data=dat)</pre>
res
### get knot positions
knots <- attr(rcs(model.matrix(res)[,2], 4), "parms")</pre>
### computed predicted values based on the model
xs <- seq(-10, 10, length=1000)</pre>
sav <- predict(res, newmods=rcspline.eval(xs, knots, inclx=TRUE))</pre>
### draw scatterplot with regression line based on the model
tmp <- regplot(res, mod=2, pred=sav,</pre>
               xvals=xs, xlab=expression("Within-Study Mean Centered BMI"~(kg/m^2)),
               ylab="Preeclampsia Prevalence (%)", las=1, bty="l",
               at=seq(0,18,by=2), olim=c(0,100), psize=2, bg="gray90",
               label=TRUE, offset=0, labsize=0.6)
abline(v=knots, lty="dotted")
points(tmp)
## End(Not run)
```

dat.pagliaro1992 Studies on the Effectiveness of Nonsurgical Treatments in Cirrhosis

# Description

Results from 26 trials examining the effectiveness of beta-blockers and sclerotherapy for the prevention of first bleeding in patients with cirrhosis

## Usage

dat.pagliaro1992

#### Format

The data frame contains the following columns:

study	numeric	study id
trt	character	either beta-blockers, sclerotherapy, or control
xi	numeric	number of patients with first bleeding
ni	numeric	number of patients treated

## Details

The dataset includes the results from 26 randomized controlled trials examining the effectiveness of nonsurgical treatments for the prevention of first bleeding in patients with cirrhosis. Patients were either treated with beta-blockers, endoscopic sclerotherapy, or with a nonactive treatment (control). Two trials included all three treatment conditions, 7 trials compared beta-blockers against control, and 17 trials compared sclerotherapy against control. The dataset has been used in various papers to illustrate methods for conducting a network meta-analysis / mixed treatment comparison.

### Concepts

medicine, odds ratios, Mantel-Haenszel method, network meta-analysis

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Pagliaro, L., D'Amico, G., Sörensen, T. I. A., Lebrec, D., Burroughs, A. K., Morabito, A., Tiné, F., Politi, F., & Traina, M. (1992). Prevention of first bleeding in cirrhosis: A meta-analysis of randomized trials of nonsurgical treatment. *Annals of Internal Medicine*, **117**(1), 59–70. https://doi.org/10.7326/0003-4819-11

```
### copy data into 'dat' and examine data
dat <- dat.pagliaro1992</pre>
dat
## Not run:
### load metafor package
library(metafor)
### restructure dataset to a contrast-based format
dat.c <- to.wide(dat, study="study", grp="trt", grpvars=3:4)</pre>
dat.c
### Mantel-Haenszel results for beta-blockers and sclerotherapy versus control, respectively
rma.mh(measure="OR", ai=xi.1, n1i=ni.1, ci=xi.2, n2i=ni.2,
       data=dat.c, subset=(trt.1=="beta-blockers"), digits=2)
rma.mh(measure="OR", ai=xi.1, n1i=ni.1, ci=xi.2, n2i=ni.2,
       data=dat.c, subset=(trt.1=="sclerotherapy"), digits=2)
### calculate log odds for each study arm
dat <- escalc(measure="PL0", xi=xi, ni=ni, data=dat)</pre>
dat
### turn treatment variable into factor and set reference level
dat$trt <- relevel(factor(dat$trt), ref="control")</pre>
### add a space before each level (this makes the output a bit more legible)
levels(dat$trt) <- paste0(" ", levels(dat$trt))</pre>
```

```
### network meta-analysis using an arm-based random-effects model with fixed study effects
### (by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons)
res <- rma.mv(yi, vi, mods = ~ 0 + factor(study) + trt, random = ~ trt | study, rho=1/2, data=dat)
res</pre>
```

### average odds ratio comparing beta-blockers and sclerotherapy versus control, respectively
predict(res, newmods=c(rep(0,26), 1, 0), transf=exp, digits=2)
predict(res, newmods=c(rep(0,26), 0, 1), transf=exp, digits=2)

### average odds ratio comparing beta-blockers versus sclerotherapy
predict(res, newmods=c(rep(0,26), 1, -1), transf=exp, digits=2)

## End(Not run)

dat.pignon2000 Studies on the Effectiveness of Locoregional Treatment plus Chemotherapy for Head and Neck Squamous-Cell Carcinoma

## Description

Results from studies examining mortality risk in patients with nonmetastatic head and neck squamouscell carcinoma receiving either locoregional treatment plus chemotherapy versus locoregional treatment alone.

#### Usage

dat.pignon2000

#### Format

The data frame contains the following columns:

id	numeric	study id number
trial	character	trial abbreviation
OmE	numeric	observed minus expected number of deaths in the locoregional treatment plus chemotherapy group
V	numeric	corresponding variance
grp	numeric	timing of chemotherapy: 1 = adjuvant, 2 = neoadjuvant, 3 = concomitant

### Details

The purpose of this meta-analysis was to examine the mortality risk in patients with nonmetastatic head and neck squamous-cell carcinoma receiving either locoregional treatment plus chemotherapy versus locoregional treatment alone. For 65 trials, the dataset provides the observed minus expected number of deaths and corresponding variances in the locoregional treatment plus chemotherapy group. Based on these values, we can estimate the log hazard ratios with OmE/V and the corresponding sampling variance with 1/V.

The trials were also divided according to the timing of the chomotherapy: (1) adjuvant, after the

## dat.pignon2000

locoregional treatment, (2) neoadjuvant, before the locoregional treatment, and (3) concomitant, chemotherapy given concomitantly or alternating with radiotherapy.

#### Concepts

medicine, oncology, hazard ratios

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Pignon, J. P., Bourhis, J., Domenge, C., & Designe, L. (2000). Chemotherapy added to locoregional treatment for head and neck squamous-cell carcinoma: Three meta-analyses of updated individual data. *Lancet*, **355**(9208), 949–955. https://doi.org/10.1016/S0140-6736(00)90011-4

```
### copy data into 'dat' and examine data
dat <- dat.pignon2000</pre>
head(dat)
## Not run:
### load metafor package
library(metafor)
### calculate log hazard ratios and sampling variances
dat$yi <- with(dat, OmE/V)</pre>
dat$vi <- with(dat, 1/V)</pre>
head(dat)
### meta-analysis based on all 65 trials
res <- rma(yi, vi, data=dat, method="EE", digits=2)</pre>
res
predict(res, transf=exp)
### only adjuvant trials
res <- rma(yi, vi, data=dat, method="EE", subset=grp==1, digits=2)</pre>
res
predict(res, transf=exp)
### only neoadjuvant trials
res <- rma(yi, vi, data=dat, method="EE", subset=grp==2, digits=2)</pre>
res
predict(res, transf=exp)
### only concomitant trials
res <- rma(yi, vi, data=dat, method="EE", subset=grp==3, digits=2)</pre>
res
predict(res, transf=exp)
```

```
## End(Not run)
```

dat.pritz1997 Studies on the Effectiveness of Hyperdynamic Therapy for Treating Cerebral Vasospasm

#### Description

Results from 14 studies on the effectiveness of hyperdynamic therapy for treating cerebral vasospasm.

## Usage

dat.pritz1997

## Format

The data frame contains the following columns:

study	numeric	study number
authors	character	study authors
xi	numeric	number of patients that improved with hyperdynamic therapy
ni	numeric	total number of patients treated

### Details

As described in Zhou et al. (1999), "hyperdynamic therapy refers to induced hypertension and hypervolaemia (volume expansion) to treat ischaemic symptoms due to vasospasm, and the success of this therapy is defined as clinical improvement in terms of neurologic deficits." For each study that was included in the meta-analysis, the dataset includes information on the number of patients that improved under this form of therapy and the total number of patients that were treated. The goal of the meta-analysis is to estimate the true (average) success rate of hyperdynamic therapy.

## Concepts

medicine, single-arm studies, proportions

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Zhou, X.-H., Brizendine, E. J., & Pritz, M. B. (1999). Methods for combining rates from several studies. *Statistics in Medicine*, **18**(5), 557–566. https://doi.org/10.1002/(SICI)1097-0258(19990315)18:5<557::AI

#### dat.pritz1997

## References

Pritz M. B., Zhou, X.-H., & Brizendine, E. J. (1996). Hyperdynamic therapy for cerebral vasospasm: A meta-analysis of 14 studies. *Journal of Neurovascular Disease*, **1**, 6–8.

Pritz, M. B. (1997). Treatment of cerebral vasospasm due to aneurysmal subarachnoid hemorrhage: Past, present, and future of hyperdynamic therapy. *Neurosurgery Quarterly*, **7**(4), 273–285.

```
### copy data into 'dat' and examine data
dat <- dat.pritz1997</pre>
dat
## Not run:
### load metafor package
library(metafor)
### computation of "weighted average" in Zhou et al. (1999), Table IV
dat <- escalc(measure="PR", xi=xi, ni=ni, data=dat, add=0)</pre>
theta.hat <- sum(dat$ni * dat$yi) / sum(dat$ni)</pre>
se.theta.hat <- sqrt(sum(dat$ni^2 * dat$vi) / sum(dat$ni)^2)</pre>
             <- theta.hat - 1.96 * se.theta.hat
ci.lb
             <- theta.hat + 1.96 * se.theta.hat
ci.ub
round(c(estimate = theta.hat, se = se.theta.hat, ci.lb = ci.lb, ci.ub = ci.ub), 4)
### this is identical to an equal-effects model with sample size weights
rma(yi, vi, weights=ni, method="EE", data=dat)
### compute sampling variances under the assumption of homogeneity
dat <- escalc(measure="PR", xi=xi, ni=ni, data=dat, add=0, vtype="AV")</pre>
dat
### fit equal-effects model (same estimate, but SE is slightly different)
rma(yi, vi, data=dat, method="EE")
### under the assumption of homogeneity, the sum of independent binomial
### counts also follows a binomial distribution; this approach yields the same
### estimate and SE as above
agg <- escalc(measure="PR", xi=sum(dat$xi), ni=sum(dat$ni))</pre>
summary(agg)
### could also compute an 'exact' CI based on the Clopper-Pearson method
binom.test(sum(dat$xi), sum(dat$ni))
### logistic regression model
res <- rma.glmm(measure="PLO", xi=xi, ni=ni, data=dat, method="EE")</pre>
res
predict(res, transf=transf.ilogit)
### the results above suggest that the true proportions may be heterogeneous
### random-effects model with raw proportions
```

```
dat <- escalc(measure="PR", xi=xi, ni=ni, data=dat)
res <- rma(yi, vi, data=dat)
predict(res)
### random-effects model with logit transformed proportions
dat <- escalc(measure="PLO", xi=xi, ni=ni, data=dat)
res <- rma(yi, vi, data=dat)
predict(res, transf=transf.ilogit)
### mixed-effects logistic regression model
res <- rma.glmm(measure="PLO", xi=xi, ni=ni, data=dat)
predict(res, transf=transf.ilogit)
### End(Not run)</pre>
```

dat.raudenbush1985 Studies on Assessing the Effects of Teacher Expectations on Pupil IQ

## Description

Results from 19 studies examining how teachers' expectations about their pupils can influence actual IQ levels.

#### Usage

dat.raudenbush1985

## Format

The data frame contains the following columns:

study	numeric	study number
author	character	study author(s)
year	numeric	publication year
weeks	numeric	weeks of contact prior to expectancy induction
setting	character	whether tests were group or individually administered
tester	character	whether test administrator was aware or blind
n1i	numeric	sample size of experimental group
n2i	numeric	sample size of control group
yi	numeric	standardized mean difference
vi	numeric	corresponding sampling variance

# Details

In the so-called 'Pygmalion study' (Rosenthal & Jacobson, 1968), "all of the predominantly poor children in the so-called Oak elementary school were administered a test pretentiously labeled the 'Harvard Test of Inflected Acquisition.' After explaining that this newly designed instrument had identified those children most likely to show dramatic intellectual growth during the coming year,

#### dat.raudenbush1985

the experimenters gave the names of these 'bloomers' to the teachers. In truth, the test was a traditional IQ test and the 'bloomers' were a randomly selected 20% of the student population. After retesting the children 8 months later, the experimenters reported that those predicted to bloom had in fact gained significantly more in total IQ (nearly 4 points) and reasoning IQ (7 points) than the control group children. Further, at the end of the study, the teachers rated the experimental children as intellectually more curious, happier, better adjusted, and less in need of approval than their control group peers'' (Raudenbush, 1984).

In the following years, a series of studies were conducted attempting to replicate this rather controversial finding. However, the great majority of those studies were unable to demonstrate a statistically significant difference between the two experimental groups in terms of IQ scores. Raudenbush (1984) conducted a meta-analysis based on 19 such studies to further examine the evidence for the existence of the 'Pygmalion effect'. The dataset includes the results from these studies.

The outcome measure used for the meta-analysis was the standardized mean difference (yi), with positive values indicating that the supposed 'bloomers' had, on average, higher IQ scores than those in the control group. The weeks variable indicates the number of weeks of prior contact between teachers and students before the expectancy induction. Testing was done either in a group setting or individually, which is indicated by the setting variable. Finally, the tester variable indicates whether the test administrators were either aware or blind to the researcher-provided designations of the children's intellectual potential.

The data in this dataset were obtained from Raudenbush and Bryk (1985) with information on the setting and tester variables extracted from Raudenbush (1984).

#### Concepts

education, standardized mean differences, meta-regression

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Raudenbush, S. W. (1984). Magnitude of teacher expectancy effects on pupil IQ as a function of the credibility of expectancy induction: A synthesis of findings from 18 experiments. *Journal of Educational Psychology*, **76**(1), 85–97. https://doi.org/10.1037/0022-0663.76.1.85

Raudenbush, S. W., & Bryk, A. S. (1985). Empirical Bayes meta-analysis. *Journal of Educational Statistics*, **10**(2), 75–98. https://doi.org/10.3102/10769986010002075

```
### copy data into 'dat' and examine data
dat <- dat.raudenbush1985
dat
## Not run:
### load metafor package
library(metafor)
### random-effects model</pre>
```

```
res <- rma(yi, vi, data = dat)
res
### create weeks variable where values larger than 3 are set to 3
dat$weeks.c <- ifelse(dat$weeks > 3, 3, dat$weeks)
### mixed-effects model with weeks.c variable as moderator
res <- rma(yi, vi, mods = ~ weeks.c, data = dat, digits = 3)
res
## End(Not run)</pre>
```

dat.riley2003 Studies on MYC-N as a Prognostic Marker for Neuroblastoma

#### Description

Results from 81 studies examining overall and disease-free survival in neuroblastoma patients with amplified versus normal MYC-N protein levels.

#### Usage

dat.riley2003

# Format

The data frame contains the following columns:

study	numeric	study number
yi	numeric	log hazard ratio of the outcome in those with amplified versus normal MYC-N protein levels
vi	numeric	sampling variance of the log hazard ratio
sei	numeric	standard error of the log hazard ratio
outcome	character	outcome (OS = overall survival; DFS = disease-free survival)

## Details

The meta-analysis by Riley et al. (2003) examined a variety of prognostic markers for overall and disease-free survival in patients with neuroblastoma. One of the markers examined was amplified levels of the MYC-N protein, with is associated with poorer outcomes.

The dataset given here was extracted from Riley (2011) and has been used in several other publications (e.g., Riley et al., 2004, 2007). The dataset provides the (log) hazard ratios (and corresponding standard errors) with respect to these two outcomes in 81 studies, with positive values indicating a greater risk of death (for OS) or disease recurrence/death (for DFS) for patients with high MYC-N levels compared to those with normal/low levels. Note that information on both outcomes could only be extracted from 17 studies (39 studies only provided sufficient information to extract the OS estimate, while 25 studies only allowed for extraction of the DFS estimate).

#### dat.riley2003

#### Concepts

medicine, oncology, hazard ratios

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

#### Source

Riley, R. D., Sutton, A. J., Abrams, K. R., & Lambert, P. C. (2004). Sensitivity analyses allowed more appropriate and reliable meta-analysis conclusions for multiple outcomes when missing data was present. *Journal of Clinical Epidemiology*, **57**(9), 911–924. https://doi.org/10.1016/j.jclinepi.2004.01.018

Riley, R. D., Abrams, K. R., Lambert, P. C., Sutton, A. J., & Thompson, J. R. (2007). An evaluation of bivariate random-effects meta-analysis for the joint synthesis of two correlated outcomes. *Statistics in Medicine*, **26**(1), 78–97. https://doi.org/10.1002/sim.2524

Riley, R. D. (2011). Erratum: An evaluation of bivariate random-effects meta-analysis for the joint synthesis of two correlated outcomes. *Statistics in Medicine*, **30**(4), 400. https://doi.org/10.1002/sim.4100

# References

Riley, R. D., Burchill, S. A., Abrams, K. R., Heney, D., Lambert, P. C., Jones, D. R., Sutton, A. J., Young, B., Wailoo, A. J., & Lewis, I. J. (2003). A systematic review and evaluation of the use of tumour markers in paediatric oncology: Ewing's sarcoma and neuroblastoma. *Health Technology Assessment*, **7**(5), 1–162. https://doi.org/10.3310/hta7050

```
### copy data into 'dat' and examine data
dat <- dat.riley2003
dat
## Not run:
### load metafor package
library(metafor)
### random-effects model analysis for outcome DFS
res <- rma(yi, sei=sei, data=dat, subset=(outcome == "DFS"), method="DL")
res
predict(res, transf=exp, digits=2)
### random-effects model analysis for outcome OS
res <- rma(yi, sei=sei, data=dat, subset=(outcome == "OS"), method="DL")
res
predict(res, transf=exp, digits=2)
### random-effects model analysis for outcome OS
res <- rma(yi, sei=sei, data=dat, subset=(outcome == "OS"), method="DL")
res
predict(res, transf=exp, digits=2)
## End(Not run)</pre>
```

dat.roever2022

## Description

12 studies investigating the occurrence of dose limiting toxicities (DLTs) at different doses of a combination therapy of Irinotecan and S-1.

#### Usage

dat.roever2022

## Format

The data frame contains the following columns:

study	character	study label
year	integer	publication year
dose	numeric	dose $(mg/m^2)$
events	integer	number of DLTs
total	integer	number of patients exposed

#### Details

A combination therapy of Irinotecan (a topoisomerase 1 inhibitor) and S-1 (a combination of three pharmacological compounds, namely, tegafur, gimeracil, and oteracil potassium) was tested in advanced colorectal and gastric cancer. This dataset contains data from twelve studies investigating this therapy in a Japanese population; it contains the doses investigated, the numbers of patients treated, and the number of dose-limiting toxicities (DLTs) observed. In general, each study investigated several doses according to some dose-escalation scheme.

## Concepts

medicine, oncology, dose-response models

#### Author(s)

Christian Roever, <christian.roever@med.uni-goettingen.de>

#### Source

Ursino, M., Roever, C., Zohar, S., & Friede T. (2021). Random-effects meta-analysis of phase I dose-finding studies using stochastic process priors. *The Annals of Applied Statistics*, **15**(1), 174–193. https://doi.org/10.1214/20-A0AS1390

Roever, C., Ursino, M., Friede, T., & Zohar, S. (2022). A straighforward meta-analysis approach for oncology phase I dose-finding studies. *Statistics in Medicine*, **41**(20), 3915–3940. https://doi.org/10.1002/sim.9484

#### dat.roever2022

#### References

European Medicines Agency (EMA) (2021). Onivyde pegylated liposomal (irinotecan hydrochloride trihydrate) EPAR summary. https://www.ema.europa.eu/en/medicines/human/EPAR/onivyde-pegylated-lipos

European Medicines Agency (EMA) (2022). Teysuno (tegafur/gimeracil/oteracil) EPAR summary. https://www.ema.europa.eu/en/medicines/human/EPAR/teysuno

Yamada, Y., Yasui, H., Goto, A., *et al.* (2003). Phase I study of irinotecan and S-1 combination therapy in patients with metastatic gastric cancer. *International Journal of Clinical Oncology*, **8**(6), 374–380. https://doi.org/10.1007/s10147-003-0359-z

Takiuchi, H., Narahara, H., Tsujinaka, T., *et al.* (2005). Phase I study of S-1 combined with irinotecan (CPT-11) in patients with advanced gastric cancer (OGSG 0002). *Japanese Journal of Clinical Oncology*, **35**(9), 520–525. https://doi.org/10.1093/jjco/hyi148

Inokuchi, M., Yamashita, T., Yamada, H., *et al.* (2006). Phase I/II study of S-1 combined with irinotecan for metastatic advanced gastric cancer. *British Journal of Cancer*, **94**(8), 11130. https://doi.org/10.1038/sj.l

Nakafusa, Y., Tanaka, M., Ohtsuka, T., *et al.* (2008). Phase I/II study of combination therapy with S-1 and CPT-11 for metastatic colorectal cancer. *Molecular Medicine Reports*, **1**(6), 925–930. https://doi.org/10.3892/mmr\_00000051

Ishimoto, O., Ishida, T., Honda, Y., Munakata, M., & Sugawara, S. (2009). Phase I study of daily S-1 combined with weekly irinotecan in patients with advanced non-small cell lung cancer. *International Journal of Clinical Oncology*, **14**(1), 43–47. https://doi.org/10.1007/s10147-008-0796-9

Ogata, Y., Sasatomi, T., Akagi, Y., Ishibashi, N., Mori, S., & Shirouzu, K. (2009). Dosage escalation study of S-1 and irinotecan in metronomic chemotherapy against advanced colorectal cancer. *The Kurume Medical Journal*, **56**(1+2), 1–7. https://doi.org/10.2739/kurumemedj.56.1

Shiozawa, M., Sugano, N., Tsuchida, K., Morinaga, S., Akaike, M., & Sugimasa, Y. (2009). A phase I study of combination therapy with S-1 and irinotecan (CPT-11) in patients with advanced colorectal cancer. *Journal of Cancer Research and Clinical Oncology*, **135**(3), 365–370. https://doi.org/10.1007/s00432-008-0480-5

Yoshioka, T., Kato, S., Gamoh, M., *et al.* (2009). Phase I/II study of sequential therapy with irinotecan and S-1 for metastatic colorectal cancer. *British Journal of Cancer*, **101**, 1972–1977. https://doi.org/10.1038/sj.bjc.6605432

Komatsu, Y., Yuki, S., Fuse, N., *et al.* (2010). Phase 1/2 clinical study of irinotecan and oral S-1 (IRIS) in patients with advanced gastric cancer. *Advances in Therapy*, **27**(7), 483–492. https://doi.org/10.1007/s12325-

Kusaba, H., Esaki, T., Futami, K., *et al.* (2010). Phase I/II study of a 3-week cycle of irinotecan and S-1 in patients with advanced colorectal cancer. *Cancer Science*, **101**(12), 2591–2595. https://doi.org/10.1111/j.1349-7006.2010.01728.x

Yoda, S., Soejima, K., Yasuda, H., *et al.* (2011). A phase I study of S-1 and irinotecan combination therapy in previously treated advanced non-small cell lung cancer patients. *Cancer Chemotherapy and Pharmacology*, **67**(3), 717–722. https://doi.org/10.1007/s00280-010-1539-y

Goya, H., Kuraishi, H., Koyama, S., *et al.* (2012). Phase I/II study of S-1 combined with biweekly irinotecan chemotherapy in previously treated advanced non-small cell lung cancer. *Cancer Chemotherapy and Pharmacology*, **70**(5), 691–697. https://doi.org/10.1007/s00280-012-1957-0

## See Also

dat.ursino2021

# Examples

```
# show (some) data
head(dat.roever2022, n=10)
## Not run:
# illustrate data
plot(NA, xlim=range(dat.roever2022$dose), ylim=0:1,
     xlab="dose (mg / m<sup>2</sup>)", ylab="proportion",
     main="dat.roever2022 (Irinotecan / S-1 data)")
studylab <- unique(dat.roever2022$study)</pre>
colvec <- rainbow(length(studylab))</pre>
for (i in 1:length(studylab)) {
  idx <- (dat.roever2022$study == studylab[i])</pre>
  lines(dat.roever2022[idx,"dose"],
        dat.roever2022[idx,"events"] / dat.roever2022[idx,"total"],
        col=colvec[i], type="b")
}
legend("topleft", studylab, col=colvec, pch=15)
## End(Not run)
```

dat.senn2013 Studies on the Effectiveness of Glucose-Lowering Agents

## Description

Results from 26 trials examining the effectiveness of glucose-lowering agents in patients with type 2 diabetes

## Usage

dat.senn2013

# Format

The data frame contains the following columns:

study	character	(first) author and year of study
ni	numeric	sample size of the study arm
treatment	character	treatment given
comment	character	whether figures given are based on raw values at outcome or on change from baseline
mi	numeric	raw mean or mean change
sdi	numeric	standard deviation

# Details

The dataset includes the results from 26 randomized controlled trials examining the effectiveness of adding various oral glucose-lowering agents to a baseline sulfonylurea therapy in patients with type

#### dat.senn2013

2 diabetes. The outcome measured in the studies was either the mean HbA1c level at follow-up or the mean change in HbA1c level from baseline to follow-up. A total of 10 different treatment types were examined in these studies: acarbose, benfluorex, metformin, miglitol, pioglitazone, placebo, rosiglitazone, sitagliptin, sulfonylurea alone, and vildagliptin. One study included three treatment arms (Willms, 1999), while the rest of the studies included two treatment arms (hence, the dataset includes the results from 53 treatment arms).

The data can be used for a network meta-analysis, either using an arm-based or a contrast-based model. See 'Examples' below.

## Concepts

medicine, raw mean differences, network meta-analysis

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Senn, S., Gavini, F., Magrez, D., & Scheen, A. (2013). Issues in performing a network metaanalysis. *Statistical Methods in Medical Research*, **22**(2), 169–189. https://doi.org/10.1177/0962280211432220

## References

Law, M., Jackson, D., Turner, R., Rhodes, K., & Viechtbauer, W. (2016). Two new methods to fit models for network meta-analysis with random inconsistency effects. *BMC Medical Research Methodology*, **16**, 87. https://doi.org/10.1186/s12874-016-0184-5

Rücker, G., & Schwarzer, G. (2015). Ranking treatments in frequentist network meta-analysis works without resampling methods. *BMC Medical Research Methodology*, **15**, 58. https://doi.org/10.1186/s12874-015

```
### copy data into 'dat' and examine data
dat <- dat.senn2013
dat
## Not run:
### load metafor package
library(metafor)
### create network graph ('igraph' package must be installed)
library(igraph, warn.conflicts=FALSE)
pairs <- data.frame(do.call(rbind,</pre>
  sapply(split(dat$treatment, dat$study), function(x) t(combn(x,2)))), stringsAsFactors=FALSE)
pairs$X1 <- factor(pairs$X1, levels=sort(unique(dat$treatment)))</pre>
pairs$X2 <- factor(pairs$X2, levels=sort(unique(dat$treatment)))</pre>
tab <- table(pairs[,1], pairs[,2])</pre>
tab # adjacency matrix
g <- graph_from_adjacency_matrix(tab, mode = "plus", weighted=TRUE, diag=FALSE)
plot(g, edge.curved=FALSE, edge.width=E(g)$weight, layout=layout_as_star(g, center="placebo"),
```

```
vertex.size=45, vertex.color="lightgray", vertex.label.color="black", vertex.label.font=2)
### table of studies versus treatments examined
print(addmargins(table(dat$study, dat$treatment)), zero.print="")
### table of frequencies with which treatment pairs were studied
print(as.table(crossprod(table(dat$study, dat$treatment))), zero.print="")
### add means and sampling variances of the means to the dataset
dat <- escalc(measure="MN", mi=mi, sdi=sdi, ni=ni, data=dat)</pre>
### turn treatment variable into factor and set reference level
dat$treatment <- relevel(factor(dat$treatment), ref="placebo")</pre>
### add a space before each level (this makes the output a bit more legible)
levels(dat$treatment) <- paste0(" ", levels(dat$treatment))</pre>
### network meta-analysis using an arm-based fixed-effects model with fixed study effects
res.fe <- rma.mv(yi, vi, mods = ~ 0 + study + treatment, data=dat, slab=paste0(study, treatment))</pre>
res.fe
### test if treatment factor as a whole is significant
anova(res.fe, btt="treatment")
### forest plot of the contrast estimates (treatments versus placebos)
forest(tail(coef(res.fe), 9), tail(diag(vcov(res.fe)), 9), slab=levels(dat$treatment)[-1],
      xlim=c(-2.5, 1.5), alim=c(-1.5, 0.5), psize=1, xlab="Estimate", header="Treatment")
### weight matrix for the estimation of the fixed effects (leaving out the study effects)
w <- t(tail(vcov(res.fe) %*% t(model.matrix(res.fe)) %*% weights(res.fe, type="matrix"), 9))</pre>
rownames(w) <- res.fe$slab</pre>
### create shade plot for the diabetes network with placebo as the reference treatment
### negative values in blue shades, positive values in red shades
cols <- colorRampPalette(c("blue", "gray95", "red"))(9)</pre>
heatmap(w, Rowv=NA, Colv=NA, scale="none", margins=c(6,11), col=cols,
        cexRow=.7, cexCol=1, labCol=levels(dat$treatment)[-1])
### network meta-analysis using an arm-based random-effects model with fixed study effects
### by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons
res.re <- rma.mv(yi, vi, mods = ~ 0 + study + treatment, random = ~ treatment | study, rho=1/2,
                 data=dat, slab=paste0(study, treatment))
res re
### test if treatment factor as a whole is significant
anova(res.re, btt="treatment")
### forest plot of the contrast estimates (treatments versus placebos)
forest(tail(coef(res.re), 9), tail(diag(vcov(res.re)), 9), slab=levels(dat$treatment)[-1],
     xlim=c(-2.5, 1.5), alim=c(-1.5, 0.5), psize=1, xlab="Estimate", header="Treatment")
```

```
### compute the contribution of each study to the overall Q-test value
qi <- sort(by((resid(res.fe) / sqrt(dat$vi))^2, dat$study, sum))</pre>
```

```
### check that the values add up
sum(ai)
res.fe$QE
### plot the values
s <- length(qi)</pre>
par(mar=c(5,10,2,1))
plot(qi, 1:s, pch=19, xaxt="n", yaxt="n", xlim=c(0,40), xlab="Chi-Square Contribution", ylab="")
axis(side=1)
axis(side=2, at=1:s, labels=names(qi), las=1, tcl=0)
segments(rep(0,s), 1:s, qi, 1:s)
### restructure dataset to a contrast-based format
dat <- dat.senn2013[c(1,4:2,5:6)] # reorder variables first</pre>
dat <- to.wide(dat, study="study", grp="treatment", ref="placebo", grpvars=4:6)</pre>
dat
### calculate mean difference and corresponding sampling variance for each treatment comparison
dat <- escalc(measure="MD", m1i=mi.1, sd1i=sdi.1, n1i=ni.1,</pre>
                            m2i=mi.2, sd2i=sdi.2, n2i=ni.2, data=dat)
dat
### calculate the variance-covariance matrix of the mean differences for the multitreatment studies
calc.v <- function(x) {</pre>
   v <- matrix(x$sdi.2[1]^2 / x$ni.2[1], nrow=nrow(x), ncol=nrow(x))</pre>
   diag(v) <- x$vi</pre>
   v
}
V <- bldiag(lapply(split(dat, dat$study), calc.v))</pre>
### add contrast matrix to dataset
dat <- contrmat(dat, grp1="treatment.1", grp2="treatment.2")</pre>
dat
### network meta-analysis using a contrast-based random-effects model
### by setting rho=1/2, tau^2 reflects the amount of heterogeneity for all treatment comparisons
### the treatment left out (placebo) becomes the reference level for the treatment comparisons
res <- rma.mv(yi, V, mods = ~ 0 + acarbose + benfluorex + metformin + miglitol + pioglitazone +
                              rosiglitazone + sitagliptin + sulfonylurea + vildagliptin,
              random = ~ comp | study, rho=1/2, data=dat)
res
### forest plot of the contrast estimates (treatments versus placebos)
forest(coef(res), diag(vcov(res)), slab=names(coef(res)), order="obs",
      xlim=c(-3.0, 2.5), alim=c(-1.5, 0.5), psize=1, xlab="Estimate", header="Treatment")
### estimate all pairwise differences between treatments
contr <- data.frame(t(combn(names(coef(res)), 2)))</pre>
contr <- contrmat(contr, "X1", "X2", last="vildagliptin")</pre>
rownames(contr) <- paste(contr$X1, "-", contr$X2)</pre>
```

```
contr <- as.matrix(contr[-c(1:2)])</pre>
sav <- predict(res, newmods=contr)</pre>
sav[["slab"]] <- rownames(contr)</pre>
sav
### fit random inconsistency effects model (see Law et al., 2016)
inc <- rma.mv(yi, V, mods = ~ 0 + acarbose + benfluorex + metformin + miglitol + pioglitazone +
                            rosiglitazone + sitagliptin + sulfonylurea + vildagliptin,
            random = list(~ comp | study, ~ comp | design), rho=1/2, phi=1/2, data=dat)
inc
### compute P-scores (see Rücker & Schwarzer, 2015)
contr <- data.frame(t(combn(c(names(coef(res)), "placebo"), 2))) # add 'placebo' to contrast matrix</pre>
contr <- contrmat(contr, "X1", "X2", last="placebo", append=FALSE)</pre>
b <- c(coef(res),0) # add 0 for 'placebo' (the reference treatment)</pre>
vb <- bldiag(vcov(res),0) # add 0 row/column for 'placebo' (the reference treatment)
pvals <- apply(contr, 1, function(x) pnorm((x%*%b) / sqrt(t(x)%*%vb%*%x)))</pre>
tab <- vec2mat(pvals, corr=FALSE)</pre>
tab[upper.tri(tab)] <- t((1 - tab)[upper.tri(tab)])</pre>
rownames(tab) <- colnames(tab) <- colnames(contr)</pre>
round(tab, 2) # like Table 2 in the article
cbind(pscore=round(sort(apply(tab, 1, mean, na.rm=TRUE), decreasing=TRUE), 3))
# note: the values are slightly different from the ones given in Table 3 of Rücker and
# Schwarzer (2015) since model 'res' above is fitted using REML estimation while the
# results shown in the article are based on the 'netmeta' package, which uses a DL-type
# estimator for the amount of heterogeneity by default
## End(Not run)
```

dat.spooner2002

Studies on Nedocromil Sodium for Preventing Exercise-Induced Bronchoconstriction

# Description

Results from 17 trials, 11 studies in children and 6 studies in adults, reporting the maximum fall in the forced expiratory volume in 1 second (FEV\_1) over the course of follow-up, expressed as a percentage.

#### Usage

dat.spooner2002

# dat.spooner2002

# Format

The data frame contains the following columns:

author	character	first author
year	character	year of publication
Ne	integer	number of participants in nedocromil sodium group
Me	numeric	maximum fall in the FEV_1 (nedocromil sodium)
Se	numeric	standard deviation (nedocromil sodium)
Nc	integer	number of participants in placebo group
Mc	numeric	maximum fall in the FEV_1 (placebo)
Sc	numeric	standard deviation (placebo)
agegroup	factor	age group (children or adults)

## Details

Spooner et al. (2002) conducted a Cochrane review comparing nedocromil sodium (experimental treatment) with placebo (control) for preventing exercise-induced bronchoconstriction. Primary outcome was the maximum fall in the forced expiratory volume in 1 second (FEV\_1) over the course of follow-up, expressed as a percentage. This outcome is available for 17 studies, 11 studies in children and 6 studies in adults. For each study, the mean value, standard deviation, and sample size are reported for both the experimental and control group. The authors conducted a random-effects meta-analysis with the mean difference as effect measure, i.e.\ mean value in the nedocromil sodium group minus mean value in the placebo group.

This data set is used as an example in Schwarzer et al. (2015).

#### Concepts

raw mean differences, subgroup analysis

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

#### Source

Spooner, C., Saunders, L. D., & Rowe, B. H. (2002). Nedocromil sodium for preventing exerciseinduced bronchoconstriction. *Cochrane Database of Systematic Reviews*, **1**, CD001183. https://doi.org/10.1002/14651

## References

Schwarzer, G., Carpenter, J. R., & Rücker, G. (2015). *Meta-analysis with R*. Cham, Switzerland: Springer.

## Examples

```
### Show first five studies
head(dat.spooner2002, 5)
```

## Not run: ### Load meta package suppressPackageStartupMessages(library("meta"))

## dat.stowe2010

dat.stowe2010 Studies on Adjuvant Treatments to Levodopa Therapy for Parkinson disease

# Description

Results from 29 trials assessing efficacy of three drug classes as adjuvant treatment to levodopa therapy in patients with Parkinson disease and motor complications.

#### Usage

dat.stowe2010

## Format

The data frame contains the following columns:

study	character	study label
id	integer	study id
t1	character	treatment 1
y1	numeric	treatment effect arm 1
sd1	numeric	standard deviation arm 1
n1	integer	sample size arm 1
t2	character	treatment 2
y2	numeric	treatment effect arm 2
sd2	numeric	standard deviation arm 2
n2	integer	sample size arm 2
t3	character	treatment 3
y3	numeric	treatment effect arm 3
sd3	numeric	standard deviation arm 3
n3	integer	sample size arm 3

## Details

This dataset contains data from a Cochrane review assessing efficacy and safety of three drug classes as adjuvant treatment to levodopa therapy in patients with Parkinson disease and motor complications (Stowe et al., 2010).

The authors conducted three pairwise meta-analyses comparing dopamine agonists, catechol-Omethyl transferase inhibitors (COMTI), and monoamine oxidase type B inhibitors (MAOBI) with placebo. The primary outcome was the mean reduction of the time spent in a relatively immobile 'off' phase (mean off-time), calculated in hours per day. Relative treatment effects were expressed as mean difference. Data on this outcome were available for 5,331 patients from 28 studies comparing an active treatment with placebo and one three-arm study comparing two active treatments with placebo.

## Concepts

medicine, raw mean differences, network meta-analysis

#### Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Stowe, R., Ives, N., Clarke, C. E., Deane, K., Hilten, V., Wheatley, K., Gray, R., Handley, K., & Furmston, A. (2010). Evaluation of the efficacy and safety of adjuvant treatment to levodopa therapy in Parkinson's disease patients with motor complications. *Cochrane Database of Systematic Reviews*, **7**, CD007166. https://doi.org/10.1002/14651858.CD007166.pub2

#### Examples

```
### Show results from three studies (including three-arm study LARGO)
dat.stowe2010[18:20, ]
```

```
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print mean differences with two digits and standard errors with 3
### digits
oldset <- settings.meta(digits = 2, digits.se = 3)
### Transform data from wide arm-based format to contrast-based
### format. Argument 'sm' must not be provided as the mean difference
### is the default in R function metacont() called internally.
pw <- pairwise(treat = list(t1, t2, t3), n = list(n1, n2, n3),
mean = list(y1, y2, y3), sd = list(sd1, sd2, sd3),
```

```
studlab = study, data = dat.stowe2010, sm = "MD")
```

```
### Show calculated mean differences (TE) for three studies
selstudy <- c("COMTI(E) INT-OZ", "LARGO", "COMTI(E) Nomecomt")
subset(pw, studlab %in% selstudy)[, c(3:7, 10, 1)]
```
```
### Conduct random effects network meta-analysis (NMA)
### with placebo as reference
net <- netmeta(pw, fixed = FALSE, ref = "plac")</pre>
### Show network graph
netgraph(net, number = TRUE, multiarm = TRUE,
  cex = 1.25, offset = 0.025,
  cex.number = 1, pos.number.of.studies = 0.3)
### Print NMA results
net
### Forest plot with NMA results
forest(net)
### Forest plot showing all network estimates of active treatments
### compared with other treatments
forest(net, ref = c("C", "D", "M"), baseline = FALSE, drop = TRUE)
### Treatment ranking using P-scores
netrank(net)
### Rankogram with all ranking probabilities
set.seed(1909)
ran <- rankogram(net)</pre>
ran
plot(ran)
### Treatment ranking using SUCRAs
netrank(ran)
### League table showing network and direct estimates
netleague(net, seq = netrank(net), ci = FALSE)
### Use previous settings
settings.meta(oldset)
## End(Not run)
```

dat.tannersmith2016 Studies on the Relationship between School Motivation and Criminal Behavior

# Description

Results from 17 studies on the correlation between school motivation/attitudes and subsequent delinquent/criminal behavior.

# Usage

dat.tannersmith2016

# Format

The data frame contains the following columns:

studyid	numeric	study identifier	
yi	numeric	r-to-z transformed correlation coefficient	
vi	numeric	corresponding sampling variance	
sei	numeric	corresponding standard error	
aget1	numeric	age at which the school motivation/attitudes were assessed	
aget2	numeric	age at which the delinquent/criminal behavior was assessed	
propmale	numeric	proportion of male participants in the sample	
sexmix	character	whether the sample consisted only of males, only of females, or a mix	

# Details

The dataset includes 113 r-to-z transformed correlation coefficients from 17 prospective longitudinal studies that examined the relationship between school motivation/attitudes and subsequent delinquent/criminal behavior.

Multiple coefficients could be extracted from the studies "given the numerous ways in which school motivation/attitudes variables could be operationalized (e.g., academic aspirations, academic self-efficacy) as well as the numerous ways in which crime/delinquency could be operationalized (e.g., property crime, violent crime)" (Tanner-Smith et al., 2016).

Since information to compute the covariance between multiple coefficients within studies is not available, Tanner-Smith et al. (2016) illustrate the use of cluster-robust inference methods for the analysis of this dataset.

Note that this dataset is only meant to be used for pedagogical and demonstration purposes and does not constitute a proper review or synthesis of the complete and current research evidence on the given topic.

## Concepts

psychology, criminology, correlation coefficients, multilevel models, cluster-robust inference, metaregression

# Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Tanner-Smith, E. E., Tipton, E. & Polanin, J. R. (2016). Handling complex meta-analytic data structures using robust variance estimates: A tutorial in R. *Journal of Developmental and Life-Course Criminology*, **2**(1), 85–112. https://doi.org/10.1007/s40865-016-0026-5

## dat.tannersmith2016

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.tannersmith2016</pre>
head(dat)
## Not run:
### load metafor package
library(metafor)
### compute mean age variables within studies
dat$aget1 <- ave(dat$aget1, dat$studyid)</pre>
dat$aget2 <- ave(dat$aget2, dat$studyid)</pre>
### construct an effect size identifier variable
dat$esid <- 1:nrow(dat)</pre>
### construct an approximate var-cov matrix assuming a correlation of 0.8
### for multiple coefficients arising from the same study
V <- vcalc(vi, cluster=studyid, obs=esid, rho=0.8, data=dat)</pre>
### fit a multivariate random-effects model using the approximate var-cov matrix V
res <- rma.mv(yi, V, random = ~ esid | studyid, data=dat)</pre>
res
### use cluster-robust inference methods
robust(res, cluster=studyid, clubSandwich=TRUE)
### note: the results obtained above and below are slightly different compared
### to those given by Tanner-Smith et al. (2016) since the approach illustrated
### here makes use a multivariate random-effects model for the 'working model'
### before applying the cluster-robust inference methods, while the results given
### in the paper are based on a somewhat simpler working model
### examine the main effects of the age variables
res <- rma.mv(yi, V, mods = ~ aget1 + aget2,</pre>
              random = ~ 1 | studyid/esid, data=dat)
robust(res, cluster=studyid, clubSandwich=TRUE)
### also examine their interaction
res <- rma.mv(yi, V, mods = ~ aget1 * aget2,</pre>
              random = ~ 1 | studyid/esid, data=dat)
robust(res, cluster=studyid, clubSandwich=TRUE)
### add the sexmix factor to the model
res <- rma.mv(yi, V, mods = ~ aget1 * aget2 + sexmix,</pre>
              random = ~ 1 | studyid/esid, data=dat)
robust(res, cluster=studyid, clubSandwich=TRUE)
## End(Not run)
```

dat.ursino2021

# Description

13 studies investigating the occurrence of dose limiting toxicities (DLTs) at different doses of Sorafenib.

## Usage

dat.ursino2021

## Format

The data frame contains the following columns:

study	character	study label
year	integer	publication year
dose	numeric	dose (mg)
events	integer	number of DLTs
total	integer	number of patients exposed

## Details

Sorafenib (BAY 43-9006, Nexavar) is a kinase inhibitor that is used in the treatment of advanced renal cell carcinoma, hepatocellular carcinoma, and radioactive iodine resistant advanced thyroid carcinoma. Thirteen trials with published results, described in eleven manuscripts, were identified in a literature search. This dataset contains the doses investigated, the numbers of patients treated, and the number of dose-limiting toxicities (DLTs) observed. In general, each study investigated several doses according to some dose-escalation scheme.

# Concepts

medicine, oncology, dose-response models

#### Author(s)

Christian Roever, <christian.roever@med.uni-goettingen.de>

#### Source

Ursino, M., Roever, C., Zohar, S., & Friede T. (2021). Random-effects meta-analysis of phase I dose-finding studies using stochastic process priors. *The Annals of Applied Statistics*, **15**(1), 174–193. https://doi.org/10.1214/20-A0AS1390

Roever, C., Ursino, M., Friede, T., & Zohar, S. (2022). A straighforward meta-analysis approach for oncology phase I dose-finding studies. *Statistics in Medicine*, **41**(20), 3915–3940. https://doi.org/10.1002/sim.9484

#### References

European Medicines Agency (EMA) (2022). Nexavar (sorafenib) EPAR summary. https://www.ema.europa.eu/en/medi

Awada, A., Hendlisz, A., Gil, T., *et al.* (2005). Phase I safety and pharmacokinetics of BAY 43-9006 administered for 21 days on / 7 days off in patients with advanced, refractory solid tumours. *British Journal of Cancer*, **92**(10), 1855. https://doi.org/10.1038/sj.bjc.6602584

Clark, J. W., Eder, J. P., Ryan, D., Lathia, C., & Lenz, H.-J. (2005). Safety and pharmacokinetics of the dual action Raf kinase and vascular endothelial growth factor receptor inhibitor, BAY 43-9006, in patients with advanced, refractory solid tumors. *Clinical Cancer Research*, **11**(15), 5472–5480. https://doi.org/10.1158/1078-0432.CCR-04-2658

Moore, M., Hirte, H. W., Siu, L., *et al.* (2005). Phase I study to determine the safety and pharmacokinetics of the novel Raf kinase and VEGFR inhibitor BAY 43-9006, administered for 28 days on / 7 days off in patients with advanced, refractory solid tumors. *Annals of Oncology*, **16**(10), 1688–1694. https://doi.org/10.1093/annonc/mdi310

Strumberg, D., Richly, H., Hilger, R. A., *et al.* (2005). Phase I clinical and pharmacokinetic study of the novel Raf kinase and vascular endothelial growth factor receptor inhibitor BAY 43-9006 in patients with advanced refractory solid tumors. *Journal of Clinical Oncology*, **23**(5), 965–972. https://doi.org/10.1200/JC0.2005.06.124

Furuse, J., Ishii, H., Nakachi, K., Suzuki, E., Shimizu, S., & Nakajima, K. (2008). Phase I study of sorafenib in Japanese patients with hepatocellular carcinoma. *Cancer Science*, **99**(1), 159–165. https://doi.org/10.1111/j.1349-7006.2007.00648.x

Minami, H., Kawada, K., Ebi, H., *et al.* (2008). Phase I and pharmacokinetic study of sorafenib, an oral multikinase inhibitor, in Japanese patients with advanced refractory solid tumors. *Cancer Science*, **99**(7), 1492–1498. https://doi.org/10.1111/j.1349-7006.2008.00837.x

Miller, A. A., Murry, D. J., Owzar, K., *et al.* (2009). Phase I and pharmacokinetic study of sorafenib in patients with hepatic or renal dysfunction: CALGB 60301. *Journal of Clinical Oncology*, **27**(11), 1800. https://doi.org/10.1200/JC0.2008.20.0931

Crump, M., Hedley, D., Kamel-Reid, S., *et al.* (2010). A randomized phase I clinical and biologic study of two schedules of sorafenib in patients with myelodysplastic syndrome or acute myeloid leukemia: A NCIC (National Cancer Institute of Canada) Clinical Trials Group Study. *Leukemia and Lymphoma*, **51**(2), 252–260. https://doi.org/10.3109/10428190903585286

Borthakur, G., Kantarjian, H., Ravandi, F., *et al.* (2011). Phase I study of sorafenib in patients with refractory or relapsed acute Leukemias. *Haematologica*, **96**(1), 62–68. https://doi.org/10.3324/haematol.2010.03045

Nabors, L. B., Supko, J. G., Rosenfeld, M., *et al.* (2011). Phase I trial of sorafenib in patients with recurrent or progressive malignant glioma. *Neuro-Oncology*, **13**(12), 1324–1330. https://doi.org/10.1093/neuonc/nor

Chen, Y.-B., Li, S., Lane, A. A., *et al.* (2014). Phase I trial of maintenance sorafenib after allogeneic hematopoietic stem cell transplantation for FMS-like tyrosine kinase 3 internal tandem duplication acute myeloid leukemia. *Biology of Blood and Marrow Transplantation*, **20**(12), 2042–2048. https://doi.org/10.1016/j.bbmt.2014.09.007

#### See Also

dat.roever2022

# Examples

```
# show (some) data
head(dat.ursino2021, n=15)
## Not run:
# illustrate data
plot(NA, xlim=range(dat.ursino2021$dose), ylim=0:1,
     xlab="dose (mg)", ylab="proportion",
     main="dat.ursino2021 (Sorafenib data)")
studylab <- unique(dat.ursino2021$study)</pre>
colvec <- rainbow(length(studylab))</pre>
for (i in 1:length(studylab)) {
  idx <- (dat.ursino2021$study == studylab[i])</pre>
  lines(dat.ursino2021[idx,"dose"],
        dat.ursino2021[idx,"events"] / dat.ursino2021[idx,"total"],
        col=colvec[i], type="b")
}
legend("topleft", studylab, col=colvec, pch=15)
## End(Not run)
```

dat.vanhowe1999 Studies on the Association between Circumcision and HIV Infection

# Description

Results from 33 studies examining the association between male circumcision and HIV infection.

## Usage

```
dat.vanhowe1999
```

## Format

The data frame contains the following columns:

study	character	study author
category	character	study type (high-risk group, partner study, or population survey)
non.pos	numeric	number of non-circumcised HIV positive cases
non.neg	numeric	number of non-circumcised HIV negative cases
cir.pos	numeric	number of circumcised HIV positive cases
cir.neg	numeric	number of circumcised HIV negative cases

# Details

The 33 studies provide data in terms of  $2 \times 2$  tables in the form:

HIV positive HIV negative

non-circumcised	non.pos	non.neg
circumcised	cir.pos	cir.neg

The goal of the meta-analysis was to examine if the risk of an HIV infection differs between noncircumcised versus circumcised men.

The dataset is interesting because it can be used to illustrate the difference between naively pooling results by summing up the counts across studies and then computing the odds ratio based on the aggregated table (as was done by Van Howe, 1999) and conducting a proper meta-analysis (as illustrated by O'Farrell & Egger, 2000). In fact, a proper meta-analysis shows that the HIV infection risk is on average higher in non-circumcised men, which is the opposite of what the naive pooling approach yields (which makes this an illustration of Simpson's paradox).

# Concepts

medicine, epidemiology, odds ratios

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

## Source

Van Howe, R. S. (1999). Circumcision and HIV infection: Review of the literature and metaanalysis. *International Journal of STD & AIDS*, **10**(1), 8–16. https://doi.org/10.1258/0956462991913015

#### References

O'Farrell, N., & Egger, M. (2000). Circumcision in men and the prevention of HIV infection: A 'meta-analysis' revisited. *International Journal of STD & AIDS*, **11**(3), 137–142. https://doi.org/10.1258/09564620019

## Examples

```
### copy data into 'dat' and examine data
dat <- dat.vanhowe1999
dat
## Not run:
### load metafor package
library(metafor)
### naive pooling by summing up the counts within categories and then
### computing the odds ratios and corresponding confidence intervals
cat1 <- with(dat[dat$category=="high-risk group",],</pre>
 escalc(measure="0R", ai=sum(non.pos), bi=sum(non.neg), ci=sum(cir.pos), di=sum(cir.neg)))
cat2 <- with(dat[dat$category=="partner study",],</pre>
 escalc(measure="0R", ai=sum(non.pos), bi=sum(non.neg), ci=sum(cir.pos), di=sum(cir.neg)))
cat3 <- with(dat[dat$category=="population survey",],</pre>
 escalc(measure="OR", ai=sum(non.pos), bi=sum(non.neg), ci=sum(cir.pos), di=sum(cir.neg)))
summary(cat1, transf=exp, digits=2)
summary(cat2, transf=exp, digits=2)
```

```
summary(cat3, transf=exp, digits=2)
### naive pooling across all studies
all <- escalc(measure="OR", ai=sum(dat$non.pos), bi=sum(dat$non.neg),</pre>
                             ci=sum(dat$cir.pos), di=sum(dat$cir.neg))
summary(all, transf=exp, digits=2)
### calculate log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=non.pos, bi=non.neg, ci=cir.pos, di=cir.neg, data=dat)</pre>
dat
### random-effects model
res <- rma(yi, vi, data=dat, method="DL")</pre>
res
predict(res, transf=exp, digits=2)
### random-effects model within subgroups
res <- rma(yi, vi, data=dat, method="DL", subset=category=="high-risk group")</pre>
predict(res, transf=exp, digits=2)
res <- rma(yi, vi, data=dat, method="DL", subset=category=="partner study")</pre>
predict(res, transf=exp, digits=2)
res <- rma(yi, vi, data=dat, method="DL", subset=category=="population survey")</pre>
predict(res, transf=exp, digits=2)
## End(Not run)
```

dat.viechtbauer2021 Studies to Illustrate Model Checking Methods

# Description

Results from 20 hypothetical randomized clinical trials examining the effectiveness of a medication for treating some disease.

# Usage

dat.viechtbauer2021

#### Format

The data frame contains the following columns:

trial	numeric	trial number
nTi	numeric	number of patients in the treatment group
nCi	numeric	number of patients in the control group
xTi	numeric	number of patients in the treatment group with remission
xCi	numeric	number of patients in the control group with remission
dose	numeric	dosage of the medication provided to patients in the treatment group (in milligrams per day)

# Details

The dataset was constructed for the purposes of illustrating the model checking and diagnostic methods described in Viechtbauer (2021). The code below provides the results for many of the analyses and plots discussed in the book chapter.

# Concepts

medicine, odds ratios, outliers, model checks

## Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

### Source

Viechtbauer, W. (2021). Model checking in meta-analysis. In C. H. Schmid, T. Stijnen, & I. R. White (Eds.), *Handbook of meta-analysis* (pp. 219-254). Boca Raton, FL: CRC Press. https://doi.org/10.1201/9781315

# Examples

```
### copy data into 'dat' and examine data
dat <- dat.viechtbauer2021</pre>
dat
## Not run:
### load metafor package
library(metafor)
### calculate log odds ratios and corresponding sampling variances
dat <- escalc(measure="OR", ai=xTi, n1i=nTi, ci=xCi, n2i=nCi, add=1/2, to="all", data=dat)</pre>
dat
### number of studies
k <- nrow(dat)</pre>
### fit models
res.CE <- rma(yi, vi, data=dat, method="CE") # same as method="EE"</pre>
res.CE
res.RE <- rma(yi, vi, data=dat, method="DL")</pre>
res.RE
res.MR <- rma(yi, vi, mods = ~ dose, data=dat, method="FE")</pre>
res.MR
res.ME <- rma(yi, vi, mods = ~ dose, data=dat, method="DL")</pre>
res.ME
```

```
### forest and bubble plot
par(mar=c(5,4,1,2))
forest(dat$yi, dat$vi, psize=0.8, efac=0, xlim=c(-4,6), ylim=c(-3,23),
       cex=1, width=c(5,5,5), xlab="Log Odds Ratio (LnOR)",
       header=c("Trial", "LnOR [95% CI]"))
addpoly(res.CE, row=-1, mlab="CE Model")
addpoly(res.RE, row=-2, mlab="RE Model")
abline(h=0)
tmp <- regplot(res.ME, xlim=c(0,250), ylim=c(-1,1.5), predlim=c(0,250), shade=FALSE, digits=1,</pre>
               xlab="Dosage (mg per day)", psize="seinv", plim=c(NA,5), bty="l", las=1,
               lty=c("solid", "dashed"), label=TRUE, labsize=0.8, offset=c(1,0.7))
res.sub <- rma(yi, vi, mods = ~ dose, data=dat, method="DL", subset=-6)</pre>
abline(res.sub, lty="dotted")
points(tmp$xi, tmp$yi, pch=21, cex=tmp$psize, col="black", bg="darkgray")
par(mar=c(5,4,4,2))
### number of standardized deleted residuals larger than +-1.96 in each model
sum(abs(rstudent(res.CE)$z) >= qnorm(0.975))
sum(abs(rstudent(res.MR)$z) >= qnorm(0.975))
sum(abs(rstudent(res.RE)$z) >= qnorm(0.975))
sum(abs(rstudent(res.ME)$z) >= qnorm(0.975))
### plot of the standardized deleted residuals for the RE and ME models
plot(NA, NA, xlim=c(1,20), ylim=c(-4,4), xlab="Study", ylab="Standardized (Deleted) Residual",
     xaxt="n", main="Random-Effects Model", las=1)
axis(side=1, at=1:20)
abline(h=c(-1.96,1.96), lty="dotted")
abline(h=0)
points(1:20, rstandard(res.RE)$z, type="o", pch=19, col="gray70")
points(1:20, rstudent(res.RE)$z, type="o", pch=19)
legend("top", pch=19, col=c("gray70","black"), lty="solid",
       legend=c("Standardized Residuals", "Standardized Deleted Residuals"), bty="n")
plot(NA, NA, xlim=c(1,20), ylim=c(-4,4), xlab="Study", ylab="Standardized (Deleted) Residual",
     xaxt="n", main="Mixed-Effects Model", las=1)
axis(side=1, at=1:20)
abline(h=c(-1.96,1.96), lty="dotted")
abline(h=0)
points(1:20, rstandard(res.ME)$z, type="o", pch=19, col="gray70")
points(1:20, rstudent(res.ME)$z, type="o", pch=19)
legend("top", pch=19, col=c("gray70","black"), lty="solid",
       legend=c("Standardized Residuals", "Standardized Deleted Residuals"), bty="n")
### Baujat plots
```

baujat(res.CE, main="Common-Effects Model", xlab="Squared Pearson Residual", ylim=c(0,5), las=1) baujat(res.ME, main="Mixed-Effects Model", ylim=c(0,2), las=1)

```
### GOSH plots (skipped because this takes quite some time to run)
if (FALSE) {
res.GOSH.CE <- gosh(res.CE, subsets=10^7)</pre>
plot(res.GOSH.CE, cex=0.2, out=6, xlim=c(-0.25,1.25), breaks=c(200,100))
res.GOSH.ME <- gosh(res.ME, subsets=10^7)</pre>
plot(res.GOSH.ME, het="tau2", out=6, breaks=50, adjust=0.6, las=1)
}
### plot of treatment dosage against the standardized residuals
plot(dat$dose, rstandard(res.ME)$z, pch=19, xlab="Dosage (mg per day)",
     ylab="Standardized Residual", xlim=c(0,250), ylim=c(-2.5,2.5), las=1)
abline(h=c(-1.96,1.96), lty="dotted", lwd=2)
abline(h=0)
title("Standardized Residual Plot")
text(dat$dose[6], rstandard(res.ME)$z[6], "6", pos=4, offset=0.4)
### guadratic polynomial model
rma(yi, vi, mods = ~ dose + I(dose^2), data=dat, method="DL")
### lack-of-fit model
resLOF <- rma(yi, vi, mods = ~ dose + factor(dose), data=dat, method="DL", btt=3:9)
resLOF
### scatter plot to illustrate the lack-of-fit model
regplot(res.ME, xlim=c(0,250), ylim=c(-1.0,1.5), xlab="Dosage (mg per day)", ci=FALSE,
       predlim=c(0,250), psize=1, pch=19, col="gray60", digits=1, lwd=1, bty="1", las=1)
dosages <- sort(unique(dat$dose))</pre>
lines(dosages, fitted(resLOF)[match(dosages, dat$dose)], type="o", pch=19, cex=2, lwd=2)
points(dat$dose, dat$yi, pch=19, col="gray60")
legend("bottomright", legend=c("Linear Model", "Lack-of-Fit Model"), pch=c(NA,19), col="black",
       lty="solid", lwd=c(1,2), pt.cex=c(1,2), seg.len=4, bty="n")
### checking normality of the standardized deleted residuals
qqnorm(res.ME, type="rstudent", main="Standardized Deleted Residuals", pch=19, label="out",
       lwd=2, pos=24, ylim=c(-4,3), lty=c("solid", "dotted"), las=1)
### checking normality of the random effects
sav <- qqnorm(ranef(res.ME)$pred, main="BLUPs of the Random Effects", cex=1, pch=19,</pre>
              xlim=c(-2.2,2.2), ylim=c(-0.6,0.6), las=1)
abline(a=0, b=sd(ranef(res.ME)$pred), lwd=2)
text(sav$x[6], sav$y[6], "6", pos=4, offset=0.4)
```

```
### hat values for the CE and RE models
plot(NA, NA, xlim=c(1,20), ylim=c(0,0.21), xaxt="n", las=1, xlab="Study", ylab="Hat Value")
axis(1, 1:20, cex.axis=1)
points(hatvalues(res.CE), type="o", pch=19, col="gray70")
points(hatvalues(res.RE), type="o", pch=19)
abline(h=1/20, lty="dotted", lwd=2)
title("Hat Values for the CE/RE Models")
legend("topright", pch=19, col=c("gray70","black"), lty="solid",
       legend=c("Common-Effects Model", "Random-Effects Model"), bty="n")
### heatmap of the hat matrix for the ME model
cols <- colorRampPalette(c("blue", "white", "red"))(101)</pre>
h <- hatvalues(res.ME, type="matrix")</pre>
image(1:nrow(h), 1:ncol(h), t(h[nrow(h):1,]), axes=FALSE,
   xlab="Influence of the Observed Effect of Study \dots", ylab="On the Fitted Value of Study \dots",
      col=cols, zlim=c(-max(abs(h)),max(abs(h))))
axis(1, 1:20, tick=FALSE)
axis(2, 1:20, labels=20:1, las=1, tick=FALSE)
abline(h=seq(0.5,20.5,by=1), col="white")
abline(v=seq(0.5,20.5,by=1), col="white")
points(1:20, 20:1, pch=19, cex=0.4)
title("Heatmap for the Mixed-Effects Model")
### plot of leverages versus standardized residuals for the ME model
plot(hatvalues(res.ME), rstudent(res.ME)$z, pch=19, cex=0.2+3*sqrt(cooks.distance(res.ME)),
     las=1, xlab="Leverage (Hat Value)", ylab="Standardized Deleted Residual",
     xlim=c(0,0.35), ylim=c(-3.5,2.5))
abline(h=c(-1.96,1.96), lty="dotted", lwd=2)
abline(h=0, lwd=2)
ids <- c(3,6,9)
text(hatvalues(res.ME)[ids] + c(0,0.013,0.010), rstudent(res.ME)$z[ids] - c(0.18,0,0), ids)
title("Leverage vs. Standardized Deleted Residuals")
### plot of the Cook's distances for the ME model
plot(1:20, cooks.distance(res.ME), ylim=c(0,1.6), type="o", pch=19, las=1, xaxt="n", yaxt="n",
     xlab="Study", ylab="Cook's Distance")
axis(1, 1:20, cex.axis=1)
axis(2, seq(0,1.6,by=0.4), las=1)
title("Cook's Distances")
### plot of the leave-one-out estimates of tau^2 for the ME model
x <- influence(res.ME)</pre>
plot(1:20, x$inf$tau2.del, ylim=c(0,0.15), type="o", pch=19, las=1, xaxt="n", xlab="Study",
   ylab=expression(paste("Estimate of ", tau^2, " without the ", italic(i), "th study")))
abline(h=res.ME$tau2, lty="dashed")
axis(1, 1:20)
title("Residual Heterogeneity Estimates")
```

```
dat.white2020
```

Studies on the Relationship between Sexual Signal Expression and Individual Quality

# Description

Results from 41 studies examining the relationship between measures of individual quality and the expression of structurally coloured sexual signals.

## Usage

dat.white2020

# Format

The object is a data frame which contains the following columns:

character character	study-level ID observation-level ID whather the study is charmotional or concrimental
	whether the study is observational or experimental whether the study did $(1)$ or did not $(0)$ include a non-served control trait.
	whether the study did (1) or did not (0) include a non-sexual control trait
character	class of the study organisms
character	class of the study organisms
character	species of the study organisms
character	sex of the study organisms
numeric	whether the colour signals were iridescent $(1)$ or not $(0)$
character	the colour variable quantified
character	whether the colour variable is chromatic or achromatic
character	the measure of individual quality used
character	the body region from which colour was sampled
numeric	study sample size
numeric	Pearson's correlation coefficient
	character character numeric character character character numeric character character character character character numeric

# Details

The 186 rows in this dataset come from 41 experimental and observational studies reporting on the correlation between measures of individual quality (age, body condition, immune function, parasite resistance) and the expression of structurally coloured sexual signals across 28 species. The purpose of this meta-analysis was to test whether structural colour signals show heightened condition-dependent expression, as predicted by evolutionary models of 'honest' signalling.

# Concepts

ecology, evolution, correlation coefficients

## Author(s)

Thomas E. White, <thomas.white@sydney.edu.au>

## Source

White, T. E. (2020). Structural colours reflect individual quality: A meta-analysis. *Biology Letters*, **16**(4), 20200001. https://doi.org/10.1098/rsbl.2020.0001

#### Examples

```
### copy data into 'dat' and examine data
dat <- dat.white2020
head(dat, 10)
## Not run:
### load metafor package
library(metafor)
### calculate r-to-z transformed correlations and corresponding sampling variances
dat <- escalc(measure="ZCOR", ri=r, ni=n, data=dat)
### fit multilevel meta-analytic model
res <- rma.mv(yi, vi, random = list(~ 1 | study_id, ~ 1 | obs), data=dat)
res
## End(Not run)</pre>
```

dat.woods2010

Studies on Treatments for Chronic Obstructive Pulmonary Disease

## Description

Results from 3 trials examining the mortality risk of three treatments and placebo in patients with chronic obstructive pulmonary disease.

# dat.woods2010

# Usage

dat.woods2010

# Format

The data frame contains the following columns:

author	character	first author / study name
treatment	character	treatment
r	integer	number of deaths
Ν	integer	number of patients

# Details

Count mortality statistics in randomised controlled trials of treatments for chronic obstructive pulmonary disease (Woods et al., 2010, Table 1).

# Concepts

medicine, odds ratios, network meta-analysis

# Author(s)

Guido Schwarzer, <guido.schwarzer@uniklinik-freiburg.de>, https://github.com/guido-s/

## Source

Woods, B. S., Hawkins, N., & Scott, D. A. (2010). Network meta-analysis on the log-hazard scale, combining count and hazard ratio statistics accounting for multi-arm trials: A tutorial. *BMC Medical Research Methodology*, **10**, 54. https://doi.org/10.1186/1471-2288-10-54

# Examples

```
### Show full dataset
dat.woods2010
## Not run:
### Load netmeta package
suppressPackageStartupMessages(library("netmeta"))
### Print odds ratios and confidence limits with two digits
oldset <- settings.meta(digits = 2)
### Change appearance of confidence intervals
cilayout("(", "-")
### Transform data from long arm-based format to contrast-based
### format. Argument 'sm' has to be used for odds ratio as summary
### measure; by default the risk ratio is used in the metabin function
### called internally.</pre>
```

```
pw <- pairwise(treatment, event = r, n = N,</pre>
```

```
studlab = author, data = dat.woods2010, sm = "OR")
pw
### Conduct network meta-analysis
net <- netmeta(pw)
net
### Show forest plot
forest(net, ref = "Placebo", drop = TRUE,
    leftlabs = "Contrast to Placebo")
### Use previous settings
settings.meta(oldset)
## End(Not run)</pre>
```

dat.yusuf1985 Studies of Beta Blockers During and After Myocardial Infarction

# Description

Results from studies examining the effectiveness of beta blockers for reducing mortality and reinfarction.

## Usage

dat.yusuf1985

#### Format

The data frame contains the following columns:

table	character	table number
id	character	trial id number
trial	character	trial name or first author
ai	numeric	number of deaths/reinfarctions in treatment group
n1i	numeric	number of patients in treatment group
ci	numeric	number of deaths/reinfarctions in control group
n2i	numeric	number of patients in control group

# Details

The dataset contains table 6 (total mortality from short-term trials of oral beta blockers), 9 (total mortality at one week from trials with an initial IV dose of a beta blocker), 10 (total mortality from long-term trials with treatment starting late and mortality from day 8 onwards in long-term trials that began early and continued after discharge), 11 (nonfatal reinfarction from long-term trials of beta blockers), 12a (sudden death in long-term beta blocker trials), and 12b (nonsudden death in long-term beta blocker trials) on the effectiveness of

# datsearch

of beta blockers for reducing mortality and reinfarction.

The article also describes what is sometimes called Peto's one-step method for meta-analyzing  $2 \times 2$  table data. This method is implemented in the rma.peto function.

# Concepts

medicine, cardiology, odds ratios, Peto's method

#### Author(s)

Wolfgang Viechtbauer, <wvb@metafor-project.org>, https://www.metafor-project.org

# Source

Yusuf, S., Peto, R., Lewis, J., Collins, R., & Sleight, P. (1985). Beta blockade during and after myocardial infarction: An overview of the randomized trials. *Progress in Cardiovascular Disease*, **27**(5), 335–371. https://doi.org/10.1016/s0033-0620(85)80003-7

## Examples

```
### copy data into 'dat'
dat <- dat.yusuf1985
dat[dat$table == 6,]
## Not run:
### load metafor package
library(metafor)
### to select a table for the analysis
tab <- "6" # either: 6, 9, 10, 11, 12a, 12b
### to double-check total counts as reported in article
apply(dat[dat$table==tab,4:7], 2, sum, na.rm=TRUE)
### meta-analysis using Peto's one-step method
res <- rma.peto(ai=ai, n1i=n1i, ci=ci, n2i=n2i, data=dat, subset=(table==tab))
res
predict(res, transf=exp, digits=2)
## End(Not run)</pre>
```

datsearch

Search Function for the Datasets

## Description

Function to search among the existing datasets.

## Usage

datsearch(pattern, concept=TRUE, matchall=TRUE, fixed=TRUE, pkgdown=FALSE)

#### Arguments

pattern	character string or vector of strings specifying the terms to search for within the datasets. Can also be left unspecified to start the function in an interactive mode.
concept	logical indicating whether the search should be confined to the concept terms (TRUE by default) or whether a full-text search should be conducted.
matchall	logical indicating whether only the datasets matching all terms (if multiple are specified) are returned (TRUE by default) or whether datasets matching any one of the terms are returned.
fixed	logical indicating whether a term is a string to be matched as is (TRUE by default). If FALSE, a search term is a regular expression that grep will search for. Only relevant when concept=FALSE (i.e., when doing a full-text search).
pkgdown	logical indicating whether the standard help file or the pkgdown docs (at https: //wviechtb.github.io/metadat/) should be shown for a chosen dataset (FALSE by default).

## Details

The function can be used to search all existing datasets in the **metadat** package based on their concept terms (see below) or based on a full-text search of their corresponding help files.

When running datsearch() without the pattern argument specified, the function starts in an interactive mode and prompts for one or multiple search terms.

Alternatively, one can specify a single search term via the pattern argument or multiple search terms by using a string vector as the pattern or by separating multiple search terms in a single string with ',', ';', or 'and'.

If matchall=TRUE (the default), only datasets matching all search terms (if multiple are specified) are returned. If matchall=FALSE, datasets matching any one of the search terms are returned.

If a single match is found, the corresponding help file is directly shown. If multiple matches are found, the user is prompted to choose one of the matching datasets of interest.

# **Concept Terms**

Each dataset is tagged with one or multiple concept terms that refer to various aspects of a dataset, such as the field/topic of research, the outcome measure used for the analysis, the model(s) used for analyzing the data, and the methods/concepts that can be illustrated with the dataset.

- In terms of 'fields/topics', the following terms have been used at least once: alternative medicine, attraction, cardiology, climate change, covid-19, criminology, dentistry, ecology, education, engineering, epidemiology, evolution, genetics, human factors, medicine, memory, obstetrics, oncology, persuasion, physics, primary care, psychiatry, psychology, smoking, social work, sociology.
- In terms of 'outcome measures', the following terms have been used at least once: correlation coefficients, Cronbach's alpha, hazard ratios, incidence rates, raw mean differences, odds ratios, proportions, ratios of means, raw means, risk differences, risk ratios, (semi-)partial correlations, standardized mean changes, standardized mean differences.

# datsearch

• In terms of 'models/methods/concepts', the following terms have been used at least once: cluster-robust inference, component network meta-analysis, cumulative meta-analysis, diagnostic accuracy studies, dose response models, generalized linear models, longitudinal models, Mantel-Haenszel method, meta-regression, model checks, multilevel models, multivariate models, network meta-analysis, outliers, Peto's method, phylogeny, publication bias, reliability generalization, single-arm studies, spatial correlation, subgroup analysis.

# Author(s)

Daniel Noble, <daniel.noble@anu.edu.au> Wolfgang Viechtbauer, <wvb@metafor-project.org>

## Examples

# note: the examples below are not run since they require interactivity

if (FALSE) {

# start the function in the interactive mode
datsearch()

# find all datasets tagged with the concept term 'standardized mean differences'
datsearch("standardized mean differences")

# find all datasets tagged with the concept terms 'odds ratio' and 'multilevel'
datsearch("odds ratio, multilevel")

# do a full-text search for the term 'infarct'
datsearch("infarct", concept=FALSE)

# do a full-text search for 'rma.mv(' (essentially finds all datasets where # the rma.mv() function was used in the examples section of a help file) datsearch("rma.mv(", concept=FALSE)

```
}
```

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