# Package: MBNMAdose (via r-universe)

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Maintainer Hugo Pedder <hugopedder@gmail.com>

Description Fits Bayesian dose-response model-based network meta-analysis (MBNMA) that incorporate multiple doses within an agent by modelling different dose-response functions, as described by Mawdsley et al. (2016) <doi:10.1002/psp4.12091>. By modelling dose-response relationships this can connect networks of evidence that might otherwise be disconnected, and can improve precision on treatment estimates. Several common dose-response functions are provided; others may be added by the user. Various characteristics and assumptions can be flexibly added to the models, such as shared class effects. The consistency of direct and indirect evidence in the network can be assessed using unrelated mean effects models and/or by node-splitting at the treatment level.

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**Depends** R (>= 3.0.2)

**Imports** grDevices, stats, graphics, utils, scales, dplyr (>= 0.7.4), R2jags (>= 0.5-7), rjags (>= 4-8), magrittr (>= 1.5), checkmate (>= 1.8.5), Rdpack (>= 0.11-0), igraph (>= 1.1.2), reshape2 (>= 1.4.3)

**Suggests** overlapping (>= 1.5.0), ggplot2 (>= 2.2.1), RColorBrewer (>= 1.1-2), mcmcplots (>= 0.4.3), coda (>= 0.19-4), testthat (>= 1.0.2), crayon (>= 1.3.4), forestplot (>= 1.10), ggdist (>= 2.4.0), zoo (>= 1.8-8), lspline (>= 1.0-0), formatR (>= 1.14), netmeta, knitr, rmarkdown

**SystemRequirements** JAGS (>= 4.3.0)

(https://mcmc-jags.sourceforge.net/)

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Encoding UTF-8 LazyData true VignetteBuilder knitr RoxygenNote 7.2.3 Roxygen list(markdown = TRUE) RdMacros Rdpack Repository https://hugaped.r-universe.dev RemoteUrl https://github.com/hugaped/mbnmadose RemoteRef HEAD RemoteSha 2aa78373f55fe55eae8e51a6f41827710d0c247b

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Index

add\_index

# Description

Adds arm indices (arms, narms) to a dataset and adds numeric identifiers for agent and class (if included in the data).

# Usage

add\_index(data.ab, agents = NULL, treatments = NULL)

# Arguments

data.ab	A data frame of arm-level data in "long" format containing the columns:
	<ul> <li>studyID Study identifiers</li> </ul>
	• dose Numeric data indicating the dose (must take positive values)
	• agent Agent identifiers (can be numeric, factor or character)
	• y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data.
	• se Numeric data indicating the standard error for a given observation. Re- quired for continuous data.
	• r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data.
	• n Numeric data indicating the total number of participants within a study arm. Required for binomial data or when modelling Standardised Mean Differences
	• E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data.
	• class An optional column indicating a particular class code. Agents with the same identifier must also have the same class code.
	• standsd An optional column of numeric data indicating reference SDs used to standardise treatment effects when modelling using Standardised Mean Differences (SMD).
agents	A character string of agent names used to force a particular agent ordering. De- fault is NULL, which automatically orders Placebo (dose=0) as agent 1 and then subsequent agents by the order given in data.ab
treatments	A character string of treatment names used to force a particular treatment order- ing. Default is NULL, which automatically orders Placebo (dose=0) as treatment 1 and then subsequent treatments by the order of agents and doses (smallest to highest) given in data.ab

#### alog\_pcfb

#### Value

A data frame similar to data. ab but with additional columns:

- arm Arm identifiers coded for each study
- narm The total number of arms in each study

If agent or class are non-numeric or non-sequential (i.e. with missing numeric codes), agents/classes in the returned data frame will be numbered and recoded to enforce sequential numbering (a warning will be shown stating this).

alog_pcfb	Studies of alogliptin for lowering blood glucose concentration in pa-
	tients with type II diabetes

# Description

A dataset from a systematic review of Randomised-Controlled Trials (RCTs) comparing different doses of alogliptin with placebo (Langford et al. 2016). The systematic review was simply performed and was intended to provide data to illustrate a statistical methodology rather than for clinical inference. Alogliptin is a treatment aimed at reducing blood glucose concentration in type II diabetes. The outcome is continuous, and aggregate data responses correspond to the mean change in HbA1c from baseline to follow-up in studies of at least 12 weeks follow-up. The dataset includes 14 Randomised-Controlled Trials (RCTs), comparing 5 different doses of alogliptin with placebo, leading to 6 different treatments (combination of dose and agent) within the network.

#### Usage

alog\_pcfb

## Format

A data frame in long format (one row per arm and study), with 46 rows and 6 variables:

- studyID Study identifiers
- · agent Character data indicating the agent to which participants were randomised
- · dose Numeric data indicating the standardised dose received
- y Numeric data indicating the mean change from baseline in blood glucose concentration (mg/dL) in a study arm
- se Numeric data indicating the standard error for the mean change from baseline in blood glucose concentration (mg/dL) in a study arm
- n Numeric data indicating the number of participants randomised

#### Details

alog\_pcfb is a data frame in long format (one row per arm and study), with the variables studyID, agent, dose, y, se, and N.

#### References

Langford O, Aronson JK, van Valkenhoef G, Stevens RJ (2016). "Methods for meta-analysis of pharmacodynamic dose-response data with application to multi-arm studies of alogliptin." *Stat Methods Med Res.* ISSN 1477-0334 (Electronic) 0962-2802 (Linking), doi:10.1177/0962280216637093.

calc.edx

Calculates values for EDx from an Emax model, the dose at which x% of the maximal response (Emax) is reached

# Description

Calculates values for EDx from an Emax model, the dose at which x% of the maximal response (Emax) is reached

# Usage

calc.edx(mbnma, x = 50)

# Arguments

mbnma	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
x	A numeric value between 0 and 100 for the dose at which $x\%$ of the maximal response (Emax) should be calculated

#### Value

A data frame of posterior EDx summary values for each agent

changepd

Update model fit statistics depending on calculation for pD

# Description

Update model fit statistics depending on calculation for pD

# Usage

```
changepd(model, jagsdata = NULL, pd = "pv", likelihood = NULL, type = "dose")
```

# check.network

# Arguments

model	A model object of class "rjags"
jagsdata	A list object containing data used to estimate model
pd	Can take either:
	<ul> <li>pv only pV will be reported (as automatically outputted by R2jags).</li> <li>plugin calculates pD by the plug-in method (Spiegelhalter et al. 2002). It is faster, but may output negative non-sensical values, due to skewed deviances that can arise with non-linear models.</li> <li>pd.kl calculates pD by the Kullback-Leibler divergence (Plummer 2008). This will require running the model for additional iterations but is a more robust calculation for the effective number of parameters in non-linear models.</li> <li>popt calculates pD using an optimism adjustment which allows for calculates.</li> </ul>
	lation of the penalized expected deviance (Plummer 2008).
likelihood	A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
type	Can take either "dose" for a dose-response MBNMA or "time" for a time- course MBNMA (this accounts for multiple observations within an arm)

# Value

A list containing pd (effective number of parameters calculated using the method specified in arguments), deviance (the posterior median of the total residual deviance) and dic (the model DIC)

check.network	Check if all nodes in the network are connected (identical to function
	<i>in</i> MBNMAtime)

# Description

Check if all nodes in the network are connected (identical to function in MBNMAtime)

# Usage

```
check.network(g, reference = 1)
```

# Arguments

g	An network plot of class("igraph")
reference	A numeric value indicating which treatment code to use as the reference treat- ment for testing that all other treatments connect to it

cumrank

# Description

Plot cumulative ranking curves from MBNMA models

#### Usage

cumrank(x, params = NULL, sucra = TRUE, ...)

#### Arguments

x	An object of class "mbnma.rank" generated by rank.mbnma()
params	A character vector of named parameters in the model that vary by either agent or class (depending on the value assigned to level). If left as NULL (the de- fault), then ranking will be calculated for all available parameters that vary by agent/class.
sucra	A logical object to indicate whether Surface Under Cumulative Ranking Curve (SUCRA) values should be calculated and returned as a data frame. Areas calculated using trapezoid approach.
	Arguments to be sent to ggplot::geom_line()

# Value

Line plots showing the cumulative ranking probabilities for each agent/class and dose-response parameter in x. The object returned is a list which contains the plot (an object of class(c("gg", "ggplot")) and a data frame of SUCRA values if sucra = TRUE.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
```

```
# Estimate rankings from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")
ranks <- rank(emax)</pre>
```

```
# Plot cumulative rankings for both dose-response parameters simultaneously
# Note that SUCRA values are also returned
cumrank(ranks)
```

default.priors

# Description

This function creates JAGS code snippets for default MBNMA model priors.

# Usage

```
default.priors(
  fun = dloglin(),
  UME = FALSE,
  regress.mat = NULL,
  regress.effect = "common",
  om = list(rel = 5, abs = 10)
)
```

# Arguments

fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
UME	A boolean object to indicate whether to fit an Unrelated Mean Effects model that does not assume consistency and so can be used to test if the consistency assumption is valid.
regress.mat	A Nstudy x Ncovariate design matrix of meta-regression covariates
regress.effect	Indicates whether effect modification should be assumed to be "common" (as- sumed to be equal versus Placebo throughout the network), "random" (assumed to be exchangeable versus Placebo throughout the network), "agent" (assumed to be equal versus Placebo within each agent), or "class" (assumed to be equal versus Placebo within each class).
om	a list with two elements that report the maximum relative ("rel") and maximum absolute ("abs") efficacies on the link scale.

# Value

A list, each element of which is a named JAGS snippet corresponding to a prior in the MBNMA JAGS code.

# Examples

```
default.priors(fun=demax())
```

# Description

Emax dose-response function

#### Usage

```
demax(emax = "rel", ed50 = "rel", hill = NULL, p.expon = FALSE)
```

# Arguments

emax	Pooling for Emax parameter. Can take "rel", "common", "random" or be assigned a numeric value (see details).
ed50	Pooling for ED50 parameter. Can take "rel", "common", "random" or be assigned a numeric value (see details).
hill	Pooling for Hill parameter. Can take "rel", "common", "random" or be assigned a numeric value (see details).
p.expon	A logical object to indicate whether ed50 and hill parameters should be expressed within the dose-response function on an exponential scale

# Details

Emax represents the maximum response. exp(ED50) represents the dose at which 50% of the maximum response is achieved. exp(Hill) is the Hill parameter, which allows for a sigmoidal function.

Without Hill parameter:

$$\frac{E_{max} \times x}{ET_{50} + x}$$

With Hill parameter:

$$\frac{E_{max} \times x^{hill}}{ET_{50} \times hill} + x^{hill}$$

#### Value

An object of class("dosefun")

# **Dose-response parameters**

Argument	Model specification
"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data but

demax

#### devdev

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

# References

There are no references for Rd macro \insertAllCites on this help page.

#### Examples

# Model without a Hill parameter

```
demax(emax="rel", ed50="common")
# Model including a Hill parameter and defaults for Emax and ED50 parameters
demax(hill="common")
```

devdev

Dev-dev plot for comparing deviance contributions from two models

#### Description

Plots the deviances of two model types for comparison. Often used to assess consistency by comparing consistency (NMA or MBNMA) and unrelated mean effects (UME) models (see Pedder et al. (2021)). Models must be run on the *same set of data* or the deviance comparisons will not be valid.

#### Usage

```
devdev(mod1, mod2, dev.type = "resdev", n.iter = 2000, n.thin = 1, ...)
```

#### Arguments

mod1	First model for which to plot deviance contributions
mod2	Second model for which to plot deviance contributions
dev.type	<i>STILL IN DEVELOPMENT FOR MBNMAdose!</i> Deviances to plot - can be either residual deviances ("resdev", the default) or deviances ("dev")
n.iter	number of total iterations per chain (including burn in; default: 2000)
n.thin	thinning rate. Must be a positive integer. Set n.thin > 1 to save memory and computation time if n.iter is large. Default is max(1, floor(n.chains * (n.iter-n.burnin) / 1000)) which will only thin if there are at least 2000 simulations.
	Arguments to be sent to ggplot2::geom_point() or ggplot2::geom_boxplot

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Run an poorly fitting linear dose-response
lin <- mbnma.run(network, fun=dpoly(degree=1))
# Run a better fitting Emax dose-response
emax <- mbnma.run(network, fun=demax())
# Run a standard NMA with unrelated mean effects (UME)
ume <- nma.run(network, UME=TRUE)
# Compare residual deviance contributions from linear and Emax
devdev(lin, emax) # Suggests model fit is very different
# Compare deviance contributions from Emax and UME
devdev(emax, ume) # Suggests model fit is similar
```

devplot

```
Plot deviance contributions from an MBNMA model
```

#### Description

Plot deviance contributions from an MBNMA model

# Usage

```
devplot(
   mbnma,
   plot.type = "box",
   facet = TRUE,
   dev.type = "resdev",
   n.iter = mbnma$BUGSoutput$n.iter/2,
   n.thin = mbnma$BUGSoutput$n.thin,
   ...
)
```

# Arguments

mbnma	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
plot.type	Deviances can be plotted either as scatter points ("scatter") or as boxplots ("box")

# devplot

facet	A boolean object that indicates whether or not to facet (by agent for MBNMAdose and by treatment for MBNMAtime) $% \label{eq:mass_state}$
dev.type	<i>STILL IN DEVELOPMENT FOR MBNMAdose!</i> Deviances to plot - can be either residual deviances ("resdev", the default) or deviances ("dev")
n.iter	number of total iterations per chain (including burn in; default: 2000)
n.thin	thinning rate. Must be a positive integer. Set n.thin > 1 to save memory and computation time if n.iter is large. Default is max(1, floor(n.chains * (n.iter-n.burnin) / 1000)) which will only thin if there are at least 2000 simulations.
	Arguments to be sent to ggplot2::geom_point() or ggplot2::geom_boxplot

# Details

Deviances should only be plotted for models that have converged successfully. If deviance contributions have not been monitored in mbnma\$parameters.to.save then additional iterations will have to be run to get results for these.

For MBNMAtime, deviance contributions cannot be calculated for models with a multivariate likelihood (i.e. those that account for correlation between observations) because the covariance matrix in these models is treated as unknown (if rho = "estimate") and deviance contributions will be correlated.

#### Value

Generates a plot of deviance contributions and returns a list containing the plot (as an object of class(c("gg", "ggplot"))), and a data.frame of posterior mean deviance/residual deviance contributions for each observation.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Run an Emax dose-response MBNMA and predict responses
emax <- mbnma.run(network, fun=demax(), method="random")
# Plot deviances
devplot(emax)
# Plot deviances using boxplots
devplot(emax, plot.type="box")
# Plot deviances on a single scatter plot (not facetted by agent)
devplot(emax, facet=FALSE, plot.type="scatter")
# A data frame of deviance contributions can be obtained from the object
#returned by `devplot`
devs <- devplot(emax)
head(devs$dev.data)
```

```
# Other deviance contributions not currently implemented but in future
#it will be possible to plot them like so
#devplot(emax, dev.type="dev")
```

dexp

Exponential dose-response function

# Description

Similar parameterisation to the Emax model but with non-asymptotic maximal effect (Emax). Can fit a 1-parameter (Emax only) or 2-parameter model (includes onset parameter that controls the curvature of the dose-response relationship)

#### Usage

dexp(emax = "rel", onset = NULL, p.expon = FALSE)

#### Arguments

emax	Pooling for Emax parameter. Can take "rel", "common", "random" or be as- signed a numeric value (see details).
onset	Pooling for onset parameter. Can take "rel", "common", "random" or be as- signed a numeric value (see details).
p.expon	A logical object to indicate whether ed50 and hill parameters should be expressed within the dose-response function on an exponential scale

# Details

1-parameter model:  $emax \times (1 - exp(-x))$ 2-parameter model:  $emax \times (1 - exp(onset * -x))$ 

where emax is the maximum efficacy of an agent and rate is the speed

Dose-response parameter arguments:

#### Argument Model specification

"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo

numeric() Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data but

#### Value

An object of class("dosefun")

# dfpoly

# References

There are no references for Rd macro \insertAllCites on this help page.

# Examples

```
# Single parameter exponential function is default
dexp()
```

```
# Two parameter exponential function
dexp(onset="rel")
```

dfpoly

# Fractional polynomial dose-response function

# Description

Fractional polynomial dose-response function

# Usage

```
dfpoly(degree = 1, beta.1 = "rel", beta.2 = "rel", power.1 = 0, power.2 = 0)
```

# Arguments

degree	The degree of the fractional polynomial as defined in Royston and Altman (1994)
beta.1	Pooling for the 1st fractional polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.2	Pooling for the 2nd fractional polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
power.1	Value for the 1st fractional polynomial power ( $\gamma_1$ ). Must take any numeric value in the set -2, -1, -0.5, 0, 0.5, 1, 2, 3.
power.2	Value for the 2nd fractional polynomial power ( $\gamma_2$ ). Must take any numeric value in the set -2, -1, -0.5, 0, 0.5, 1, 2, 3.

# Details

- $\beta_1$  represents the 1st coefficient.
- $\beta_2$  represents the 2nd coefficient.
- $\gamma_1$  represents the 1st fractional polynomial power

dfpoly

•  $\gamma_2$  represents the 2nd fractional polynomial power

For a polynomial of degree=1:

 $\beta_1 x^{\gamma_1}$ 

For a polynomial of degree=2:

$$\beta_1 x^{\gamma_1} + \beta_2 x^{\gamma_2}$$

 $x^{\gamma}$  is a regular power except where  $\gamma = 0$ , where  $x^{(0)} = ln(x)$ . If a fractional polynomial power  $\gamma$  repeats within the function it is multiplied by another ln(x).

# Value

An object of class("dosefun")

#### **Dose-response parameters**

Argument	Model specification
"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data bu

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

#### References

Royston P, Altman D (1994). "Regression Using Fractional Polynomials of Continuous Covariates: Parsimonious Parametric Modelling." *Journal of the Royal Statistical Society: Series C*, **43**(3), 429-467.

#### Examples

```
# 1st order fractional polynomial a value of 0.5 for the power
dfpoly(beta.1="rel", power.1=0.5)
# 2nd order fractional polynomial with relative effects for coefficients
# and a value of -0.5 and 2 for the 1st and 2nd powers respectively
dfpoly(degree=2, beta.1="rel", beta.2="rel",
    power.1=-0.5, power.2=2)
```

#### Description

Similar parameterisation to the Emax model but with non-asymptotic maximal effect (Emax). Proposed by proposed by Fu and Manner (2010)

Integrated Two-Component Prediction (ITP) function

#### Usage

ditp(emax = "rel", rate = "rel", p.expon = FALSE)

# Arguments

emax	Pooling for Emax parameter. Can take "rel", "common", "random" or be assigned a numeric value (see details).
rate	Pooling for Rate parameter. Can take "rel", "common", "random" or be assigned a numeric value (see details).
p.expon	A logical object to indicate whether ed50 and hill parameters should be expressed within the dose-response function on an exponential scale

#### Details

Emax represents the maximum response. Rate represents the rate at which a change in the dose of the drug leads to a change in the effect

$$E_{max} \times \frac{(1 - exp(-rate \times x))}{(1 - exp(-rate \times max(x)))}$$

#### Value

An object of class("dosefun")

#### **Dose-response parameters**

# Argument Model specification

"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the number of the second
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data but

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

# ditp

# References

Fu H, Manner D (2010). "Bayesian adaptive dose-finding studies with delayed responses." *J Biopharm Stat*, **20**(5), 1055-1070. doi:10.1080/10543400903315740.

#### Examples

# Model a common effect on rate
ditp(emax="rel", rate="common")

dloglin

Log-linear (exponential) dose-response function

# Description

Modelled assuming relative effects ("rel")

# Usage

dloglin()

#### Details

 $rate \times log(x+1)$ 

Dose-response parameter arguments:

#### Argument Model specification

"rel"	Implies that relative effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data but

#### Value

An object of class("dosefun")

# References

There are no references for Rd macro \insertAllCites on this help page.

#### Examples

dloglin()

dmulti

# Description

Function combines different dose-response functions together to create an object containing parameters for multiple dose-response functions.

#### Usage

dmulti(funs = list())

# Arguments

# funs

A list of objects of class("dosefun"), each element of which corresponds to an agent in the dataset to be modelled. The list length must be equal to the number of agents in network\$agents used in mbnma.run(), and the order of the dose-response functions in the list is assumed to correspond to the same order of agents in network\$agents.

# Value

An object of class("dosefun")

#### Examples

```
funs <- c(rep(list(demax()),3),
            rep(list(dloglin()),2),
            rep(list(demax(ed50="common")),3),
            rep(list(dexp()),2))</pre>
```

dmulti(funs)

dnonparam

Non-parameteric dose-response functions

#### Description

Used to fit monotonically increasing non-parametric dose-response relationship following the method of Owen et al. (2015))

#### Usage

dnonparam(direction = "increasing")

#### Arguments

direction

Can take either "increasing" or "decreasing" to indicate the monotonic direction of the dose-response relationship

#### Value

```
An object of class("dosefun")
```

# References

Owen RK, Tincello DG, Keith RA (2015). "Network meta-analysis: development of a three-level hierarchical modeling approach incorporating dose-related constraints." *Value Health*, **18**(1), 116-26. ISSN 1524-4733 (Electronic) 1098-3015 (Linking), doi:10.1016/j.jval.2014.10.006, https://pubmed.ncbi.nlm.nih.gov/25595242/.

# Examples

```
# Monotonically increasing dose-response
dnonparam(direction="increasing")
```

```
# Monotonically decreasing dose-response
dnonparam(direction="decreasing")
```

dpoly

Polynomial dose-response function

#### Description

Polynomial dose-response function

#### Usage

```
dpoly(
    degree = 1,
    beta.1 = "rel",
    beta.2 = "rel",
    beta.3 = "rel",
    beta.4 = "rel"
)
```

# Arguments

degree	The degree of the polynomial - e.g. degree=1 for linear, degree=2 for quadratic, degree=3 for cubic.
beta.1	Pooling for the 1st polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).

#### dpoly

beta.2	Pooling for the 2nd polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.3	Pooling for the 3rd polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.4	Pooling for the 4th polynomial coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).

# Details

- $\beta_1$  represents the 1st coefficient.
- $\beta_2$  represents the 2nd coefficient.
- $\beta_3$  represents the 3rd coefficient.
- $\beta_4$  represents the 4th coefficient.

Linear model:

 $\beta_1 x$ 

Quadratic model:

 $\beta_1 x + \beta_2 x^2$ 

Cubic model:

 $\beta_1 x + \beta_2 x^2 + \beta_3 x^3$ 

Quartic model:

 $\beta_1 x + \beta_2 x^2 + \beta_3 x^3 + \beta_4 x^4$ 

#### Value

An object of class("dosefun")

# **Dose-response parameters**

# Argument Model specification

"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data bu

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

# References

There are no references for Rd macro \insertAllCites on this help page.

# Examples

```
# Linear model with random effects
dpoly(beta.1="rel")
# Quadratic model dose-response function
# with an exchangeable (random) absolute parameter estimated for the 2nd coefficient
dpoly(beta.1="rel", beta.2="random")
```

DR. comparisons Adds placebo comparisons for dose-response relationship

# Description

Function adds additional rows to a data.frame of comparisons in a network that account for the relationship between placebo and other agents via the dose-response relationship.

# Usage

```
DR.comparisons(data.ab, level = "treatment", doselink = NULL)
```

# Arguments

data.ab	A data frame stored in an <code>mbnma.network</code> object ( <code>mbnma.network\$data.ab</code> )
level	A character that can take either "treatment" or "agent" to indicate the level of the network for which to identify dose-response
doselink	If given an integer value it indicates that connections via the dose-response re- lationship with placebo should be plotted. The integer represents the minimum number of doses from which a dose-response function could be estimated and is equivalent to the number of parameters in the desired dose-response function plus one. If left as NULL (the default), connections to placebo via dose-response relationships will not be included.

drop.comp

Drop treatments from multi-arm (>2) studies for node-splitting

# Description

Drops arms in a way which preserves connectivity and equally removes data from each treatment in a nodesplit comparison (so as to maximise precision)

# Usage

```
drop.comp(ind.df, drops, comp, start = 1)
```

# drop.disconnected

#### Arguments

ind.df	A data frame in long format (one arm per row) from which to drop treatments
drops	A vector of study identifiers from which to drop treatments
comp	A numeric vector of length 2 that contains treatment codes corresponding to the comparison for node-splitting
start	Can take either 0 or 1 to indicate whether to drop the treatment in $comp[1](0)$ or $comp[2](1)$

drop.disconnected Drop studies that are not connected to the network reference treatment

#### Description

Drop studies that are not connected to the network reference treatment

#### Usage

```
drop.disconnected(network, connect.dose = FALSE)
```

#### Arguments

network	An object of class mbnma.network.
connect.dose	A boolean object to indicate whether treatments should be kept in the network if they connect via the simplest possible dose-response relationship (TRUE) or not (FALSE). Simplest possible dose-response relationship is any function with
	a single dose-response parameter (e.g. linear, exponential)

# Value

A list containing a single row per arm data frame containing only studies that are connected to the network reference treatment, and a character vector of treatment labels

# Examples

```
# Using the triptans headache dataset
network <- mbnma.network(triptans)
drops <- drop.disconnected(network)</pre>
```

```
# No studies have been dropped since network is fully connected
length(unique(network$data.ab$studyID))==length(unique(drops$data.ab$studyID))
```

```
# Make data with no placebo
noplac.df <- network$data.ab[network$data.ab$narm>2 & network$data.ab$agent!=1,]
net.noplac <- mbnma.network(noplac.df)</pre>
```

# Studies are dropped as some only connect via the dose-response function

```
drops <- drop.disconnected(net.noplac, connect.dose=FALSE)
length(unique(net.noplac$data.ab$studyID))==length(unique(drops$data.ab$studyID))
# Studies are not dropped if they connect via the dose-response function
drops <- drop.disconnected(net.noplac, connect.dose=TRUE)
length(unique(net.noplac$data.ab$studyID))==length(unique(drops$data.ab$studyID))</pre>
```

dspline

#### Spline dose-response functions

# Description

Used to fit B-splines, natural cubic splines, and piecewise linear splines(Perperoglu et al. 2019).

# Usage

```
dspline(
   type = "bs",
   knots = 1,
   degree = 1,
   beta.1 = "rel",
   beta.2 = "rel",
   beta.3 = "rel",
   beta.4 = "rel",
   beta.5 = "rel",
   beta.6 = "rel"
)
```

#### Arguments

type	The type of spline. Can take "bs" (B-spline), "ns" (natural cubic spline), or "1s" (piecewise linear spline)
knots	The number/location of spline internal knots. If a single number is given it indicates the number of knots (they will be equally spaced across the range of doses <i>for each agent</i> ). If a numeric vector is given it indicates the location of the knots.
degree	The degree of the piecewise B-spline polynomial - e.g. degree=1 for linear, degree=2 for quadratic, degree=3 for cubic.
beta.1	Pooling for the 1st coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.2	Pooling for the 2nd coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.3	Pooling for the 3rd coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).

#### dspline

beta.4	Pooling for the 4th coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.5	Pooling for the 5th coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.6	Pooling for the 6th coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).

# Value

An object of class("dosefun")

# **Dose-response parameters**

Argument	Model specification
"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data bu

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

#### References

Perperoglu A, Sauerbrei W, Abrahamowicz M, Schmid M (2019). "A review of spline function procedures in R." *BMC Medical Research Methodology*, **19**(46), 1-16. doi:10.1186/s12874019-06663.

#### Examples

```
# Second order B spline with 2 knots and random effects on the 2nd coefficient
dspline(type="bs", knots=2, degree=2,
    beta.1="rel", beta.2="rel")
# Piecewise linear spline with knots at 0.1 and 0.5 quantiles
# Single parameter independent of treatment estimated for 1st coefficient
#with random effects
dspline(type="ls", knots=c(0.1,0.5),
    beta.1="random", beta.2="rel")
```

# Description

User-defined dose-response function

#### Usage

```
duser(fun, beta.1 = "rel", beta.2 = "rel", beta.3 = "rel", beta.4 = "rel")
```

#### Arguments

fun	A formula specifying any relationship including dose and one/several of: beta.1, beta.2, beta.3, beta.4.
beta.1	Pooling for the 1st coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.2	Pooling for the 2nd coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.3	Pooling for the 3rd coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).
beta.4	Pooling for the 4th coefficient. Can take "rel", "common", "random" or be assigned a numeric value (see details).

# Value

An object of class("dosefun")

#### **Dose-response parameters**

# Argument Model specification

"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data bu

When relative effects are modelled on more than one dose-response parameter, correlation between them is automatically estimated using a vague inverse-Wishart prior. This prior can be made slightly more informative by specifying the scale matrix omega and by changing the degrees of freedom of the inverse-Wishart prior using the priors argument in mbnma.run().

# References

There are no references for Rd macro \insertAllCites on this help page.

fitplot

# Examples

```
dr <- ~ beta.1 * (1/(dose+1)) + beta.2 * dose^2
duser(fun=dr,
    beta.1="common", beta.2="rel")</pre>
```

```
fitplot
```

#### Plot fitted values from MBNMA model

# Description

Plot fitted values from MBNMA model

# Usage

```
fitplot(
  mbnma,
  disp.obs = TRUE,
  n.iter = mbnma$BUGSoutput$n.iter,
  n.thin = mbnma$BUGSoutput$n.thin,
   ...
)
```

#### Arguments

mbnma	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
disp.obs	A boolean object to indicate whether raw data responses should be plotted as points on the graph
n.iter	number of total iterations per chain (including burn in; default: 2000)
n.thin	thinning rate. Must be a positive integer. Set n.thin > 1 to save memory and computation time if n.iter is large. Default is max(1, floor(n.chains * (n.iter-n.burnin) / 1000)) which will only thin if there are at least 2000 simulations.
	Arguments to be sent to ggplot2::geom_point() or ggplot2::geom_line()

# Details

Fitted values should only be plotted for models that have converged successfully. If fitted values (theta) have not been monitored in mbnma\$parameters.to.save then additional iterations will have to be run to get results for these.

# Value

Generates a plot of fitted values from the MBNMA model and returns a list containing the plot (as an object of class(c("gg", "ggplot"))), and a data.frame of posterior mean fitted values for each observation.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Run an Emax dose-response MBNMA and predict responses
emax <- mbnma.run(network, fun=demax(), method="random")
# Plot fitted values and observed values
fitplot(emax)
# Plot fitted values only
fitplot(emax, disp.obs=FALSE)
# A data frame of fitted values can be obtained from the object
#returned by `fitplot`
fits <- fitplot(emax)</pre>
```

gen.parameters.to.save

head(fits\$fv)

Automatically generate parameters to save for a dose-response MB-NMA model

# Description

Automatically generate parameters to save for a dose-response MBNMA model

### Usage

```
gen.parameters.to.save(fun, model, regress.mat = NULL)
```

#### Arguments

fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
model	A JAGS model written as a character object
regress.mat	A Nstudy x Ncovariate design matrix of meta-regression covariates

genspline

# Description

Generates spline basis matrices for fitting to dose-response function

# Usage

```
genspline(
    x,
    spline = "bs",
    knots = 1,
    degree = 1,
    max.dose = max(x),
    boundaries = NULL
)
```

# Arguments

х	A numeric vector indicating all time points available in the dataset
spline	Indicates the type of spline function. Can be either a piecewise linear spline ("1s"), natural cubic spline ("ns"), or B-spline ("bs").
knots	The number/location of internal knots. If a single integer is given it indicates the number of knots (they will be equally spaced across the range of doses <i>for each agent</i> ). If a numeric vector is given it indicates the quantiles of the knots as a proportion of the maximum dose in the dataset. For example, if the maximum dose in the dataset is $100 \text{ mg/d}$ , knots=c(0.1,0.5) would indicate knots should be fitted at $10 \text{ mg/d}$ and $50 \text{ mg/d}$ .
degree	a positive integer giving the degree of the polynomial from which the spline function is composed (e.g. degree=3 represents a cubic spline).
max.dose	A number indicating the maximum dose between which to calculate the spline function.
boundaries	A positive numeric vector of length 2 that represents the doses at which to anchor the B-spline or natural cubic spline basis matrix. This allows data to extend beyond the boundary knots, or for the basis parameters to not depend on x. The default (boundaries=NULL) is the range of x.

# Value

A spline basis matrix with number of rows equal to length(x) and the number of columns equal to the number of coefficients in the spline.

#### Examples

x <- 0:100
genspline(x)
# Generate a quadratic B-spline with 1 equally spaced internal knot
genspline(x, spline="bs", knots=2, degree=2)
# Generate a natural cubic spline with 3 knots at selected quantiles
genspline(x, spline="ns", knots=c(0.1, 0.5, 0.7))
# Generate a piecewise linear spline with 3 equally spaced knots
genspline(x, spline="ls", knots=3)</pre>

get.prior

Get current priors from JAGS model code

#### Description

Identical to get.prior() in MBNMAtime package. This function takes JAGS model presented as a string and identifies what prior values have been used for calculation.

# Usage

get.prior(model)

#### Arguments

model A character object of JAGS MBNMA model code

# Details

Even if an MBNMA model that has not initialised successfully and results have not been calculated, the JAGS model for it is saved in mbnma\$model.arg\$jagscode and therefore priors can still be obtained. This allows for priors to be changed even in failing models, which may help solve issues with compiling or updating.

#### Value

A character vector, each element of which is a line of JAGS code corresponding to a prior in the JAGS code.

# get.relative

#### Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Run an Emax dose-response MBNMA
result <- mbnma.run(network, fun=demax(), method="random")
# Obtain model prior values
print(result$model.arg$priors)
# Priors when using mbnma.run with an exponential function
result <- mbnma.run(network, fun=dexp(), method="random")
print(result$model.arg$priors)
```

get.relative

Calculates league table of effects between treatments in MBNMA and/or NMA models

#### Description

Calculates league table of effects between treatments in MBNMA and/or NMA models

# Usage

```
get.relative(
    lower.diag,
    upper.diag = lower.diag,
    treatments = list(),
    lower.direction = "colvrow",
    upper.direction = "rowvcol",
    regress.vals = NULL,
    eform = FALSE,
    lim = "cred"
)
```

# Arguments

lower.diag	An S3 object either of class "mbnma" generated by running a dose-response MB- NMA model (using mbnma.run()), or of class "nma" generated by running an NMA model (using nma.run()). Indicates that this model should be used to estimate treatment effects in the <b>lower-left diagonal</b> of the league table.
upper.diag	Same as for lower.diag, but indicates that this model should be used to estimate treatment effects in the <b>upper-right diagonal</b> of the league table.

treatments	A list whose elements each represent different treatments. Treatment is defined as a combination of agent and dose. Only agents specified in lower.diag and upper.diag can be included. Each element in treatments is named corre- sponding to the agent and contains a numeric vector of doses. Relative ef- fects will be calculated between all treatments specified in treatments. If treatments is left empty then the set of treatments in the network on which lower.diag is estimated will be used as the default.	
lower.direction		
	Whether treatment effects should be presented as the column versus the row treatment for each cell in the <b>lower-left diagonal</b> of the league table ("colvrow") or as the row versus the column treatment ("rowvcol").	
upper.direction		
	Same as for upper.direction but for results shown in the <b>upper-right diago-</b> <b>nal</b> of the league table.	
regress.vals	A named numeric vector of effect modifier values at which relative effects should be estimated. Named elements must match variable names specified in regres- sion design matrix (mbnma\$model.arg\$regress.mat), corresponding to vari- ables in regress within the MBNMA model (or <b>both</b> MBNMA models if dif- ferent models are used for upper.diag and lower.diag. Meta-regression is not currently possible for NMA models, so only unadjusted results will be shown for these (and a warning given if they are compared to adjusted MBNMA results).	
eform	Whether outputted results should be presented in their exponential form (e.g. for models with log or logit link functions)	
lim	Specifies calculation of either 95% credible intervals (lim="cred") or 95% prediction intervals (lim="pred").	

# Value

An array of length(treatments) x length(treatments) x nsims, where nsims is the number of iterations monitored in lower.diag. The array contains the individual MCMC values for each relative effect calculated between all treatments on the link scale specified in the lower.diag and upper.diag models.

# Examples

```
# Using the osteoarthritis data
network <- mbnma.network(osteopain)
# Run an MBNMA model
expon <- mbnma.run(network, fun=dexp(), method="random")
# Calculate relative effects for MBNMA between:
# Celebrex 100mg/d, Celebrex 200mg/d, Tramadol 100mg/d
rel.eff <- get.relative(lower.diag=expon,
    treatments=list("Celebrex"=c(100,200), "Tramadol"=100))
# Run an NMA model
nma <- nma.run(network, method="random")</pre>
```

```
# Compare results between MBNMA and NMA models
rel.eff <- get.relative(lower.diag=expon, upper.diag=nma,
    treatments=list("Celebrex"=c(100,200), "Tramadol"=100),
    upper.direction="colvrow")</pre>
```

getjagsdata

# Prepares data for JAGS

#### Description

Converts MBNMA data frame to a list for use in JAGS model

#### Usage

```
getjagsdata(
  data.ab,
  class = FALSE,
  sdscale = FALSE,
  regress = NULL,
  regress.effect = "common",
  likelihood = check.likelink(data.ab)$likelihood,
  link = check.likelink(data.ab)$link,
  level = "agent",
  fun = NULL,
  nodesplit = NULL
)
```

#### Arguments

data.ab

A data frame of arm-level data in "long" format containing the columns:

- studyID Study identifiers
- dose Numeric data indicating the dose (must take positive values)
- agent Agent identifiers (can be numeric, factor or character)
- y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data.
- se Numeric data indicating the standard error for a given observation. Required for continuous data.
- r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data.
- n Numeric data indicating the total number of participants within a study arm. Required for binomial data or when modelling Standardised Mean Differences
- E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data.

	• class An optional column indicating a particular class code. Agents with the same identifier must also have the same class code.
	• standsd An optional column of numeric data indicating reference SDs used to standardise treatment effects when modelling using Standardised Mean Differences (SMD).
class	A boolean object indicating whether or not data.ab contains information on different classes of treatments
sdscale	Logical object to indicate whether to write a model that specifies a reference SD for standardising when modelling using Standardised Mean Differences. Specifying sdscale=TRUE will therefore only modify the model if link function is set to SMD (link="smd").
regress	A formula of effect modifiers (variables that interact with the treatment effect) to incorporate using Network Meta-Regression (E.g. ~ Population + Age). All variables in the formula are modelled as interacting with the treatment effect (i.e. prognostic variables cannot be included in this way). Effects modifiers must be named variables in network\$data.ab and must be identical within a study. Factor and character effect modifiers will be converted to a series of named dummy variables.
regress.effect	Indicates whether effect modification should be assumed to be "common" (as- sumed to be equal versus Placebo throughout the network), "random" (assumed to be exchangeable versus Placebo throughout the network), "agent" (assumed to be equal versus Placebo within each agent), or "class" (assumed to be equal versus Placebo within each class).
likelihood	A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
link	A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.
level	Can take either "agent" to indicate that data should be at the agent- level (for MBNMA) or "treatment" to indicate that data should be at the treatment- level (for NMA)
fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
nodesplit	A numeric vector of length 2 containing treatment codes on which to perform an MBNMA nodesplit (see mbnma.nodesplit).

# Value

A named list of numbers, vector, matrices and arrays to be sent to JAGS. List elements are:

• If likelihood="normal":

- y An array of mean responses for each arm within each study

#### getjagsdata

- se An array of standard errors for each arm within each study
- If likelihood="binomial":
  - r An array of the number of responses/count for each each arm within each study
  - n An array of the number of participants for each arm within each study
- If likelihood="poisson":
  - r An array of the number of responses/count for each each arm within each study
  - E An array of the total exposure time for each arm within each study
- dose A matrix of doses for each arm within each study (if level="agent")
- narm A numeric vector with the number of arms per study
- NS The total number of studies in the dataset
- Nagent The total number of agents in the dataset (if level="agent")
- agent A matrix of agent codes within each study (if level="agent")
- NT The total number of treatment in the dataset (if level="treatment")
- treatment A matrix of treatment codes within each study (if level="treatment")
- Nclass Optional. The total number of classes in the dataset
- · class Optional. A matrix of class codes within each study
- classkey Optional. A vector of class codes that correspond to agent codes. Same length as the number of agent codes.
- split.ind Optional. A matrix indicating whether a specific arm contributes evidence to a nodesplit comparison.

#### Examples

```
# Using the triptans headache dataset
network <- mbnma.network(triptans)
jagsdat <- getjagsdata(network$data.ab, likelihood="binomial", link="logit")
# Get JAGS data with class
netclass <- mbnma.network(osteopain)
jagsdat <- getjagsdata(netclass$data.ab, class=TRUE)
# Get JAGS data at the treatment level for split Network Meta-Analysis
network <- mbnma.network(triptans)
jagsdat <- getjagsdata(network$data.ab, level="treatment")</pre>
```

gout

Studies of treatments for Serum Uric Acid reduction in patients with gout

#### Description

A dataset from a systematic review of interventions for lowering Serum Uric Acid (SUA) concentration in patients with gout (**not published previously**). The outcome is continuous, and aggregate data responses correspond to the mean change from baseline in SUA in mg/dL at 2 weeks followup. The dataset includes 10 Randomised-Controlled Trials (RCTs), comparing 5 different agents, and placebo. Data for one agent (RDEA) arises from an RCT that is not placebo-controlled, and so is not connected to the network directly. In total there were 19 different treatments (combination of dose and agent).

#### Usage

gout

# Format

A data frame in long format (one row per arm and study), with 27 rows and 5 variables:

- studyID Study identifiers
- y Numeric data indicating the mean change from baseline in SUA in a study arm
- se Numeric data indicating the standard error for the mean change from baseline in SUA in a study arm
- · agent Character data indicating the agent to which participants were randomised
- · dose Numeric data indicating the standardised dose received

#### Source

Pfizer Ltd.

inconsistency.loops Identify comparisons in loops that fulfill criteria for node-splitting

#### Description

Identify comparisons informed by both direct and indirect evidence from independent sources, which therefore fulfill the criteria for testing for inconsistency via node-splitting.

#### Usage

```
inconsistency.loops(df, checkindirect = TRUE, incldr = FALSE)
```

#### Arguments

df	A data frame containing variables studyID and treatment (as numeric codes) that indicate which treatments are used in which studies. If checkindirect = TRUE then variables agent and dose are also required.
checkindirect	A boolean object to indicate whether or not to perform an additional check to ensure network remains connected even after dropping direct evidence on a comparison. Default is TRUE and should be kept as TRUE if working with dose- response data, though this requires further computational iterations to confirm. If set to FALSE, additional comparisons may be identified, though computation will be much more rapid.
incldr	A boolean object indicating whether or not to allow for indirect evidence contributions via the dose-response relationship. This can be used when node-splitting in dose-response MBNMA to allow for a greater number of potential loops in which to check for consistency.

### Details

Similar to gemtc::mtc.nodesplit.comparisons() but uses a fixed reference treatment and therefore identifies fewer loops in which to test for inconsistency. Heterogeneity can also be parameterised as inconsistency and so testing for inconsistency in additional loops whilst changing the reference treatment would also be identifying heterogeneity. Depends on igraph.

## Value

A data frame of comparisons that are informed by direct and indirect evidence from independent sources. Each row of the data frame is a different treatment comparison. Numerical codes in t1 and t2 correspond to treatment codes. path indicates the treatment codes that connect the shortest path of indirect evidence.

If incldr=TRUE then path may indicate doseresp for some comparisons. These are comparisons for which indirect evidence is only available via the dose-response relationship. The two numbers given after (e.g. 3 2) indicate the number of doses available in the indirect evidence with which to estimate the dose-response function for the treatments in t1 and t2 respectively/

#### References

There are no references for Rd macro \insertAllCites on this help page.

### Examples

```
# Identify comparisons informed by direct and indirect evidence
#in triptans dataset
network <- mbnma.network(triptans)
inconsistency.loops(network$data.ab)
```

# Include indirect evidence via dose-response relationship inconsistency.loops(network\$data.ab, incldr=TRUE)

mbnma.comparisons *Identify unique comparisons within a network* 

#### Description

Identify unique contrasts within a network that make up all the head-to-head comparisons. Repetitions of the same treatment comparison are grouped together.

## Usage

```
mbnma.comparisons(df)
```

### Arguments

df

A data frame containing variables studyID and treatment (as numeric codes) that indicate which treatments are used in which studies.

#### Value

A data frame of unique comparisons in which each row represents a different comparison. t1 and t2 indicate the treatment codes that make up the comparison. nr indicates the number of times the given comparison is made within the network.

If there is only a single follow-up observation for each study within the dataset (i.e. as for standard network meta-analysis) nr will represent the number of studies that compare treatments t1 and t2.

If there are multiple observations for each study within the dataset (as in time-course MBNMA) nr will represent the number of time points in the dataset in which treatments t1 and t2 are compared.

#### Examples

```
df <- data.frame(studyID=c(1,1,2,2,3,3,4,4,5,5,5),
    treatment=c(1,2,1,3,2,3,3,4,1,2,4)
  )
# Identify unique comparisons within the data
mbnma.comparisons(df)
```

```
# Using the triptans headache dataset
network <- mbnma.network(triptans) # Adds treatment identifiers
mbnma.comparisons(network$data.ab)
```

mbnma.nodesplit

*Node-splitting model for testing consistency at the treatment level using MBNMA* 

# Description

Splits contributions for a given set of treatment comparisons into direct and indirect evidence. A discrepancy between the two suggests that the consistency assumption required for NMA and MB-NMA may violated.

## Usage

```
mbnma.nodesplit(
   network,
   fun = dpoly(degree = 1),
   method = "common",
   comparisons = NULL,
   incldr = TRUE,
   ...
)
### S3 method for class 'nodesplit'
plot(x, plot.type = "forest", ...)
```

network	An object of class mbnma.network.
fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
method	Can take either "common" or "random" to indicate whether relative effects should be modelled with between-study heterogeneity or not (see details).
comparisons	A matrix specifying the comparisons to be split (one row per comparison). The matrix must have two columns indicating each treatment for each comparison. Values can either be character (corresponding to the treatment names given in network) or numeric (corresponding to treatment codes within the network - note that these may change if drop.discon = TRUE).
incldr	A boolean object indicating whether or not to allow for indirect evidence contributions via the dose-response relationship. This can be used when node-splitting in dose-response MBNMA to allow for a greater number of potential loops in which to check for consistency.
	Arguments to be sent to ggplot2::ggplot() or forestplot::forestplot()
x	An object of class("nodesplit")
plot.type	A character string that can take the value of "forest" to plot forest plots or "density" to plot posterior density plots.

#### Details

The S3 method plot() on an nodesplit object generates either forest plots of posterior medians and 95\% credible intervals, or density plots of posterior densities for direct and indirect evidence.

## Value

Plots the desired graph if plot.type="forest" and plots and returns an object of class(c("gg", "ggplot")) if plot.type="density".

# Functions

• plot(nodesplit): Plot outputs from treatment-level nodesplit MBNMA models

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
```

```
split <- mbnma.nodesplit(network, fun=demax(), likelihood = "binomial", link="logit",
    method="common")
```

#### To perform nodesplit on selected comparisons ####

```
# Check for closed loops of treatments with independent evidence sources
# Including indirect evidence via the dose-response relationship
loops <- inconsistency.loops(network$data.ab, incldr=TRUE)</pre>
```

```
# Plot results
plot(split, plot.type="density") # Plot density plots of posterior densities
plot(split, txt_gp=forestplot::fpTxtGp(cex=0.5)) # Plot forest plots (with smaller label size)
# Print and summarise results
print(split)
summary(split) # Generate a data frame of summary results
```

mbnma.run

#### Description

Fits a Bayesian dose-response for model-based network meta-analysis (MBNMA) that can account for multiple doses of different agents by applying a desired dose-response function. Follows the methods of Mawdsley et al. (2016).

### Usage

```
mbnma.run(
  network,
  fun = dpoly(degree = 1),
 method = "common",
 regress = NULL,
  regress.effect = "common",
  class.effect = list(),
 UME = FALSE,
  sdscale = FALSE,
  cor = FALSE,
 omega = NULL,
  parameters.to.save = NULL,
  pd = "pd.kl",
  likelihood = NULL,
  link = NULL,
  priors = NULL,
 n.iter = 20000,
 n.chains = 3,
 n.burnin = floor(n.iter/2),
 n.thin = max(1, floor((n.iter - n.burnin)/1000)),
 autojags = FALSE,
 Rhat = 1.05,
 n.update = 10,
 model.file = NULL,
  jagsdata = NULL,
  . . .
)
```

network	An object of class mbnma.network.
fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
method	Can take either "common" or "random" to indicate whether relative effects should be modelled with between-study heterogeneity or not (see details).

regress	A formula of effect modifiers (variables that interact with the treatment effect) to incorporate using Network Meta-Regression (E.g. ~ Population + Age). All variables in the formula are modelled as interacting with the treatment effect (i.e. prognostic variables cannot be included in this way). Effects modifiers must be named variables in network\$data.ab and must be identical within a study. Factor and character effect modifiers will be converted to a series of named dummy variables.
regress.effect	Indicates whether effect modification should be assumed to be "common" (as- sumed to be equal versus Placebo throughout the network), "random" (assumed to be exchangeable versus Placebo throughout the network), "agent" (assumed to be equal versus Placebo within each agent), or "class" (assumed to be equal versus Placebo within each class).
class.effect	A list of named strings that determines which dose-response parameters to model with a class effect and what that effect should be ("common" or "random"). Element names should match dose-response parameter names. Note that assuming class effects on some dose-response parameters may be unreasonable if the range of doses differ substantially across agents within a class.
UME	A boolean object to indicate whether to fit an Unrelated Mean Effects model that does not assume consistency and so can be used to test if the consistency assumption is valid.
sdscale	Logical object to indicate whether to write a model that specifies a reference SD for standardising when modelling using Standardised Mean Differences. Specifying sdscale=TRUE will therefore only modify the model if link function is set to SMD (link="smd").
cor	A boolean object that indicates whether correlation should be modelled between relative effect dose-response parameters. This is automatically set to FALSE if class effects are modelled or if multiple dose-response functions are fitted.
omega	A scale matrix for the inverse-Wishart prior for the covariance matrix used to model the correlation between dose-response parameters (see Details for dose-response functions). omega must be a symmetric positive definite matrix with dimensions equal to the number of dose-response parameters modelled using relative effects ("rel"). If left as NULL (the default) a diagonal matrix with elements equal to 100 is used.
parameters.to.	
	A character vector containing names of parameters to monitor in JAGS
pd	Can take either:
	<ul> <li>pv only pV will be reported (as automatically outputted by R2jags).</li> <li>plugin calculates pD by the plug-in method (Spiegelhalter et al. 2002). It is faster, but may output negative non-sensical values, due to skewed deviances that can arise with non-linear models.</li> <li>pd.kl calculates pD by the Kullback-Leibler divergence (Plummer 2008). This will require running the model for additional iterations but is a more robust calculation for the effective number of parameters in non-linear mod-</li> </ul>
	els.
	• popt calculates pD using an optimism adjustment which allows for calculation of the penalized expected deviance (Plummer 2008).

- likelihood A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
- link A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.
- priors A named list of parameter values (without indices) and replacement prior distribution values given as strings **using distributions as specified in JAGS syntax** (see Plummer (2017)). Note that normal distributions in JAGS are specified as

 $N(\mu, prec)$ 

, where

 $prec = 1/\sigma^2$ 

n.iter	number of total iterations per chain (including burn in; default: 20000)
n.chains	number of Markov chains (default: 3)
n.burnin	length of burn in, i.e. number of iterations to discard at the beginning. Default is 'n.iter/2", that is, discarding the first half of the simulations. If n.burnin is 0, jags() will run 100 iterations for adaption.
n.thin	thinning rate. Must be a positive integer. Set n. thin > 1`` to save memory and computation time if floor(n.chains * (n.iter-n.burnin) / 1000))" which will only thin if there are at least 2000 simulations.
autojags	A boolean value that indicates whether the model should be continually updated until it has converged below a specific cutoff of Rhat
Rhat	A cutoff value for the Gelman-Rubin convergence diagnostic(Gelman and Rubin 1992). Unless all parameters have Rhat values lower than this the model will continue to sequentially update up to a maximum of n.update. Default is 1.05.
n.update	The maximum number of updates. Each update is run for 1000 iterations, after which the Rhat values of all parameters are checked against Rhat. Default maximum updates is 10 (i.e. 10,000 additional iterations in total).
model.file	The file path to a JAGS model (.jags file) that can be used to overwrite the JAGS model that is automatically written based on the specified options in MBNMAdose. Useful for adding further model flexibility.
jagsdata	A named list of the data objects to be used in the JAGS model. Only required if users are defining their own JAGS model using model.file. Format should match that of standard models fitted in MBNMAdose (see mbnma\$model.arg\$jagsdata)
	Arguments to be sent to R2jags.

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#### Details

When relative effects are modelled on more than one dose-response parameter and cor = TRUE, correlation between the dose-response parameters is automatically estimated using a vague Wishart prior. This prior can be made slightly more informative by specifying the relative scale of variances between the dose-response parameters using omega. cor will automatically be set to FALSE if class effects are modelled.

# Value

An object of S3 class(c("mbnma", "rjags")) containing parameter results from the model. Can be summarized by print() and can check traceplots using R2jags::traceplot() or various functions from the package mcmcplots.

Nodes that are automatically monitored (if present in the model) have the following interpretation:

### Parameters(s)/Parameter Prefix

```
<named dose-response parameter>(e.g. emax)
sd
sd.<named dose-response parameter>(e.g. sd.emax)
<named capitalized dose-response parameter>(e.g. EMAX)
sd.<named capitalized dose-response parameter>(e.g. sd.EMAX)
totresdev
deviance
```

#### Interpretation

The pooled effect for each dose-response parameter The between-study SD (heterogeneity) for relative Between-study SD (heterogeneity) for absolute do The class effect within each class for a given dose-The within-class SD for different agents within the The residual deviance of the model The deviance of the model

If there are errors in the JAGS model code then the object will be a list consisting of two elements - an error message from JAGS that can help with debugging and model.arg, a list of arguments provided to mbnma.run() which includes jagscode, the JAGS code for the model that can help users identify the source of the error.

#### **Dose-response parameter arguments**

#### Argument Model specification

8	1
"rel"	Implies that <i>relative</i> effects should be pooled for this dose-response parameter separately for each agent in the n
"common"	Implies that all agents share the same common effect for this dose-response parameter.
"random"	Implies that all agents share a similar (exchangeable) effect for this dose-response parameter. This approach allo
numeric()	Assigned a numeric value, indicating that this dose-response parameter should not be estimated from the data bu

## **Dose-response function**

Several general dose-response functions are provided, but a user-defined dose-response relationship can instead be used.

As of version 0.4.0 dose-response functions are specified as an object of class("dosefun"). See help details for each of the functions below for the interpretation of specific dose-response parameters.

Built-in dose-response functions are:

#### mbnma.run

- dpoly(): polynomial (e.g. for a linear model dpoly(degree=1))
- dloglin(): log-linear
- dexp(): exponential
- demax(): (emax with/without a Hill parameter)
- dspline(): splines (can fit B-splines (type="bs"), restricted cubic splines (type="rcs"), natural splines (type="ns"), or piecewise linear splines (type="ls"))
- dfpoly(): fractional polynomials
- dnonparam(): Non-parametric monotonic function (direction can be either "increasing" or "decreasing") following the method of Owen et al. (2015)
- duser(): user-defined function
- dmulti(): allows agent-specific dose-response functions to be fitted. A separate function must be provided for each agent in the network.

#### References

Gelman A, Rubin DB (1992). "Inference from iterative simulation using multiple sequences." *Statistical Science*, **7**(4), 457-511. https://projecteuclid.org/journals/statistical-science/volume-7/issue-4/Inference-from-Iterative-Simulation-Using-Multiple-Sequences/10. 1214/ss/1177011136.full.

Mawdsley D, Bennetts M, Dias S, Boucher M, Welton NJ (2016). "Model-Based Network Meta-Analysis: A Framework for Evidence Synthesis of Clinical Trial Data." *CPT Pharmacometrics Syst Pharmacol*, **5**(8), 393-401. ISSN 2163-8306 (Electronic) 2163-8306 (Linking), doi:10.1002/ psp4.12091, https://pubmed.ncbi.nlm.nih.gov/27479782/.

Owen RK, Tincello DG, Keith RA (2015). "Network meta-analysis: development of a three-level hierarchical modeling approach incorporating dose-related constraints." *Value Health*, **18**(1), 116-26. ISSN 1524-4733 (Electronic) 1098-3015 (Linking), doi:10.1016/j.jval.2014.10.006, https://pubmed.ncbi.nlm.nih.gov/25595242/.

Plummer M (2008). "Penalized loss functions for Bayesian model comparison." *Biostatistics*, **9**(3), 523-39. ISSN 1468-4357 (Electronic) 1465-4644 (Linking), https://pubmed.ncbi.nlm. nih.gov/18209015/.

Plummer M (2017). JAGS user manual. https://people.stat.sc.edu/hansont/stat740/ jags\_user\_manual.pdf.

Spiegelhalter DJ, Best NG, Carlin BP, van der Linde A (2002). "Bayesian measures of model complexity and fit." *J R Statistic Soc B*, **64**(4), 583-639.

# Examples

# Using the triptans data
network <- mbnma.network(triptans)</pre>

####### Dose-response functions ########

```
# Fit a dose-response MBNMA with a linear function
# with common treatment effects
result <- mbnma.run(network, fun=dpoly(degree=1), method="common")</pre>
# Fit a dose-response MBNMA with a log-linear function
# with random treatment effects
result <- mbnma.run(network, fun=dloglin(), method="random")</pre>
# Fit a dose-response MBNMA with a fractional polynomial function
# with random treatment effects
# with a probit link function
result <- mbnma.run(network, fun=dfpoly(), method="random", link="probit")
# Fit a user-defined function (quadratic)
fun.def <- ~ (beta.1 * dose) + (beta.2 * (dose^2))</pre>
result <- mbnma.run(network, fun=duser(fun=fun.def), method="common")</pre>
# Fit an Emax function
# with a single random (exchangeable) parameter for ED50
# with common treatment effects
result <- mbnma.run(network, fun=demax(emax="rel", ed50="random"),</pre>
              method="common")
# Fit an Emax function with a Hill parameter
# with a fixed value of 5 for the Hill parameter
# with random relative effects
result <- mbnma.run(network, fun=demax(hill=5), method="random")</pre>
# Fit a model with natural cubic splines
# with 3 knots at 10% 30% and 60% quartiles of dose ranges
depnet <- mbnma.network(ssri) # Using the sSRI depression dataset</pre>
result <- mbnma.run(depnet, fun=dspline(type="ns", knots=c(0.1,0.3,0.6)))</pre>
# Fit a model with different dose-response functions for each agent
multifun <- dmulti(list(dloglin(), # for placebo (can be any function)</pre>
                       demax(), # for eletriptan
                       demax(), # for sumatriptan
                       dloglin(), # for frovatriptan
                       demax(), # for almotriptan
                       demax(), # for zolmitriptan
                       dloglin(), # for naratriptan
                       demax())) # for rizatriptan
multidose <- mbnma.run(network, fun=multifun)</pre>
# Using the osteoarthritis dataset
 pain.df <- osteopain</pre>
 # Set a shared class (NSAID) only for Naproxcinod and Naproxen
```

```
pain.df <- pain.df %>% dplyr::mutate(
              class = dplyr::case_when(agent %in% c("Naproxcinod", "Naproxen") ~ "NSAID",
                        !agent %in% c("Naproxcinod", "Naproxen") ~ agent
                        )
                )
 # Run an Emax MBNMA with a common class effect on emax
 painnet <- mbnma.network(pain.df)</pre>
result <- mbnma.run(painnet, fun = demax(),</pre>
                class.effect = list(emax = "common"))
####### Priors #######
# Obtain priors from a fractional polynomial function
result <- mbnma.run(network, fun=dfpoly(degree=1), method="random")</pre>
print(result$model.arg$priors)
# Change the prior distribution for the power
newpriors <- list(power.1 = "dnorm(0,0.001) T(0,)")</pre>
newpriors <- list(sd = "dnorm(0,0.5) T(0,)")</pre>
result <- mbnma.run(network, fun=dfpoly(degree=1), method="random",</pre>
              priors=newpriors)
# Change the number of MCMC iterations, the number of chains, and the thin
result <- mbnma.run(network, fun=dloglin(), method="random",</pre>
              n.iter=5000, n.thin=5, n.chains=4)
# Calculate effective number of parameters via plugin method
result <- mbnma.run(network, fun=dloglin(), method="random",</pre>
              pd="plugin")
# Calculate effective number of parameters using penalized expected deviance
result <- mbnma.run(network, fun=dloglin(), method="random",</pre>
              pd="popt")
####### Examine MCMC diagnostics (using mcmcplots or coda packages) #######
# Density plots
mcmcplots::denplot(result)
# Traceplots
mcmcplots::traplot(result)
# Caterpillar plots
mcmcplots::caterplot(result, "rate")
# Autocorrelation plots (using the coda package)
```

mbnma.update

Update MBNMA to monitor deviance nodes in the model

## Description

Useful for obtaining deviance contributions or fitted values. Same function used in MBNMAdose and MBNMAtime packages.

# Usage

```
mbnma.update(
   mbnma,
   param = "theta",
   armdat = TRUE,
   n.iter = mbnma$BUGSoutput$n.iter,
   n.thin = mbnma$BUGSoutput$n.thin
)
```

mbnma	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
param	Used to indicate which node to monitor in the model. Can be any parameter in the model code that varies by all arms within all studies. These are some typical parameters that it might be of interest to monitor, provided they are in the original model code:

	• "theta" for fitted values
	• "psi" for fitted values on natural scale (e.g. probabilities)
	• "dev" for deviance contributions
	<ul> <li>"resdev" for residual deviance contributions</li> </ul>
	<ul> <li>"delta" for within-study relative effects versus the study reference treat- ment</li> </ul>
armdat	Include raw arm-level data for each data point (agent, dose, study grouping)
n.iter	number of total iterations per chain (including burn in; default: 2000)
n.thin	thinning rate. Must be a positive integer. Set n.thin > 1 to save memory and computation time if n.iter is large. Default is max(1, floor(n.chains * (n.iter-n.burnin) / 1000)) which will only thin if there are at least 2000 simulations.

### Value

A data frame containing the posterior mean of the updates by arm and study, with arm and study identifiers.

For MBNMAdose:

- · facet indicates the agent identifier in the given arm of a study
- fupdose indicates the dose in the given arm of a study

For MBNMAtime:

- · facet indicates the treatment identifier in the given arm of the study
- fupdose indicates the follow-up time at the given observation in the given arm of the study

### Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Fit a dose-response MBNMA, monitoring "psi" and "resdev"
result <- mbnma.run(network, fun=dloglin(), method="random",
parameters.to.save=c("psi", "resdev"))
mbnma.update(result, param="theta") # monitor theta
mbnma.update(result, param="rhat") # monitor rhat
mbnma.update(result, param="delta") # monitor delta
```

mbnma.validate.data Validates that a dataset fulfills requirements for MBNMA

### Description

Validates that a dataset fulfills requirements for MBNMA

### Usage

```
mbnma.validate.data(data.ab, single.arm = FALSE)
```

#### Arguments

data.ab A data frame of arm-level data in "long" format containing the columns: studyID Study identifiers • dose Numeric data indicating the dose (must take positive values) • agent Agent identifiers (can be numeric, factor or character) • y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data. • se Numeric data indicating the standard error for a given observation. Required for continuous data. • r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data. • n Numeric data indicating the total number of participants within a study arm. Required for binomial data or when modelling Standardised Mean Differences • E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data. • class An optional column indicating a particular class code. Agents with the same identifier must also have the same class code. • standsd An optional column of numeric data indicating reference SDs used to standardise treatment effects when modelling using Standardised Mean Differences (SMD). single.arm A boolean object to indicate whether to allow single arm studies in the dataset (TRUE) or not (FALSE)

# Details

Checks done within the validation:

- · Checks data.ab has required column names
- · Checks there are no NAs
- Checks that all SEs are >0 (if variables are included in dataset)
- Checks that all doses are >=0

### mbnma.write

- Checks that all r and n are positive (if variables are included in dataset)
- Checks that all y, se, r, n and E are numeric
- · Checks that class codes are consistent within each agent
- · Checks that agent/class names do not contain restricted characters
- Checks that studies have at least two arms (if single.arm = FALSE)
- · Checks that each study includes at least two treatments
- · Checks that agent names do not include underscores
- · Checks that standsd values are consistent within a study

# Value

An error if checks are not passed. Runs silently if checks are passed

mbnma.write

### Write MBNMA dose-response model JAGS code

#### Description

Writes JAGS code for a Bayesian time-course model for model-based network meta-analysis (MB-NMA).

## Usage

```
mbnma.write(
  fun = dpoly(degree = 1),
 method = "common",
  regress.mat = NULL,
  regress.effect = "common",
  sdscale = FALSE,
  cor = FALSE,
  cor.prior = "wishart",
  omega = NULL,
  om = list(rel = 5, abs = 10),
  class.effect = list(),
 UME = FALSE,
 likelihood = "binomial",
  link = NULL
```

```
)
```

fun	An object of class("dosefun") that specifies a functional form to be assigned to the dose-response. See Details.
method	Can take either "common" or "random" to indicate whether relative effects should be modelled with between-study heterogeneity or not (see details).

regress.mat	A Nstudy x Ncovariate design matrix of meta-regression covariates
regress.effect	Indicates whether effect modification should be assumed to be "common" (as- sumed to be equal versus Placebo throughout the network), "random" (assumed to be exchangeable versus Placebo throughout the network), "agent" (assumed to be equal versus Placebo within each agent), or "class" (assumed to be equal versus Placebo within each class).
sdscale	Logical object to indicate whether to write a model that specifies a reference SD for standardising when modelling using Standardised Mean Differences. Specifying sdscale=TRUE will therefore only modify the model if link function is set to SMD (link="smd").
cor	A boolean object that indicates whether correlation should be modelled between relative effect dose-response parameters. This is automatically set to FALSE if class effects are modelled or if multiple dose-response functions are fitted.
cor.prior	NOT CURRENTLY IN USE - indicates the prior distribution to use for the correlation/covariance between relative effects. Must be kept as "wishart"
omega	A scale matrix for the inverse-Wishart prior for the covariance matrix used to model the correlation between dose-response parameters (see Details for dose-response functions). omega must be a symmetric positive definite matrix with dimensions equal to the number of dose-response parameters modelled using relative effects ("rel"). If left as NULL (the default) a diagonal matrix with elements equal to 100 is used.
om	a list with two elements that report the maximum relative ("rel") and maximum absolute ("abs") efficacies on the link scale.
class.effect	A list of named strings that determines which dose-response parameters to model with a class effect and what that effect should be ("common" or "random"). El- ement names should match dose-response parameter names. Note that assum- ing class effects on some dose-response parameters may be unreasonable if the range of doses differ substantially across agents within a class.
UME	A boolean object to indicate whether to fit an Unrelated Mean Effects model that does not assume consistency and so can be used to test if the consistency assumption is valid.
likelihood	A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
link	A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.

# Details

When relative effects are modelled on more than one dose-response parameter and cor = TRUE, correlation between the dose-response parameters is automatically estimated using a vague Wishart prior. This prior can be made slightly more informative by specifying the relative scale of variances

#### mbnma.write

between the dose-response parameters using omega. cor will automatically be set to FALSE if class effects are modelled.

# Value

A single long character string containing the JAGS model generated based on the arguments passed to the function.

# Examples

```
# Write model code for a model with an exponential dose-response function,
# with random treatment effects
model <- mbnma.write(fun=dexp(),</pre>
             method="random",
             likelihood="binomial",
             link="logit"
             )
names(model) <- NULL</pre>
print(model)
# Write model code for a model with an Emax dose-response function,
# relative effects modelled on Emax with a random effects model,
# a single parameter estimated for ED50 with a common effects model
model <- mbnma.write(fun=demax(emax="rel", ed50="common"),</pre>
             likelihood="normal",
             link="identity"
             )
names(model) <- NULL</pre>
print(model)
# Write model code for a model with an Emax dose-response function,
# relative effects modelled on Emax and ED50.
# Class effects modelled on ED50 with common effects
model <- mbnma.write(fun=demax(),</pre>
             likelihood="normal",
             link="identity",
             class.effect=list("ed50"="common")
             )
names(model) <- NULL</pre>
print(model)
# Write model code for a model with an Emax dose-response function,
# relative effects modelled on Emax and ED50 with a
# random effects model that automatically models a correlation between
# both parameters.
model <- mbnma.write(fun=demax(),</pre>
             method="random".
             likelihood="normal",
             link="identity",
             )
names(model) <- NULL</pre>
print(model)
```

nma.nodesplit

# Description

Splits contributions for a given set of treatment comparisons into direct and indirect evidence. A discrepancy between the two suggests that the consistency assumption required for NMA (and subsequently MBNMA) may violated.

# Usage

```
nma.nodesplit(
    network,
    likelihood = NULL,
    link = NULL,
    method = "common",
    comparisons = NULL,
    drop.discon = TRUE,
    ...
)
```

```
## S3 method for class 'nma.nodesplit'
plot(x, plot.type = NULL, ...)
```

network	An object of class mbnma.network.
likelihood	A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
link	A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.
method	Can take either "common" or "random" to indicate whether relative effects should be modelled with between-study heterogeneity or not (see details).
comparisons	A matrix specifying the comparisons to be split (one row per comparison). The matrix must have two columns indicating each treatment for each comparison. Values can either be character (corresponding to the treatment names given in network) or numeric (corresponding to treatment codes within the network - note that these may change if drop.discon = TRUE).
drop.discon	A boolean object that indicates whether to drop treatments that are disconnected at the treatment level. Default is TRUE. If set to FALSE then this could lead to

	identification of nodesplit comparisons that are not connected to the network reference treatment, or lead to errors in running the nodesplit models, though it can be useful for error checking.
	Arguments to be sent to ggplot2::ggplot()
х	An object of class("nma.nodesplit")
plot.type	A character string that can take the value of "forest" to plot only forest plots, "density" to plot only density plots, or left as NULL (the default) to plot both types of plot.

# Details

The S3 method plot() on an nma.nodesplit object generates either forest plots of posterior medians and 95% credible intervals, or density plots of posterior densities for direct and indirect evidence.

# Value

Plots the desired graph(s) and returns an object (or list of object if plot.type=NULL) of class(c("gg", "ggplot"))

# Methods (by generic)

• plot(nma.nodesplit): Plot outputs from treatment-level nodesplit models

## Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
split <- nma.nodesplit(network, likelihood = "binomial", link="logit",
    method="common")</pre>
```

#### To perform nodesplit on selected comparisons ####

```
# Check for closed loops of treatments with independent evidence sources
loops <- inconsistency.loops(network$data.ab)</pre>
```

```
# Plot results
plot(split, plot.type="density") # Plot density plots of posterior densities
plot(split, plot.type="forest") # Plot forest plots of direct and indirect evidence
```

```
# Print and summarise results
print(split)
summary(split) # Generate a data frame of summary results
```

norm21norm

*Convert normal distribution parameters to corresponding log-normal distribution parameters* 

# Description

Converts mean and variance of normal distribution to the parameters for a log-normal distribution with the same mean and variance

# Usage

norm2lnorm(m, v)

# Arguments

m	Mean of the normal distribution
v	Variance of the normal distribution

# Value

A vector of length two. The first element is the mean and the second element is the variance of the log-normal distribution

# Examples

```
norm <- rnorm(1000, mean=5, sd=2)
params <- norm2lnorm(5, 2^2)
lnorm <- rlnorm(1000, meanlog=params[1], sdlog=params[2]^0.5)
# Mean and SD of lnorm is equivalent to mean and sd of norm
mean(lnorm)
sd(lnorm)</pre>
```

### Description

A dataset from a systematic review of interventions for pain relief in osteoarthritis, used previously in Pedder et al. (2019). The outcome is continuous, and aggregate data responses correspond to the mean WOMAC pain score at 2 weeks follow-up. The dataset includes 18 Randomised-Controlled Trials (RCTs), comparing 8 different agents with placebo. In total there were 26 different treatments (combination of dose and agent). The active treatments can also be grouped into 3 different classes, within which they have similar mechanisms of action.

#### Usage

osteopain

### Format

A data frame in long format (one row per arm and study), with 74 rows and 7 variables:

- studyID Study identifiers
- · agent Character data indicating the agent to which participants were randomised
- · dose Numeric data indicating the standardised dose received
- · class Character data indicating the drug class to which the agent belongs to
- y Numeric data indicating the mean pain score on the WOMAC scale in a study arm
- se Numeric data indicating the standard error for the mean pain score on the WOMAC scale in a study arm
- n Numeric data indicating the number of participants randomised

### Source

Pfizer Ltd.

#### References

Pedder H, Dias S, Bennetts M, Boucher M, Welton NJ (2019). "Modelling time-course relationships with multiple treatments: Model-Based Network Meta-Analysis for continuous summary outcomes." *Res Synth Methods*, **10**(2), 267-286. pDcalc

Calculate plugin pD from a JAGS model with univariate likelihood for studies with repeated measurements

# Description

Uses results from MBNMA JAGS models to calculate pD via the plugin method (Spiegelhalter et al. 2002). Can only be used for models with known standard errors or covariance matrices. Currently only functions with univariate likelihoods. Function is identical in MBNMAdose and MBNMAtime packages.

# Usage

```
pDcalc(
  obs1,
  obs2,
  fups = NULL,
  narm,
  NS,
  theta.result,
  resdev.result,
  likelihood = "normal",
  type = "time"
)
```

obs1	A matrix (study x arm) or array (study x arm x time point) containing observed data for y (normal likelihood) or r (binomial or poisson likelihood) in each arm of each study. This will be the same array used as data for the JAGS model.
obs2	A matrix (study x arm) or array (study x arm x time point) containing observed data for se (normal likelihood), n (binomial likelihood) or E (poisson likelihood) in each arm of each study. This will be the same array used as data for the JAGS model.
fups	A numeric vector of length equal to the number of studies, containing the number of follow-up mean responses reported in each study. Required for time-course MBNMA models (if type="time")
narm	A numeric vector of length equal to the number of studies, containing the number of arms in each study.
NS	A single number equal to the number of studies in the dataset.
theta.result	A matrix (study x arm) or array (study x arm x time point) containing the pos- terior mean predicted means/probabilities/rate in each arm of each study. This will be estimated by the JAGS model.
resdev.result	A matrix (study x arm) or array (study x arm x time point) containing the pos- terior mean residual deviance contributions in each arm of each study. This will be estimated by the JAGS model.

#### pDcalc

likelihood	A character object of any of the following likelihoods:
	• normal
	<ul> <li>binomial (does not work with time-course MBNMA models)</li> </ul>
	• poisson (does not work with time-course MBNMA models)
type	The type of MBNMA model fitted. Can be either "time" or "dose"

# Details

Method for calculating pD via the plugin method proposed by Spiegelhalter (Spiegelhalter et al. 2002). Standard errors / covariance matrices must be assumed to be known. To obtain values for theta.result and resdev.result these parameters must be monitored when running the MB-NMA model (using parameters.to.save).

For non-linear time-course MBNMA models residual deviance contributions may be skewed, which can lead to non-sensical results when calculating pD via the plugin method. Alternative approaches are to use pV as an approximation or pD calculated by Kullback-Leibler divergence (Plummer 2008).

### Value

A single numeric value for pD calculated via the plugin method.

# References

Plummer M (2008). "Penalized loss functions for Bayesian model comparison." *Biostatistics*, **9**(3), 523-39. ISSN 1468-4357 (Electronic) 1465-4644 (Linking), https://pubmed.ncbi.nlm.nih. gov/18209015/.

Spiegelhalter DJ, Best NG, Carlin BP, van der Linde A (2002). "Bayesian measures of model complexity and fit." *J R Statistic Soc B*, **64**(4), 583-639.

#### Examples

resdevs <- result\$BUGSoutput\$median\$resdev</pre>

plot.mbnma

Forest plot for results from dose-response MBNMA models

### Description

Generates a forest plot for dose-response parameters.

# Usage

## S3 method for class 'mbnma'
plot(x, params = NULL, ...)

#### Arguments

х	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
params	A character vector of dose-response parameters to plot. Parameters must be given the same name as monitored nodes in mbnma and must be modelled as relative effects ("rel"). Can be set to NULL to include all available dose-response parameters estimated by mbnma.
	Arguments to be passed to methods, such as graphical parameters

# Value

A forest plot of class c("gg", "ggplot") that has separate panels for different dose-response parameters. Results are plotted on the link scale.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
```

```
# Run an exponential dose-response MBNMA and generate the forest plot
exponential <- mbnma.run(network, fun=dexp())
plot(exponential)
```

```
# Plot only Emax parameters from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")
plot(emax, params=c("emax"))</pre>
```

plot.mbnma.network Create an mbnma.network object

# Description

Creates an object of class("mbnma.network"). Various MBNMA functions can subsequently be applied to this object.

# Usage

```
## S3 method for class 'mbnma.network'
plot(
  х,
  level = "treatment",
  v.color = "connect",
  doselink = NULL,
  layout = igraph::in_circle(),
  remove.loops = FALSE,
  edge.scale = 1,
  v.scale = NULL,
  label.distance = 0,
  legend = TRUE,
  legend.x = "bottomleft",
  legend.y = NULL,
  . . .
)
```

mbnma.network(data.ab, description = "Network")

х	An object of class mbnma.network.
level	A string indicating whether nodes/facets should represent "treatment" or "agent" in the plot. Can be used to examine the expected impact of modelling dose-response in terms of network connectivity.

v.color	Can take either "connect" (the default) to indicate that nodes should only be coloured if they are connected to the network reference treatment (indicates network connectivity) or "agent" to colour nodes by agent.
doselink	If given an integer value it indicates that connections via the dose-response re- lationship with placebo should be plotted. The integer represents the minimum number of doses from which a dose-response function could be estimated and is equivalent to the number of parameters in the desired dose-response function plus one. If left as NULL (the default), connections to placebo via dose-response relationships will not be included.
layout	An igraph layout specification. This is a function specifying an igraph layout that determines the arrangement of the vertices (nodes). The default igraph::as_circle() arranged vertices in a circle. Two other useful layouts for network plots are: igraph::as_star(), igraph::with_fr(). Others can be found in layout_
remove.loops	A boolean value indicating whether to include loops that indicate comparisons within a node.
edge.scale	A number to scale the thickness of connecting lines (edges). Line thickness is proportional to the number of studies for a given comparison. Set to 0 to make thickness equal for all comparisons.
v.scale	A number with which to scale the size of the nodes. If the variable N (to indicate the numbers of participants in each study arm) is included in the dataset then the size of the nodes will be proportional to the number of participants within a treatment/agent in the network.
label.distance	A number scaling the distance of labels from the nodes to improve readability. The labels will be directly on top of the nodes if the default of 0 is used. Option only applicable if layout_in_circle is set to TRUE.
legend	A boolean object to indicate whether or not to plot a legend to indicate which node colour corresponds to which agent if v.color="agent". Default is TRUE.
legend.x, legend	. у
	The x and y co-ordinates to be used to position the legend. They can be specified by keyword or in any way which is accepted by xy.coords.
	Options for plotting in igraph.
data.ab	A data frame of arm-level data in "long" format containing the columns:
	studyID Study identifiers
	<ul> <li>dose Numeric data indicating the dose (must take positive values)</li> </ul>
	• agent Agent identifiers (can be numeric, factor or character)
	<ul> <li>y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data.</li> </ul>
	• se Numeric data indicating the standard error for a given observation. Re- quired for continuous data.
	<ul> <li>r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data.</li> </ul>
	<ul> <li>n Numeric data indicating the total number of participants within a study arm. Required for binomial data or when modelling Standardised Mean Differences</li> </ul>

- E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data.
- class An optional column indicating a particular class code. Agents with the same identifier must also have the same class code.
- standsd An optional column of numeric data indicating reference SDs used to standardise treatment effects when modelling using Standardised Mean Differences (SMD).

description Optional. Short description of the network.

# Details

The S3 method plot() on an mbnma.network object generates a network plot that shows how different treatments are connected within the network via study comparisons. This can be used to identify how direct and indirect evidence are informing different treatment comparisons. Depends on igraph.

Agents/classes for arms that have dose = 0 will be relabelled as "Placebo". Missing values (NA) cannot be included in the dataset. Single arm studies cannot be included.

## Value

plot(): An object of class("igraph") - any functions from the igraph package can be applied to this object to change its characteristics.

mbnma.network(): An object of class("mbnma.network") which is a list containing:

- description A short description of the network
- data.ab A data frame containing the arm-level network data (treatment identifiers will have been recoded to a sequential numeric code)
- studyID A character vector with the IDs of included studies
- agents A character vector indicating the agent identifiers that correspond to the new agent codes.
- treatments A character vector indicating the treatment identifiers that correspond to the new treatment codes.
- classes A character vector indicating the class identifiers (if included in the original data) that correspond to the new class codes.

# Methods (by generic)

• plot(mbnma.network): Generate a network plot

#### Examples

```
# Create an mbnma.network object from the data
network <- mbnma.network(triptans)</pre>
```

```
# Generate a network plot from the data
plot(network)
```

```
# Generate a network plot at the agent level that removes loops indicating comparisons
#within a node
plot(network, level="agent", remove.loops=TRUE)
# Generate a network plot at the treatment level that colours nodes by agent
plot(network, v.color="agent", remove.loops=TRUE)
# Generate a network plot that includes connections via the dose-response function
# For a one parameter dose-response function (e.g. exponential)
plot(network, level="treatment", doselink=1, remove.loops=TRUE)
# For a two parameter dose-response function (e.g. Emax)
plot(network, level="treatment", doselink=2, remove.loops=TRUE)
# Arrange network plot in a star with the reference treatment in the centre
plot(network, layout=igraph::as_star(), label.distance=3)
#### Plot a network with no placebo data included ####
# Make data with no placebo
noplac.df <- network$data.ab[network$data.ab$narm>2 & network$data.ab$agent!=1,]
net.noplac <- mbnma.network(noplac.df)</pre>
# Plotting network automatically plots connections to Placebo via dose-response
plot(net.noplac)
# Using the triptans headache dataset
print(triptans)
# Define network
network <- mbnma.network(triptans, description="Example network")</pre>
summary(network)
plot(network)
```

plot.mbnma.predict Plots predicted responses from a dose-response MBNMA model

# Description

Plots predicted responses on the natural scale from a dose-response MBNMA model.

#### Usage

```
## S3 method for class 'mbnma.predict'
plot(
    x,
    disp.obs = FALSE,
    overlay.split = FALSE,
    method = "common",
    agent.labs = NULL,
```

```
scales = "free_x",
...
)
```

### Arguments

x	An object of class "mbnma.predict" generated by predict("mbnma")
disp.obs	A boolean object to indicate whether to show the location of observed doses in the data on the 95\% credible intervals of the predicted dose-response curves as shaded regions (TRUE) or not (FALSE). If set to TRUE the original network object used for the model <b>must</b> be specified in network.
overlay.split	A boolean object indicating whether to overlay a line showing the split (treatment- level) NMA results on the plot (TRUE) or not (FALSE). This will require automatic running of a split NMA model. For overlay.split=TRUE the original network object used for the model <b>must</b> be specified in network.
method	Indicates the type of split (treatment-level) NMA to perform when overlay.split=TRUE. Can take either "common" or "random".
agent.labs	A character vector of agent labels to display on plots. If left as NULL (the default) the names of agents will be taken from predict. The position of each label corresponds to each element of predict. The number of labels must equal the number of active agents in predict. If placebo / dose=0 data is included in the predictions then a label for placebo <b>should not</b> be included in agent.labs. It will not be shown in the final plot since placebo is the point within each plot at which dose = 0 (rather than a separate agent).
scales	Should scales be fixed ("fixed", the default), free ("free"), or free in one dimension ("free_x", "free_y")?
	Arguments for ggplot2

# Details

For the S3 method plot(), it is advisable to ensure predictions in predict are estimated using a sufficient number of doses to ensure a smooth predicted dose-response curve. If disp.obs = TRUE it is advisable to ensure predictions in predict are estimated using an even sequence of time points to avoid misrepresentation of shaded densities.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
```

```
# Run an Emax dose-response MBNMA and predict responses
emax <- mbnma.run(network, fun=demax(), method="random")
pred <- predict(emax, E0 = 0.5)
plot(pred)
# Display observed doses on the plot</pre>
```

```
# Display split NMA results on the plot
plot(pred, overlay.split=TRUE)
# Split NMA results estimated using random treatment effects model
plot(pred, overlay.split=TRUE, method="random")
# Add agent labels
plot(pred, agent.labs=c("Elet", "Suma", "Frov", "Almo", "Zolmi",
      "Nara", "Riza"))
# These labels will throw an error because "Placebo" is included in agent.labs when
#it will not be plotted as a separate panel
#### ERROR ####
#plot(pred, agent.labs=c("Placebo", "Elet", "Suma", "Frov", "Almo", "Zolmi",
       "Nara", "Riza"))
#
# If insufficient predictions are made across dose-response function
# then the plotted responses are less smooth and can be misleading
pred <- predict(emax, E0 = 0.5, n.doses=3)</pre>
plot(pred)
```

plot.mbnma.rank Plot histograms of rankings from MBNMA models

# Description

Plot histograms of rankings from MBNMA models

#### Usage

```
## S3 method for class 'mbnma.rank'
plot(x, params = NULL, treat.labs = NULL, ...)
```

# Arguments

х	An object of class "mbnma.rank" generated by rank.mbnma()
params	A character vector of named parameters in the model that vary by either agent or class (depending on the value assigned to level). If left as NULL (the de- fault), then ranking will be calculated for all available parameters that vary by agent/class.
treat.labs	A vector of treatment labels in the same order as treatment codes. Easiest to use treatment labels stored by mbnma.network()
	Arguments to be sent to ggplot::geom_bar()

### plot.nma

#### Value

A series of histograms that show rankings for each treatment/agent/prediction, with a separate panel for each parameter. The object returned is a list containing a separate element for each parameter in params which is an object of class(c("gg", "ggplot")).

#### Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
# Estimate rankings from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")</pre>
ranks <- rank(emax)</pre>
# Plot rankings for both dose-response parameters (in two separate plots)
plot(ranks)
# Plot rankings just for ED50
plot(ranks, params="ed50")
# Plot rankings from prediction
doses <- list("eletriptan"=c(0,1,2,3), "rizatriptan"=c(0.5,1,2))</pre>
pred <- predict(emax, E0 = "rbeta(n, shape1=1, shape2=5)",</pre>
            exact.doses=doses)
rank <- rank(pred)</pre>
plot(rank)
# Trying to plot a parameter that has not been ranked will return an error
#### ERROR ####
# plot(ranks, params="not.a.parameter")
```

plot.nma

Run an NMA model

### Description

Used for calculating treatment-level NMA results, either when comparing MBNMA models to models that make no assumptions regarding dose-response, or to estimate split results for overlay.split. Results can also be compared between consistency (UME=FALSE) and inconsistency (UME=TRUE) models to test the validity of the consistency assumption at the treatment-level.

#### Usage

```
## S3 method for class 'nma'
plot(x, bydose = TRUE, scales = "free_x", ...)
```

nma.run(

# plot.nma

```
network,
method = "common",
likelihood = NULL,
link = NULL,
priors = NULL,
sdscale = FALSE,
warn.rhat = TRUE,
n.iter = 20000,
drop.discon = TRUE,
UME = FALSE,
pd = "pd.kl",
parameters.to.save = NULL,
....)
```

# Arguments

х	An object of class("nma")
bydose	A boolean object indicating whether to plot responses with dose on the x-axis (TRUE) to be able to examine potential dose-response shapes, or to plot a conventional forest plot with all treatments on the same plot (FALSE)
scales	Should scales be fixed ("fixed", the default), free ("free"), or free in one dimension ("free_x", "free_y")?
	Arguments to be sent to ggplot2::ggplot()
network	An object of class mbnma.network.
method	Indicates the type of split (treatment-level) NMA to perform when overlay.split=TRUE. Can take either "common" or "random".
likelihood	A string indicating the likelihood to use in the model. Can take either "binomial", "normal" or "poisson". If left as NULL the likelihood will be inferred from the data.
link	A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.
priors	A named list of parameter values (without indices) and replacement prior distribution values given as strings <b>using distributions as specified in JAGS syntax</b> (see Plummer (2017)). Note that normal distributions in JAGS are specified as
	$N(\mu, prec)$

, where

.

 $prec = 1/\sigma^2$ 

sdscale	Logical object to indicate whether to write a model that specifies a reference SD for standardising when modelling using Standardised Mean Differences. Specifying sdscale=TRUE will therefore only modify the model if link function is set to SMD (link="smd").
warn.rhat	A boolean object to indicate whether to return a warning if Rhat values for any monitored parameter are >1.02 (suggestive of non-convergence).
n.iter	number of total iterations per chain (including burn in; default: 20000)
drop.discon	A boolean object that indicates whether or not to drop disconnected studies from the network.
UME	A boolean object to indicate whether to fit an Unrelated Mean Effects model that does not assume consistency and so can be used to test if the consistency assumption is valid.
pd	Can take either:
	• pv only pV will be reported (as automatically outputted by R2jags).
	• plugin calculates pD by the plug-in method (Spiegelhalter et al. 2002). It is faster, but may output negative non-sensical values, due to skewed deviances that can arise with non-linear models.
	• pd.kl calculates pD by the Kullback-Leibler divergence (Plummer 2008). This will require running the model for additional iterations but is a more robust calculation for the effective number of parameters in non-linear mod- els.
	• popt calculates pD using an optimism adjustment which allows for calculation of the penalized expected deviance (Plummer 2008).
parameters.to.save	

A character vector containing names of parameters to monitor in JAGS

# Functions

• plot(nma): Plot outputs from treatment-level NMA models

Results can be plotted either as a single forest plot, or facetted by agent and plotted with increasing dose in order to identify potential dose-response relationships. If Placebo (or any agents with dose=0) is included in the network then this will be used as the reference treatment, but if it is not then results will be plotted versus the network reference used in the NMA object (x).

# Examples

```
# Run random effects NMA on the alogliptin dataset
alognet <- mbnma.network(alog_pcfb)
nma <- nma.run(alognet, method="random")
print(nma)
plot(nma)
# Run common effects NMA keeping treatments that are disconnected in the NMA
goutnet <- mbnma.network(gout)
nma <- nma.run(goutnet, method="common", drop.discon=FALSE)</pre>
```

```
# Run an Unrelated Mean Effects (UME) inconsistency model on triptans dataset
tripnet <- mbnma.network(triptans)
ume <- nma.run(tripnet, method="random", UME=TRUE)</pre>
```

predict.mbnma

Predict responses for different doses of agents in a given population based on MBNMA dose-response models

# Description

Used to predict responses for different doses of agents or to predict the results of a new study. This is calculated by combining relative treatment effects with a given reference treatment response (specific to the population of interest).

### Usage

```
## S3 method for class 'mbnma'
predict(
   object,
   n.doses = 30,
   exact.doses = NULL,
   E0 = 0.2,
   synth = "fixed",
   lim = "cred",
   regress.vals = NULL,
   ...
)
```

# Arguments

object	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
n.doses	A number indicating the number of doses at which to make predictions within each agent. The default is 30.
exact.doses	A list of numeric vectors. Each named element in the list corresponds to an agent (either named similarly to agent names given in the data, or named correspondingly to the codes for agents given in mbnma) and each number within the vector for that element corresponds to a dose of the agent for which to predict responses. Doses can only take positive values. For models fitted using dspline() making predictions at only a very small number of doses for each agent may throw an error since it can make the spline difficult to identify.
EØ	An object to indicate the value(s) to use for the response at dose = 0 (i.e. placebo) in the prediction. This can take a number of different formats depending on how it will be used/calculated. The default is $0.2$ since a default of 0 will typically lead to non-sensical predictions unless an identify link function has been used for the MBNMA model in object.

	<ul> <li>numeric() A single numeric value representing the deterministic response at dose = 0, given on the natural scale - so for binomial data, proportions should be given and for Poisson data, a rate should be given.</li> <li>character() A single string representing a stochastic distribution for the response at dose = 0, given on the natural scale - so for binomial data, proportions should be given and for Poisson data, a rate should be given. This is specified as a random number generator (RNG) given as a string, and can take any RNG distribution for which a function exists in R. For example: "rnorm(n, 7, 0.5)".</li> <li>data.frame() A data frame containing data in the long format (one row per string).</li> </ul>
	study arm) to be meta-analysed to estimate the dose = $0$ (placebo) response. This could be a set of observational studies that are specific to the popula- tion on which to make predictions, or it can be a subset of the study arms within the MBNMA dataset that investigate placebo. See ref.synth()
synth	A character object that can take the value "fixed" or "random" to specify the the type of pooling to use for synthesis of E0 if a data frame has been provided for it. Using "random" rather than "fixed" for synth will result in wider 95\% CrI for predictions.
lim	Specifies calculation of either 95% credible intervals (lim="cred") or 95% pre- diction intervals (lim="pred").
regress.vals	A named numeric vector of effect modifier values at which results should be pre- dicted. Named elements must match variable names specified in regress.vars within the MBNMA model.
	Arguments to be sent to R2jags::jags() for synthesis of the network reference treatment effect (using ref.synth())

# Details

The range of doses on which to make predictions can be specified in one of two ways:

- 1. Use max.dose and n.doses to specify the maximum dose for each agent and the number of doses within that agent for which to predict responses. Doses will be chosen that are equally spaced from zero to the maximum dose for each agent. This is useful for generating plots of predicted responses (using [plot-mbnma.predict]) as it will lead to fitting a smooth doseresponse curve (provided n.doses is sufficiently high).
- 2. Use exact.doses to specify the exact doses for which to predict responses for each agent. This may be more useful when ranking different predicted responses using [rank-mbnma.predict]

## Value

An S3 object of class mbnma.predict that contains the following elements:

- predicts A named list of matrices. Each matrix contains the MCMC results of predicted responses at follow-up times specified in times for each treatment specified in treats
- likelihood The likelihood used in the MBNMA model object
- link The link function used in the MBNMA model object
- network The dataset in mbnma.network format
- E0 A numeric vector of value(s) used for E0 in the prediction, on the link scale.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
```

```
# Run an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")</pre>
```

```
#### Predict responses using deterministic value for E0 ####
# Data is binomial so we specify E0 on the natural scale as a probability
pred <- predict(emax, E0 = 0.2)</pre>
```

```
# Specifying non-sensical values will return an error
#pred <- predict(emax, E0 = -10)
### ERROR ###
```

```
#### Predict responses using stochastic value for E0 ####
# Data is binomial so we might want to draw from a beta distribution
pred <- predict(emax, E0 = "rbeta(n, shape1=1, shape2=5)")</pre>
```

```
# Misspecifying the RNG string will return an error
#pred <- predict(emax, E0 = "rbeta(shape1=1, shape2=5)")
### ERROR ###
```

```
#### Predict responses using meta-analysis of dose = 0 studies ####
```

```
# E0 is assigned a data frame of studies to synthesis
# Can be taken from placebo arms in triptans dataset
ref.df <- network$data.ab[network$data.ab$agent==1,]</pre>
```

```
# Synthesis can be fixed/random effects
pred <- predict(emax, E0 = ref.df, synth="random")</pre>
```

```
# Change the number of predictions for each agent
pred <- predict(emax, E0 = 0.2, n.doses=20)
pred <- predict(emax, E0 = 0.2, n.doses=3)</pre>
```

## print.mbnma.network

```
# Print and summarise `mbnma.predict` object
print(pred)
summary(pred)
# Plot `mbnma.predict` object
plot(pred)
```

print.mbnma.network Print mbnma.network information to the console

## Description

Print mbnma.network information to the console

### Usage

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

## Arguments

Х	An object of class mbnma.network.
	further arguments passed to or from other methods

print.mbnma.predict Print summary information from an mbnma.predict object

## Description

Print summary information from an mbnma.predict object

#### Usage

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

х	An object of class("mbnma.predict") generated by predict.mbnma()
	further arguments passed to or from other methods

print.mbnma.rank Prints summary information about an mbnma.rank object

## Description

Prints summary information about an mbnma.rank object

# Usage

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

# Arguments

х	An object of class "mbnma.rank" generated by rank.mbnma()
	further arguments passed to or from other methods

print.nma.nodesplit Prints summary results from an nma.nodesplit object

# Description

Prints summary results from an nma.nodesplit object

# Usage

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

х	An object of class("nma.nodesplit")
	further arguments passed to or from other methods

print.nodesplit Prints summary results from a nodesplit object

## Description

Prints summary results from a nodesplit object

## Usage

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

## Arguments

x	An object of class("nodesplit")
	further arguments passed to or from other methods

print.relative.array	Print posterior medians (95% credible intervals) for table of relative
	effects/mean differences between treatments/classes

## Description

Print posterior medians (95% credible intervals) for table of relative effects/mean differences between treatments/classes

# Usage

```
## S3 method for class 'relative.array'
print(x, digits = 2, ...)
```

х	An object of class "relative.array" generated by get.relative()
digits	An integer indicating the number of significant digits to be used.
	further arguments passed to knitr::kable

psoriasis100

Studies of biologics for treatment of moderate-to-severe psoriasis (100% improvement)

#### Description

A dataset from a systematic review of Randomised-Controlled Trials (RCTs) comparing biologics at different doses and placebo (Warren et al. 2019). The outcome is the number of patients experiencing 100% improvement on the Psoriasis Area and Severity Index (PASI) measured at 12 weeks follow-up. The dataset includes 19 Randomised-Controlled Trials (RCTs), comparing 8 different biologics at different doses with placebo.

#### Usage

psoriasis100

## Format

A data frame in long format (one row per arm and study), with 81 rows and 9 variables:

- studyID Study identifiers
- · agent Character data indicating the agent to which participants were randomised
- · dose\_mg Numeric data indicating the dose to which participants were randomised in mg
- freq Character data indicating the frequency of the dose to which participants were randomised
- dose Numeric data indicating the dose in mg/week to which the participants were randomised
- n Numeric data indicating the number of participants randomised
- r Numeric data indicating the number of participants who achieved 100% improvement in PASI score after 12 weeks

#### References

Warren RB, Gooderham M, Burge R, Zhu B, Amato D, Liu KH, Shrom D, Guo J, Brnabic A, Blauvelt A (2019). "Comparison of cumulative clinical benefits of biologics for the treatment of psoriasis over 16 weeks: Results from a network meta-analysis." *J Am Acad Dermatol*, **82**(5), 1138-1149.

psoriasis75

Studies of biologics for treatment of moderate-to-severe psoriasis (>=75% improvement)

### Description

A dataset from a systematic review of Randomised-Controlled Trials (RCTs) comparing biologics at different doses and placebo (Warren et al. 2019). The outcome is the number of patients experiencing >=75% improvement on the Psoriasis Area and Severity Index (PASI) measured at 12 weeks follow-up. The dataset includes 28 Randomised-Controlled Trials (RCTs), comparing 9 different biologics at different doses with placebo.

#### Usage

psoriasis75

## Format

A data frame in long format (one row per arm and study), with 81 rows and 9 variables:

- studyID Study identifiers
- · agent Character data indicating the agent to which participants were randomised
- · dose\_mg Numeric data indicating the dose to which participants were randomised in mg
- freq Character data indicating the frequency of the dose to which participants were randomised
- · dose Numeric data indicating the dose in mg/week to which the participants were randomised
- n Numeric data indicating the number of participants randomised
- r Numeric data indicating the number of participants who achieved >=75% improvement in PASI score after 12 weeks

#### References

Warren RB, Gooderham M, Burge R, Zhu B, Amato D, Liu KH, Shrom D, Guo J, Brnabic A, Blauvelt A (2019). "Comparison of cumulative clinical benefits of biologics for the treatment of psoriasis over 16 weeks: Results from a network meta-analysis." *J Am Acad Dermatol*, **82**(5), 1138-1149.

psoriasis90

Studies of biologics for treatment of moderate-to-severe psoriasis (>=90% improvement)

#### Description

A dataset from a systematic review of Randomised-Controlled Trials (RCTs) comparing biologics at different doses and placebo (Warren et al. 2019). The outcome is the number of patients experiencing >=90% improvement on the Psoriasis Area and Severity Index (PASI) measured at 12 weeks follow-up. The dataset includes 24 Randomised-Controlled Trials (RCTs), comparing 9 different biologics at different doses with placebo.

#### Usage

psoriasis90

## Format

A data frame in long format (one row per arm and study), with 81 rows and 9 variables:

- studyID Study identifiers
- · agent Character data indicating the agent to which participants were randomised
- · dose\_mg Numeric data indicating the dose to which participants were randomised in mg
- freq Character data indicating the frequency of the dose to which participants were randomised
- · dose Numeric data indicating the dose in mg/week to which the participants were randomised
- n Numeric data indicating the number of participants randomised
- r Numeric data indicating the number of participants who achieved >=90% improvement in PASI score after 12 weeks

#### References

Warren RB, Gooderham M, Burge R, Zhu B, Amato D, Liu KH, Shrom D, Guo J, Brnabic A, Blauvelt A (2019). "Comparison of cumulative clinical benefits of biologics for the treatment of psoriasis over 16 weeks: Results from a network meta-analysis." *J Am Acad Dermatol*, **82**(5), 1138-1149.

rank

# Description

Set rank as a method

## Usage

rank(x, ...)

# Arguments

х	An object on which to apply the rank method
	Arguments to be passed to methods

rank.r	nbnma
--------	-------

Rank parameter estimates

# Description

Only parameters that vary by agent/class can be ranked.

## Usage

```
## S3 method for class 'mbnma'
rank(
    x,
    params = NULL,
    lower_better = TRUE,
    level = "agent",
    to.rank = NULL,
    ...
)
```

х	An object on which to apply the rank method
params	A character vector of named parameters in the model that vary by either agent or class (depending on the value assigned to level). If left as NULL (the de- fault), then ranking will be calculated for all available parameters that vary by agent/class.
lower_better	Indicates whether negative responses are better (TRUE) or positive responses are better (FALSE)

level	Can be set to "agent" to rank across different agents or "class" to rank across different classes.
to.rank	A numeric vector containing the codes for the agents/classes you wish to rank. If left NULL then all agents/classes (depending on the value assigned to level) in the model will be ranked. Included codes must be greater than 2 if placebo has been modelled, since placebo cannot be included in the ranking
	Arguments to be passed to methods

#### Details

Ranking cannot currently be performed on non-parametric dose-response MBNMA

#### Value

An object of class("mbnma.rank") which is a list containing a summary data frame, a matrix of rankings for each MCMC iteration, a matrix of probabilities that each agent has a particular rank, and a matrix of cumulative ranking probabilities for each agent, for each parameter that has been ranked.

#### Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
# Rank selected agents from a log-linear dose-response MBNMA
loglin <- mbnma.run(network, fun=dloglin())</pre>
ranks <- rank(loglin, to.rank=c("zolmitriptan", "eletriptan", "sumatriptan"))</pre>
summary(ranks)
# Rank only ED50 parameters from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")</pre>
ranks <- rank(emax, params="ed50")</pre>
summary(ranks)
#### Ranking by class ####
# Generate some classes for the data
class.df <- triptans</pre>
class.df$class <- ifelse(class.df$agent=="placebo", "placebo", "active1")</pre>
class.df$class <- ifelse(class.df$agent=="eletriptan", "active2", class.df$class)</pre>
netclass <- mbnma.network(class.df)</pre>
emax <- mbnma.run(netclass, fun=demax(), method="random",</pre>
             class.effect=list("ed50"="common"))
# Rank by class, with negative responses being worse
ranks <- rank(emax, level="class", lower_better=FALSE)</pre>
print(ranks)
# Print and generate summary data frame for `mbnma.rank` object
summary(ranks)
print(ranks)
```

```
# Plot `mbnma.rank` object
plot(ranks)
```

rank.mbnma.predict Rank predicted doses of different agents

## Description

Ranks predictions at different doses from best to worst.

#### Usage

## S3 method for class 'mbnma.predict'
rank(x, lower\_better = TRUE, rank.doses = NULL, ...)

### Arguments

х	An object on which to apply the rank method
lower_better	Indicates whether negative responses are better (TRUE) or positive responses are better (FALSE)
rank.doses	A list of numeric vectors. Each named element corresponds to an agent (as named/coded in predict), and each number within the vector for that element corresponds to the dose for that agent. Doses of agents specified in rank.doses <b>must</b> be a subset of those for which responses have been predicted in predict. If left as NULL (the default) then all doses of all agents in predict will be ranked.
	Arguments to be passed to methods

#### Details

If predict contains multiple predictions at dose=0, then only the first of these will be included, to avoid duplicating rankings.

#### Value

An object of class("mbnma.rank") which is a list containing a summary data frame, a matrix of rankings for each MCMC iteration, and a matrix of probabilities that each agent has a particular rank, for each parameter that has been ranked.

## Examples

```
# Using the triptans data
network <- mbnma.network(triptans)</pre>
# Rank all predictions from a log-linear dose-response MBNMA
loglin <- mbnma.run(network, fun=dloglin())</pre>
pred <- predict(loglin, E0 = 0.5)</pre>
rank <- rank(pred)</pre>
summary(rank)
# Rank selected predictions from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")</pre>
doses <- list("eletriptan"=c(0,1,2,3), "rizatriptan"=c(0.5,1,2))</pre>
pred <- predict(emax, E0 = "rbeta(n, shape1=1, shape2=5)",</pre>
             exact.doses=doses)
rank <- rank(pred,</pre>
             rank.doses=list("eletriptan"=c(0,2), "rizatriptan"=2))
# Print and generate summary data frame for `mbnma.rank` object
summary(rank)
print(rank)
# Plot `mbnma.rank` object
plot(rank)
```

rank.relative.array Rank relative effects obtained between specific doses

#### Description

Ranks "relative.table" objects generated by get.relative().

#### Usage

```
## S3 method for class 'relative.array'
rank(x, lower_better = TRUE, ...)
```

## Arguments

Х	An object on which to apply the rank method
lower_better	Indicates whether negative responses are better (TRUE) or positive responses are better (FALSE)
	Arguments to be passed to methods

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## recode.agent

## Value

An object of class("mbnma.rank") which is a list containing a summary data frame, a matrix of rankings for each MCMC iteration, and a matrix of probabilities that each agent has a particular rank, for each parameter that has been ranked.

#### Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Rank selected predictions from an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")
rels <- get.relative(emax)
rank <- rank(rels, lower_better=TRUE)
# Print and generate summary data frame for `mbnma.rank` object
summary(rank)
print(rank)
# Plot `mbnma.rank` object
plot(rank)
```

recode.agent	Assigns agent or class variables	numeric identifiers
--------------	----------------------------------	---------------------

## Description

Assigns agent or class variables numeric identifiers

#### Usage

```
recode.agent(data.ab, level = "agent")
```

data.ab	A data frame of arm-level data in "long" format containing the columns:
	<ul> <li>studyID Study identifiers</li> </ul>
	• dose Numeric data indicating the dose (must take positive values)
	• agent Agent identifiers (can be numeric, factor or character)
	• y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data.
	• se Numeric data indicating the standard error for a given observation. Re- quired for continuous data.
	• r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data.

	• n Numeric data indicating the total number of participants within a study arm. Required for binomial data or when modelling Standardised Mean Differences
	• E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data.
	• class An optional column indicating a particular class code. Agents with the same identifier must also have the same class code.
	• standsd An optional column of numeric data indicating reference SDs used to standardise treatment effects when modelling using Standardised Mean Differences (SMD).
level	Can take either "agent" or "class"

#### Details

Also relabels the agent for any arms in which dose = 0 to "Placebo\_0"

## Value

A list containing a data frame with recoded agent/class identifiers and a character vector of original agent/class names

ref.synth

Synthesise single arm dose = 0 / placebo studies to estimate E0

## Description

Synthesises single arm studies to estimate E0. Used in predicting responses from a dose-response MBNMA.

#### Usage

```
ref.synth(
   data.ab,
   mbnma,
   synth = "fixed",
   n.iter = mbnma$BUGSoutput$n.iter,
   n.burnin = mbnma$BUGSoutput$n.burnin,
   n.thin = mbnma$BUGSoutput$n.thin,
   n.chains = mbnma$BUGSoutput$n.chains,
   ...
)
```

## ref.synth

#### Arguments

data.ab	A data frame of arm-level data in "long" format containing the columns:
	• studyID Study identifiers
	• y Numeric data indicating the aggregate response for a continuous outcome. Required for continuous data.
	• se Numeric data indicating the standard error for a given observation. Re- quired for continuous data.
	• r Numeric data indicating the number of responders within a study arm. Required for binomial or poisson data.
	• n Numeric data indicating the total number of participants within a study arm. Required for binomial data
	• E Numeric data indicating the total exposure time for participants within a study arm. Required for poisson data.
mbnma	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
synth	A character object that can take the value "fixed" or "random" to specify the the type of pooling to use for synthesis of E0 if a data frame has been provided for it. Using "random" rather than "fixed" for synth will result in wider 95\% CrI for predictions.
n.iter	number of total iterations per chain (including burn in; default: 2000)
n.burnin	length of burn in, i.e. number of iterations to discard at the beginning. Default is n.iter/2, that is, discarding the first half of the simulations. If n.burnin is 0, jags() will run 100 iterations for adaption.
n.thin	thinning rate. Must be a positive integer. Set n.thin > 1 to save memory and computation time if n.iter is large. Default is max(1, floor(n.chains * (n.iter-n.burnin) / 1000)) which will only thin if there are at least 2000 simulations.
n.chains	number of Markov chains (default: 3)
	Arguments to be sent to R2jags::jags() for synthesis of the network reference treatment effect (using ref.synth())

# Details

data.ab can be a collection of studies that closely resemble the population of interest intended for the prediction, which could be different to those used to estimate the MBNMA model, and could include single arms of RCTs or observational studies. If other data is not available, the data used to estimate the MBNMA model can be used by selecting only the studies and arms that investigate dose = 0 (placebo).

Defaults for n.iter, n.burnin, n.thin and n.chains are those used to estimate mbnma.

## Value

A list of named elements corresponding to E0 and the between-study standard deviation for E0 if synth="random". Each element contains the full MCMC results from the synthesis.

# Examples

```
# Using the triptans data
network <- mbnma.network(triptans)
# Run an Emax dose-response MBNMA
emax <- mbnma.run(network, fun=demax(), method="random")
# Data frame for synthesis can be taken from placebo arms
ref.df <- triptans[triptans$agent=="placebo",]
# Meta-analyse placebo studies using fixed treatment effects
E0 <- ref.synth(ref.df, emax, synth="fixed")
names(E0)
# Meta-analyse placebo studies using random treatment effects
E0 <- ref.synth(ref.df, emax, synth="random")
names(E0)
```

rescale.link

Rescale data depending on the link function provided

#### Description

Rescale data depending on the link function provided

#### Usage

```
rescale.link(x, direction = "link", link = "logit")
```

### Arguments

х	A numeric vector of data to be rescaled
direction	Can take either "link" to convert data to a particular scale as defined by the link function, or "natural" to return it to the natural scale.
link	A string indicating the link function to use in the model. Can take any link function defined within JAGS (e.g. "logit", "log", "probit", "cloglog"), be assigned the value "identity" for an identity link function, or be assigned the value "smd" for modelling Standardised Mean Differences using an identity link function. If left as NULL the link function will be automatically assigned based on the likelihood.

## Value

A rescaled numeric vector

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ssi\_closure

Studies of wound closure methods to reduce Surgical Site Infections (SSI)

#### Description

A dataset from an ongoing systematic review examining the efficacy of different wound closure methods to reduce surgical site infections (SSI). The outcome is binary and represents the number of patients who experienced a SSI. The dataset includes 129 RCTs comparing 16 different interventions in 6 classes. This dataset is primarily used to illustrate how MBNMAdose can be used to perform different types of network meta-analysis without dose-response information.

#### Usage

ssi\_closure

#### Format

A data frame in long format (one row per arm and study), with 281 rows and 6 variables:

- studyID Study identifiers
- Year Year of publication
- n Numeric data indicating the number of participants randomised
- r Numeric data indicating the number of participants who achieved >50% improvement in depression symptoms
- trt Treatment names, given as character data
- · class Class names, given as character data

ssri

Studies of Selective Serotonin Reuptake Inhibitors (SSRIs) for major depression

#### Description

A dataset from a systematic review examining the efficacy of different doses of SSRI antidepressant drugs and placebo (Furukawa et al. 2019). The response to treatment is defined as a 50% reduction in depressive symptoms after 8 weeks (4-12 week range) follow-up. The dataset includes 60 RCTs comparing 5 different SSRIs with placebo.

#### Usage

ssri

#### Format

A data frame in long format (one row per arm and study), with 145 rows and 8 variables:

- studyID Study identifiers
- bias Risk of bias evaluated on 6 domains
- age Mean participant age
- · weeks Duration of study follow-up
- · agent Character data indicating the agent to which participants were randomised
- · dose Numeric data indicating the dose to which participants were randomised in mg
- n Numeric data indicating the number of participants randomised
- r Numeric data indicating the number of participants who achieved >50% improvement in depression symptoms

#### References

Furukawa TA, Cipriani A, Cowen PJ, Leucht S, Egger M, Salanti G (2019). "Optimal dose of selective serotonin reuptake inhibitors, venlafaxine, and mirtazapine in major depression: a systematic review and dose-response meta-analysis." *Lancet Psychiatry*, **6**, 601-609.

summary.mbnma Print summary of MBNMA results to the console

## Description

Print summary of MBNMA results to the console

#### Usage

```
## S3 method for class 'mbnma'
summary(object, digits = 4, ...)
```

#### Arguments

object	An S3 object of class "mbnma" generated by running a dose-response MBNMA model
digits	The maximum number of digits for numeric columns
	additional arguments affecting the summary produced

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summary.mbnma.network *Print summary mbnma.network information to the console* 

## Description

Print summary mbnma.network information to the console

## Usage

```
## S3 method for class 'mbnma.network'
summary(object, ...)
```

## Arguments

object	An object of class mbnma.network.
	further arguments passed to or from other methods

summary.mbnma.predict Produces a summary data frame from an mbnma.predict object

## Description

Produces a summary data frame from an mbnma.predict object

# Usage

```
## S3 method for class 'mbnma.predict'
summary(object, ...)
```

# Arguments

object	An object of class("mbnma.predict)" generated by predict("mbnma")
	additional arguments affecting the summary produced.

## Value

A data frame containing posterior summary statistics from predicted responses from a dose-response MBNMA model

summary.mbnma.rank Generates summary data frames for an mbnma.rank object

## Description

Generates summary data frames for an mbnma.rank object

# Usage

```
## S3 method for class 'mbnma.rank'
summary(object, ...)
```

# Arguments

object	An object of class("mbnma.rank") generated by rank.mbnma()
	additional arguments affecting the summary produced

#### Value

A list in which each element represents a parameter that has been ranked in mbnma.rank and contains a data frame of summary ranking results.

summary.nma.nodesplit Generates a summary data frame for nma.nodesplit objects

## Description

Generates a summary data frame for nma.nodesplit objects

#### Usage

```
## S3 method for class 'nma.nodesplit'
summary(object, ...)
```

object	An object of class("nma.nodesplit")
	further arguments passed to or from other methods

summary.nodesplit Generates a summary data frame for nodesplit objects

#### Description

Generates a summary data frame for nodesplit objects

#### Usage

```
## S3 method for class 'nodesplit'
summary(object, ...)
```

#### Arguments

object	An object of class("nodesplit")
	further arguments passed to or from other methods

triptans

Studies of triptans for headache pain relief

#### Description

A dataset from a systematic review of interventions for pain relief in migraine (Thorlund et al. 2014). The outcome is binary, and represents (as aggregate data) the proportion of participants who were headache-free at 2 hours. Data are from patients who had had at least one migraine attack, who were not lost to follow-up, and who did not violate the trial protocol. The dataset includes 70 Randomised-Controlled Trials (RCTs), comparing 7 triptans with placebo. Doses are standardised as relative to a "common" dose, and in total there are 23 different treatments (combination of dose and agent).

#### Usage

triptans

#### Format

A data frame in long format (one row per arm and study), with with 181 rows and 6 variables:

- studyID Study identifiers
- AuthorYear The author and year published of the study
- n Numeric data indicating the number of participants in a study arm
- r Numeric data indicating the number of responders (headache free at 2 hours) in a study arm
- · dose Numeric data indicating the standardised dose received
- · agent Factor data indicating the agent to which participants were randomised

# Source

There are no references for Rd macro \insertAllCites on this help page.

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