

Package ‘remiod’

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Type Package

Title Reference-Based Multiple Imputation for Ordinal/Binary Response

Version 1.0.0

Description Reference-based multiple imputation of ordinal and binary responses under Bayesian framework, as described in Wang and Liu (2022) <[arXiv:2203.02771](https://arxiv.org/abs/2203.02771)>. Methods for missing-not-at-random include Jump-to-Reference (J2R), Copy Reference (CR), and Delta Adjustment which can generate tipping point analysis.

License GPL (>= 2)

URL <https://github.com/xsswang/remiod>

RoxygenNote 7.1.2

LazyData true

Imports JointAI, rjags, coda, foreach, data.table, future, doFuture, mathjaxr, survival, ggplot2, ordinal, progressr, Matrix, mcmcse, reshape2

SystemRequirements JAGS (<http://mcmc-jags.sourceforge.net/>)

Suggests knitr, rmarkdown, bookdown, ggpubr, testthat (>= 3.0.0), spelling

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Language en-US

Depends R (>= 2.10)

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R topics documented:

clm_MI_CR	2
clm_MI_delta	3
clm_MI_J2R	4
extract_MIdata	5
get_MI_RB	6
get_Mlist	7
glm_MI_CR	7
glm_MI_delta	8
glm_MI_J2R	9
list.models	10
mcmcplot	11
remiod	12
schizo	15
schizob	16
schizow	17
summary	17
tang_MI_RB	19

Index	21
--------------	-----------

clm_MI_CR	<i>Apply Copy-Reference(CR) Method to Update JAGS MCMC outputs under MAR for Cumulative Logistic Model</i>
-----------	--

Description

Internal function to obtain Copy-Reference(CR) MCMC from an MAR object.

Usage

```
clm_MI_CR(object, treatment, start = NULL, end = NULL, thin = NULL,
  exclude_chains = NULL, subset = FALSE, seed = NULL, mess = FALSE,
  ...)
```

Arguments

object	an object of class remoid
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
start	first iteration to be used.
end	last iteration to be used.
thin	thinning to be applied.
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output.

subset	subset of parameters (columns of the mcmc object) to be used.
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters. A subset of the MCMC sample can be selected using start, end and thin.

clm_MI_delta	<i>Apply Delta adjustment to Update JAGS MCMC outputs under MAR for Cumulative Logistic Model</i>
--------------	---

Description

Internal function to obtain delta-adjusted MCMC from an MAR object.

Usage

```
clm_MI_delta(object, treatment, delta = 0, start = NULL, end = NULL,
  thin = NULL, exclude_chains = NULL, subset = FALSE, seed = NULL,
  mess = FALSE, ...)
```

Arguments

object	an object of class remiod
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
delta	specific value used for Delta adjustment, applicable only for method="delta".
start	first iteration to be used.
end	last iteration to be used.
thin	thinning to be applied.
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output.
subset	subset of parameters (columns of the mcmc object) to be used.
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters. A subset of the MCMC sample can be selected using start, end and thin.

clm_MI_J2R	<i>Apply Jump-to-Reference(J2R) Method to Update JAGS MCMC outputs under MAR for Cumulative Logistic Model</i>
------------	--

Description

Internal function to obtain Jump-to-Reference(J2R) MCMC from an MAR object.

Usage

```
clm_MI_J2R(object, treatment, start = NULL, end = NULL, thin = NULL,
  exclude_chains = NULL, subset = FALSE, seed = NULL, mess = FALSE,
  ...)
```

Arguments

object	an object of class remiod
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
start	first iteration to be used.
end	last iteration to be used.
thin	thinning to be applied.
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output.
subset	subset of parameters (columns of the mcmc object) to be used.
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters .A subset of the MCMC sample can be selected using start, end and thin.

extract_MIdata	<i>Extract a specified number of multiple imputed datasets</i>
----------------	--

Description

Extract a specified number of multiple imputed datasets

Usage

```
extract_MIdata(object, method = c("MAR", "J2R", "CR", "delta"), delta = 0,
  mi.setting = NULL, M = 10, minspace = 2, mess = FALSE)
```

Arguments

object	object inheriting from class 'remoid'
method	a method for obtaining multiple-imputed dataset. Options include MAR, J2R, CR, and delta adjustment. Default is MAR.
delta	specific value used for Delta adjustment, applicable only for method="delta".
mi.setting	a list of arguments related to multiple imputation, including trtvar, algorithm, method, include, exclude_chains, thin, start, end, and seed.
M	number of imputed datasets
minspace	minimum number of iterations between iterations to be chosen as imputed values (to prevent strong correlation between imputed datasets in the case of high autocorrelation of the MCMC chains).
mess	logical; should messages be given? Default is TRUE.

Value

A data.frame in which the imputed datasets are stacked onto each other.

Examples

```
# data(schizow)

test = remiod(formula = y6 ~ tx + y0 + y1 + y3, data = schizow,
  trtvar = 'tx', algorithm = 'jags', method="MAR",
  ord_cov_dummy = FALSE, n.adapt = 10, n.chains = 1,
  n.iter = 100, thin = 2, warn = FALSE, seed = 1234)

extdt = extract_MIdata(object=test, method="J2R",mi.setting=NULL, M=10, minspace=2)
```

get_MI_RB	<i>Create multiple imputed datasets based on assigned imputation method.</i>
-----------	--

Description

Internal function, creates multiple imputed datasets based on assigned imputation method returns multiple imputed datasets stacked onto each other (i.e., long format; optionally including the original, incomplete data).

Usage

```
get_MI_RB(object, treatment, method = c("MAR", "J2R", "CR", "delta"),
  delta = 0, exclude_chains = NULL, start = NULL, end = NULL,
  seed = NULL, thin = NULL, subset = FALSE, include = TRUE,
  mess = TRUE, ...)
```

Arguments

object	an object of class JointAI
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
method	a method for obtaining multiple-imputed dataset. Options include MAR, J2R, CR, and Delta adjustment.
delta	specific value used for Delta adjustment, applicable only for method="delta".
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output.
start	first iteration to be used.
end	last iteration to be used.
seed	optional seed value.
thin	thinning to be applied.
subset	subset of parameters (columns of the mcmc object) to be used.
include	should the original, incomplete data be included? Default is TRUE.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A data.frame in which the original data (if include = TRUE) and the imputed datasets are stacked onto each other.

The variable Imputation_ indexes the imputation, while .rownr links the rows to the rows of the original data. In cross-sectional datasets the variable .id is added as subject identifier.

get_Mlist	<i>Prepare imputation-model-related information</i>
-----------	---

Description

Internal function to extract information of imputation models.

Usage

```
get_Mlist(object)
```

Arguments

object object inheriting from class remiod.

Value

a list include raw data, imputation models, model types, fixed effects, random effects if any, reference categories corresponding to categorical variables in models, and interaction terms.

glm_MI_CR	<i>Apply Copy-Reference(CR) Method to Update JAGS MCMC outputs under MAR for Generalized Linear Model</i>
-----------	---

Description

Internal function to obtain Copy-Reference(CR) MCMC from an MAR object.

Usage

```
glm_MI_CR(object, treatment, start = NULL, end = NULL, thin = NULL,
  exclude_chains = NULL, subset = FALSE, seed = 5432, mess = FALSE,
  ...)
```

Arguments

object an object of class remiod

treatment the variable name of treatment. Reference level of treatment should be coded as 0.

start first iteration to be used

end last iteration to be used

thin thinning to be applied

exclude_chains optional vector of numbers, indexing MCMC chains to be excluded from the output

subset	subset of parameters (columns of the mcmc object) to be used
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters. A subset of the MCMC sample can be selected using start, end and thin.

glm_MI_delta	<i>Apply Delta adjustment to Update JAGS MCMC outputs under MAR for Generalized Linear Model</i>
--------------	--

Description

Internal function to obtain delta-adjusted MCMC from an MAR object.

Usage

```
glm_MI_delta(object, treatment, delta = 0, start = NULL, end = NULL,
             thin = NULL, exclude_chains = NULL, subset = FALSE, seed = 5432,
             mess = FALSE, ...)
```

Arguments

object	an object of class remiod
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
delta	specific value used for Delta adjustment, applicable only for method="delta".
start	first iteration to be used
end	last iteration to be used
thin	thinning to be applied
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output
subset	subset of parameters (columns of the mcmc object) to be used
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters. A subset of the MCMC sample can be selected using start, end and thin.

glm_MI_J2R	<i>Apply Jump-to-Reference(J2R) Method to Update JAGS MCMC outputs under MAR for Generalized Linear Model</i>
------------	---

Description

Internal function to obtain Jump-to-Reference(J2R) MCMC from an MAR object.

Usage

```
glm_MI_J2R(object, treatment, start = NULL, end = NULL, thin = NULL,
  exclude_chains = NULL, subset = FALSE, seed = 5432, mess = FALSE,
  ...)
```

Arguments

object	an object of class remiod
treatment	the variable name of treatment. Reference level of treatment should be coded as 0.
start	first iteration to be used
end	last iteration to be used
thin	thinning to be applied
exclude_chains	optional vector of numbers, indexing MCMC chains to be excluded from the output
subset	subset of parameters (columns of the mcmc object) to be used
seed	optional seed value.
mess	logical, should messages be displayed?
...	optional arguments pass from main function.

Value

A matrix of MCMC samples with all monitored parameters. A subset of the MCMC sample can be selected using start, end and thin.

list.models	<i>Listing the sequence of models used for imputation</i>
-------------	---

Description

Listing the sequence of models used for imputation

Usage

```
list.models(object, details = FALSE, print = TRUE)
```

Arguments

object	an object of class remiod
details	logical. Default is FALSE, where listing all models in formula format. If TRUE, details of each models will be presented.
print	logical. Default is TRUE to print all imputation models or detailed imputation models.

Value

a list of formula of imputation models. If details=TRUE, information on the conditional distributions of the covariates in each imputation models. Note: the sequence of conditional models together specifies the joint distribution.

Examples

```
# data(schizow)

test = remiod(formula = y6 ~ tx + y0 + y1 + y3, data = schizow,
              trtvar = 'tx', algorithm = 'jags', method="MAR",
              ord_cov_dummy = FALSE, n.adapt = 10, n.chains = 1,
              n.iter = 10, thin = 2, warn = FALSE, seed = 1234)

list.models(test)
```

mcmcplot	<i>Visualizing the posterior sample Creates a set of plots for visually evaluating convergence and mixing of the chains from the MCMC sample of an object of class 'remoid'.</i>
----------	--

Description

Visualizing the posterior sample Creates a set of plots for visually evaluating convergence and mixing of the chains from the MCMC sample of an object of class 'remoid'.

Usage

```
mcmcplot(object, what = c("trace", "density"), subset = c(analysis_main =
  TRUE), outcome = NULL, mi.setting = NULL, nrow = NULL, ncol = NULL,
  use_ggplot = TRUE, mess = TRUE, warn = FALSE, ...)
```

Arguments

object	an object inheriting from class 'remoid'
what	select either trace or density plots from MCMC samples
subset	subset of parameters/variables/nodes (columns in the MCMC sample). Follows the same principle as the argument <code>monitor_params</code> in <code>*_imp</code> .
outcome	optional; vector identifying a subset of sub-models included in the output, either by specifying their indices (using the order used in the list of model formulas), or their names (LHS of the respective model formula as character string)
mi.setting	a list of arguments for extracting MI data set, which will be used to update the one in remoid object. Default is NULL, meaning no update to the mi.setting in remoid object.
nrow	optional; number of rows in the plot layout; automatically chosen if unspecified
ncol	optional; number of columns in the plot layout; automatically chosen if unspecified
use_ggplot	logical; Should ggplot be used instead of the base graphics?
mess	logical; should messages be given? Default is TRUE.
warn	logical; should warnings be given? Default is TRUE.
...	Arguments passed on to <code>graphics::matplot</code>
lty	vector of line types, widths, and end styles. The first element is for the first column, the second element for the second column, etc., even if lines are not plotted for all columns. Line types will be used cyclically until all plots are drawn.
lwd	vector of line types, widths, and end styles. The first element is for the first column, the second element for the second column, etc., even if lines are not plotted for all columns. Line types will be used cyclically until all plots are drawn.

lend vector of line types, widths, and end styles. The first element is for the first column, the second element for the second column, etc., even if lines are not plotted for all columns. Line types will be used cyclically until all plots are drawn.

col vector of colors. Colors are used cyclically.

cex vector of character expansion sizes, used cyclically. This works as a multiple of `par("cex")`. NULL is equivalent to 1.0.

bg vector of background (fill) colors for the open plot symbols given by `pch = 21:25` as in `points`. The default NA corresponds to the one of the underlying function `plot.xy`.

add logical. If TRUE, plots are added to current one, using `points` and `lines`.

verbose logical. If TRUE, write one line of what is done.

Value

plots of traces or densities of MCMC samples for selected parameters in imputation models.

Examples

```
# data(schizow)

test = remiod(formula = y6 ~ tx + y0 + y1 + y3, data = schizow,
             trtvar = 'tx', algorithm = 'jags', method="MAR",
             ord_cov_dummy = FALSE, n.adapt = 10, n.chains = 1,
             n.iter = 10, thin = 2, warn = FALSE, seed = 1234)

p1 = mcmcplot(object=test, what="trace")
```

remiod

Reference-Based Multiple Imputation for Ordinal/Binary Response

Description

Reference-Based Multiple Imputation for Ordinal/Binary Response

Usage

```
remiod(formula, data, trtvar, refcats = NULL, family = NULL,
       method = "MAR", delta = 0, algorithm = c("tang_seq", "jags"),
       rinv = 1e-04, scheme = 2, model_order = NULL, models = NULL,
       ord_cov_dummy = TRUE, n.chains = 2, n.adapt = 100, n.iter = 1000,
       thin = 2, start = NULL, end = NULL, seed = 1234,
       exclude_chains = NULL, subset = NULL, include = FALSE, mess = TRUE,
       warn = FALSE, progress.bar = TRUE, ...)
```

Arguments

formula	a two sided model formula (see formula) or a list of such formulas; (more details below).
data	a data.frame containing the original data (more details below)
trtvar	the name of treatment variable. When necessary, its reference category, i.e. control arm, can be set in <code>refcats</code> argument.
refcats	optional; either one of "first", "last", "largest" (which sets the category for all categorical variables) or a named list specifying which category should be used as reference category per categorical variable. Options are the category label, the category number, or one of "first" (the first category), "last" (the last category) or "largest" (chooses the category with the most observations). Default is "first". If reference categories are specified for a subset of the categorical variables the default will be used for the remaining variables. (See also set_refcat)
family	only for <code>glm_imp</code> and <code>glmm_imp/glmer_imp</code> : a description of the distribution and link function to be used in the model. This can be a character string naming a family function, a family function or the result of a call to a family function. (For more details see below and family .)
method	a method for obtaining multiple-imputed dataset. Options include MAR, J2R, CR, and delta adjustment. Default is MAR.
delta	specific value used for Delta adjustment, applicable only for <code>method="delta"</code> .
algorithm	either algorithm <code>tang_seq</code> proposed by Tang (2018) or <code>jags</code> the original method inherited in JAGS (Plummer 2003).
rinv	a small number used to adjusting Fish information matrix
scheme	scheme of distribution used for proposing coefficients of imputation models. <code>scheme=1</code> : $\beta \sim N(\text{mean} + \text{inv}(I) \cdot \text{score}, \text{inv}(I))$; <code>scheme=2</code> : $\beta \sim N(\text{mean}, \text{inv}(I))$.
model_order	optional. manually specify an order for imputation models.
models	optional; named vector specifying the types of models for (incomplete) covariates. This arguments replaces the argument <code>meth</code> used in earlier versions. If NULL (default) models will be determined automatically based on the class of the respective columns of data.
ord_cov_dummy	optional. specify whether ordinal variables should be treated as categorical variables or continuous variables when they are included as covariates in the sequential imputation models. Default is TRUE, dummy variables will be created accordingly.
n.chains	number of MCMC chains
n.adapt	number of iterations for adaptation of the MCMC samplers (see adapt)
n.iter	number of iterations of the MCMC chain (after adaptation; see coda.samples)
thin	thinning interval (integer; see window.mcmc). For example, <code>thin = 1</code> (default) will keep the MCMC samples from all iterations; <code>thin = 5</code> would only keep every 5th iteration.
start	the first iteration of interest (see window.mcmc)

end	the last iteration of interest (see window.mcmc)
seed	optional; seed value (for reproducibility)
exclude_chains	optional vector of the index numbers of chains that should be excluded
subset	subset of parameters/variables/nodes (columns in the MCMC sample). Follows the same principle as the argument <code>monitor_params</code> and <code>selected_params</code> .
include	logical, if TRUE, raw data will be included in imputed data sets with imputation ID = 0.
mess	logical; should messages be given? Default is TRUE.
warn	logical; should warnings be given? Default is TRUE.
progress.bar	character string specifying the type of progress bar. Possible values are "text" (default), "gui", and "none" (see update). Note: when sampling is performed in parallel it is not possible to display a progress bar.
...	additional, optional arguments
trunc	named list specifying limits of truncation for the distribution of the named incomplete variables (see the vignette ModelSpecification)
hyperpars	list of hyper-parameters, as obtained by <code>default_hyperpars()</code>
scale_vars	named vector of (continuous) variables that will be centred and scaled (such that mean = 0 and sd = 1) when they enter a linear predictor to improve convergence of the MCMC sampling. Default is that all numeric variables and integer variables with >20 different values will be scaled. If set to FALSE no scaling will be done.
custom	named list of JAGS model chunks (character strings) that replace the model for the given variable.
append_data_list	list that will be appended to the list containing the data that is passed to <code>rjags</code> (<code>data_list</code>). This may be necessary if additional data / variables are needed for custom (covariate) models.
progress.bar	character string specifying the type of progress bar. Possible values are "text" (default), "gui", and "none" (see update). Note: when sampling is performed in parallel it is not possible to display a progress bar.
quiet	logical; if TRUE then messages generated by <code>rjags</code> during compilation as well as the progress bar for the adaptive phase will be suppressed, (see jags.model)
keep_scaled_mcmc	should the "original" MCMC sample (i.e., the scaled version returned by <code>coda.samples()</code>) be kept? (The MCMC sample that is re-scaled to the scale of the data is always kept.)
modelname	character string specifying the name of the model file (including the ending, either <code>.R</code> or <code>.txt</code>). If unspecified a random name will be generated.
modeldir	directory containing the model file or directory in which the model file should be written. If unspecified a temporary directory will be created.
overwrite	logical; whether an existing model file with the specified <code><modeldir>/<modelname></code> should be overwritten. If set to FALSE and a model already exists, that model will be used. If unspecified (NULL) and a file exists, the user is asked for input on how to proceed.
keep_model	logical; whether the created JAGS model file should be saved or removed from (FALSE; default) when the sampling has finished.

Value

A list includes (1) Information from JAGS modeling and MCMC samples and (2) A data.frame in which the original data (if include = TRUE) and the imputed datasets are stacked onto each other. The variable Imputation_ indexes the imputation, while .rownr links the rows to the rows of the original data. In cross-sectional datasets the variable .id is added as subject identifier.

Examples

```
data(schizow)

test = remiod(formula = y6 ~ tx + y0 + y1 + y3, data = schizow,
              trtvar = 'tx', algorithm = 'jags', method="MAR",
              ord_cov_dummy = FALSE, n.adapt = 10, n.chains = 1,
              n.iter = 10, thin = 2, warn = FALSE, seed = 1234)
```

schizo

National Institute of Mental Health shizophrenia study

Description

Schizophrenia data from a randomized controlled trial with patients assigned to either drug or placebo group. "Severity of Illness" was measured, at weeks 0, 1, ..., 6, on a four category ordered scale. Most of the observations were made on weeks 0, 1, 3, and 6.

Usage

```
data(schizo)
```

Format

A data frame with 1603 observations on 437 subjects. Five vectors contain information on

id patient ID.

imps79 original response measurements on a numerical scale.

imps79b binary response based on the cut-off value of 3.5 to the measurements on a numerical scale: 0 = normal to mildly ill and 1 = moderately to extremely ill

imps79o ordinal response on a 4 category scale, "normal or borderline mentally ill" < "mildly or moderately ill", "markedly ill", "severely or among the most extremely ill".

tx treatment indicator: 1 for drug, 0 for placebo.

week week.

Details

The documentation file was copied from the **mixcat** package and slightly modified.

Source

<https://hedeker.people.uic.edu/ml.html>

References

Hedeker, D. and R. Gibbons (2006). *Longitudinal Data Analysis*. New Jersey, USA: John Wiley & Sons.

schizob	<i>wide format of binary response of Schizophrenia data.</i>
---------	--

Description

A dataset containing the treatment and binary responses measured at baseline and 3 post-baseline visits

Usage

schizob

Format

A data frame with 437 rows and 5 variables:

tx treatment, 1 for treated and 0 for placebo

y0 binary response at the baseline

y1, y3, y6 binary response at the post-baseline week 1, 3, and 6.

Source

long-to-wide transformation of schizo data, i.e. `schizob = reshape2::dcast(schizo, id + tx ~ week, value.var = "imps79b")`

schizow	<i>wide format of ordinal response of Schizophrenia data.</i>
---------	---

Description

A dataset containing the treatment and ordinal responses measured at baseline and 3 post-baseline visits

Usage

```
schizow
```

Format

A data frame with 437 rows and 5 variables:

tx treatment, 1 for treated and 0 for placebo

y0 ordinal response at the baseline

y1, y3, y6 ordinal response at the post-baseline week 1, 3, and 6.

Source

long-to-wise transformation of schizo data, i.e. `schizow = reshape2::dcast(schizo, id + tx ~ week, value.var = "imps79o")`

summary	<i>Summarize the results from an object of class remiod</i>
---------	---

Description

Obtain and print the summary, (fixed effects) coefficients (`coef`) and credible interval (`confint`).

Usage

```
summary(object, ...)
```

```
## S3 method for class 'remiod'
summary(object, start = NULL, end = NULL, thin = NULL,
  quantiles = c(0.025, 0.975), outcome = NULL, exclude_chains = NULL,
  warn = TRUE, mess = TRUE, ...)
```

```
## S3 method for class 'summary.remiod'
print(x, digits = 3, ...)
```

```
## S3 method for class 'summary.remiod'
coef(object, start = NULL, end = NULL,
  thin = NULL, subset = NULL, exclude_chains = NULL, warn = TRUE,
  mess = TRUE, ...)
```

Arguments

object	object inheriting from class 'remoid'
...	additional, optional arguments
trunc	named list specifying limits of truncation for the distribution of the named incomplete variables (see the vignette ModelSpecification)
hyperpars	list of hyper-parameters, as obtained by default_hyperpars()
scale_vars	named vector of (continuous) variables that will be centred and scaled (such that mean = 0 and sd = 1) when they enter a linear predictor to improve convergence of the MCMC sampling. Default is that all numeric variables and integer variables with >20 different values will be scaled. If set to FALSE no scaling will be done.
custom	named list of JAGS model chunks (character strings) that replace the model for the given variable.
append_data_list	list that will be appended to the list containing the data that is passed to rjags (data_list). This may be necessary if additional data / variables are needed for custom (covariate) models.
progress.bar	character string specifying the type of progress bar. Possible values are "text" (default), "gui", and "none" (see update). Note: when sampling is performed in parallel it is not possible to display a progress bar.
quiet	logical; if TRUE then messages generated by rjags during compilation as well as the progress bar for the adaptive phase will be suppressed, (see jags.model)
keep_scaled_mcmc	should the "original" MCMC sample (i.e., the scaled version returned by <code>coda.samples()</code>) be kept? (The MCMC sample that is re-scaled to the scale of the data is always kept.)
modelName	character string specifying the name of the model file (including the ending, either .R or .txt). If unspecified a random name will be generated.
modeldir	directory containing the model file or directory in which the model file should be written. If unspecified a temporary directory will be created.
overwrite	logical; whether an existing model file with the specified <code><modeldir>/<modelName></code> should be overwritten. If set to FALSE and a model already exists, that model will be used. If unspecified (NULL) and a file exists, the user is asked for input on how to proceed.
keep_model	logical; whether the created JAGS model file should be saved or removed from (FALSE; default) when the sampling has finished.
start	the first iteration of interest (see window.mcmc)
end	the last iteration of interest (see window.mcmc)
thin	thinning interval (integer; see window.mcmc). For example, <code>thin = 1</code> (default) will keep the MCMC samples from all iterations; <code>thin = 5</code> would only keep every 5th iteration.
quantiles	posterior quantiles
outcome	specify outcome variable to select imputation model(s) to summarize. Default generates summaries for all models.
exclude_chains	optional vector of the index numbers of chains that should be excluded

warn	logical; should warnings be given? Default is TRUE.
mess	logical; should messages be given? Default is TRUE.
x	an object of class <code>summary.remiod</code>
digits	the minimum number of significant digits to be printed in values.
subset	subset of parameters/variables/nodes (columns in the MCMC sample). Follows the same principle as the argument <code>monitor_params</code> and <code>selected_parms</code> .

Value

summary information, including parameter posterior mean, posterior SD, quantiles, tail probability `tail-prob`, Gelman-Rubin criterion `GR-crit`, the ratio of the Monte Carlo error and posterior standard deviation) for specified parameters `MCE/SD`.

Examples

```
# data(schizow)

test = remiod(formula = y6 ~ tx + y0 + y1 + y3, data = schizow,
              trtvar = 'tx', algorithm = 'jags', method="MAR",
              ord_cov_dummy = FALSE, n.adapt = 50, n.chains = 1,
              n.iter = 50, thin = 2, warn = FALSE, seed = 1234)

summary(object = test, outcome = c("y6", "y3"))
```

tang_MI_RB	<i>Implement controlled multiple imputation algorithms proposed by Tang</i>
------------	---

Description

Internal function, creates multiple imputed datasets based on assigned imputation method with the algorithm of Tang's sequential modeling.

Usage

```
tang_MI_RB(object, dtimp, treatment, method = "MAR", delta = 0,
           ord_cov_dummy = FALSE, exclude_chains = NULL, include = FALSE)
```

Arguments

object	object inheriting from class 'remiod'
dtimp	imputed complete data sets from <code>remiod</code> function.
treatment	treatment variable.

method	a method for obtaining multiple-imputed dataset. Options include MAR, J2R, CR, and delta adjustment. Default is MAR.
delta	specific value used for Delta adjustment, applicable only for method="delta".
ord_cov_dummy	optional. specify whether ordinal variables should be treated as categorical variables or continuous variables when they are included as covariates in the sequential imputation models. Default is TRUE, dummy variables will be created accordingly.
exclude_chains	optional vector of the index numbers of chains that should be excluded
include	logical, if TRUE, raw data will be included in imputed data sets with imputation ID = 0.

Value

multiple imputed datasets stacked onto each other (i.e., long format; optionally including the original incomplete data).

The variable `Imputation_` indexes the imputation, while `.rownr` links the rows to the rows of the original data. In cross-sectional datasets the variable `.id` is added as subject identifier.

Index

- * **datasets**
 - schizo, [15](#)
 - schizob, [16](#)
 - schizow, [17](#)
- *_imp, [11](#)
- adapt, [13](#)
- clm_MI_CR, [2](#)
- clm_MI_delta, [3](#)
- clm_MI_J2R, [4](#)
- coda.samples, [13](#)
- coef.summary.remiod (summary), [17](#)
- default_hyperpars, [14, 18](#)
- extract_MIdata, [5](#)
- family, [13](#)
- formula, [13](#)
- get_MI_RB, [6](#)
- get_Mlist, [7](#)
- glm_MI_CR, [7](#)
- glm_MI_delta, [8](#)
- glm_MI_J2R, [9](#)
- graphics::matplot, [11](#)
- jags.model, [14, 18](#)
- lines, [12](#)
- list.models, [10](#)
- mcmcplot, [11](#)
- par, [12](#)
- plot.xy, [12](#)
- points, [12](#)
- print.summary.remiod (summary), [17](#)
- remiod, [12](#)
- schizo, [15](#)
- schizob, [16](#)
- schizow, [17](#)
- set_refcat, [13](#)
- summary, [17](#)
- tang_MI_RB, [19](#)
- update, [14, 18](#)
- window.mcmc, [13, 14, 18](#)