Package 'MatchThem'

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Title Matching and Weighting Multiply Imputed Datasets

Description Provides essential tools for the pre-processing techniques of matching and weighting multiply imputed datasets. The package includes functions for matching within and across multiply imputed datasets using various methods, estimating weights for units in the imputed datasets using multiple weighting methods, calculating causal effect estimates in each matched or weighted dataset using parametric or non-parametric statistical models, and pooling the resulting estimates according to Rubin's rules (please see <<u>https:</u> //journal.r-project.org/archive/2021/RJ-2021-073/> for more details).

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as.mimids

Create a mimids object

Description

Creates a mimids object from a list of matchit objects and an imputed dataset.

Usage

```
as.mimids(x, ...)
## Default S3 method:
as.mimids(x, datasets, ...)
```

Arguments

х	A list of matchit objects, each the output of a call to MatchIt::matchit() on an imputed dataset.
	Ignored.
datasets	This argument specifies the datasets containing the exposure and the potential confounders called in the formula. This argument must be an object of the mids or amelia class, which is typically produced by a previous call to mice() function from the mice package or to amelia() function from the Amelia package (the Amelia package is designed to impute missing data in a single cross-sectional dataset or in a time-series dataset, currently, the MatchThem package only supports the former datasets).

as.wimids

Details

The matched datasets are stored as though matchthem() was called with approach = "within".

Value

A mimids object.

See Also

matchthem(), mimids, MatchIt::matchit()

Examples

as.wimids

Create a wimids object

Description

Creates a wimids object from a list of weightit objects and an imputed dataset.

Usage

```
as.wimids(x, ...)
## Default S3 method:
as.wimids(x, datasets, ...)
```

Arguments

х	A list of weightit objects, each the output of a call to WeightIt::weightit() on an imputed dataset.
	Ignored.
datasets	The datasets containing the exposure and covariates mentioned in the formula. This argument must be an object of the mids or amelia class, which is typically produced by a previous call to mice() from the mice package or to amelia() from the Amelia package (the Amelia package is designed to impute missing data in a single cross-sectional dataset or in a time-series dataset, currently, the MatchThem package only supports the former datasets).

Details

The weighted datasets are stored as though weightthem() was called with approach = "within".

Value

A wimids object.

See Also

weightthem(), wimids, WeightIt::weightit()

Examples

cbind

Description

This function combines a mimids or wimids object columnwise with additional datasets or variables. Typically these would be variables not included in the original multiple imputation and therefore absent in the mimids or wimids object. with() can then be used on the output to run models with the added variables.

Usage

cbind(..., deparse.level = 1)
S3 method for class 'mimids'
cbind(..., deparse.level = 1)

S3 method for class 'wimids'
cbind(..., deparse.level = 1)

Arguments

• • •

Objects to combine columnwise. The first argument should be a mimids or wimids object. Additional data.frames, matrixes, factors, or vectors can be supplied. These can be given as named arguments.

deparse.level Ignored.

Value

An object with the same class as the first input object with the additional datasets or variables added to the components.

Author(s)

Farhad Pishgar and Noah Greifer

See Also

cbind()

Examples

#Loading libraries
library(survey)

#Loading the dataset
data(osteoarthritis)

```
complete
```

```
complete
```

Extracts Multiply Imputed Datasets

Description

complete() extracts data from an object of the mimids or wimids class.

Usage

```
## S3 method for class 'mimids'
complete(data, action = 1, include = FALSE, mild = FALSE, all = TRUE, ...)
## S3 method for class 'wimids'
complete(data, action = 1, include = FALSE, mild = FALSE, all = TRUE, ...)
```

Arguments

data	A mimids or wimids object; the output of a call to matchthem() or weightthem().
action	The imputed dataset number, intended to extract its data, or an action. The input must be a positive integer or a keyword. The keywords include "all" (produces a mild object of the multiply imputed datasets), "long" (produces a dataset with multiply imputed datasets stacked vertically), and "broad" (produces a dataset with multiply imputed datasets stacked horizontally). The default is 1.
include	Whether the original data with the missing values should be included. The input must be a logical value. The default is FALSE.
mild	Whether the return value should be an object of mild class. Please note that set- ting mild = TRUE overrides action keywords of "long", "broad", and "repeated". The default is FALSE.

complete

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all	Whether to include observations with a zero estimated weight. The default is
	TRUE.
	Ignored.

Details

complete() works by running mice::complete() on the mids object stored within the mimids or wimids object and appending the outputs of the matching or weighting procedure. For mimids objects, the appended outputs include the matching weights, the propensity score (if included), pair membership (if included), and whether each unit was discarded. For wimids objects, the appended output is the estimated weights.

Value

This function returns the imputed dataset within the supplied mimids or wimids objects.

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

mimids
wimids
mice::complete()

Examples

is.mimids

Description

is.mimids() function checks whether class of objects is mimids or not.

Usage

```
is.mimids(object)
```

Arguments

object This argument specifies the object that should be checked to see if it is of the mimids class or not.

Details

The class of objects is checked to be of the mimids.

Value

This function returns a logical value indicating whether object is of the mimids class.

Author(s)

Farhad Pishgar

See Also

matchthem()
mimids
inherits()

Examples

```
#Loading the dataset
data(osteoarthritis)
```

```
#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>
```

is.mimipo

```
#Checking the 'matched.datasets' object
is.mimids(matched.datasets)
```

is.mimipo

Checks for the mimipo Class

Description

is.mimipo() function checks whether class of objects is mimipo or not.

Usage

```
is.mimipo(object)
```

Arguments

object This argument specifies the object that should be checked to see if it is of the mimipo class or not.

Details

The class of objects is checked to be of the mimipo.

Value

This function returns a logical value indicating whether object is of the mimipo class.

Author(s)

Farhad Pishgar

See Also

pool()
mimipo()

Examples

#Loading libraries
library(survey)

#Loading the dataset
data(osteoarthritis)

#Multiply imputing the missing values imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>

#Estimating weights of observations in the multiply imputed datasets weighted.datasets <- weightthem(OSP ~ AGE + SEX + BMI + RAC + SMK,</pre>

is.mimira

is.mimira

Checks for the mimira Class

Description

is.mimira() function checks whether class of objects is mimira or not.

Usage

is.mimira(object)

Arguments

object This argument specifies the object that should be checked to see if it is of the mimira class or not.

Details

The class of objects is checked to be of the mimira.

Value

This function returns a logical value indicating whether object is of the mimira class.

Author(s)

Farhad Pishgar

See Also

with()
mimira()

is.wimids

Examples

```
#Loading libraries
library(survey)
#Loading the dataset
data(osteoarthritis)
#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>
#Estimating weights of observations in the multiply imputed datasets
weighted.datasets <- weightthem(OSP ~ AGE + SEX + BMI + RAC + SMK,</pre>
                                 imputed.datasets,
                                 approach = 'within',
                                 method = 'glm',
                                 estimand = "ATT")
#Analyzing the weighted datasets
models <- with(weighted.datasets,</pre>
               svyglm(KOA ~ OSP, family = binomial))
#Checking the 'models' object
is.mimira(models)
```

is.wimids Checks for the wimids Class

Description

is.wimids() function checks whether class of objects is wimids or not.

Usage

is.wimids(object)

Arguments

object	This argument specifies the object that should be checked to see if it is of the
	wimids class or not.

Details

The class of objects is checked to be of the wimids.

Value

This function returns a logical value indicating whether object is of the wimids class.

Author(s)

Farhad Pishgar

See Also

weightthem()
wimids()

Examples

#Loading the dataset
data(osteoarthritis)

#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>

#Checking the 'weighted.datasets' object
is.wimids(weighted.datasets)

matchthem

Matches Multiply Imputed Datasets

Description

matchthem() performs matching in the supplied multiply imputed datasets, given as mids or amelia objects, by running MatchIt::matchit() on each of the multiply imputed datasets with the supplied arguments.

Usage

```
matchthem(
   formula,
   datasets,
   approach = "within",
   method = "nearest",
   distance = "glm",
   link = "logit",
   distance.options = list(),
   discard = "none",
   reestimate = FALSE,
   ....
)
```

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matchthem

Arguments

- formula A formula of the form z ~ x1 + x2, where z is the exposure and x1 and x2 are the covariates to be balanced, which is passed directly to MatchIt::matchit() to specify the propensity score model or treatment and covariates to be used in matching. See MatchIt::matchit() for details.
- datasets This argument specifies the datasets containing the exposure and the potential confounders called in the formula. This argument must be an object of the mids or amelia class, which is typically produced by a previous call to mice() function from the **mice** package or to amelia() function from the **Amelia** package (the **Amelia** package is designed to impute missing data in a single cross-sectional dataset or in a time-series dataset, currently, the **MatchThem** package only supports the former datasets).
- approach The approach that should be used to combine information in multiply imputed datasets. Currently, "within" (performing matching within each dataset) and "across" (estimating propensity scores within each dataset, averaging them across datasets, and performing matching using the averaged propensity scores in each dataset) approaches are available. The default is "within", which has been shown to have superior performance in most cases.
- method This argument specifies a matching method. Currently, "nearest" (nearest neighbor matching), "exact" (exact matching), "full" (optimal full matching), "genetic" (genetic matching), "subclass" (subclassication), "cem" (coarsened exact matching), "optimal" (optimal pair matching), "quick" (generalized full matching), and ("cardinality") (cardinality and profile matching) methods are available. Only methods that produce a propensity score ("nearest", "full", "genetic", "subclass", "optimal", and "quick") are compatible with the "across" approach. The default is "nearest" for nearest neighbor matching. See MatchIt::matchit() for details.
- distance The method used to estimate the distance measure (e.g., propensity scores) used in matching, if any. Only options that specify a method of estimating propensity scores (i.e., not "mahalanobis") are compatible with the "across" approach. The default is "glm" for estimating propensity scores using logistic regression. See MatchIt::matchit() and MatchIt::distance for details and allowable options.

Details

If an amelia object is supplied to datasets, it will be transformed into a mids object for further use. matchthem() works by calling mice::complete() on the mids object to extract a complete dataset, and then calls MatchIt::matchit() on each one, storing the output of each matchit() call and the mids in the output. All arguments supplied to matchthem() except datasets and approach are passed directly to matchit(). With the "across" approach, the estimated propensity scores are averaged across multiply imputed datasets and re-supplied to another set of calls to matchit().

Value

An object of the mimids() (matched multiply imputed datasets) class, which includes the supplied mids object (or an amelia object transformed into a mids object if supplied) and the output of the calls to matchit() on each multiply imputed dataset.

Author(s)

Farhad Pishgar and Noah Greifer

References

Daniel Ho, Kosuke Imai, Gary King, and Elizabeth Stuart (2007). Matching as Nonparametric Preprocessing for Reducing Model Dependence in Parametric Causal Inference. *Political Analysis*, 15(3): 199-236. https://gking.harvard.edu/files/abs/matchp-abs.shtml

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

Gary King, James Honaker, Anne Joseph, and Kenneth Scheve (2001). Analyzing Incomplete Political Science Data: An Alternative Algorithm for Multiple Imputation. *American Political Science Review*, 95: 49–69. https://gking.harvard.edu/files/abs/evil-abs.shtml

See Also

mimids
with()
pool()
weightthem()
MatchIt::matchit()

Examples

#1

#Loading the dataset
data(osteoarthritis)

```
#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>
```

#2

#Loading the dataset
data(osteoarthritis)

mimids

mimids

Matched Multiply Imputed Datasets

Description

mimids object contains data of matched multiply imputed datasets. mimids objects are generated by calls to matchthem().

Details

```
mimids objects have methods for print(), summary(), plot(), and cbind().
```

Note

The **MatchThem** package does not use the S4 class definitions and instead relies on the S3 list equivalents.

Author(s)

Farhad Pishgar

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

matchthem(), as.mimids()

mimipo

Description

mimipo object contains data of multiply imputed pooled outcome. mimipo objects are generated by calls to pool().

Details

mimipo objects has methods for the print() and summary() functions (please see **mice** package reference manual for details).

Note

The **MatchThem** package does not use the S4 class definitions and instead relies on the S3 list equivalents.

Author(s)

Farhad Pishgar

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

pool()

mimira

Multiply Imputed Repeated Analyses

Description

mimira object contains data of multiply imputed repeated analyses. mimira objects are generated by calls to with().

Details

mimira objects has methods for the print() and summary() functions (please see **mice** package reference manual for details).

osteoarthritis

Note

The **MatchThem** package does not use the S4 class definitions and instead relies on the S3 list equivalents.

Author(s)

Farhad Pishgar

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

with()

osteoarthritis	Data of 2,585	Participants	in th	e Osteoarthritis	Initiative (OAI)
	Project				

Description

osteoarthritis includes demographic data of 2,585 units (individuals) with or at risk of knee osteoarthritis. The recorded data has missing values in body mass index (BMI, a quantitative variable), race (RAC, a categorical qualitative variable), smoking status (SMK, a binary qualitative variable), and knee osteoarthritis status at follow-up (KOA, a binary qualitative variable).

Usage

osteoarthritis

Format

This dataset contains 2,585 rows and 7 columns. Each row presents data of an unit (individual) and each column presents data of a characteristic of that unit. The columns are:

AGE Age of each unit (individual);

- SEX Gender of each unit (individual), coded as 0 (female) and 1 (male);
- **BMI** Estimated body mass index of each unit (individual);
- **RAC** Race of each unit (individual), coded as 0 (other), 1 (Caucasian), 2 (African American), and 3 (Asian);
- **SMK** The smoking status of each unit (individual), coded as 0 (non-smoker) and 1 (smoker);
- **OSP** Osteoporosis status of each unit (individual) at baseline, coded as 0 (negative) and 1 (positive); and
- **KOA** Knee osteoarthritis status of each unit (individual) in the follow-up, coded as 0 (at risk) and 1 (diagnosed).

Source

The information presented in the osteoarthritis dataset is based on the publicly available data of the Osteoarthritis Initiative (OAI) project (see https://nda.nih.gov/oai/ for details), with changes.

pool

Pools Estimates by Rubin's Rules

Description

pool() pools estimates from the analyses done within each multiply imputed dataset. The typical sequence of steps to do a matching or weighting procedure on multiply imputed datasets are:

- Multiply impute the missing values using the mice() function (from the mice package) or the amelia() function (from the Amelia package), resulting in a multiply imputed dataset (an object of the mids or amelia class);
- 2. Match or weight each multiply imputed dataset using matchthem() or weightthem(), resulting in an object of the mimids or wimids class;
- 3. Check the extent of balance of covariates in the datasets (using functions from the **cobalt** package);
- 4. Fit the statistical model of interest on each dataset by the with() function, resulting in an object of the mimira class; and
- 5. Pool the estimates from each model into a single set of estimates and standard errors, resulting in an object of the mimipo class.

Usage

pool(object, dfcom = NULL)

Arguments

object	An object of the mimira class (produced by a previous call to with()).
dfcom	A positive number representing the degrees of freedom in the data analysis. The
	default is NULL, which means to extract this information from the fitted model
	with the lowest number of observations or the first fitted model (when that fails
	the parameter is set to 999999).

Details

pool() function averages the estimates of the model and computes the total variance over the repeated analyses by Rubin's rules. It calls mice::pool() after computing the model degrees of freedom.

Value

This function returns an object from the mimipo class. Methods for mimipo objects (e.g., print(), summary(), etc.) are imported from the **mice** package.

trim

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

with()
mice::pool()

Examples

#Pooling results obtained from analyzing the datasets
results <- pool(models)
summary(results)</pre>

trim

Trim Weights

Description

Trims (i.e., truncates) large weights by setting all weights higher than that at a given quantile to the weight at the quantile. This can be useful in controlling extreme weights, which can reduce effective sample size by enlarging the variability of the weights.

Usage

```
## S3 method for class 'wimids'
trim(x, at = 0, lower = FALSE, ...)
```

Arguments

х	A wimids object; the output of a call to weightthem().
at	numeric; either the quantile of the weights above which weights are to be trimmed. A single number between .5 and 1, or the number of weights to be trimmed (e.g., $at = 3$ for the top 3 weights to be set to the 4th largest weight).
lower	logical; whether also to trim at the lower quantile (e.g., for at = .9, trimming at both .1 and .9, or for at = 3, trimming the top and bottom 3 weights). Default is FALSE to only trim the higher weights.
	Ignored.

Details

trim.wimids() works by calling WeightIt::trim() on each weightit object stored in the models component of the wimids object. Because trim() itself is not exported from MatchThem, it must be called using WeightIt::trim() or by attaching WeightIt (i.e., running library(WeightIt)) before use.

Value

An object from the wimids class, identical to the original object except with trim() applied to each of the weightit objects in the models component.

Author(s)

Noah Greifer

See Also

WeightIt::trim()

Examples

#Loading the dataset
data(osteoarthritis)

```
#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>
```

#Trimming the top 10% of weights in each dataset #to the 90th percentile trimmed.datasets <- trim(weighted.datasets, at = 0.9)</pre> weightthem

Description

weightthem() performs weighting in the supplied multiply imputed datasets, given as mids or amelia objects, by running WeightIt::weightit() on each of the multiply imputed datasets with the supplied arguments.

Usage

```
weightthem(formula, datasets, approach = "within", method = "glm", ...)
```

Arguments

formula	A formula of the form $z \sim x1 + x2$, where z is the exposure and x1 and x2 are the covariates to be balanced, which is passed directly to WeightIt::weightit() to specify the propensity score model or treatment and covariates to be used to estimate the weights. See WeightIt::weightit() for details.
datasets	The datasets containing the exposure and covariates mentioned in the formula. This argument must be an object of the mids or amelia class, which is typically produced by a previous call to mice() from the mice package or to amelia() from the Amelia package (the Amelia package is designed to impute missing data in a single cross-sectional dataset or in a time-series dataset, currently, the MatchThem package only supports the former datasets).
approach	The approach used to combine information in multiply imputed datasets. Cur- rently, "within" (estimating weights within each dataset), "across" (estimat- ing propensity scores within each dataset, averaging them across datasets, and computing a single set of weights based on that to be applied to all datasets), and "apw" (or averaging the probability weights, estimating weights within each dataset and averaging them across datasets) approaches are available. The de- fault is "within", which has been shown to have superior performance in most cases.
method	The method used to estimate weights. See WeightIt::weightit() for allow- able options. Only methods that produce a propensity score ("glm", "gbm", "ipt" "cbps", "super", and "bart") are compatible with the "across" ap- proach). The default is "glm" propensity score weighting using logistic regres- sion propensity scores.
	Additional arguments to be passed to weightit(). see WeightIt::weightit() for more details.

Details

If an amelia object is supplied to datasets, it will be transformed into a mids object for further use. weightthem() works by calling mice::complete() on the mids object to extract a complete dataset, and then calls WeightIt::weightit() on each dataset, storing the output of each

weightit() call and the mids in the output. All arguments supplied to weightthem() except datasets and approach are passed directly to weightit(). With the "across" approach, the estimated propensity scores are averaged across imputations and re-supplied to another set of calls to weightit().

Value

An object of the wimids() (weighted multiply imputed datasets) class, which includes the supplied mids object (or an amelia object transformed into a mids object if supplied) and the output of the calls to weightit() on each multiply imputed dataset.

Author(s)

Farhad Pishgar and Noah Greifer

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

wimids
with()
pool()
matchthem()
WeightIt::weightit()

Examples

#1

```
#Loading the dataset
data(osteoarthritis)
```

```
#Multiply imputing the missing values
imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>
```

#2

#Loading the dataset
data(osteoarthritis)

wimids

wimids

Weighted Multiply Imputed Datasets

Description

wimids object contains data of weighted multiply imputed datasets. The wimids object is generated by calls to the weightthem().

Details

wimids objects have methods for print(), summary(), and cbind().

Note

The **MatchThem** package does not use the S4 class definitions and instead relies on the S3 list equivalents.

Author(s)

Farhad Pishgar

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

weightthem(), as.wimids()

Evaluates an Expression in Matched or Weighted Multiply Imputed Datasets

Description

with() runs a model on the n multiply imputed datasets of the supplied mimids or wimids object. The typical sequence of steps to do a matching or weighting procedure on multiply imputed datasets are:

- Multiply impute the missing values using the mice() function (from the mice package) or the amelia() function (from the Amelia package), resulting in a multiply imputed dataset (an object of the mids or amelia class);
- 2. Match or weight each multiply imputed dataset using matchthem() or weightthem(), resulting in an object of the mimids or wimids class;
- 3. Check the extent of balance of covariates in the datasets (using functions from the **cobalt** package);
- 4. Fit the statistical model of interest on each dataset by the with() function, resulting in an object of the mimira class; and
- 5. Pool the estimates from each model into a single set of estimates and standard errors, resulting in an object of the mimipo class.

Usage

S3 method for class 'mimids'
with(data, expr, cluster, ...)

```
## S3 method for class 'wimids'
with(data, expr, ...)
```

Arguments

data	A mimids or wimids object, typically produced by a previous call to the matchthem() or weightthem().
expr	An expression (usually a call to a modeling function like glm(), coxph(), svyglm(), etc.) to evaluate in each (matched or weighted) multiply imputed dataset. See Details.
cluster	When a function from survey (e.g., survey :: svyglm()) is supplied in expr, whether the standard errors should incorporate clustering due to dependence between matched pairs. This is done by supplying the variable containing pair membership to the ids argument of link[survey:svydesign]{svydesign()}. If unspecified, it will be set to TRUE if subclasses (i.e., pairs) are present in the output and there are 20 or more unique subclasses. It will be ignored for match- ing methods that don't return subclasses (e.g., matching with replacement).
	Additional arguments to be passed to expr.

with

with

Details

with() applies the supplied model in expr to the (matched or weighted) multiply imputed datasets, automatically incorporating the (matching) weights when possible. The argument to expr should be of the form $glm(y \sim z, family = quasibinomial)$, for example, excluding the data or weights argument, which are automatically supplied.

Functions from the **survey** package, such as svyglm(), are treated a bit differently. No svydesign object needs to be supplied because with() automatically constructs and supplies it with the imputed dataset and estimated weights. When cluster = TRUE (or with() detects that pairs should be clustered; see the cluster argument above), pair membership is supplied to the ids argument of svydesign().

After weighting using weightthem(), glm_weightit() should be used as the modeling function to fit generalized lienar models. It correctly produces robust standard errors that account for estimation of the weights, if possible. See WeightIt::glm_weightit() for details. Otherwise, svyglm() should be used rather than glm() in order to correctly compute standard errors. For Cox models, coxph() will produce approximately correct standard errors when used with weighting but svycoxph() will produce more accurate standard errors when matching is used.

Value

An object from the mimira class containing the output of the analyses.

Author(s)

Farhad Pishgar and Noah Greifer

References

Stef van Buuren and Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. *Journal of Statistical Software*, 45(3): 1-67. doi:10.18637/jss.v045.i03

See Also

matchthem()
weightthem()
mice::with.mids()

Examples

#Loading libraries
library(survey)

#Loading the dataset
data(osteoarthritis)

#Multiply imputing the missing values imputed.datasets <- mice::mice(osteoarthritis, m = 5)</pre>

```
#Matching in the multiply imputed datasets
matched.datasets <- matchthem(OSP ~ AGE + SEX + BMI + RAC + SMK,</pre>
```

```
imputed.datasets,
                               approach = 'within',
                               method = 'nearest')
#Analyzing the matched datasets
models <- with(matched.datasets,</pre>
               svyglm(KOA ~ OSP, family = binomial),
               cluster = TRUE)
#Weghting in the multiply imputed datasets
weighted.datasets <- weightthem(OSP ~ AGE + SEX + BMI + RAC + SMK,</pre>
                                imputed.datasets,
                                approach = 'within',
                                method = 'glm')
#Analyzing the matched datasets
models <- with(weighted.datasets,</pre>
               WeightIt::glm_weightit(KOA ~ OSP,
                                       family = binomial))
```

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