# logicFS

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## R topics documented:

<table>
<thead>
<tr>
<th>Function</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>data.logicfs</td>
<td>1</td>
</tr>
<tr>
<td>getMatEval</td>
<td>2</td>
</tr>
<tr>
<td>logic.bagging</td>
<td>2</td>
</tr>
<tr>
<td>logicFS-internal</td>
<td>6</td>
</tr>
<tr>
<td>logicFS</td>
<td>6</td>
</tr>
<tr>
<td>logic.oob</td>
<td>9</td>
</tr>
<tr>
<td>logic.pimp</td>
<td>10</td>
</tr>
<tr>
<td>make.snp.dummy</td>
<td>11</td>
</tr>
<tr>
<td>minDNF</td>
<td>12</td>
</tr>
<tr>
<td>mlogreg</td>
<td>12</td>
</tr>
<tr>
<td>plot.logicFS</td>
<td>14</td>
</tr>
<tr>
<td>predict.logicBagg</td>
<td>16</td>
</tr>
<tr>
<td>predict.mlogreg</td>
<td>17</td>
</tr>
<tr>
<td>print.logicFS</td>
<td>17</td>
</tr>
<tr>
<td>vim.chisq</td>
<td>18</td>
</tr>
<tr>
<td>vim.ebam</td>
<td>19</td>
</tr>
<tr>
<td>vim.individual</td>
<td>21</td>
</tr>
<tr>
<td>vim.logicFS</td>
<td>22</td>
</tr>
<tr>
<td>vim.norm</td>
<td>24</td>
</tr>
<tr>
<td>vim.set</td>
<td>25</td>
</tr>
</tbody>
</table>

**Example Data of logicFS**

- **data.logicfs** contains two objects: a simulated matrix `data.logicfs` of 400 observations (rows) and 15 variables (columns) and a vector `cl.logicfs` of length 400 containing the class labels of the observations.

Each variable is categorical with realizations 1, 2 and 3. The first 200 observations are cases, the remaining are controls. If one of the following expression is `TRUE`, then the corresponding observation is a case:

```
SNP1 == 3
```
SNP2 == 1 AND SNP4 == 3
SNP3 == 3 AND SNP5 == 3 AND SNP6 == 1

where SNP1 is in the first column of data.logicfs, SNP2 in the second, and so on.

See Also

logic.bagging, logicFS

getMatEval  Evaluate Prime Implicants

Description

Computes the values of prime implicants for observations for which the values of the variables composing the prime implicants are available.

Usage

getMatEval(data, vec.primes, check = TRUE)

Arguments

data a data frame in which each row corresponds to an observation, and each column to a binary variable.
vec.primes a character vector naming the prime implicants that should be evaluated. Each of the variables composing these prime implicants must be represented by one column of data.
check should some checks be done before the evaluation is performed? It is highly recommended not to change the default check = TRUE.

Value

a matrix in which each row corresponds to an observation (the same observations in the same order as in data, and each column to one of the prime implicants.

Author(s)

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logic.bagging

Description

A bagging and subsampling version of logic regression. Currently available for the classification, the linear regression, and the logistic regression approach of logreg. Additionally, an approach based on multinomial logistic regressions as implemented in mlogreg can be used if the response is categorical.

Usage

## S3 method for class 'formula':
logic.bagging(formula, data, recdom = TRUE, ...)

## Default S3 method:
logic.bagging(x, y, B = 100, useN = TRUE, ntrees = 1, nleaves = 8,
glm.if.1tree = FALSE, replace = TRUE, sub.frac = 0.632,
anneal.control = logreg.anneal.control(), oob = TRUE,
onlyRemove = FALSE, prob.case = 0.5, importance = TRUE,
addMatImp = FALSE, fast = FALSE, rand = NULL, ...)

Arguments

formula

a data frame containing the variables in the model. Each row of data must correspond to an observation, and each column to a binary variable (coded by 0 and 1) or a factor (for details, see recdom) except for the column comprising the response. The response must be either binary (coded by 0 and 1), categorical or continuous. If continuous, a linear model is fitted in each of the B iterations of logic.bagging. If categorical, the column of data specifying the response must be a factor. In this case, multinomial logic regressions are performed as implemented in mlogreg. Otherwise, depending on ntrees (and glm.if.1tree) the classification or the logistic regression approach of logic regression is used.

data

a matrix consisting of 0’s and 1’s. Each column must correspond to a binary variable and each row to an observation.

trees and nleaves

A matrix consisting of 0's and 1's. Each column must correspond to a binary variable and each row to an observation.
a numeric vector or a factor specifying the values of a response for all the observations represented in \( x \). If a numeric vector, then \( y \) either contains the class labels (coded by 0 and 1) or the values of a continuous response depending on whether the classification or logistic regression approach of logic regression, or the linear regression approach, respectively, should be used. If the response is categorical, then \( y \) must be a factor naming the class labels of the observations.

\( y \)

B

an integer specifying the number of iterations.

useN

logical specifying if the number of correctly classified out-of-bag observations should be used in the computation of the importance measure. If FALSE, the proportion of correctly classified oob observations is used instead. Ignored if importance = FALSE.

ntrees

an integer indicating how many trees should be used.

For a binary response: If ntrees is larger than 1, the logistic regression approach of logic regression will be used. If ntrees is 1, then by default the classification approach of logic regression will be used (see glm.if.1tree).

For a continuous response: A linear regression model with ntrees trees is fitted in each of the \( B \) iterations.

For a categorical response: \( n.lev - 1 \) logic regression models with ntrees trees are fitted, where \( n.lev \) is the number of levels of the response (for details, see mlogreg).

nleaves

a numeric value specifying the maximum number of leaves used in all trees combined. See the help page of the function logreg of the package LogicReg for details.

glm.if.1tree

if ntrees is 1 and glm.if.1tree is TRUE the logistic regression approach of logic regression is used instead of the classification approach. Ignored if ntrees is not 1 or the response is not binary.

replace

should sampling of the cases be done with replacement? If TRUE, a bootstrap sample of size length(cl) is drawn from the length(cl) observations in each of the \( B \) iterations. If FALSE, ceiling(sub.frac * length(cl)) of the observations are drawn without replacement in each iteration.

sub.frac

a proportion specifying the fraction of the observations that are used in each iteration to build a classification rule if replace = FALSE. Ignored if replace = TRUE.

anneal.control

a list containing the parameters for simulated annealing. See the help page of logreg.anneal.control in the LogicReg package.

oob

should the out-of-bag error rate (classification and logistic regression) or the out-of-bag root mean square prediction error (linear regression), respectively, be computed?

onlyRemove

should in the single tree case the multiple tree measure be used? If TRUE, the prime implicants are only removed from the trees when determining the importance in the single tree case. If FALSE, the original single tree measure is computed for each prime implicant, i.e. a prime implicant is not only removed from the trees in which it is contained, but also added to the trees that do not contain this interaction. Ignored in all other than the classification case.

prob.case

a numeric value between 0 and 1. If the outcome of the logistic regression, i.e. the class probability, for an observation is larger than prob.case, this observations will be classified as case (or 1).

importance

should the measure of importance be computed?
addMatImp should the matrix containing the improvements due to the prime implicants in each of the iterations be added to the output? (For each of the prime implicants, the importance is computed by the average over the B improvements.) Must be set to TRUE, if standardized importances should be computed using vim.norm, or if permutation based importances should be computed using vim.perm.

fast should a greedy search (as implemented in logreg) be used instead of simulated annealing?

rand numeric value. If specified, the random number generator will be set into a reproducible state.

... for the formula method, optional parameters to be passed to the low level function logic.bagging.default. Otherwise, ignored.

Value

logic.bagging returns an object of class logicBagg containing

logreg.model a list containing the B logic regression models,
inbagg a list specifying the B Bootstrap samples,
vim an object of class logicFS (if importance = TRUE),
oob.error the out-of-bag error (if oob = TRUE),
... further parameters of the logic regression.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

predict.logicBagg, plot.logicBagg, logicFS

Examples

## Not run:
# Load data.
data(data.logicfs)

data(data.logicfs)

# For logic regression and hence logic.bagging, the variables must
# be binary. data.logicfs, however, contains categorical data
# with realizations 1, 2 and 3. Such data can be transformed
# into binary data by
bin.snps<-make.snp.dummy(data.logicfs)

# To speed up the search for the best logic regression models
# only a small number of iterations is used in simulated annealing.
my.anneal<-logreg.anneal.control(start=2,end=-2,iter=10000)
# Bagged logic regression is then performed by
bagg.out<-logic.bagging(bin.snps,cl.logicfs,B=20,nleaves=10,
    rand=123,anneal.control=my.anneal)

# The output of logic.bagging can be printed
bagg.out

# By default, also the importances of the interactions are
# computed
bagg.out$vim

# and can be plotted.
plot(bagg.out)

# The original variable names are displayed in
plot(bagg.out,coded=FALSE)

# New observations (here we assume that these observations are
# in data.logicfs) are assigned to one of the classes by
predict(bagg.out,data.logicfs)

## End(Not run)

---

**logicFS-internal**  
*Internal logicFS functions*

**Description**

Internal logicFS functions.

**Details**

These functions are not meant to be directly called by the user.

**Author(s)**

Holger Schwender, ⟨holger.schwender@udo.edu⟩

---

**logicFS**  
*Feature Selection with Logic Regression*

**Description**

Identification of interesting interactions between binary variables using logic regression. Currently available for the classification, the linear regression and the logistic regression approach of *logreg* and for a multinomial logic regression as implemented in *mlogreg*. 
Usage

```r
## S3 method for class 'formula':
logicFS(formula, data, recdom = TRUE, ...)

## Default S3 method:
logicFS(x, y, B = 100, useN = TRUE, ntrees = 1, nleaves = 8,
       glm.if.1tree = FALSE, replace = TRUE, sub.frac = 0.632,
       anneal.control = logreg.anneal.control(), onlyRemove = FALSE,
       prob.case = 0.5, addMatImp = TRUE, fast = FALSE, rand = NULL, ...)
```

Arguments

- `formula`: an object of class `formula` describing the model that should be fitted.
- `data`: a data frame containing the variables in the model. Each row of `data` must correspond to an observation, and each column to a binary variable (coded by 0 and 1) or a factor (for details, see `recdom`) except for the column comprising the response. The response must be either binary (coded by 0 and 1), categorical or continuous. If continuous, a linear model is fitted in each of the `B` iterations of `logicFS`. If categorical, the column of `data` specifying the response must be a factor. In this case, multinomial logic regressions are performed as implemented in `mlogreg`. Otherwise, depending on `ntrees` and `glm.if.1tree` the classification or the logistic regression approach of logic regression is used.
- `recdom`: a logical value or vector of length `ncol(data)` comprising whether a SNP should be transformed into two binary dummy variables coding for a recessive and a dominant effect. If TRUE (logical value), then all factors (variables) with three levels will be coded by two dummy variables as described in `make.snp.dummy`. Each level of each of the other factors (also factors specifying a SNP that shows only two genotypes) is coded by one indicator variable. If `recdom` is a logical vector, all factors corresponding to an entry in `recdom` that is TRUE are assumed to be SNPs and transformed into the two binary variables described above. Each variable that corresponds to an entry of `recdom` that is TRUE (no matter whether `recdom` is a vector or a value) must be coded by the integers 1 (coding for the homozygous reference genotype), 2 (heterozygous), and 3 (homozygous variant).
- `x`: a matrix consisting of 0’s and 1’s. Each column must correspond to a binary variable and each row to an observation.
- `y`: a numeric vector or a factor specifying the values of a response for all the observations represented in `x`. If a numeric vector, then `y` either contains the class labels (coded by 0 and 1) or the values of a continuous response depending on whether the classification or logistic regression approach of logic regression, or the linear regression approach, respectively, should be used. If the response is categorical, then `y` must be a factor naming the class labels of the observations.
- `B`: an integer specifying the number of iterations.
- `useN`: logical specifying if the number of correctly classified out-of-bag observations should be used in the computation of the importance measure. If FALSE, the proportion of correctly classified oob observations is used instead.
- `ntrees`: an integer indicating how many trees should be used.

For a binary response: If `ntrees` is larger than 1, the logistic regression approach of logic regression will be used. If `ntrees` is 1, then by default the classification approach of logic regression will be used (see `glm.if.1tree`).
For a continuous response: A linear regression model with \( ntrees \) trees is fitted in each of the \( B \) iterations.

For a categorical response: \( n.lev - 1 \) logic regression models with \( ntrees \) trees are fitted, where \( n.lev \) is the number of levels of the response (for details, see \texttt{mlogreg}).

\textbf{nleaves} a numeric value specifying the maximum number of leaves used in all trees combined. For details, see the help page of the function \texttt{logreg} of the package \texttt{LogicReg}.

\textbf{glm.if.1tree} if \( ntrees \) is 1 and \texttt{glm.if.1tree} is \texttt{TRUE} the logistic regression approach of logic regression is used instead of the classification approach. Ignored if \( ntrees \) is not 1, or the response is not binary.

\textbf{replace} should sampling of the cases be done with replacement? If \texttt{TRUE}, a Bootstrap sample of size \( \text{length(cl)} \) is drawn from the \( \text{length(cl)} \) observations in each of the \( B \) iterations. If \texttt{FALSE}, \( \text{ceiling(sub.frac * length(cl))} \) of the observations are drawn without replacement in each iteration.

\textbf{sub.frac} a proportion specifying the fraction of the observations that are used in each iteration to build a classification rule if \texttt{replace} = \texttt{FALSE}. Ignored if \texttt{replace} = \texttt{TRUE}.

\textbf{anneal.control} a list containing the parameters for simulated annealing. See the help of the function \texttt{logreg.anneal.control} in the \texttt{LogicReg} package.

\textbf{onlyRemove} should in the single tree case the multiple tree measure be used? If \texttt{TRUE}, the prime implicants are only removed from the trees when determining the importance in the single tree case. If \texttt{FALSE}, the original single tree measure is computed for each prime implicant, i.e. a prime implicant is not only removed from the trees in which it is contained, but also added to the trees that do not contain this interaction. Ignored in all other than the classification case.

\textbf{prob.case} a numeric value between 0 and 1. If the outcome of the logistic regression, i.e. the predicted probability, for an observation is larger than \texttt{prob.case} this observation will be classified as case (or 1).

\textbf{addMatImp} should the matrix containing the improvements due to the prime implicants in each of the iterations be added to the output? (For each of the prime implicants, the importance is computed by the average over the \( B \) improvements.) Must be set to \texttt{TRUE}, if standardized importances should be computed using \texttt{vim.norm}, or if permutation based importances should be computed using \texttt{vim.perm}.

\textbf{fast} should a greedy search (as implemented in \texttt{logreg}) be used instead of simulated annealing?

\textbf{rand} numeric value. If specified, the random number generator will be set into a reproducible state.

\textbf{...} for the \texttt{formula} method, optional parameters to be passed to the low level function \texttt{logicFS.default}. Otherwise, ignored.

\textbf{Value}

An object of class \texttt{logicFS} containing

\textbf{primes} the prime implicants,

\textbf{vim} the importance of the prime implicants,

\textbf{prop} the proportion of logic regression models that contain the prime implicants,
type the type of model (1: classification, 2: linear regression, 3: logistic regression),
param further parameters (if addInfo = TRUE),
mat.imp the matrix containing the improvements if addMatImp = TRUE, otherwise, NULL,
measure the name of the used importance measure,
useN the value of useN,
threshold NULL,
mu NULL.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

plot.logicFS, logic.bagging

Examples

## Not run:
# Load data.
data(data.logicfs)

# For logic regression and hence logic.fs, the variables must be binary. data.logicfs, however, contains categorical data with realizations 1, 2 and 3. Such data can be transformed into binary data by
bin.snps<-make.snp.dummy(data.logicfs)

# To speed up the search for the best logic regression models only a small number of iterations is used in simulated annealing.
my.anneal<-logreg.anneal.control(start=2,end=-2,iter=10000)

# Feature selection using logic regression is then done by
log.out<-logicFS(bin.snps,cl.logicfs,B=20,nleaves=10,
rand=123,anneal.control=my.anneal)

# The output of logic.fs can be printed
log.out

# One can specify another number of interactions that should be printed, here, e.g., 15.
print(log.out,topX=15)

# The variable importance can also be plotted.
plot(log.out)
logic.oob  

**Prime Implicants**

Description

Computes the out-of-bag error of the classification rule comprised by a `logicBagg` object.

Usage

```r
logic.oob(log.out, prob.case = 0.5)
```

Arguments

- `log.out`: an object of class `logicBagg`, i.e. the output of `logic.bagging`.
- `prob.case`: a numeric value between 0 and 1. If the logic regression models are logistic regression models, i.e. if in `logic.bagging` `ntree` is set to a value larger than 1, or `glm.if.1tree` is set to `TRUE`, then an observation will be classified as case (or more exactly, as 1) if the class probability is larger than `prob.case`.

Value

The out-of-bag error estimate.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

See Also

`logic.bagging`

logic.pimp  

**Prime Implicants**

Description

Determines the prime implicants contained in the logic regression models comprised in an object of class `logicBagg`.

Usage

```r
logic.pimp(log.out)
```

Arguments

- `log.out`: an object of class `logicBagg`, i.e. the output of `logic.bagging`. 
Details

Since we are interested in all potentially interested interactions and not in a minimum set of them, logic.pimp and returns all prime implicants and not a minimum number of them.

Value

A list consisting of the prime implicants for each of the B logic regression models of log.out.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

See Also

logic.bagging, logicFS, prime.implicants

make.snp.dummy

SNPs to Dummy Variables

Description

Transforms SNPs into binary dummy variables.

Usage

make.snp.dummy(data)

Arguments

data a matrix containing only 1’s, 2’s and 3’s (see details). Each column of data corresponds to a SNP and each row to an observation.

Details

make.snp.dummy assumes that the homozygous dominant genotype is coded by 1, the heterozygous genotype by 2, and the homozygous recessive genotype by 3. For each SNP, 2 dummy variables are generated:

SNP.1 At least one of the bases explaining the SNP are of the recessive type.
SNP.2 Both bases are of the recessive type.

Value

A matrix with 2*ncol(data) columns containing 2 dummy variables for each SNP.

Note

See the R package scrime for more general functions for recoding SNPs.

Author(s)

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minDNF  

Minimum Disjunctive Normal Form

Description

Computes the prime implicants or the minimal disjunctive form, respectively, of a given truth table.

Usage

```r
prime.implicants(mat)
minDNF(mat)
```

Arguments

- `mat`  
a matrix containing only 0’s and 1’s. Each column of `mat` corresponds to a binary variable and each row to a combination of the variables for which the logic expression is TRUE.

Value

Either an object of class `minDNF` or of class `primeImp`. Both contain a vector of (a minimum number of) prime implicants. The `primeImp` additionally contains the prime implicant table.

Author(s)

Holger Schwender, (holger.schwender@udo.edu)

References


See Also

- `logic.pimp`

mlogreg  

Multinomial Logic Regression

Description

Performs a multinomial logic regression for a nominal response by fitting a logic regression model (with logit as link function) for each of the levels of the response except for the level with the smallest value which is used as reference category.
Usage

## S3 method for class 'formula':
mlogreg(formula, data, recdom = TRUE, ...)

## Default S3 method:
mlogreg(x, y, ntrees = 1, nleaves = 8, anneal.control = logreg.anneal.control(),
      select = 1, rand = NA, ...)

Arguments

- **formula**: an object of class formula describing the model that should be fitted.
- **data**: a data frame containing the variables in the model. Each column of data must correspond to a binary variable (coded by 0 and 1) or a factor (for details on factors, see recdom) except for the column comprising the response, and each row to an observation. The response must be a categorical variable with less than 10 levels. This response can be either a factor or of type numeric or character.
- **recdom**: a logical value or vector of length ncol(data) comprising whether a SNP should be transformed into two binary dummy variables coding for a recessive and a dominant effect. If TRUE (logical value), then all factors (variables) with three levels will be coded by two dummy variables as described in make.snp.dummy. Each level of each of the other factors (also factors specifying a SNP that shows only two genotypes) is coded by one indicator variable. If FALSE (logical value), each level of each factor is coded by an indicator variable. If recdom is a logical vector, all factors corresponding to an entry in recdom that is TRUE are assumed to be SNPs and transformed into the two binary variables described above. Each variable that corresponds to an entry of recdom that is TRUE (no matter whether recdom is a vector or a value) must be coded by the integers 1 (coding for the homozygous reference genotype), 2 (heterozygous), and 3 (homozygous variant).
- **x**: a matrix consisting of 0’s and 1’s. Each column must correspond to a binary variable and each row to an observation.
- **y**: either a factor or a numeric or character vector specifying the values of the response. The length of y must be equal to the number of rows of x.
- **ntrees**: an integer indicating how many trees should be used in the logic regression models. For details, see logreg in the LogicReg package.
- **nleaves**: a numeric value specifying the maximum number of leaves used in all trees combined. See the help page of the function logreg in the LogicReg package for details.
- **anneal.control**: a list containing the parameters for simulated annealing. For details, see the help page of logreg.anneal.control in the LogicReg package.
- **select**: numeric value. Either 0 for a stepwise greedy selection (corresponds to select = 6 in logreg) or 1 for simulated annealing.
- **rand**: numeric value. If specified, the random number generator will be set into a reproducible state.
- **...**: for the formula method, optional parameters to be passed to the low level function mlogreg.default. Otherwise, ignored.
Value

An object of class `mlogreg` composed of

- **model**: a list containing the logic regression models,
- **data**: a matrix containing the binary predictors,
- **cl**: a vector comprising the class labels,
- **ntrees**: a numeric value naming the maximum number of trees used in the logic regressions,
- **nleaves**: a numeric value comprising the maximum number of leaves used in the logic regressions,
- **fast**: a logical value specifying whether the faster search algorithm, i.e. the greedy search, has been used.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

`predict.mlogreg`, `logic.bagging`, `logicFS`

---

**plot.logicFS**  
**Variable Importance Plot**

Description

Generates a dotchart of the importance of the most important interactions for an object of class `logicFS` or `logicBagg`.

Usage

```r
## S3 method for class 'logicFS'
plot(x, topX = 15, cex = 0.9, pch = 16, col = 1, show.prop = FALSE,
     force.topX = FALSE, coded = TRUE, add.thres = TRUE, thres = NULL,
     include0 = TRUE, add.v0 = TRUE, v0.col = "grey50", main = NULL, ...)

## S3 method for class 'logicBagg'
plot(x, topX = 15, cex = 0.9, pch = 16, col = 1, show.prop = FALSE,
     force.topX = FALSE, coded = TRUE, include0 = TRUE, add.v0 = TRUE,
     v0.col = "grey50", main = NULL, ...)
```
Arguments

\( x \)  
an object of either class `logicFS` or `logicBagg`.

\( \text{topX} \)  
integer specifying how many interactions should be shown. If \( \text{topX} \) is larger than the number of interactions contained in \( x \) all the interactions are shown. For further information, see `force.topX`.

\( \text{cex} \)  
a numeric value specifying the relative size of the text and symbols.

\( \text{pch} \)  
specifies the used symbol. See the help of `par` for details.

\( \text{col} \)  
the color of the text and the symbols. See the help of `par` for how colors can be specified.

\( \text{show.prop} \)  
if TRUE the proportions of models that contain the interactions of interest are shown. If FALSE (default) the importances of the interactions are shown.

\( \text{force.topX} \)  
if TRUE exactly \( \text{topX} \) interactions are shown. If FALSE (default) all interactions up to the \( \text{topX} \)th most important one and all interactions having the same importance as the \( \text{topX} \)th most important one are shown.

\( \text{coded} \)  
should the coded variable names be displayed? Might be useful if the actual variable names are pretty long. The coded variable name of the \( j \)-th variable is \( X_j \).

\( \text{add.thres} \)  
should a vertical line marking the threshold for a prime implicant to be called important be drawn in the plot? If TRUE, this vertical line will be drawn at NULL.

\( \text{thres} \)  
non-negative numeric value specifying the threshold for a prime implicant to be called important. If NULL and \( \text{add.thres} = \text{TRUE} \), the suggested threshold from \( x \) will be used.

\( \text{include0} \)  
should the x-axis include zero regardless whether the importances of the shown interactions are much higher than 0?

\( \text{add.v0} \)  
should a vertical line be drawn at \( x = 0 \)? Ignored if \( \text{include0} = \text{FALSE} \) and all importances are larger than zero.

\( \text{v0.col} \)  
the color of the vertical line at \( x = 0 \). See the help page of `par` for how colors can be specified.

\( \text{main} \)  
character string naming the title of the plot. If NULL, the name of the importance measure is used.

\( \ldots \)  
Ignored.

Author(s)

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See Also

`logicFS`, `logic.bagging`
predict.logicBagg  Predict Method for logicBagg objects

Description
Prediction for test data using an object of class logicBagg.

Usage
## S3 method for class 'logicBagg':
predict(object, newData, prob.case = 0.5,
         type = c("class", "prob"), ...)

Arguments

object      an object of class logicBagg.
newData     a matrix or data frame containing new data. If omitted object$data, i.e. the original training data, are used. Each row of newData must correspond to a new observation. Each row of newData must contain the same variable as the corresponding column of the data matrix used in logic.bagging, i.e. x if the default method of logic.bagging has been used, or data without the column containing the response if the formula method has been used.
prob.case   a numeric value between 0 and 1. A new observation will be classified as case (or more exactly, as 1) if the class probability, i.e. the average of the predicted probabilities of the models (if the logistic regression approach of logic regression has been used), or the percentage of votes for class 1 (if the classification approach of logic regression has been used) is larger than prob.case. Ignored if type = "prob" or the response is quantitative.
type        character vector indicating the type of output. If "class", a numeric vector of zeros and ones containing the predicted classes of the observations (using the specification of prob.case) will be returned. If "prob", the class probabilities or percentages of votes for class 1, respectively, for all observations are returned. Ignored if the response is quantitative.
...
         Ignored.

Value
A numeric vector containing the predicted classes (if type = "class") or the class probabilities (if type = "prob") of the new observations if the classification or the logistic regression approach of logic regression is used. If the response is quantitative, the predicted value of the response for all observations in the test data set is returned.

Author(s)
Holger Schwender, ⟨holger.schwender@udo.edu⟩

See Also
logic.bagging
predict.mlogreg  

Predict Method for mlogreg Objects

Description
Prediction for test data using an object of class mlogreg.

Usage
## S3 method for class 'mlogreg':
predict(object, newData, type = c("class", "prob"), ...)

Arguments
- object: an object of class mlogreg, i.e. the output of the function mlogreg.
- newData: a matrix or data frame containing new data. If omitted object$data, i.e. the original training data, are used. Each row of newData must correspond to a new observation. Each row of newData must contain the same variable as the corresponding column of the data matrix used in mlogreg, i.e. x if the default method of mlogreg has been used, or data without the column containing the response if the formula method has been used.
- type: character vector indicating the type of output. If "class", a vector containing the predicted classes of the observations will be returned. If "prob", the class probabilities for each level and all observations are returned.
- ...: Ignored.

Value
A numeric vector containing the predicted classes (if type = "class"), or a matrix composed of the class probabilities (if type = "prob").

Author(s)
Holger Schwender, (holger.schwender@udo.edu)

See Also
mlogreg

Print a logicFS object

Description
Prints an object of class logicFS.

Usage
## S3 method for class 'logicFS':
print(x, topX = 5, show.prop = TRUE, coded = FALSE, digits = 2, ...)
Arguments

- `x`: an object of either class `logicFS`.
- `topX`: integer indicating how many interactions should be shown. Additionally to the topX most important interactions, any interaction having the same importance as the topX most important one are also shown.
- `show.prop`: should the proportions of models containing the interactions of interest also be shown?
- `coded`: should the coded variable names be displayed? Might be useful if the actual variable names are pretty long. The coded variable name of the j-th variable is $X_j$.
- `digits`: number of digits used in the output. Ignored.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

See Also

`logicFS, vim.logicFS`

---

**vim.chisq**

*ChiSquare Based Importance*

Description

Determining the importance of interactions found by `logic.bagging` or `logicFS` by Pearson’s ChiSquare Statistic. Only available for the classification and the logistic regression approach of logic regression.

Usage

```r
vim.chisq(object, data = NULL, cl = NULL)
```

Arguments

- `object`: either an object of class `logicFS` or the output of an application of `logic.bagging` with importance = TRUE.
- `data`: a data frame or matrix consisting of 0’s and 1’s in which each column corresponds to one of the explanatory variables used in the original analysis with `logic.bagging` or `logicFS`, and each row corresponds to an observation. Must be specified if `object` is an object of class `logicFS`, or `cl` is specified. If `object` is an object of class `logicBarg` and neither `data` nor `cl` is specified, `data` and `cl` stored in `object` is used to compute the ChiSquare statistics. It is, however, highly recommended to use new `data` to test the interactions contained in `object`, as they have been found using the `data` stored in `object`, and it is very likely that most of them will show up as interesting if they are tested on the same data set.
- `cl`: a numeric vector of 0’s and 1’s specifying the class labels of the observations in `data`. Must be specified either if `object` is an object of class `logicFS`, or if `data` is specified.
Details

Currently Pearson’s ChiSquare statistic is computed without continuity correction.

Contrary to vim.logicFS (and vim.norm and vim.perm), vim.chisq does neither take the
logic regression models into account nor uses the out-of-bag observations for computing the importances of the identified interactions. It “just” tests each of the found interactions on the whole data set by calculating Pearson’s ChiSquare statistic for each of these interactions. It is, therefore, highly recommended to use an independent data set for specifying the importances of these interactions with vim.chisq.

Value

An object of class logicFS containing

primes the prime implicants
vim the values of Pearson’s ChiSquare statistic,
prop NULL,
type NULL,
param further parameters (if object is the output of logicFS or vim.logicFS
with addInfo = TRUE),
mat.imp NULL,
measure "ChiSquare Based",
threshold the 1 - 0.05/m quantile of the ChiSquare distribution with one degree of freedom,
u null

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

See Also

logic.bagging, logicFS, vim.logicFS, vim.norm, vim.ebam

vim.ebam

EBAM Based Importance

Description

Determines the importance of interactions found by logic.bagging or logicFS by an Empirical Bayes Analysis of Microarrays (EBAM). Only available for the classification and the logistic regression approach of logic regression.

Usage

vim.ebam(object, data = NULL, cl = NULL, nameEBAM = NULL, ...)
vim.ebam

Arguments

object  either an object of class logicFS or the output of an application of logic.bagging with importance = TRUE.
data  a data frame or matrix consisting of 0’s and 1’s in which each column corresponds to one of the explanatory variables used in the original analysis with logic.bagging or logicFS, and each row corresponds to an observation. Must be specified if object is an object of class logicFS, or cl is specified. If object is an object of class logicBagg and neither data nor cl is specified, data and cl stored in object is used to compute the ChiSquare statistics. It is, however, highly recommended to use new data to test the interactions contained in object, as they have been found using the data stored in object, and it is very likely that most of them will show up as interesting if they are tested on the same data set.
cl  a numeric vector of 0’s and 1’s specifying the class labels of the observations in data. Must be specified either if object is an object of class logicFS, or if data is specified.
nameEBAM  a character string. If specified, then the output of the EBAM analysis is stored under this name in the global environment.
...  further arguments of ebam and cat.ebam. For details, see the help files of these functions from the package siggenes.

Details

For each interaction found by logic.bagging or logicFS, the posterior probability that this interaction is significant is computed using the Empirical Bayes Analysis of Microarrays (EBAM). These posterior probabilities are used as the EBAM based importances of the interactions.

The test statistic underlying this EBAM analysis is Pearson’s ChiSquare statistic. Currently, the value of this statistic is computed without continuity correction.

Contrary to vim.logicFS (and vim.norm and vim.perm), vim.ebam does neither take the logic regression models into account nor uses the out-of-bag observations for computing the importances of the identified interactions. It “just” tests each of the found interactions on the whole data set by calculating Pearson’s ChiSquare statistic for each of these interactions and performing an EBAM analysis. It is, therefore, highly recommended to use an independent data set for specifying the importances of these interactions with vim.ebam.

Value

An object of class logicFS containing

primes  the prime implicants,
vim  the posterior probabilities of the interactions,
prop  NULL,
type  NULL,
param  further parameters (if object is the output of logicFS or vim.logicFS with addInfo = TRUE),
mat.imp  NULL,
measure  "EBAM Based",
threshold  the value of delta used in the EBAM analysis (see help files for ebam); by default: 0.9,
mu  NULL.
Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

logic.bagging, logicFS, vim.logicFS, vim.norm, vim.chisq

vim.individual  VIM for Individual Variables

Description

Quantifies the importance of each individual variable occurring in at least one of the logic regression models found in the application of logic.bagging.

Usage

vim.individual(object, useN = NULL, iter = NULL, prop = TRUE, standardize = FALSE, mu = 0, addMatImp = FALSE, prob.case = 0.5, rand = NA)

Arguments

object  an object of class logicBagg, i.e. the output of logic.bagging
useN  logical specifying if the number of correctly classified out-of-bag observations should be used in the computation of the importance measure. If FALSE, the proportion of correctly classified oob observations is used instead. If NULL (default), then the specification of useN in object is used.
iter  integer specifying the number of times the values of the considered variable are permuted in the computation of its importance. If NULL (default), the values of the variable are not permuted, but the variable is removed from the model.
prop  should the proportion of logic regression models containing the respective variable also be computed?
standardize  should a standardized version of the individual variable importance measure be returned? For details, see mu.
mu  a non-negative numeric value. Ignored if standardize = FALSE. Otherwise, a t-statistic for testing the null hypothesis that the importance of the respective variable is equal to mu is computed.
addMatImp  should the matrix containing the improvements due to each of the variables in each of the logic regression models be added to the output?
prob.case  a numeric value between 0 and 1. If the logistic regression approach of logic regression has been used in logic.bagging, then an observation will be classified as a case (or more exactly, as 1), if the class probability of this observation is larger than prob.case. Otherwise, prob.case is ignored.
rnd  an integer for setting the random number generator in a reproducible case.
vim.logicFS

Value

An object of class logicFS containing

- `vim`: the importances of the variables,
- `prop`: the proportion of logic regression models containing the respective variable (if `prop = TRUE`) or NULL (if `prop = FALSE`),
- `primes`: the names of the variables,
- `type`: the type of model (1: classification, 2: linear regression, 3: logistic regression),
- `param`: further parameters (if `addInfo = TRUE` in the previous call of logic.bagging),
- `mat.imp`: either a matrix containing the improvements due to the variables for each of the models (if `addMatImp = TRUE`), or NULL (if `addMatImp = FALSE`),
- `measure`: the name of the used importance measure,
- `useN`: the value of `useN`,
- `threshold`: NULL if `standardize = FALSE`, otherwise the \(1 - 0.05/m\) quantile of the t-distribution with \(B - 1\) degrees of freedom, where \(m\) is the number of variables and \(B\) is the number of logic regression models composing object,
- `mu`: `mu` (if `standardize = TRUE`), or NULL (otherwise),
- `iter`: `iter`.

Author(s)

Holger Schwender, ⟨holger.schw@udo.edu⟩

References


See Also

logic.bagging, logicFS, vim.logicFS, vim.set, vim.ebam, vim.chisq

vim.logicFS  Importance Measures

Description

Computes the value of the single or the multiple tree measure, respectively, for each prime implicant contained in a logic bagging model to specify the importance of the prime implicant for classification, if the response is binary. If the response is quantitative, the importance is specified by a measure based on the mean square prediction error.

Usage

vim.logicFS(log.out, useN = TRUE, onlyRemove = FALSE, prob.case = 0.5, addInfo = FALSE, addMatImp = TRUE)
Arguments

log.out: an object of class logicBagg, i.e. the output of logic.bagging.
useN: logical specifying if the number of correctly classified out-of-bag observations should be used in the computation of the importance measure. If FALSE, the proportion of correctly classified oob observations is used instead.
onlyRemove: should in the single tree case the multiple tree measure be used? If TRUE, the prime implicants are only removed from the trees when determining the importance in the single tree case. If FALSE, the original single tree measure is computed for each prime implicant, i.e. a prime implicant is not only removed from the trees in which it is contained, but also added to the trees that do not contain this interaction. Ignored in all other than the classification case.
prob.case: a numeric value between 0 and 1. If the logistic regression approach of logic regression is used (i.e. if the response is binary, and in logic.bagging ntrees is set to a value larger than 1, or glm.if.1tree is set to TRUE), then an observation will be classified as a case (or more exactly as 1), if the class probability of this observation estimated by the logic bagging model is larger than prob.case.
addInfo: should further information on the logic regression models be added?
addMatImp: should the matrix containing the improvements due to the prime implicants in each of the iterations be added to the output? (For each of the prime implicants, the importance is computed by the average over the B improvements.) Must be set to TRUE, if standardized importances should be computed using vim.norm, or if permutation based importances should be computed using vim.perm.

Value

An object of class logicFS containing

primes: the prime implicants,
vim: the importance of the prime implicants,
prop: the proportion of logic regression models containing the prime implicants,
type: the type of model (1: classification, 2: linear regression, 3: logistic regression),
param: further parameters (if addInfo = TRUE),
mat.imp: the matrix containing the improvements if addMatImp = TRUE, otherwise, NULL,
measure: the name of the used importance measure,
useN: the value of useN,
threshold: NULL,
mu: NULL.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References

See Also

logic.bagging, logicFS, vim.norm, vim.perm

vim.norm

Standardized and Permutation Based Importance Measure

Description

Computes a standarized or a permutation based version of either the Single Tree Measure, the Quantitative Response Measure, or the Multiple Tree Measure.

Usage

vim.norm(object, mu = 0)

vim.perm(object, mu = 0, n.perm = 10000, n.subset = 1000,
        adjust = "bonferroni", rand = NA)

Arguments

object either the output of logicFS or vim.logicFS with addMatImp = TRUE, or the output of logic.bagging with importance = TRUE and addMatImp = TRUE.
mu a non-negative numeric value. Default is zero. However, mu should actually be set to a value larger than zero. See Details.
n.perm the number of (sign) permutations used in vim.perm.
n.subset an integer specifying how many permutations should be considered at once.
adjust character vector naming the method with which the raw permutation based p-values are adjusted for multiplicity. If "qvalue", the function qvalue.cal from the package siggenes is used to compute q-values. Otherwise, p.adjust is used to adjust for multiple comparisons. See p.adjust for all other possible specifications of adjust. If "none", the raw p-values will be used. For more details, see Details.
rand an integer for setting the random number generator in a reproducible case.

Details

In both vim.norm and vim.perm, an one-sample t-statistic is computed for each prime implicant, where the numerator is given by $VIM - \mu$ with VIM being the single or the multiple tree importance, and the denominator is the corresponding standard error computed by employing the B improvements of the considered prime implicant in the B logic regression models. (Note that VIM is the mean over these B improvements.)

As using $\mu = 0$ might lead to calling a prime implicant important, even though it actually shows only improvements of 1 or 0, $\mu$ should be set to a value larger than zero.

In vim.norm, the value of this t-statistic is returned as the standardized importance of a prime implicant. The larger this value, the more important is the prime implicant. (This applies to all importance measures – at least for those contained in this package.) Assuming normality, a possible threshold for a prime implicant to be considered as important is the $1 - 0.05/m$ quantile of the t-distribution with $B - 1$ degrees of freedom, where $m$ is the number of prime implicants.
In `vim.perm`, the sign permutation is used to determine `n.perm` permuted values of the one-sample t-statistic, and to compute the raw p-values for each of the prime implicants. Afterwards, these p-values are adjusted for multiple comparisons using the method specified by `adjust`. The permutation based importance of a prime implicant is then given by $1 - \text{these adjusted p-values}$. Here, a possible threshold for calling a prime implicant important is 0.95.

Value

An object of class `logicFS` containing

- `primes`: the prime implicants,
- `vim`: the respective importance of the prime implicants,
- `prop`: NULL,
- `type`: the type of model (1: classification, 2: linear regression, 3: logistic regression),
- `param`: further parameters (if `addInfo = TRUE`),
- `mat.imp`: NULL,
- `measure`: the name of the used importance measure,
- `useN`: the value of `useN` from the original analysis with, e.g., `logicFS`,
- `threshold`: the threshold suggested in Details,
- `mu`: `mu`.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

- `logic.bagging`, `logicFS`, `vim.logicFS`, `vim.chisq`, `vim.ebam`

---

`vim.set` *VIM for Sets of Variables*

Description

Quantifies the importances of sets of variables contained in a logic bagging model.

Usage

```r
vim.set(object, set = NULL, useN = NULL, iter = NULL, standardize = FALSE, mu = 0, addMatImp = FALSE, prob.case = 0.5, rand = NA)
```
Arguments

object

an object of class logicBagg, i.e. the output of logic.bagging.

set

either a list or a character or numeric vector.

If NULL (default), then it will be assumed that data, i.e. the data set used in the application of logic.bagging, has been generated using make.snp.dummy or similar functions for coding variables by binary variables, i.e. with a function that splits a variable, say SNPx, into the dummy variables SNPx.1, SNPx.2, ... (where the “.” can also be any other sign, e.g., an underscore).

If a character or a numeric vector, then the length of set must be equal to the number of variables used in object, i.e. the number of columns of data in the logicBagg object, and must specify the set to which a variable belongs either by an integer between 1 and the number of sets, or by a set name. If a variable should not be included in any of the sets, set the corresponding entry of set to NA. Using this specification of set it is not possible to assign a variable to more than one sets. For such a case, set set to a list (as follows).

If set is a list, then each object in this list represents a set of variables. Therefore, each object must be either a character or a numeric vector specifying either the names of the variables that belongs to the respective set or the columns of data that contains these variables. If names(set) is NULL, generic names will be employed as names for the sets. Otherwise, names(set) are used.

useN

logical specifying if the number of correctly classified out-of-bag observations should be used in the computation of the importance measure. If FALSE, the proportion of correctly classified oob observations is used instead. If NULL (default), then the specification of useN in object is used.

iter

integer specifying the number of times the values of the variables in the respective set are permuted in the computation of the importance of this set. If NULL (default), the values of the variables are not permuted, but all variables belonging to the set are removed from the model.

standardize

should a standardized version of the importance measure for a set of variables be returned? For details, see mu.

mu

a non-negative numeric value. Ignored if standardize = FALSE. Otherwise, a t-statistic for testing the null hypothesis that the importance of the respective set is equal to mu is computed.

addMatImp

should the matrix containing the improvements due to each of the sets in each of the logic regression models be added to the output?

prob.case

a numeric value between 0 and 1. If the logistic regression approach of logic regression has been used in logic.bagging, then an observation will be classified as a case (or more exactly, as 1), if the class probability of this observation is larger than prob.case. Otherwise, prob.case is ignored.

rand

an integer for setting the random number generator in a reproducible case.

Value

An object of class logicFS containing

vim

the importances of the sets of variables,

prop

NULL,

primes

the names of the sets of variables,

type

the type of model (1: classification, 2: linear regression, 3: logistic regression),
param further parameters (if addInfo = TRUE in the previous call of logic.bagging), or NULL (otherwise),

mat.imp either a matrix containing the improvements due to the sets of variables for each of the models (if addMatImp = TRUE), or NULL (if addMatImp = FALSE),

measure the name of the used importance measure,

threshold NULL if standardize = FALSE, otherwise the $1 - 0.05/m$ quantile of the t-distribution with $B - 1$ degrees of freedom, where $m$ is the number of sets and $B$ is the number of logic regression models composing object,

mu $\mu$ (if standardize = TRUE), or NULL (otherwise),

iter iter.

Author(s)

Holger Schwender, ⟨holger.schwender@udo.edu⟩

References


See Also

logic.bagging, logicFS, vim.logicFS, vim.set, vim.ebam, vim.chisq
Index

*Topic datasets
   data.logicfs, 1
*Topic hplot
   plot.logicFS, 14
*Topic htest
   logic.oob, 9
   vim.chisq, 17
   vim.ebam, 19
   vim.individual, 20
   vim.logicFS, 22
   vim.norm, 23
   vim.set, 25
*Topic internal
   logicFS-internal, 5
*Topic logic
   logic.pimp, 10
   minDNF, 11
   vim.chisq, 17
   vim.ebam, 19
   vim.individual, 20
   vim.logicFS, 22
   vim.norm, 23
   vim.set, 25
*Topic manip
   make.snp.dummy, 10
*Topic multivariate
   logicFS, 6
*Topic optimize
   minDNF, 11
*Topic print
   minDNF, 11
   print.logicFS, 17
*Topic regression
   logic.bagging, 2
   logicFS, 6
   mlogreg, 12
   predict.logicBagg, 15
   predict.mlogreg, 16
*Topic tree
   logic.bagging, 2
   logicFS, 6
   mlogreg, 12
*Topic utilities
   getMatEval, 1
   logic.oob, 9
   logic.pimp, 10
   check.listprimes (logicFS-internal), 5
   check.mat.imp (logicFS-internal), 5
   checkDataCl (logicFS-internal), 5
   checkNewTree (logicFS-internal), 5
   checkSet (logicFS-internal), 5
   cl.logicfs (data.logicfs), 1
   compLarger (logicFS-internal), 5
   compMatImpIndividual1 (logicFS-internal), 5
   compMatImpIndividual3 (logicFS-internal), 5
   compMatImpSet1 (logicFS-internal), 5
   compMatImpSet3 (logicFS-internal), 5
   compMatProbMLR (logicFS-internal), 5
   contr.none (logicFS-internal), 5
   contr.snps (logicFS-internal), 5
   correctPreds1 (logicFS-internal), 5
   correctPredsPermute1 (logicFS-internal), 5
   correctPredsPermute2 (logicFS-internal), 5
   correctPredsPermute3 (logicFS-internal), 5
   correctPredsRemove1 (logicFS-internal), 5
   correctPredsRemove2 (logicFS-internal), 5
   correctPredsRemove3 (logicFS-internal), 5
   correctSetPermute1 (logicFS-internal), 5
   correctSetPermute2 (logicFS-internal), 5
   correctSetPermute3 (logicFS-internal), 5
   cl.logicfs (data.logicfs), 1
INDEX

correctSetPermute3 (logicFS-internal), 5
correctSetRemove1 (logicFS-internal), 5
correctSetRemove2 (logicFS-internal), 5
correctSetRemove3 (logicFS-internal), 5
cyclic.covering (logicFS-internal), 5
data.logicfs, 1
generateTruthTab (logicFS-internal), 5
getKnots (logicFS-internal), 5
getMatPrime (logicFS-internal), 5
getNames (logicFS-internal), 5
getNewProbsMLR (logicFS-internal), 5
getNewTree (logicFS-internal), 5
getPerms (logicFS-internal), 5
getPimps (logicFS-internal), 5
getTree (logicFS-internal), 5
getVarInTree (logicFS-internal), 5
getXy (logicFS-internal), 5
getXyPred (logicFS-internal), 5
getY (logicFS-internal), 5
ia.samp (logicFS-internal), 5
logic.bagging, 1, 2, 8–10, 13, 15, 16, 18, 20, 22–26
logic.oob, 9
logic.pimp, 10, 12
logicFS, 1, 5, 6, 10, 13, 15, 17, 18, 20, 22–26
logicFS-internal, 5
make.snp.dummy, 3, 6, 10, 12, 25
makeContrastList (logicFS-internal), 5
minDNF, 11
minimizePI (logicFS-internal), 5
mlogreg, 3, 6, 7, 12, 16
mlogreg.factor (logicFS-internal), 5
modelMat (logicFS-internal), 5
oobMLR (logicFS-internal), 5
pimpMLR (logicFS-internal), 5
plot.logicBagg (plot.logicFS), 14
plot.logicFS, 8, 14
predict.logicBagg, 5, 15
predict.logregmodel (logicFS-internal), 5
predict.mlogreg, 13, 16
predictMLB (logicFS-internal), 5
prime.implicants, 10
prime.implicants (minDNF), 11
print.logicBagg (logic.bagging), 2
print.logicFS, 17
print.minDNF (minDNF), 11
print.mlogreg (mlogreg), 12
print.primeImp (minDNF), 11
rm.dom (logicFS-internal), 5
standardizeMatImp (logicFS-internal), 5
vim.chisq, 17, 20, 22, 25, 26
vim.ebam, 18, 19, 22, 25, 26
vim.individual, 20
vim.lm (logicFS-internal), 5
vim.logicFS, 17–20, 22, 24–26
vim.MLR (logicFS-internal), 5
vim.multiple (logicFS-internal), 5
vim.norm, 4, 7, 18–20, 22, 23
vim.perm, 4, 7, 18, 19, 22, 23
vim.perm (vim.norm), 23
vim.set, 22, 25, 26
vim.singleBoth (logicFS-internal), 5
vim.singleRemove (logicFS-internal), 5