Package ‘bbl’

January 28, 2022

Title  Boltzmann Bayes Learner
Version  1.0.0
Maintainer  Jun Woo <junwoo035@gmail.com>
Depends  R (>= 3.6.0)
Imports  methods, stats, utils, Rcpp (>= 0.12.16), pROC, RColorBrewer
Description  Supervised learning using Boltzmann Bayes model inference, which extends naive Bayes model to include interactions. Enables classification of data into multiple response groups based on a large number of discrete predictors that can take factor values of heterogeneous levels. Either pseudo-likelihood or mean field inference can be used with L2 regularization, cross-validation, and prediction on new data.
<doi:10.18637/jss.v101.i05>.
License  GPL (>= 2)
LinkingTo  Rcpp
Encoding  UTF-8
RoxygenNote  7.1.2
NeedsCompilation  yes
Suggests  glmnet, BiocManager, Biostrings
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Repository  CRAN
Date/Publication  2022-01-27 23:20:12 UTC

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Main driver for bbl inference

Usage

```r
bbl(
    formula,
    data,
    weights,
    xlevels = NULL,
    verbose = 1,
    method = "pseudo",
    novarOk = FALSE,
    testNull = TRUE,
    prior.count = 1,
    ...
)
```
Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>formula</code></td>
<td>Formula for modeling</td>
</tr>
<tr>
<td><code>data</code></td>
<td>Data for fitting</td>
</tr>
<tr>
<td><code>weights</code></td>
<td>Vector of weights for each instance in data. Restricted to non-negative integer frequencies, recoding the number of times each row of data must be repeated. If <code>NULL</code>, assumed to be all 1. Fractional weights are not supported. Can be a named column in <code>data</code></td>
</tr>
<tr>
<td><code>xlevels</code></td>
<td>List of factor levels for predictors. If <code>NULL</code>, will be inferred from data with factor levels ordered alphanumerically.</td>
</tr>
<tr>
<td><code>verbose</code></td>
<td>Output verbosity level. Will be send to down-stream function calls with one level lower</td>
</tr>
<tr>
<td><code>method</code></td>
<td>BB inference algorithm; pseudo-likelihood inference ('pseudo') or mean field ('mf')</td>
</tr>
<tr>
<td><code>novarOk</code></td>
<td>If <code>TRUE</code>, will proceed with predictors having only one level</td>
</tr>
<tr>
<td><code>testNull</code></td>
<td>Repeat the inference for the ‘pooled’ sample; i.e., under the null hypothesis of all rows in data belonging to a single group</td>
</tr>
<tr>
<td><code>prior.count</code></td>
<td>Prior count for computing single predictor and pairwise frequencies</td>
</tr>
<tr>
<td>...</td>
<td>Other parameters to <code>mestimate</code></td>
</tr>
</tbody>
</table>

Details

Formula argument and data are used to tabulate `xlevels` unless explicitly given as list. Data are expected to be factors or integers. This function is a driver interpreting formula and calls `bbi.fit`. Will stop with error if any predictor has only one level unless `novarOk='TRUE'`. Use `removeConst` to remove the non-varying predictors before calling if this happens.

Value

A list of class `bb1` with the following elements:

<table>
<thead>
<tr>
<th>Element</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>coefficients</code></td>
<td>List of inferred coefficients with elements <code>h</code>, <code>J</code>, <code>h0</code>, and <code>J0</code>. The bias parameter <code>h</code> is a list of length equal to no. of response groups, each of which is a list of the same structure as <code>xlevels</code>: length equal to no. of predictors, containing vectors of length equal to each predictor factor levels: $h_i^{(y)}(x)$ represented by $h[[y]][[i]][x]$. The interaction parameter <code>J</code> is a list of lists of dimension $m \times m$, where $m$ is the number of predictors. Each element is a matrix of dimension $L_i \times L_j$, where $L_i$ and $L_j$ are numbers of factor levels in predictor $i$ and $j$: $J_{ij}^{(y)}(x_1, x_2)$ represented by $J[[y]][[i]][[j]][x1, x2]$. All elements of lists are named. The pooled parameters <code>h0</code> and <code>J0</code>, if computed, are of one less dimension, omitting response group argument.</td>
</tr>
<tr>
<td><code>xlevels</code></td>
<td>List of vectors containing predictor levels.</td>
</tr>
<tr>
<td><code>terms</code></td>
<td>The terms of <code>formula</code> input.</td>
</tr>
<tr>
<td><code>groups</code></td>
<td>Vector of response groups.</td>
</tr>
<tr>
<td><code>groupname</code></td>
<td>Name of the response variable.</td>
</tr>
</tbody>
</table>
Matrix of logicals whose elements record whether formula includes interaction between the two predictors.

Model data frame derived from formula and data.

Log likelihood.

Vector log partition function. Used in predict.

Vector of integral weights (frequencies).

Function call.

Degrees of freedom.

Author(s)

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References

doi: 10.18637/jss.v101.i05

Examples

titanic <- as.data.frame(Titanic)
b <- bbl(Survived ~ (Class + Sex + Age)^2, data = titanic, weights = Freq)
b

bbl.fit

bbl Inference with model matrix

Description

Performs bbl inference using response vector and predictor matrix

Usage

bbl.fit(
  x,
  y,
  qJ = NULL,
  weights = NULL,
  xlevels = NULL,
  verbose = 1,
  method = "pseudo",
  prior.count = 1,
  ...
)
crossVal

Arguments

- **x**: Data frame of factors with each predictor in columns.
- **y**: Vector of response variables.
- **qJ**: Matrix of logicals indicating which predictor combinations are interacting.
- **weights**: Vector of non-negative integer frequencies, recoding the number of times each row of data must be repeated. If NULL, assumed to be all 1. Fractional weights are not supported.
- **xlevels**: List of factor levels for predictors. If NULL, will be inferred from data with factor levels ordered alphanumerically.
- **verbose**: Verbosity level of output. Will be propagated to mestimate with one level down.
- **method**: c('pseudo', 'mf'); inference method.
- **prior.count**: Prior count for computing single predictor and pairwise frequencies
- **...**: Other arguments to mestimate.

Details

This function would normally be called by bbl rather than directly. Expects the predictor data x and response vector y instead of formula input to bbl.

Value

List of named components h, J, lkh, and lz; see bbl for information regarding these components.

Examples

titanic <- as.data.frame(Titanic)
freq <- titanic$Freq
x <- titanic[,1:3]
y <- titanic$Survived
b <- bbl.fit(x=x, y=y, weights=freq)
b

---

crossVal  

Cross-Validation of BB Learning

Description

Run multiple fittings of bbl model with training/validation division of data
Usage

crossVal(
  formula,
  data,
  weights,
  novarOk = FALSE,
  lambda = 1e-05,
  lambdad = 0,
  eps = 0.9,
  nfold = 5,
  method = "pseudo",
  use.auc = TRUE,
  verbose = 1,
  progress.bar = FALSE,
  storeOpt = TRUE,
  ...
)

Arguments

formula Formula for model. Note that intercept has no effect.
data Data frame of data. Column names must match formula.
weights Frequency vector of how many times each row of data must be repeated. If NULL, defaults to vector of 1s. Fractional values are not supported.
novarOk Proceed even when there are predictors with only one factor level.
lambda Vector of L2 penalizer values for method = 'pseudo'. Inferences will be repeated for each value. Restricted to non-negative values.
lambdad L2 penalizer in method = 'pseudo' applied to parameter h. In contrast to lambda, only a single value is allowed.
eps Vector of regularization parameters, \( \epsilon \in [0, 1] \), for method = 'mf'. Inference will be repeated for each value.
nfold Number of folds for training/validation split.
method c('pseudo', 'mf') for pseudo-likelihood maximization or mean field.
use.auc Use AUC as the measure of prediction accuracy. Only works if response groups are binary. If FALSE, mean prediction group accuracy will be used as score.
verbose Verbosity level. Downgraded when relayed into bbl.
progress.bar Display progress bar in predict.
storeOpt Store the optimal fitted object of class bbl.

Details

The data slot of object is split into training and validation subsets of (nfold-1):1 ratio. The model is trained with the former and validated on the latter. Individual division/fold results are combined into validation result for all instances in the data set and prediction score is evaluated using the known response group identity.
Value

Object of class cv.bbl extending bbl, a list with extra components: regstar, Value of regularization parameter, lambda and eps for method=’pseudo’ and method=’mf’, respectively, at which the accuracy score is maximized; maxscore, Value of maximum accuracy; cvframe, Data frame of regularization parameters and scores scanned. If use.auc=TRUE, also contains 95

Examples

```r
set.seed(513)
m <- 5
n <- 100
predictors <- list()
for(i in 1:m) predictors[[i]] <- c('a','c','g','t')
names(predictors) <- paste0('v',1:m)
par <- list(randompar(predictors), randompar(predictors, h0=0.1, J0=0.1))
dat <- randomsamp(predictors, response=c('ctrl','case'), par=par, nsample=n)
cv <- crossVal(y ~ .^2, data=dat, method='mf', eps=seq(0.1,0.9,0.1))
cv
```

fitted.bbl  

Fitted Response Group Probabilities

Description

Response group probabilities from BBL fit

Usage

```r
## S3 method for class 'bbl'
fitted(object, ...)
```

Arguments

- object Object of class bbl.
- ... Other arguments

Details

This method returns predicted response group probabilities of training data

Value

Matrix of response group probabilities with data points in rows and response groups in columns

Examples

```r
titanic <- as.data.frame(Titanic)
fit <- bbl(Survived ~ Class + Sex + Age, data=titanic, weights=titanic$Freq)
```
**formula.bbl**  
*Formula in BBL Fitting*

**Description**

Returns the formula used in BBL fit

**Usage**

```r
## S3 method for class 'bbl'
formula(x, ...)
```

**Arguments**

- `x`: Object of class `bbl`
- `...`: Other arguments

**Value**

Formula object

**Examples**

```r
titanic <- as.data.frame(Titanic)
fit <- bbl(Survived ~ Class + Sex + Age, data=titanic, weights=titanic$Freq)
formula(fit)
```

---

**freq2raw**  
*Convert Frequency Table into Raw Data*

**Description**

Data with unique rows and a frequency column is converted into data with duplicate rows.

**Usage**

```r
freq2raw(data, freq)
```

**Arguments**

- `data`: Data frame with factors in columns
- `freq`: Vector of frequency of each row in `data`; can be a named column in `data`; if missing, the column `Freq` is looked for in `data`
Details

The output data frame can be used as input to \texttt{bb1}.

Value

Data frame with one row per instances

Examples

Titanic
\begin{verbatim}
x <- as.data.frame(Titanic)
head(x)
titanic <- freq2raw(data=x[,1:3], freq=x$Freq)
head(titanic)
\end{verbatim}

Description

Compute log likelihood from a fitted \texttt{bb1} object

Usage

\begin{verbatim}
## S3 method for class 'bb1'
logLik(object, ...)
\end{verbatim}

Arguments

\begin{verbatim}
object Object of class \texttt{bb1}
...
Other arguments to methods
\end{verbatim}

Details

This method uses inferred parameters from calls to \texttt{bb1} and data to compute the log likelihood.

Value

An object of class \texttt{logLik}, the Log likelihood value and the attribute "df" (degrees of freedom), the number of parameters.
**mcSample**  
*Sample Predictor Distributions*

**Description**

Uses fitted BBL model to explore predictor distributions

**Usage**

```r
mcSample(object, nsteps = 1000, verbose = 1, progress.bar = TRUE)
```

**Arguments**

- `object`: Object of class `bbl`
- `nsteps`: Total number of MC steps
- `verbose`: Verbosity level of output
- `progress.bar`: Display progress bar

**Details**

After `bbl` fit, the resulting model is used by this function to sample predictor distributions in each response group and find the most likely predictor set using MCMC.

**Examples**

```r
titanic <- as.data.frame(Titanic)
b <- bbl(Survived~., data=titanic[,1:4], weights=titanic$Freq)
pxy <- mcSample(b)
pxy
```

---

**mlestimate**  
*Maximum likelihood estimate*

**Description**

Perform inference of bias and interaction parameters for a single response group
Usage

mlestimate(
  xi,
  weights = NULL,
  qJ = NULL,
  method = "pseudo",
  L = NULL,
  lambda = 1e-05,
  lambdah = 0,
  symmetrize = TRUE,
  eps = 0.9,
  nprint = 100,
  itmax = 1e+05,
  tolerance = 1e-04,
  verbose = 1,
  prior.count = 1,
  naive = FALSE,
  lz.half = FALSE
)

Arguments

xi Data matrix; expected to be numeric with elements ranging from zero to positive integral upper bound L-1.
weights Frequency vector of number of times each row of xi is to be repeated. If NULL, defaults to 1. Expected to be non-negative integers.
qJ Matrix of logicals indicating which predictor pairs are interacting. If NULL, all are allowed.
method c(‘pseudo’, ‘mf’) for pseudo-likelihood maximization or mean field inference.
L Vector of number of factor levels in each predictor. If NULL, will be inferred from xi.
lambda Vector of L2 regularization parameters for method = ‘pseudo’. Applies to interaction parameters J.
lambdah L2 parameters for h in ‘pseudo’. If NULL, it is set equal to lambda. lambdah = 0 will free h from penalization.
symmetrize Enforce the symmetry of interaction parameters by taking mean values of the matrix and its trace: $J^{(y)}_{ij}(x_1, x_2) = J^{(y)}_{ji}(x_2, x_1)$.
eps Vector of regularization parameters for mf. Must be within the range of $\epsilon \in [0, 1]$.
nprint Frequency of printing iteration progress under ‘pseudo’.
itmax Maximum number of iterations for ‘pseudo’.
tolerance Upper bound for fractional changes in pseudo-likelihood values before terminating iteration in ‘pseudo’.
verbose   Verbosity level.
prior.count Prior count for method = 'mf' to reduce numerical instability.
a naive    Naive Bayes inference. Equivalent to method = 'mf' together with eps = 0.
lz.half Divide interaction term in approximation to ln Z_lij in 'pseudo'.

Details

Given numeric data matrix, either pseudo-likelihood of mean-field theory is used to find the maximum likelihood estimate of bias h and interaction J parameters. Normally called by bbl rather than directly.

Value

List of inferred parameters h and J. See bbl for parameter structures.

Examples

set.seed(535)
predictors <- list()
for(i in 1:5) predictors[[i]] <- c('a','c','g','t')
par <- randompar(predictors)
par
xi <- sample_xi(nsample=5000, predictors=predictors, h=par$h, J=par$J,
                  code_out=TRUE)
head(xi)
ps <- mlestimate(xi=xi, method='pseudo', lambda=0)
ps$h
ps$J[[1]]
mf <- mlestimate(xi=xi, method='mf', eps=0.9)
plot(x=unlist(par$h), y=unlist(ps$h), xlab='True', ylab='Inferred')
segments(x0=-2, x1=2, y0=-2, y1=2, lty=2)
points(x=unlist(par$J), y=unlist(ps$J), col='red')
points(x=unlist(par$h), y=unlist(mf$h), col='blue')
points(x=unlist(par$J), y=unlist(mf$J), col='green')
nobs.bbl

Arguments

formula Object of class bbl
... Other arguments

Value

Data frame used for fitting

Examples

titanic <- as.data.frame(Titanic)
fit <- bbl(Survived ~ Class + Sex + Age, data=titanic[,1:4], weights=titanic$Freq)
head(model.frame(fit))

nobs.bbl

Number of Observations in BBL Fit

Description

Returns the number of observations from a BBL fit

Usage

## S3 method for class 'bbl'
nobs(object, ...)

Arguments

object Object of class bbl
... Other arguments

Value

An integer of number of observations

Examples

titanic <- as.data.frame(Titanic)
fit <- bbl(Survived ~ Class + Sex + Age, data=titanic[,1:4], weights=titanic$Freq)
nobs(fit)
plot.bbl  

Plot bbl object

Description

Visualize bias and interaction parameters

Usage

```r
## S3 method for class 'bbl'
plot(x, layout = NULL, hcol = NULL, Jcol = NULL, npal = 100, ...)
```

Arguments

- `x`: Object of class `bbl`
- `layout`: Matrix of layouts for arrangement of linear and interaction parameters. If `NULL`, the top half will be used for linear parameter barplot and bottom half will be divided into interaction heatmaps for each response group.
- `hcol`: Color for linear barplots. Grayscale if `NULL`.
- `Jcol`: Color for interaction heatmaps. Default (NULL) is RdBu from RColorBrewer.
- `npal`: Number of color scales.
- `...`: Other graphical parameters for `plot`.

Details

This method displays a barplot of bias parameters and heatmaps (one per response group) of interaction parameters. All parameters are offset by the pooled values (single group inference) unless missing.

plot.cv.bbl  

Plot Cross-validation Outcome

Description

Plot cross-validation score as a function of regularization parameter
### predict.bbl

**Usage**

```r
## S3 method for class 'cv.bbl'
plot(
  x,
  type = "b",
  log = "x",
  pch = 21,
  bg = "white",
  xlab = NULL,
  ylab = NULL,
  las = 1,
  ...
)
```

**Arguments**

- `x` Object of class `cv.bbl` from a call to `crossVal`
- `type` Symbol type in `plot`, present here to set default.
- `log` Log scale argument to `plot`.
- `pch` Symbol type code in `par`.
- `bg` Symbol background color in `par`.
- `xlab` X axis label
- `ylab` Y axis label
- `las` Orientation of axis labels in `par`.
- `...` Other arguments to `plot`.

**Details**

This function will plot accuracy score as a function of regularization parameter from a call to `crossVal`.

---

**Description**

Make prediction of response group identity based on trained model

**Usage**

```r
## S3 method for class 'bbl'
predict(object, newdata, type = "link", verbose = 1, progress.bar = FALSE, ...)
```
predict.bbl

Arguments

object Object of class bbl containing trained model
newdata Data frame of new data for which prediction is to be made. Columns must contain all of those in model@data. If column names are present, the columns will be matched based on them. Extra columns will be ignored. If column names are not provided, the columns should exactly match model@data predictor parts. If NULL, replaced by model@data (self-prediction).
type Return value type. If 'link', the logit scale probabilities. If 'prob' the probability itself.
verbose Verbosity level
progress.bar Display progress of response group probability. Useful for large samples.
... Other arguments to methods

Details

This method uses a new data set for predictors and trained bbl model parameters to compute posterior probabilities of response group identity.

Value

Data frame of predicted posterior probabilities with samples in rows and response groups in columns. The last column is the predicted response group with maximum probability.

Examples

```r
set.seed(154)

m <- 5
L <- 3
n <- 1000

predictors <- list()
for(i in 1:m) predictors[[i]] <- seq(0,L-1)
names(predictors) <- paste0('v',1:m)
par <- list(randompar(predictors=predictors, dJ=0.5),
            randompar(predictors=predictors, h0=0.1, J0=0.1, dJ=0.5))
dat <- randomsamp(predictors=predictors, response=c('ctrl','case'), par=par,
                  nsample=n)
dat <- dat[sample(n),]
dtrain <- dat[seq(n/2),]
dtest <- dat[seq(n/2+1,n),]

model <- bbl(y ~ .^2, data=dtrain)
pred <- predict(model, newdata=dtest)
score <- mean(dtest$y==pred$yhat)
score

auc <- pROC::roc(response=dtest$y, predictor=pred$case, direction='<')$auc
auc
```
predict.cv.bbl

Description
Use the optimal fitted model from cross-validation run to make prediction

Usage

```r
## S3 method for class 'cv.bbl'
predict(object, ...)
```

Arguments

- `object`: Object of class `cv.bbl`
- `...`: Other parameters to `predict.bbl`

Details
This method will use the fitted model with maximum accuracy score returned by a call to `crossVal` to make prediction on new data

Value
Data frame of prediction; see `predict.bbl`.

print.bbl

Description
This method displays model structure and first elements of coefficients

Usage

```r
## S3 method for class 'bbl'
print(x, showcoeff = TRUE, maxcoeff = 3L, ...)
```

Arguments

- `x`: An object of class `bbl`, usually dervied from a call to `bbl`
- `showcoeff`: Display first few fit coefficients
- `maxcoeff`: Maximum number of coefficients to display
- `...`: Further arguments passed to or from other methods
print.summary.bbl

Details
Displays the call to bbl, response variable and its levels, predictors and their levels, and the first few fit coefficients.

print.cv.bbl
Display Cross-validation Result

Description
Print cross-validation optimal result and data frame

Usage
## S3 method for class 'cv.bbl'
print(x, ...)

Arguments
x  Object of class cv.bbl
... Other arguments to methods

Details
This method prints crossVal object with the optimal regularization condition and maximum accuracy score on top and the entire score profile as a data frame below.

print.summary.bbl
Print Summary of Boltzmann Bayes Learning

Description
This method prints the summary of bbl object

Usage
## S3 method for class 'summary.bbl'
print(x, ...)

Arguments
x  Object of class summary.bbl
... Other arguments to methods

Details
The naive Bayes summary of summary.bbl object is displayed.
randompar

Generate Random Parameters

Description

Random values of bias and interaction parameters are generated using either uniform or normal distributions.

Usage

randompar(predictors, distr = "unif", h0 = 0, dh = 1, J0 = 0, dJ = 1)

Arguments

predictors List of predictor factor levels. See bbl.
distr c('unif', 'norm') for uniform or normal distributions.
h0 Mean of bias parameters
dh sd of bias if distr = 'unif'. If distr = 'norm', h = [h0 - dh, h0 + dh].
J0 Mean of interaction parameters.
dJ sd of interactions if distr = 'unif'. If distr = 'norm', J = [J0 - dJ, J0 + dJ].

Details

Input argument predictors is used to set up proper list structures of parameters.

Value

List of parameters, h and J.

Examples

```
set.seed(311)
predictors <- list()
for(i in 1:5) predictors[[i]] <- c('a', 'c')
par <- randompar(predictors=predictors)
par
```
Generate Random Boltzmann Bayes Model Data

**Description**

Predictor-response paired data are generated

**Usage**

```r
randomsamp(predictors, response, prob = NULL, par, nsample = 100)
```

**Arguments**

- `predictors`: List of vectors of predictor levels
- `response`: Vector of response variables
- `prob`: Vector of probabilities for sampling each response group
- `par`: List of `bbl` parameters for each response group; e.g., generated from calls to `randompar`.
- `nsample`: Sample size

**Details**

The argument `response` is used to set up all possible levels of response groups and likewise for `predictors`. The parameter argument `par` must have the appropriate structure consistent with `response` and `predictors`. This function is a wrapper calling `sample_xi` multiple times.

**Value**

Data frame of response and predictor variables.

---

**Read FASTA File**

**Description**

Read nucleotide sequence files in FASTA format

**Usage**

```r
readFasta(file, rownames = FALSE)
```

**Arguments**

- `file`: File name of FASTA input.
- `rownames`: Use the sequence annotation line in file (starts with '>' ) as the row names. Will fail if there are duplicate items.
**Details**

Sequence data in FASTA files are converted into data frame suitable as input to `bb1`. If sequence lengths are different, instances longer than those already read will be truncated. Empty sequences are skipped.

**Value**

Data frame of each sequence in rows.

**Examples**

```r
file <- tempfile('data')
write('>seq1', file)
write('atgcc', file, append=TRUE)
write('>seq2', file, append=TRUE)
write('gccaa', file, append=TRUE)
system(paste0('cat ',file))
x <- readFasta(file)
x
```

---

### removeConst

**Remove Non-varying Predictors**

**Description**

Constant predictor is identified and removed

**Usage**

```r
removeConst(x)
```

**Arguments**

- `x` Data frame containing discrete factor variables in each column

**Details**

Variables with only one factor level is removed from data. Intended for use before calling `bb1`.

**Value**

Data frame omitting non-varying variables from `x`. 
Examples

```r
set.seed(351)
nt <- c('a','c','g','t')
x <- data.frame(v1=sample(nt,size=50,replace=TRUE),
               v2=rep('a',50),v3=sample(nt,size=50,replace=TRUE))
y <- sample(c('case','ctrl'),size=50,replace=TRUE)
dat <- cbind(data.frame(y=y), x)
summary(dat)
dat <- removeConst(dat)
summary(dat)
```

---

**residuals.bbl**

*Residuals of BBL fit*

**Description**

Binary-valued vector of fitted vs. true response group

**Usage**

```r
## S3 method for class 'bbl'
residuals(object, ...)
```

**Arguments**

- `object` Object of class `bbl`
- `...` Other arguments

**Details**

Discrete response group identity for each data point is compared with the fitted group and 0 (discordant) or 1 (concordant) is returned

**Value**

Vector binary values for each data point

**Examples**

```r
titanic <- as.data.frame(Titanic)
dat <- freq2raw(titanic[,1:4], freq=titanic$Freq)
fit <- bbl(Survived ~ .^2, data=dat)
x <- residuals(fit)
table(x)
```
sample_xi  

Generate Random Samples from Boltzmann Distribution

Description

Random samples are drawn from Boltzmann distribution

Usage

sample_xi(nsample = 1, predictors = NULL, h, J, code_out = FALSE)

Arguments

nsample  Sample size
predictors  List of predictor factor levels.
h  Bias parameter; see bbl.
J  Interaction parameters; see bbl.
code_out  Output in integer codes; \( a_i = 0, \cdots, L_i - 1 \). If FALSE, output in factors in predictors.

Details

All possible factor states are enumerated exhaustively using input argument predictors. If the number of predictors \( m \) or the number of factor levels \( L_i \) for each predictor \( i \) are even moderately large (\( m \geq 10 \) or \( L_i \geq 5 \)), this function will likely hang because the number of all possible states grows exponentially.

Value

Data frame of samples in rows and predictors in columns.

Examples

```r
set.seed(512)
m <- 5
n <- 1000
predictors <- list()
for(i in 1:m) predictors[[i]] <- c('a','c','g','t')
par <- randompar(predictors)
xi <- sample_xi(nsample=n, predictors=predictors, h=par$h, J=par$J)
head(xi)
```
Naive Bayes Summary

Description

Estimate significant of predictor-group association using naive Bayes model

Usage

```r
## S3 method for class 'bbl'
summary(object, prior.count = 0, ...)
```

Arguments

- `object`: Object of class `bbl`
- `prior.count`: Prior count to be used for computing naive Bayes coefficients and test results. If 0, will produce NAs for factor levels without data points.
- `...`: Other arguments to methods.

Details

This `summary.bbl` method gives a rough overview of associations within a `bbl` fit object via naive Bayes coefficients and test p-values. Note that naive Bayes results displayed ignore interactions even when interactions are present in the model being displayed. This feature is because simple analytic results exist for naive Bayes coefficients and test p-values. The likelihood ratio test is with respect to the null hypothesis that coefficients are identical for all response groups.

Value

Object of class `summary.bbl` extending `bbl` class; a list with extra components

- `h`: List of bias coefficients of response groups under naive Bayes approximation
- `h0`: Bias coefficients of pooled group under naive Bayes
- `chisqNaive`: Vector of chi-square statistics for likelihood ratio test for each predictor
- `dfNaive`: Vector of degrees of freedom for likelihood ratio test for each predictor
- `pvNaive`: Vector p-values for each predictor
weights.bbl

Weights in BBL Fit

Description

This method returns weights used in BBL fit.

Usage

```r
## S3 method for class 'bbl'
weights(object, ...)
```

Arguments

- `object` Object of class `bbl`.
- `...` Other arguments

Details

Note that weights are integral frequency values specifying repeat number of each instance in `bbl`. If no weights were used (default of 1s), `NULL` is returned.

Value

Vector of weights for each instance
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