Package ‘pvLRT’

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pvLRT-package

pvLRT: An R package implementing various Likelihood Ratio Test-based approaches to pharmacovigilance

Description

pvLRT is an R package that implements a suite of likelihood ratio test based methods to use in pharmacovigilance. It can handle adverse effects data on several simultaneous drugs, with possibly zero inflated report counts. Several testing and post-processing functions are implemented.

as.matrix.pvlrt

Casting a pvlrt object as a matrix of log LR statistics

Description

Casting a pvlrt object as a matrix of log LR statistics

Usage

```r
## S3 method for class 'pvlrt'
as.matrix(x, ...)
```

Arguments

- `x`: a pvlrt object; an output of function pvlrt().
- `...`: other input parameters. Currently unused.
### extract_AE_names

**Value**

Returns a matrix with the same dimensions as the input contingency table in the original `pvlrt` call, with each cell providing the corresponding value of the observed log-likelihood ratio test statistic.

**See Also**

`pvlrt`

**Examples**

```r
# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- pvlrt(statin46, nsim = 500)
as.matrix(test1)
```

---

### extract_AE_names

*Extracting and setting AE and Drug names from a pvlrt object*

**Description**

Extracting and setting AE and Drug names from a `pvlrt` object

**Usage**

```r
extract_AE_names(object)
extract_Drug_names(object)
set_AE_names(object, old, new)
set_Drug_names(object, old, new)
```

**Arguments**

- **object**: a `pvlrt` object, which is the output of the function `pvlrt` or one of its wrappers such as `lrt_zi_poisson`, `lrt_poisson` and `lrt_vanilla_poisson`.
- **old**: character vector containing the old names
- **new**: character vector containing the new names
Value

- extract_AE_names returns a character vector of the names of the AEs in the input pvlrt object
- extract_Drug_names returns a character vector of the names of the Drugs in the input pvlrt object
- set_AE_names returns a new pvlrt object with updated AE names as specified through the arguments old and new.
- set_Drug_names returns a new pvlrt object with updated Drug names as specified through the arguments old and new.

Note

Because a pvlrt object is simply a reclassified matrix, the AE (rows) and Drug (columns) names can also be extracted/modified through rownames and colnames respectively.

See Also

pvlrt

Examples

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- pvlrt(statin46, test_zi = TRUE, nsim = 500)
exttract_AE_names(test1)
extrack_Drug_names(test1)
set_AE_names(test1, old = "Rhabdomyolysis", new = "Rhabdo")
set_Drug_names(test1, old = "Other", new = "Other-Drugs")

---

extract_lrstat_matrix  Extract various summary measures from a pvlrt object

Description

Extract various summary measures from a pvlrt object

Usage

eextract_lrstat_matrix(object, ...)

extract_p_value_matrix(object, ...)
extract_lrstat_matrix

extract_zi_probability(object, ...)

extract_n_matrix(object, ...)

extract_significant_pairs(object, significance_level = 0.05, ...)

extract_run_time(object, ...)

Arguments

object: a pvlrt object, which is the output of the function pvlrt or one of its wrappers such as lrt_zi_poisson, lrt_poisson and lrt_vanilla_poisson.

other input parameters. Currently unused.

significance_level: numeric. Level of significance.

Value

- extract_lrstat_matrix returns the matrix of the computed log-likelihood ratio test statistics for signals. This produces a result identical to applying as.matrix.
- extract_p_value_matrix returns the matrix of computed p-values associated with the likelihood ratio tests.
- extract_zi_probability returns a vector of (estimated) zero-inflation probabilities.
- extract_n_matrix returns the original contingency table (matrix) used.
- extract_significant_pairs returns a data.table listing the AE/drug pairs determined to be significant at the provided significance level. This is essentially a subset of the data.table obtained through summary.pvlrt() that satisfies the provided significance threshold.
- extract_run_time returns a difftime object measuring the total CPU time needed to run the original pvlrt call.

See Also

pvlrt

Examples

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- pvlrt(statin46, test_zi = TRUE, nsim = 500)
exttract_lrstat_matrix(test1)
extrack_p_value_matrix(test1)
extrack_zi_probability(test1)
extrack_n_matrix(test1)
extrack_significant_pairs(test1)
**gbca**

*FDA GBCA dataset with all observed 1707 adverse events*

**Description**

A Drug-Adverse event (AE) count dataset (contingency table) obtained from the FDA FAERS database for the quarters 2014Q3 - 2020Q4

**Usage**

`gbca`

**Format**

An object of class `matrix` (inherits from `array`) with 1707 rows and 10 columns.

**Details**

Data are stored in the form of a contingency table, with drugs listed across the columns and the 1707 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that drug/AE pair and detected in the FDA FAERS database during 2014Q3 - 2020Q4.

The dataset contains 6 Gadolinium-Based Contrast Agents (GBCAs) as drugs:

gadobenate, gadobutrol, gadodiamide, gadofosveset, gadopentetate, gadoterate, gadoteridol, gadoversetamide, gadoxetate

Corresponding to all 1707 observed adverse events (AEs) as curated in FAERS database.

**Source**


---

**heatmap_pvlrt**

*Heatmap, barplot and bubbleplot displaying likelihood ratio test results in a pvlrt object*

**Description**

Heatmap, barplot and bubbleplot displaying likelihood ratio test results in a pvlrt object
Usage

heatmap_pvlrt(
  object,
  AE = NULL,
  Drug = NULL,
  grep = FALSE,
  fill_measure = "p_value",
  show_n = FALSE,
  show_lrstat = FALSE,
  show_p_value = FALSE,
  p_value_lower = 0,
  p_value_upper = 1,
  lrstat_lower = 0,
  lrstat_upper = Inf,
  n_lower = 0,
  n_upper = Inf,
  arrange_alphabetical = FALSE,
  remove_outside = FALSE,
  digits = 2,
  border_color = "black",
  fill_gradient_range = c("darkred", "white"),
  ...
)

## S3 method for class 'pvlrt'
barplot(
  height,
  AE = NULL,
  Drug = NULL,
  grep = FALSE,
  x_axis_measure = "lrstat",
  fill_measure = "p_value",
  show_n = FALSE,
  arrange_alphabetical = FALSE,
  show_p_value = FALSE,
  show_lrstat = FALSE,
  p_value_lower = 0,
  p_value_upper = 1,
  lrstat_lower = 0,
  lrstat_upper = Inf,
  n_lower = 0,
  n_upper = Inf,
  remove_outside = FALSE,
  digits = 2,
  Drug_nrow = 1,
  border_color = "black",
  x_axis_logscale = TRUE,
  fill_gradient_range = c("darkred", "white"),
  ...)
bubbleplot_pvlrt(
  object,
  AE = NULL,
  Drug = NULL,
  grep = FALSE,
  x_axis_measure = "lrstat",
  fill_measure = "p_value",
  size_measure = "n",
  show_n = FALSE,
  arrange_alphabetical = FALSE,
  show_p_value = FALSE,
  show_lrstat = FALSE,
  p_value_lower = 0,
  p_value_upper = 1,
  lrstat_lower = 0,
  lrstat_upper = Inf,
  n_lower = 0,
  n_upper = Inf,
  remove_outside = FALSE,
  digits = 2,
  Drug_nrow = 1,
  border_color = "black",
  x_axis_logscale = TRUE,
  size_logscale = TRUE,
  fill_gradient_range = c("darkred", "white"),
  ...
)

Arguments

object, height  pvlrt object; output of pvlrt()

AE  input parameter determining which adverse effects to show in the plot. This can be a numeric scalar specifying the number of top (in terms of computed LRT values) adverse effects to show. Alternatively, it can be a character vector, specifying the exact adverse effects to show. It can also be a vector of patterns to match (ignores cases) against the full names of all available adverse effects, provided grep is set to TRUE. Defaults to adverse effects corresponding to the top M pairs where M = max(number of possible pairs, 50). Set AE = Inf to force display of all adverse effects.

Drug  input parameter determining which drugs to show in the plot. This can be a numeric scalar specifying the number of top (in terms of computed LRT values) drugs to show. Alternatively, it can be a character vector, specifying the exact drugs to show. It can also be a vector of patterns to match (ignores cases) against the full names of all available drugs, provided grep is set to TRUE. Defaults to drugs corresponding to the top M pairs where M = max(number of possible
heatmap_pvlrt

pairs, 50). Set Drug = Inf to force display all drugs.

grep logical. Match patterns against the supplied AE or Drug names? Ignores if
neither AE nor Drug is a character vector.

fill_measure Measure to govern the filling color in each cell (in heatmap) or bar (in barplot)
or circle/bubble (in bubbleplot) for each drug/AE combination. Defaults to
"p.value". Available choices are: "p.value", "lrstat", and "n".

show_n logical. Show the sample size as inscribed text on each cell?

show_lrstat logical. Show the computed LRT statistic (on log-scale) inscribed text on each
cell?

show_p_value logical. Show the computed p-value as inscribed text on each cell?

p_value_lower, p_value_upper lower and upper limits on the computed p-values to display on the plot.

lrstat_lower, lrstat_upper lower and upper limits on the computed LRT values to display on the plot.

n_lower, n_upper lower and upper limits on the computed sample sizes to display on the plot.

arrange_alphabetical logical. Should the rows (AEs) and columns (Drugs) be arranged in alphabetical
orders? Defaults to FALSE, in which case the orderings of the computed LRT
values are used.

remove_outside logical. Should the values for pairs with p-value, LRT statistics or sample sizes
falling outside of the provided ranges through p_value_lower, p_value_upper
etc., be replaced with NA? Defaults to FALSE. Setting this to TRUE may help
distinguish drugs or AEs which has some pairs falling within and some pairs
falling outside of the provided ranges better.

digits numeric. Number of decimal places to show on the inscribed texts on the plot.

border_color character string. Specifies the border color of cells/bars.

fill_gradient_range character vector. Specifies the range of gradient colors used for fill_measure.
Passed into the colours argument of scale_fill_gradientn from ggplot2.

... Other arguments. Currently ignored

x_axis_measure measure to show on the x-axis of the (horizontal) bar plots. Defaults to "lrstat"
available choices are "lrstat", "p_value" and "n".

Drug.nrow Number of rows in the panels for Drugs for the barplots.

x_axis_logscale logical. Should the x axis measure in the bar plot or the bubble plot be log
transformed (more precisely, "log1p" transformed with the function f(x) = log(1 + x))? Defaults to TRUE.

size_measure measure to govern sizes of the circles in the bubble plot. Defaults to "n". Available
choices are "lrstat", "p.value" and "n".

size_logscale logical. Should the circle size measure in the bubble plot be log transformed
(more precisely, "log1p" transformed with the function f(x) = log(1 + x))? Defaults to TRUE.
logLik.pvlrt

Overall Log-likelihood for a pvlrt object

Description
Overall Log-likelihood for a pvlrt object

Usage
## S3 method for class 'pvlrt'
logLik(object, type = "full-zip", ...)

Arguments

- **object**: a pvlrt object, which is the output of the function pvlrt or one of its wrappers such as lrt_zi_poisson, lrt_poisson and lrt_vanilla_poisson.
- **type**: Type of model and hypothesis combination. Available choices are "full-poisson", "null-poisson", "full-zip" (default), and "null-zip". See details.
- **...**: other input parameters. Currently unused.

Details
The function extracts the overall log-likelihood and degrees of freedom (the number of estimated parameters) from a pvlrt object. There are four possible choices of distribution and hypothesis combinations, supplied through the argument type, with the default being type = "full-zip". In a "full" model the signal parameters lambdas are estimated for all cells in the contingency table from the data (subject to the condition lambda >= 1), whereas under a "null" model each lambda is fixed at 1 for each cell. In a "zip" model (type = "full-zip" and type = "null-zip") the log-likelihood

Value
A ggplot object.

See Also
pvlrt

Examples
# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger
test1 <- pvlrt(statin46, nsim = 500)
bubbleplot_pvlrt(test1)
heatmap_pvlrt(test1)
barplot(test1)
under a zero-inflated Poisson model with estimated or supplied zero inflation parameters (through `zi_prob` in `pvlrt`) is returned. The degrees of freedom reflects whether the zero-inflation parameters are estimated. Note that if an ordinary Poisson LRT is run either by setting `zi_prob = 0` in `pvlrt` or equivalently through `lrt_poisson` then "full-zip" and "null-zip" refer to zero-inflated poisson models with presepecified zero-inflation probabilities equal to 0. Thus, in such cases the results with type = "full-zip" and type = "null-zip" are identical to type = "full-poisson" and type = "null-poisson" respectively. See examples.

**Value**

An object of class `logLik`. See Details.

**Note**

The overall log likelihood must be computed during the original `pvlrt` run. This is ensured by setting `return_overall_loglik = TRUE`, and parametrization = "lambda" (or parametrization = "rrr") while running `pvlrt()`.

**See Also**

`pvlrt`; `AIC`

**Examples**

```r
# 500 bootstrap iterations (nsim) in each example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

set.seed(100)
# estimates zero inflation probabilities
test1 <- pvlrt(statin46, nsim = 500)
logLik(test1)
AIC(test1)
BIC(test1)

# compare with and without zero inflation
BIC(logLik(test1, type = "full-zip"))
BIC(logLik(test1, type = "full-poisson"))

# ordinary poisson model
## equivalent to pvlrt(statin46, zi_prob = 0, nsim = 500)
test2 <- lrt_poisson(statin46, nsim = 500)

all.equal(logLik(test2, "full-zip"), logLik(test2, "full-poisson"))
```
lovastatin  

**Description**

A Drug-Adverse event (AE) count dataset (contingency table) obtained from the FDA FAERS database for the quarters 2014Q3 - 2020Q4

**Usage**

lovastatin

**Format**

An object of class matrix (inherits from array) with 47 rows and 3 columns.

**Details**

Data are stored in the form of a contingency table, with drugs listed across the columns and the 46 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that drug/AE pair and detected in the FDA FAERS database during 2014Q3 - 2020Q4.

The dataset contains 1 column for the lovastatin drug, and one column for all other drugs combined.

**Source**


---

lrt_poisson  

**Description**

Likelihood Ratio Test under the (vanilla, non-zero-inflated) Poisson model

**Usage**

lrt_poisson(contin_table, nsim = 10000, parametrization = "rrr", ...)

lrt_vanilla_poisson(contin_table, nsim = 10000, parametrization = "rrr", ...)
**lrzzi_poisson**

**Arguments**

- `contin_table`: IxJ contingency table showing pairwise counts of adverse effects for I AE and J Drugs
- `nsim`: Number of simulated null contingency table to use for computing the p-value of the test
- `parametrization`: Type of parametrization to use in the LR test. Available choices are "rrr", "lambda", "rr", and "p-q". The relative reporting ratio (default) parametrization of the test is used when `parametrization %in% c("rrr","lambda")`, and the reporting rate parametrization is used otherwise. NOTE: zero inflation can be handled only for the relative reporting ratio parametrization.

... additional arguments. Currently unused.

**Value**

Returns a `pvlrt` object. See `pvlrt` for more details.

**Note**

`lrt_poisson()` and `lrt_vanilla_poisson()` are both wrappers for `pvlrt()` with `omega_vec = rep(0, ncol(contin_table))`

**See Also**

`pvlrt`

**Examples**

```r
data("statin46")

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

# no grouping -- each drug forms its own class
test1 <- lrt_poisson(lovasartatin, nsim = 500)
```

---

**Description**

Pseudo Likelihood Ratio Test under the zero-inflated Poisson model with relative reporting rate parametrization
Usage

```r
lrt_zi_poisson(contin_table, nsim = 10000, ...)
```

Arguments

- `contin_table`: IxJ contingency table showing pairwise counts of adverse effects for I AE and J Drugs
- `nsim`: Number of simulated null contingency table to use for computing the p-value of the test
- `...`: additional arguments passed to `pvlrt`

Value

Returns a `pvlrt` object. See `pvlrt` for more details.

Note

`lrt_zi_poisson()` is a wrapper for `pvlrt()` with parametrization = "rrr".

See Also

- `pvlrt`

Examples

```r
data("statin46")

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- lrt_zi_poisson(statin46, nsim = 500)

test1
```

---

**plot.pvlrt**

*Plotting method for a pvlrt object*

Description

Plotting method for a pvlrt object

Usage

```r
## S3 method for class 'pvlrt'
plot(x, type = "bubbleplot", ...)```

---
Arguments

x  a pvlrt object; an output of function pvlrt().

type  character string determining the type of plot to show. Available choices are
"bubbleplot" which calls bubbleplot_pvlrt, "heatmap" which calls heatmap_pvlrt,
and "barplot" which calls barplot.pvlrt, with the additional arguments supplied in ...

...  additional arguments passed to heatmap_pvlrt or barplot.pvlrt depending on type.

Value

A ggplot object.

See Also

pvlrt; bubbleplot_pvlrt; heatmap_pvlrt; barplot.pvlrt

Examples

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- pvlrt(statin46, nsim = 500)
plot(test1, type = "bubbleplot")
plot(test1, type = "barplot")
plot(test1, type = "heatmap")
Arguments

- **x**: a `pvlrt` object; an output of function `pvlrt()`.
- **significance_level**: numeric. Level of significance.
- **topn**: number of top (with respect to likelihood ratio statistic value) pairs to show at the given significance level.
- **digits**: number of digits to show after the decimal place.
- **show_test_summary**: logical. Should a brief summary showing the top few test results be displayed? defaults to FALSE.
- ... other input parameters. Currently unused.

Value

Invisibly returns the input `pvlrt` object.

See Also

`pvlrt`

Examples

```r
# 500 bootstrap iterations (nsim) in the example below # are for quick demonstration only -- # we recommended setting nsim to 10000 (default) or bigger
test1 <- pvlrt(statin46, nsim = 500)
print(test1)
```

---

**pvlrt**

*Pseudo Likelihood Ratio Test for determining significant AE-Drug pairs under Poisson and zero-inflated Poisson models for pharmacovigilance*

Description

Pseudo Likelihood Ratio Test for determining significant AE-Drug pairs under Poisson and zero-inflated Poisson models for pharmacovigilance
Usage

\[ \text{pvlrt}( \text{contin_table}, \]  
\[ \text{nsim} = 10000, \]  
\[ \text{omega_vec} = \text{NULL}, \]  
\[ \text{zi_prob} = \text{NULL}, \]  
\[ \text{no_zi_idx} = \text{NULL}, \]  
\[ \text{drug_class_idx} = \text{as.list}(1: \text{ncol(contin_table)}), \]  
\[ \text{test_drug_idx} = 1: \text{ncol(contin_table)}, \]  
\[ \text{grouped_omega_est} = \text{FALSE}, \]  
\[ \text{test_zi} = \text{NULL}, \]  
\[ \text{test_omega} = \text{NULL}, \]  
\[ \text{pval_ineq_strict} = \text{FALSE}, \]  
\[ \text{parametrization} = \text{"rrr"}, \]  
\[ \text{null_boot_type} = \text{"parametric"}, \]  
\[ \text{is_zi_structural} = \text{TRUE}, \]  
\[ \text{return_overall_loglik} = \text{TRUE}, \]  
\[ ... \]  
\)

Arguments

- **contin_table**: IxJ contingency table showing pairwise counts of adverse effects for I AE and J Drugs
- **nsim**: Number of simulated null contingency table to use for computing the p-value of the test
- **zi_prob**, **omega_vec**: Alias, determining zero inflation probabilities in the model. Can be a vector, providing different zero inflation probabilities for different drugs, or a scalar, providing the common zero inflation probability for all drugs. If NULL (default), then is estimated from the data. See also the description of the argument grouped_omega_est. If omega_vec = \text{rep}(0, \text{ncol(contin_table)}), then test reduces to an ordinary (non-zero inflated) Poisson test. NOTE: zi_prob and omega_vec are alias.
- **no_zi_idx**: List of pairs (i, j) where zero inflation is not allowed. To specify the entirety i-th row (or j-th column) use c(i, 0) (or c(0, j)). See examples.
- **drug_class_idx**: a list, with the h-th entry providing the h-th group/class of drugs. By default, each drug forms its own class. If more than one drug is present in a class, an extended LRT is performed. See examples.
- **test_drug_idx**: integer vector representing the columns (drug indices) of contin_table to be tested for signal. Defaults to all columns.
- **grouped_omega_est**: Logical. When performing a test with grouped drug classes (extended LRT), should the estimated zero-inflation parameter "omega" reflect the corresponding grouping? If TRUE, then the estimated omegas are obtained by combining columns from the same group, and if FALSE (default), then omegas are estimated separately for each drug (column) irrespective of the groups specified.
through drug_class_idx. Ignored if omega_vec is supplied/non-NULL (i.e., not estimated).

\textbf{test\_zi, test\_omega}

logical indicators specifying whether to perform a pseudo likelihood ratio test for zero inflation. Defaults to FALSE. Ignored if omega_vec is supplied (is non-NULL). Defaults to FALSE. NOTE: test\_omega and test\_zi are aliases.

\textbf{pval\_ineq\_strict}

logical. Use a strict inequality in the definition of the p-values? Defaults to FALSE.

\textbf{parametrization}

Type of parametrization to use in the LR test. Available choices are "rrr", "lambda", "rr", and "p-q". The relative reporting ratio (default) parametrization of the test is used when when parametrization \%in\% c("rrr","lambda"), and the reporting rate parametrization is used otherwise. NOTE: zero inflation can be handled only for the relative reporting ratio parametrization.

\textbf{null\_boot\_type}

Type of bootstrap sampling to perform for generating null resamples. Available choices are "parametric" (default) and "non-parametric". NOTE: zero inflation is not handled properly in a non-parametric bootstrap resampling.

\textbf{is\_zi\_structural}

logical. Do the inflated zeros correspond to structural zeros (indicating impossible AE-Drug combination)? This determines how the bootstrap null zero-inflation indicators are generated. If TRUE (default), then then the null zero-inflation random indicators are randomly generated using the (estimated) conditional probabilities of zero inflation given observed counts. If FALSE, then they are generated using the marginal (drug-specific) estimated probabilities of zero-inflation.

\textbf{return\_overall\_loglik}

logical. Return overall log-likelihood for the table? This is needed if logLik method is to be used.

\ldots

additional arguments. Currently unused.

\textbf{Value}

The function returns an S3 object of class \texttt{pvlrt} containing test results. A \texttt{pvlrt} object is simply a re-classified matrix containing log likelihood ratio test statistics for cells in the input contingency table, with various other test and input data information (including p-values, estimated zero inflation parameters, overall log-likelihood etc.) embedded as attributes. The following S3 methods and functions are available for an \texttt{pvlrt} object:

Various postprocessing methods for \texttt{pvlrt} objects are available. This includes:

- \texttt{bubbleplot\_pvlrt}
- \texttt{extract\_AE\_names}
- \texttt{extract\_Drug\_names}
- \texttt{extract\_lrstat\_matrix}
- \texttt{extract\_n\_matrix}
- \texttt{extract\_p\_value\_matrix}
Examples

data("statin46")

# 500 bootstrap iterations (nsim) in each example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

# no grouping -- each drug forms its own class,
# default "rrr" (lambda) parametrization, possible zi,
# null distribution evaluated using parametric bootstrap

test1 <- pvlrt(statin46, nsim = 500)
test1
## extract the observed LRT statistic
extract_lrstat_matrix(test1)
## extract the estimated omegas
extract_zi_probability(test1)

# grouped drugs --
# group 1: drug 1, drug 2
# group 2: drug 3, drug 4
# drug 5, 6, 7 in their own groups
drug_groups <- list(c(1, 2), c(3, 4), 5, 6, 7)
test2 <- pvlrt(statin46, drug_class_idx = drug_groups, nsim = 500)
test2

# specify no zero inflation at the entirety of the last row and the last column
no_zi_idx <- list(c(nrow(statin46), 0), c(0, ncol(statin46)))
test3 <- pvlrt(statin46, no_zi_idx = no_zi_idx, nsim = 500)
test3
# use non-parametric bootstrap to evaluate the null distribution
# takes longer, due to computational costs with non-parametric
# contingency table generation

test4 <- pvlrt(statin46, null_boot_type = "non-parametric", nsim = 500)
test4

# test zi probabilities (omegas)
test5 <- pvlrt(statin46, test_omega = TRUE, nsim = 500)
test5

---

**rv**

*FDA rotavirus vaccine dataset with 794 adverse events observed among combined old (age >= 1 year) and young (age < 1 year) individuals*

---

**Description**

A vaccine-Adverse event (AE) count dataset (contingency table) obtained from the FDA VAERS database for the year 1999

**Usage**

`rv`

**Format**

An object of class `matrix` (inherits from `array`) with 794 rows and 2 columns.

**Details**

Data are stored in the forms of contingency table, with the vaccines listed across the columns and the 794 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that vaccine/AE pair and detected in the FDA VAERS database for the year 1999.

The dataset contains two columns – one for the rotavirus vaccine, and another for other (37 vaccines combined).

**Source**

[https://vaers.hhs.gov/data/datasets.html](https://vaers.hhs.gov/data/datasets.html)
rvold

*FDA rotavirus vaccine dataset with 727 adverse events observed among "old" (non-infant; age >= 1 year) individuals*

**Description**

A vaccine-Adverse event (AE) count dataset (contingency table) obtained from the FDA VAERS database for the year 1999

**Usage**

rvold

**Format**

An object of class `matrix` (inherits from `array`) with 727 rows and 2 columns.

**Details**

Data are stored in the forms of contingency table, with the vaccines listed across the columns and the 727 AEs presented across the rows. Each cell in the contingency table represents the total report counts (from "old" individuals with age >= 1 year) associated with that vaccine/AE pair and detected in the FDA VAERS database for the year 1999.

The dataset contains two columns – one for the rotavirus vaccine, and another for other (37 vaccines combined).

**Source**

https://vaers.hhs.gov/data/datasets.html

rvyoung

*FDA rotavirus vaccine dataset with 346 adverse events observed among young (infant – 1 year) individuals*

**Description**

A vaccine-Adverse event (AE) count dataset (contingency table) obtained from the FDA VAERS database for the year 1999

**Usage**

rvyoung

**Format**

An object of class `matrix` (inherits from `array`) with 346 rows and 2 columns.
Details

Data are stored in the forms of contingency table, with the vaccines listed across the columns and the 346 AEs presented across the rows. Each cell in the contingency table represents the total report counts from young individuals with age < 1 year associated with that vaccine/AE pair and detected in the FDA VAERS database for the year 1999.

The dataset contains two columns – one for the rotavirus vaccine, and another for other (37 vaccines combined).

Source

https://vaers.hhs.gov/data/datasets.html

r_contin_table_zip

Generate random contingency tables for adverse effect (across rows) and drug (across columns) combinations given row and column marginal totals, embedded signals, and possibly with zero inflation

Description

Generate random contingency tables for adverse effect (across rows) and drug (across columns) combinations given row and column marginal totals, embedded signals, and possibly with zero inflation

Usage

r_contin_table_zip(
  n = 1,
  row_marginals,
  col_marginals,
  signal_mat = matrix(1, length(row_marginals), length(col_marginals)),
  omega_vec = rep(0, length(col_marginals)),
  no_zi_idx = NULL,
  ...
)

Arguments

n number of random matrices to generate.
row_marginals, col_marginals (possibly named) vector of row and column marginal totals. Must add up to the same total. If named, the names are passed on to the randomly generated matrix/matrices.
signal_mat numeric matrix of dimension length(row_marginals) x length(col_marginals). The (i,j)-th entry of signal_mat determines the signal strength of the i-th adverse effect and j-th drug pair. The default is 1 for each pair, which means no signal for the pair.
omega_vec vector of zero inflation probabilities. Must be of the same length as col_marginals.

no_zi_idx List of pairs (i, j) where zero inflation is not allowed. To specify the entirety i-th row (or j-th column) use c(i, 0) (or c(0, j)). See examples.

... Additional arguments. Currently unused.

Value

A list of length n, with each entry being a length(row_marginal) by length(col_marginal) matrix.

Examples

```r
set.seed(42)

# first load the 46 statin data
data(statin46)
# zero inflation probabilities
omega_tru <- c(rep(0.15, ncol(statin46) - 1), 0)

# signal matrix
signal_mat <- matrix(1, nrow(statin46), ncol(statin46))

# "positive" signal at the (1, 1) entry
# the first column
signal_mat[1, 1] <- 10

# Now simulate data with the same row/column marginals
# as in statin46, with embedded signals as specified in
# the above signal_mat

# no zero inflation at (1, 1)
# (where signal is elicited) and the last row ("Other PT")
# and at the last column ("Other drugs") of the simulated matrix

sim_statin <- r_contin_table_zip(
  n = 1,
  row_marginals = rowSums(statin46),
  col_marginals = colSums(statin46),
  signal_mat = signal_mat,
  omega_vec = omega_tru,
  no_zi_idx = list(
    c(1, 1),
    c(nrow(statin46), 0), # the entire last row
    c(0, ncol(statin46)) # the entire last column
  )
)[[1]]

# now analyze the above simulated data

# using a pseudo LRT with a ZIP model
test1 <- pvlrt(
  sim_statin, FALSE, TRUE, TRUE, TRUE
)
```
contin_table = sim_statin,
nsim = 500
# set to 500 for demonstration purposes only,
# we recommend the default 10000 or bigger)
extract_zi_probability(test1)
extract_significant_pairs(test1)

# LRT with a poisson model
test2 <- lrt_poisson(
    contin_table = sim_statin,
    nsim = 500
    # set to 500 for demonstration purposes only,
    # we recommend the default 10000 or bigger)
)
extract_zi_probability(test2)
extract_significant_pairs(test2)

# LRT with true omega supplied
test3 <- pvlrt(
    contin_table = sim_statin,
    zi_prob = omega_tru,
    nsim = 500
    # set to 500 for demonstration purposes only,
    # we recommend the default 10000 or bigger)
)
extract_zi_probability(test3)
extract_significant_pairs(test3)

#### statin

**FDA Statin dataset with 6039 adverse events**

**Description**

A Drug-Adverse event (AE) count dataset (contingency table) obtained from the FDA FAERS database for the quarters 2014Q3 - 2020Q4

**Usage**

statin

**Format**

An object of class matrix (inherits from array) with 6039 rows and 7 columns.
Details
Data are stored in the form of a contingency table, with drugs listed across the columns and the 6039 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that drug/AE pair and detected in the FDA FAERS database during 2014Q3 - 2020Q4.
The dataset catalogs 6 statin drugs (across columns):
Atorvastatin, Fluvastatin, Lovastatin, Pravastatin, Rosuvastatin, Simvastatin
Corresponding to all 6039 observed adverse events (AEs) observed in statins

Source

Description
A Drug-Adverse event (AE) count dataset (contingency table) obtained from the FDA FAERS database for the quarters 2014Q3 - 2020Q4

Usage
statin1491

Format
An object of class matrix (inherits from array) with 1491 rows and 7 columns.

Details
Data are stored in the form of a contingency table, with drugs listed across the columns and the 1491 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that drug/AE pair and detected in the FDA FAERS database during 2014Q3 - 2020Q4.
The dataset catalogs 6 statin drugs (across columns):
Atorvastatin, Fluvastatin, Lovastatin, Pravastatin, Rosuvastatin, Simvastatin
The 1491 AEs stored in the dataset represent the intersection of adverse events of the statin class of drugs and the GBCA drugs

Source

See Also
statin46, statin, gbca
**statin46**  
*FDA Statin dataset with 46 adverse events*

**Description**
A Drug-Adverse event (AE) count dataset (contingency table) obtained from the FDA FAERS database for the quarters 2014Q3 - 2020Q4

**Usage**
statin46

**Format**
An object of class `matrix` (inherits from `array`) with 47 rows and 7 columns.

**Details**
Data are stored in the form of a contingency table, with drugs listed across the columns and the 46 AEs presented across the rows. Each cell in the contingency table represents the total report counts associated with that drug/AE pair and detected in the FDA FAERS database during 2014Q3 - 2020Q4.

The dataset catalogs 6 statin drugs (across columns):
Atorvastatin, Fluvastatin, Lovastatin, Pravastatin, Rosuvastatin, Simvastatin

The 46 adverse events presented across the rows are considered significant by FDA.

**Source**

**See Also**
statin, statin1491, gbca

---

**summary.pvlrt**  
*Summary method for a pvlrt object*

**Description**
Summary method for a pvlrt object

**Usage**
```r
## S3 method for class 'pvlrt'
summary(object, show_zi = FALSE, ...)
```
Arguments

object  a pvlrt object, which is the output of the function pvlrt or one of its wrappers such as lrt_zi_poisson, lrt_poisson and lrt_vanilla_poisson.

show_zi  logical. Should summary of the estimates and tests (if performed) of the zero inflation parameters be returned? Defaults to FALSE. If TRUE, then the zero inflation summary is included as an attribute with name “zi”. See examples.

Value

Returns a data.table with rows corresponding to all possible AE/Drug pairs as obtained from the input contingency table, and columns titled "AE", "Drug", "n", "lrstat" (log-likelihood ratio test statistic) and "p_value". Additionally, if show_zi is set to TRUE, then as an attribute named "zi" a data.table with rows corresponding to Drugs (columns in the input contingency table), and columns titled "AE", "zi", "lrstat" (log-likelihood ratio test statistic for zero-inflation), "p_value" and "q_value" (Benjamini-Hochberg adjusted p-values, as obtained through p.adjust) is returned.

See Also

pvlrt

Examples

# 500 bootstrap iterations (nsim) in the example below
# are for quick demonstration only --
# we recommended setting nsim to 10000 (default) or bigger

test1 <- pvlrt(statin46, test_zi = TRUE, nsim = 500)
summary(test1)
tmp <- summary(test1, show_zi = TRUE)
print(tmp)
tmp_zi <- attr(tmp, "zi")
print(tmp_zi)
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