Package 'ZIprop'

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Type Package

Title Permutations Tests and Performance Indicator for Zero-Inflated Proportions Response

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Description Permutations tests to identify factor correlated to zero-inflated proportions response. Provide a performance indicator based on Spearman correlation to quantify the part of correlation explained by the selected set of factors. See details for the method at the following preprint e.g.: <https://hal.archives-ouvertes.fr/hal-02936779v3>.

URL https://gitlab.paca.inrae.fr/meribaud/ziprop

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Encoding UTF-8

LazyData true

Depends R (>= 3.5.0), rgenoud, purrr, data.table, parallel

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`delta` *The scalar delta*

Description

Calculate the scalar delta. This parameter comes from the optimal Spearman's correlation when the rank of two vectors `X` and `proba` are equal except on a given set of indices. In our context, this set correspond to the zero-values of the vector `proba`.

Usage

```
delta(X, proba)
```

Arguments

`X`	a vector.
`proba`	a zero-inflated proportions response.

Value

`Delta` the scalar Delta calculated for the vector `x` and the vector `proba`.

Examples

```
X = rnorm(100)
proba = runif(100)
proba[sample(1:100,80)]=0
Delta = delta(X,proba)
print(Delta)
```

`diffFactors` *diffFactors*

Description

Data for the comparison of COVID-19 mortality in European and North American geographic entities

Usage

```
data(diffFactors)
```

Format

A data frame with 483 rows and 32 variables

Details

- geographic_entity_receptor are the entity receptor
- geographic_entity_source are the entity source
- proba is the probability that the receptor follows the mortality dynamics of the source
- other columns are the difference between factors

Author(s)

Melina Ribaud, Davide Martinetti and Samuel Soubeyrand

References

doi: 10.5281/zenodo.4769671

`equineDiffFactors` *equineDiffFactors*

Description

Equine Influenza dataset

Usage

```
data(equineDiffFactors)
```

Format

A data frame with 2256 rows and 8 variables

Details

- ID.source are the ID of source hosts
- ID.recep are the ID of receiver hosts
- y are the vector of transmission probabilities source -> receiver
- other columns are the factors

Author(s)

Melina Ribaud and Joseph Hughes

References

doi: /10.5281/zenodo.4837560

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example_data *Zero-inflated proportions dataset*

Description

A dataset example to test the package functions. The factor X1 to X5 and F1 to F5 are correlated to the responses y.

Usage

```
data(example_data)
```

Format

A data frame with 440 rows and 23 variables

Details

- ID.source are the ID of source hosts
- ID.recep are the ID of receiver hosts
- y are the vector of transmission probabilities source -> receiver
- X1 to X10 are continuous factor
- F1 to F10 are discrete factor

fact2mat *Turn factor into multiple column*

Description

Turns a factor with several levels into a matrix with several columns composed of zeros and ones.

Usage

```
fact2mat(x)
```

Arguments

x	a vector.

Value

Columns with zeros and ones.

Examples

```
x = sample(1:3,100,replace = TRUE)
fact2mat(x)
```

indicator *The performance indicator*

Description

Calculate the indicator for a vector X and a zero-inflated proportions response proba.

Usage

```
indicator(X, proba)
```

Arguments

X	a vector.
proba	a zero-inflated proportions response.

Value

a scalar represents the performance indicator and the vector proba.

Examples

```
X = rnorm(100)
proba = runif(100)
proba[sample(1:100,80)]=0
print(indicator(X,proba))
```

`indicator_max` *The max performance indicator*

Description

Search for the set of parameters that maximize the indicator (equivalent to Spearman correlation). For a given set of factors scaled between 0 and 1 and a zero-inflated proportions response.

Usage

```
indicator_max(
  DT,
  ColNameFactor,
  ColNameWeight = "weight",
  bounds = c(-10, 10),
  max_generations = 200,
  hard_limit = TRUE,
  wait_generations = 50,
  other_class = NULL
)
```

Arguments

`DT`	a data table contains the factors and the response.
`ColNameFactor`	a char vector with the name of the selected factor.
`ColNameWeight`	a char with the name of the ZI response.
`bounds`	default is $[-10;10]$. Upper and Lower bounds.
`max_generations`	default is 200 see genoud for more information.
`hard_limit`	default is TRUE see genoud for more information.
`wait_generations`	default is 50 see genoud for more information.
`other_class`	a char vector with the name of other classes than numeric (factor or char).

Value

Return a list of two elements with the value of the indicator and the associate set of parameters (beta).

Examples

```
library(data.table)
data(example_data)
# For real cases increase max_generations and wait_generations
I_max = indicator_max(example_data,
names(example_data)[c(4:8, 14:18)],
ColNameWeight = "proba",
max_generations = 20,
wait_generations = 5)
print(I_max)
```

`model_matrix` *Construct Design Matrix*

Description

Creates a design matrix by expanding factors to a set of dummy variables.

Usage

```
model_matrix(DT, ColNameFactor, other_class)
```

Arguments

`DT`	a data table contains the factors and the response.
`ColNameFactor`	a char vector with the name of the selected factor.
`other_class`	a char vector with the name of other classes than numeric (factor or char).

Value

return the value.

Examples

```
library(data.table)
data(example_data)
m = model_matrix (example_data,
colnames(example_data)[-c(1:3)],
other_class = colnames(example_data)[14:23])
print(m)
```

permDT *Permutations tests*

Description

Permutations tests to identify factor correlated to a zero-inflated proportions response. The statistic are the Spearman's correlation for numeric factor and mean by level for other factor.

Usage

```
permDT(
  DT,
  ColNameFactor,
  B = 1000,
  nclust = 1,
  ColNameWeight = "weight",
  ColNameRecep = "ID.recep",
  ColNameSource = "ID.source",
  seed = NULL,
  no_const = FALSE,
  num_class = ColNameFactor,
  other_class = NULL,
  multiple_test = FALSE,
  adjust_method = "none",
  alpha = 0.05
)
```

Arguments

DT	a data table contains the factors and the response.
ColNameFactor	a char vector with the name of the selected factor.
B	number of permutations (use at least B=1000 permutations to get a correct accuracy of the p-value.)
nclust	number of proc for parallel computation.
ColNameWeight	a char with the name of the ZI response.
ColNameRecep	colname of the column with the target names
ColNameSource	colname of the column with the contributor names
seed	vector with the seed for the permutations: size(seed)=B
no_const	FALSE for receiver block constraint for permutations: TRUE no constraint.
num_class	a char vector with the name of numeric factor.
other_class	a char vector with the name of other classes than numeric (factor or char).
multiple_test	useful option only for discrete factors: Set TRUE to calculate multiple tests.
adjust_method	p-values adjusted methods (default "none"). c("holm", "hochberg", "hommel", "bonferroni", "BH", "BY","fdr", "none").
alpha	significant level (default 0.05).

Value

A data frame with two columns. One for the statistics and the other one for the p-value.

Examples

```
library(data.table)
data(example_data)
res = permDT (example_data,
colnames(example_data)[c(4,10,14,20)],
B = 10,
nclust = 1,
ColNameWeight = "y",
ColNameRecep = "ID.recep",
ColNameSource = "ID.source",
seed = NULL,
num_class = colnames(example_data)[c(4,10)],
other_class = colnames(example_data)[c(14,20)])
print(res)
```

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scale_01 *Scale vector*

Description

Scale a vector between 0 and 1.

Usage

```
scale_01(x)
```

Arguments

x	a vector.

Value

the scaled vector of x.

Examples

```
x = runif(100,-10,10)
x_scale = scale_01(x)
range(x_scale)
```

T_stat_discr *Statistic for non-numeric factor tests*

Description

Statistic for non-numeric factor tests (same statistic as H-test).

Usage

```
T_stat_discr(permu, al)
```

Arguments

permu	the response vector.
al	the factor.

Value

the statistic.

Examples

```
permu = runif(100,-10,10)
al = as.factor(sample(1:3,100,replace=TRUE))
T_stat_discr(permu, al)
```

T_stat_multi *Statistic for non-numeric factor multiple tests*

Description

Statistic for non-numeric factor multiple tests (difference in mean ranks).

Usage

```
T_stat_multi(permu, al)
```

Arguments

permu	the response vector.
al	the factor.

Value

the means difference of two levels for a discrete factor.

Examples

```
permu = runif(100,-10,10)
al = as.factor(sample(1:3,100,replace=TRUE))
T_stat_multi(permu, al)
```

ZIprop *ZIprop: A package for Zero-Inflated Proportions data (ZIprop)*

Description

We propose a by block-permutation-based methodology (i) to identify factors (discrete or continuous) that are potentially significant, (ii) to define a performance indicator to quantify the percentage of correlation explained by the significant factors subset for Zero-Inflated Proportions data (ZIprop).

References

Melina Ribaud, Edith Gabriel, Joseph Hughes, Samuel Soubeyrand. Identifying potential significant factors impacting zero-inflated proportions data. 2020. hal-02936779

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