

Package ‘ShapleyOutlier’

October 21, 2022

Type Package

Title Multivariate Outlier Explanations using Shapley Values and Mahalanobis Distances

Version 0.1.0

Depends R (>= 4.0.0)

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Description Based on Shapley values to explain multivariate outlyingness and to detect and impute cellwise outliers.
Includes implementations of methods described in Mayrhofer and Filzmoser (2022) <[doi:10.48550/ARXIV.2210.10063](https://doi.org/10.48550/ARXIV.2210.10063)>.

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Imports dplyr, Rdpack, stats, tibble, tidyr, robustbase, forcats, egg, ggplot2, gridExtra, RColorBrewer, magrittr

Suggests grDevices, cellWise, robustHD, tidyverse, knitr, MASS, rmarkdown

RdMacros Rdpack

Encoding UTF-8

LazyData true

RoxygenNote 7.2.1

VignetteBuilder knitr

Repository CRAN

NeedsCompilation no

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Date/Publication 2022-10-21 15:15:07 UTC

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MOE	<i>Detecting cellwise outliers using Shapley values based on local outlyingness.</i>
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Description

The MOE function indicates outlying cells for a data vector with p entries or data matrix with $n \times p$ entries containing only numeric entries x for a given center μ and covariance matrix Σ using the Shapley value. It is a more sophisticated alternative to the SCD algorithm, which uses the information of the regular cells to derive an alternative reference point (Mayrhofer and Filzmoser 2022).

Usage

```
MOE(
  x,
  mu,
  Sigma,
  Sigma_inv = NULL,
  step_size = 0.1,
  min_deviation = 0,
  max_step = NULL,
  local = TRUE,
  max_iter = 1000,
  q = 0.99,
  check_outlyingness = FALSE,
  check = TRUE,
  cells = NULL,
  method = "cellMCD"
)
```

Arguments

<code>x</code>	Data vector with p entries or data matrix with $n \times p$ entries containing only numeric entries.
<code>mu</code>	Either NULL (default) or mean vector of <code>x</code> . If NULL, method is used for parameter estimation.
<code>Sigma</code>	Either NULL (default) or covariance matrix $p \times p$ of <code>x</code> . If NULL, method is used for parameter estimation.
<code>Sigma_inv</code>	Either NULL (default) or <code>Sigma</code> 's inverse $p \times p$ matrix. If NULL, the inverse of <code>Sigma</code> is computed using <code>solve(Sigma)</code> .
<code>step_size</code>	Numeric. Step size for the imputation of outlying cells, with <code>step_size</code> $\in [0, 1]$. Defaults to 0.1.
<code>min_deviation</code>	Numeric. Detection threshold, with <code>min_deviation</code> $\in [0, 1]$. Defaults to 0.2
<code>max_step</code>	Either NULL (default) or an integer. The maximum number of steps in each iteration. If NULL, <code>max_step</code> = p .
<code>local</code>	Logical. If TRUE (default), the non-central Chi-Squared distribution is used to determine the cutoff value based on <code>mu_tilde</code> .
<code>max_iter</code>	Integer. The maximum number of iterations.
<code>q</code>	Numeric. The quantile of the Chi-squared distribution for detection and imputation of outliers. Defaults to 0.99.
<code>check_outlyingness</code>	Logical. If TRUE (default), the outlyingness is rechecked after applying <code>min_deviation</code> .
<code>check</code>	Logical. If TRUE (default), inputs are checked before running the function and an error message is returned if one of the inputs is not as expected.
<code>cells</code>	Either NULL (default) or a vector/matrix of the same dimension as <code>x</code> , indicating the outlying cells. The matrix must contain only zeros and ones, or TRUE/FALSE.
<code>method</code>	Either "cellMCD" (default) or "MCD". Specifies the method used for parameter estimation if <code>mu</code> and/or <code>Sigma</code> are not provided.

Value

A list of class `shapley_algorithm` ([new_shapley_algorithm](#)) containing the following:

<code>x</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the imputed data.
<code>phi</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the Shapley values (outlyingness-scores) of <code>x</code> ; see shapley .
<code>mu_tilde</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the alternative reference points based on the regular cells of the original observations.
<code>x_original</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the original data.
<code>x_original</code>	The non-centrality parameters for the Chi-Squared distribution
<code>x_history</code>	A list with n elements, each containing the path of how the original data vector was modified.
<code>phi_history</code>	A list with n elements, each containing the Shapley values corresponding to <code>x_history</code> .

mu_tilde_history	A list with n elements, each containing the alternative reference points corresponding to <code>x_history</code> .
S_history	A list with n elements, each containing the indices of the outlying cells in each iteration.

References

Mayrhofer M, Filzmoser P (2022). “Multivariate outlier explanations using Shapley values and Mahalanobis distances.” [doi:10.48550/ARXIV.2210.10063](https://doi.org/10.48550/ARXIV.2210.10063).

Examples

```
p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
MOE_x <- MOE(x = x, mu = mu, Sigma = Sigma)
plot(MOE_x)

library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
MOE_X <- MOE(X, mu, Sigma)
plot(MOE_X, subset = 20)
```

new_shapley

Class constructor for class shapley.

Description

This function creates an object of class `shapley` that is returned by the `shapley` function.

Usage

```
new_shapley(phi = numeric(), mu_tilde = NULL, non centrality = NULL)
```

Arguments

phi	A p -dimensional vector (or a $n \times p$ matrix) containing the Shapley values (outlyingness-scores) of a p -dimensional data vector (or a $n \times p$ data matrix).
mu_tilde	Optional. A p -dimensional vector (or a $n \times p$ matrix) containing the alternative reference points based on the regular cells of the original observations.
non centrality	Optional. The non-centrality parameters for the Chi-Squared distribution, which are given by <code>mahlanobis(mu_tilde, mu, Sigma)</code> .

Value

Named list of class `shapley`, containing the input parameters.

`new_shapley_algorithm` *Class constructor for class `shapley_algorithm`.*

Description

This function creates an object of class `shapley_algorithm` that is returned by the [SCD](#) and [MOE](#) functions.

Usage

```
new_shapley_algorithm(
  x = numeric(),
  phi = numeric(),
  x_original = numeric(),
  mu_tilde = NULL,
  non centrality = NULL,
  x_history = NULL,
  phi_history = NULL,
  mu_tilde_history = NULL,
  S_history = NULL
)
```

Arguments

<code>x</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the imputed data.
<code>phi</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the Shapley values (outlyingness-scores) of a p -dimensional data vector (or a $n \times p$ data matrix).
<code>x_original</code>	A p -dimensional vector (or a $n \times p$ matrix) containing the original data.
<code>mu_tilde</code>	Optional. A p -dimensional vector (or a $n \times p$ matrix) containing the alternative reference points based on the regular cells of the original observations.
<code>non centrality</code>	Optional. The non-centrality parameters for the Chi-Squared distribution, which are given by <code>mahlanobis(mu_tilde, mu, Sigma)</code> .
<code>x_history</code>	Optional. A list with n elements, each containing the path of how the original data vector was modified.
<code>phi_history</code>	Optional. A list with n elements, each containing the Shapley values corresponding to <code>x_history</code> .
<code>mu_tilde_history</code>	Optional. A list with n elements, each containing the alternative reference points corresponding to <code>x_history</code> .
<code>S_history</code>	Optional. A list with n elements, each containing the indices of the outlying cells in each iteration.

Value

Named list of class `shapley_algorithm`, containing the input parameters.

`new_shapley_interaction`

Class constructor for class `shapley_interaction`.

Description

This function creates an object of class `shapley_interaction` that is returned by the [shapley_interaction](#) function.

Usage

```
new_shapley_interaction(PHI = numeric())
```

Arguments

PHI A $p \times p$ matrix containing the decomposition of the squared Mahalanobis distance of a p -dimensional numeric vector into outlyingness scores for pairs of variables.

Value

Matrix of class `shapley_interaction`, containing input matrix `PHI`.

`plot.shapley`

Barplot of Shapley values

Description

Barplot of Shapley values

Usage

```
## S3 method for class 'shapley'
plot(
  x,
  subset = NULL,
  chi2.q = 0.99,
  abbrev.var = 3,
  abbrev.obs = 10,
  sort.var = FALSE,
  sort.obs = FALSE,
  plot_md = TRUE,
```

```

    md_squared = TRUE,
    rotate_x = TRUE,
    ...
  )

```

Arguments

x	A list of class shapley.
subset	Either an integer, "chi2", or NULL (default) to select which rows of phi should be displayed. If NULL, all n rows are displayed, for a single integer the subset rows with the highest Mahalanobis distance are displayed, for an integer vector the subset selected rows are displayed, and for "chi2" all outlying rows are displayed (Mahalanobis distance greater than $\sqrt{q\text{chisq}(\text{chi2},q,p)}$).
chi2.q	Quantile, only used if subset == "chi2".
abbrev.var	Integer. If abbrev.var > 0, column names are abbreviated using abbreviate with minlength = abbrev.var.
abbrev.obs	Integer. If abbrev.obs > 0, row names are abbreviated using abbreviate with minlength = abbrev.obs.
sort.var	Logical. If TRUE (default), variables are sorted according to the distance
sort.obs	Logical. If TRUE (default), observations are sorted according to their Mahalanobis distance.
plot_md	Logical. If TRUE (default), the Mahalanobis distance will be included in the plot.
md_squared	Logical. If TRUE (default), the squared Mahalanobis distance is plotted otherwise the (not-squared) Mahalanobis distance.
rotate_x	Logical. If TRUE (default), the x-axis labels are rotated.
...	Optional arguments passed to methods.

Value

Returns a barplot that displays the Shapley values ([shapley](#)) for each observation and optionally (plot_md = TRUE) includes the squared Mahalanobis distance (black bar) and the corresponding (non-)central chi-square quantile (dotted line).

Examples

```

library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X_clean <- X
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
call_shapley <- shapley(X, mu, Sigma)
plot(call_shapley, subset = 1:20)
plot(call_shapley, subset = 5, rotate_x = FALSE)
plot(call_shapley, subset = 5, md_squared = FALSE, rotate_x = FALSE)

```

```
plot.shapley_algorithm
```

Barplot and tileplot of Shapley values.

Description

Barplot and tileplot of Shapley values.

Usage

```
## S3 method for class 'shapley_algorithm'
plot(
  x,
  type = "both",
  subset = NULL,
  abbrev.var = FALSE,
  abbrev.obs = FALSE,
  sort.var = FALSE,
  sort.obs = FALSE,
  n_digits = 2,
  rotate_x = TRUE,
  continuous_rowname = FALSE,
  ...
)
```

Arguments

x	A list of class shapley_algorithm.
type	Either "both" (default), "bar", or "cell". If "both" (default) a barplot and a tileplot are created, otherwise only the selected plot is created.
subset	Either an integer, "chi2", or NULL (default) to select which rows of phi should be displayed. If NULL, all n rows are displayed, for a single integer the subset rows with the highest Mahalanobis distance are displayed, for an integer vector the subset selected rows are displayed, and for "chi2" all outlying rows are displayed (Mahalanobis distance greater than $\sqrt{q\text{chisq}(\text{chi2}, q, p)}$).
abbrev.var	Integer. If abbrev.var > 0, column names are abbreviated using abbreviate with minlength = abbrev.var.
abbrev.obs	Integer. If abbrev.obs > 0, row names are abbreviated using abbreviate with minlength = abbrev.obs.
sort.var	Logical. If TRUE (default), variables are sorted according to the distance
sort.obs	Logical. If TRUE (default), observations are sorted according to their Mahalanobis distance.
n_digits	Integer. If n_digits > 0, the original values of the variables are given in each cell with n_digits decimals places.

rotate_x Logical. If TRUE (default), the x-axis labels are rotated.
 continuous_rowname Logical. If TRUE, the rownames are converted to a numeric vector.
 ... Arguments passed on to `plot.shapley`.

Value

Returns plots for a list of class `shapley_algorithm`. If type is "bar", a barplot is generated. It displays the Shapley values (`shapley`) for each observation and optionally (`plot_md = TRUE`) includes the squared Mahalanobis distance (black bar) and the corresponding (non-)central chi-square quantile (dotted line). If type is "cell" a tileplot is generated. It displays each cells of the dataset and shows the original value from the observations, color coding indicates whether those values were higher (red) or lower (blue) than the imputed values, and the color intensity is based on the magnitude of the Shapley value. If type is "both", the barplot and the tileplot are generated.

Examples

```
library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
MOE_X <- MOE(X, mu, Sigma)
plot(MOE_X, subset = 20, n_digits = 0)
```

plot.shapley_interaction

Plot of Shapley interaction indices

Description

Plot of Shapley interaction indices

Usage

```
## S3 method for class 'shapley_interaction'
plot(
  x,
  abbrev = 4,
  title = "Shapley Interaction",
  legend = TRUE,
  text_size = 22,
  ...
)
```

Arguments

<code>x</code>	A $p \times p$ matrix containing the Shapley interaction indices (<code>shapley_interaction</code>) of a single observation.
<code>abbrev</code>	Integer. If <code>abbrev.var > 0</code> , variable names are abbreviated using <code>abbreviate</code> with <code>minlength = abbrev</code> .
<code>title</code>	Character. Title of the plot.
<code>legend</code>	Logical. If TRUE (default), a legend is plotted.
<code>text_size</code>	Integer. Size of the text in the plot
<code>...</code>	Optional arguments passed to methods.

Value

Returns a figure consisting of two panels. The upper panel shows the Shapley values, and the lower panel the Shapley interaction indices.

Examples

```
p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
PHI <- shapley_interaction(x, mu, Sigma)
plot(PHI)
```

`print.shapley` *Print function for class shapley.*

Description

Print function for class `shapley`.

Usage

```
## S3 method for class 'shapley'
print(x, ...)
```

Arguments

<code>x</code>	List of class <code>shapley</code> .
<code>...</code>	Optional arguments passed to methods.

Value

Prints the list entries of `x` that are not NULL.

```
print.shapley_algorithm
```

Print function for class shapley_algorithm.

Description

Print function for class shapley_algorithm.

Usage

```
## S3 method for class 'shapley_algorithm'  
print(x, ...)
```

Arguments

x List of class shapley_algorithm.
... Optional arguments passed to methods.

Value

Prints the imputed data and the Shapley values.

```
print.shapley_interaction
```

Print function for class shapley_interaction.

Description

Print function for class shapley_interaction.

Usage

```
## S3 method for class 'shapley_interaction'  
print(x, ...)
```

Arguments

x Matrix of class shapley_interaction.
... Optional arguments passed to methods.

Value

Prints the Shapley interaction indices.

SCD

*Detecting cellwise outliers using Shapley values.***Description**

The SCD function indicates outlying cells for a data vector with p entries or data matrix with $n \times p$ entries containing only numeric entries x for a given center μ and covariance matrix Σ using the Shapley value (Mayrhofer and Filzmoser 2022).

Usage

```
SCD(
  x,
  mu,
  Sigma,
  Sigma_inv = NULL,
  step_size = 0.1,
  min_deviation = 0,
  max_step = NULL,
  max_iter = 1000,
  q = 0.99,
  method = "cellMCD",
  check = TRUE,
  cells = NULL
)
```

Arguments

<code>x</code>	Data vector with p entries or data matrix with $n \times p$ entries containing only numeric entries.
<code>mu</code>	Either NULL (default) or mean vector of x . If NULL, method is used for parameter estimation.
<code>Sigma</code>	Either NULL (default) or covariance matrix $p \times p$ of x . If NULL, method is used for parameter estimation.
<code>Sigma_inv</code>	Either NULL (default) or Σ 's inverse $p \times p$ matrix. If NULL, the inverse of Σ is computed using <code>solve(Sigma)</code> .
<code>step_size</code>	Numeric. Step size for the imputation of outlying cells, with <code>step_size</code> $\in [0, 1]$. Defaults to 0.1.
<code>min_deviation</code>	Numeric. Detection threshold, with <code>min_deviation</code> $\in [0, 1]$. Defaults to 0.2
<code>max_step</code>	Either NULL (default) or an integer. The maximum number of steps in each iteration. If NULL, <code>max_step</code> = p .
<code>max_iter</code>	Integer. The maximum number of iterations.
<code>q</code>	Numeric. The quantile of the Chi-squared distribution for detection and imputation of outliers. Defaults to 0.99.

method	Either "cellMCD" (default) or "MCD". Specifies the method used for parameter estimation if mu and/or Sigma are not provided.
check	Logical. If TRUE (default), inputs are checked before running the function and an error message is returned if one of the inputs is not as expected.
cells	Either NULL (default) or a vector/matrix of the same dimension as x, indicating the outlying cells. The matrix must contain only zeros and ones, or TRUE/FALSE.

Value

A list of class `shapley_algorithm` (`new_shapley_algorithm`) containing the following:

x	A p -dimensional vector (or a $n \times p$ matrix) containing the imputed data.
phi	A p -dimensional vector (or a $n \times p$ matrix) containing the Shapley values (outlyingness-scores) of x; see <code>shapley</code> .
x_original	A p -dimensional vector (or a $n \times p$ matrix) containing the original data.
x_history	The path of how the original data vector was modified.
phi_history	The Shapley values corresponding to x_history.
S_history	The indices of the outlying cells in each iteration.

References

Mayrhofer M, Filzmoser P (2022). "Multivariate outlier explanations using Shapley values and Mahalanobis distances." [doi:10.48550/ARXIV.2210.10063](https://doi.org/10.48550/ARXIV.2210.10063).

Examples

```
p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
SCD_x <- SCD(x = x, mu = mu, Sigma = Sigma)
plot(SCD_x)

library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
SCD_X <- SCD(X, mu, Sigma)
plot(SCD_X, subset = 20)
```

shapley	<i>Decomposition of squared Mahalanobis distance using Shapley values.</i>
---------	----------------------------------------------------------------------------

Description

The `shapley` function computes a p -dimensional vector containing the decomposition of the squared Mahalanobis distance of x (with respect to μ and Σ) into outlyingness contributions of the individual variables (Mayrhofer and Filzmoser 2022). The value of the j -th coordinate of this vector represents the average marginal contribution of the j -th variable to the squared Mahalanobis distance of the individual observation x .

If `cells` is provided, Shapley values of x are computed with respect to a local reference point, that is based on a cellwise prediction of each coordinate, using the information of the regular cells of x , see (Mayrhofer and Filzmoser 2022).

If x is a $n \times p$ matrix, a $n \times p$ matrix is returned, containing the decomposition for each row.

Usage

```
shapley(
  x,
  mu = NULL,
  Sigma = NULL,
  inverted = FALSE,
  method = "cellMCD",
  check = TRUE,
  cells = NULL
)
```

Arguments

<code>x</code>	Data vector with p entries or data matrix with $n \times p$ entries containing only numeric entries.
<code>mu</code>	Either NULL (default) or mean vector of x . If NULL, <code>method</code> is used for parameter estimation.
<code>Sigma</code>	Either NULL (default) or covariance matrix $p \times p$ of x . If NULL, <code>method</code> is used for parameter estimation.
<code>inverted</code>	Logical. If TRUE, <code>Sigma</code> is supposed to contain the inverse of the covariance matrix.
<code>method</code>	Either "cellMCD" (default) or "MCD". Specifies the method used for parameter estimation if <code>mu</code> and/or <code>Sigma</code> are not provided.
<code>check</code>	Logical. If TRUE (default), inputs are checked before running the function and an error message is returned if one of the inputs is not as expected.
<code>cells</code>	Either NULL (default) or a vector/matrix of the same dimension as x , indicating the outlying cells. The matrix must contain only zeros and ones, or TRUE/FALSE.

Value

phi	A p -dimensional vector (or a $n \times p$ matrix) containing the Shapley values (outlyingness-scores) of x .
mu_tilde	A p -dimensional vector (or a $n \times p$ matrix) containing the alternative reference points based on the regular cells of the original observations.
non centrality	The non-centrality parameters for the Chi-Squared distribution, given by mahlanobis(mu_tilde, mu, Sigma)

References

Mayrhofer M, Filzmoser P (2022). "Multivariate outlier explanations using Shapley values and Mahalanobis distances." doi:[10.48550/ARXIV.2210.10063](https://doi.org/10.48550/ARXIV.2210.10063).

Examples

```
## Without outlying cells as input in the 'cells' argument#
# Single observation
p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
shapley(x, mu, Sigma)
phi <- shapley(x, mu, Sigma_inv, inverted = TRUE)
plot(phi)

# Multiple observations
library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X_clean <- X
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
call_shapley <- shapley(X, mu, Sigma)
plot(call_shapley, subset = 20)

## Giving outlying cells as input in the 'cells' argument
# Single observation
p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
call_shapley <- shapley(x, mu, Sigma_inv, inverted = TRUE,
method = "cellMCD", check = TRUE, cells = c(1,1,0,0,0))
plot(call_shapley)

# Multiple observations
```

```

library(MASS)
set.seed(1)
n <- 100; p <- 10
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
X <- mvrnorm(n, mu, Sigma)
X_clean <- X
X[sample(1:(n*p), 100, FALSE)] <- rep(c(-5,5),50)
call_shapley <- shapley(X, mu, Sigma, cells = (X_clean - X)!=0)
plot(call_shapley, subset = 20)

```

shapley_interaction *Decomposition of squared Mahalanobis distance using Shapley interaction indices.*

Description

The `shapley_interaction` function computes a $p \times p$ matrix containing pairwise outlyingness scores based on Shapley interaction indices. It decomposes the squared Mahalanobis distance of x (with respect to μ and Σ) into outlyingness contributions of pairs of variables (Mayrhofer and Filzmoser 2022).

Usage

```
shapley_interaction(x, mu, Sigma, inverted = FALSE)
```

Arguments

<code>x</code>	Data vector with p entries containing only numeric entries.
<code>mu</code>	Either NULL (default) or mean vector of x . If NULL, method is used for parameter estimation.
<code>Sigma</code>	Either NULL (default) or covariance matrix $p \times p$ of x . If NULL, method is used for parameter estimation.
<code>inverted</code>	Logical. If TRUE, <code>Sigma</code> is supposed to contain the inverse of the covariance matrix.

Value

A $p \times p$ matrix containing the decomposition of the squared Mahalanobis distance of x into outlyingness scores for pairs of variables with respect to μ and Σ .

References

Mayrhofer M, Filzmoser P (2022). “Multivariate outlier explanations using Shapley values and Mahalanobis distances.” [doi:10.48550/ARXIV.2210.10063](https://doi.org/10.48550/ARXIV.2210.10063).

Examples

```

p <- 5
mu <- rep(0,p)
Sigma <- matrix(0.9, p, p); diag(Sigma) = 1
Sigma_inv <- solve(Sigma)
x <- c(0,1,2,2.3,2.5)
shapley_interaction(x, mu, Sigma)
PHI <- shapley_interaction(x, mu, Sigma_inv, inverted = TRUE)
plot(PHI)

```

WeatherVienna

Weather data from Vienna

Description

Monthly data from the weather station Hohe Warte since April 1872 - Vienna (Stadt Wien 2022).

Usage

```
WeatherVienna
```

Format

A data frame with 1,804 rows and 25 columns:

year Year

month Month

t Daily mean air temperature in °C, $(t7 \text{ mean} + t19 \text{ mean} + t_{\max} \text{ mean} + t_{\min} \text{ mean})/4$; before 1971: $t7 \text{ mean} + t14 \text{ mean} + 2 \times t21 \text{ mean}$

t_max Absolute maximum air temperature in °C

t_min Absolute air temperature minimum in °C

avg_t_max Mean daily maximum air temperature in °C

avg_t_min Mean daily minimum air temperature in °C

num_frost Number of frost days (days with a temperature maximum $t_{\min} < 0.0$ °C)

num_ice Number of ice days (days with a temperature maximum $t_{\max} < 0.0$ °C)

num_summer Number of summer days (days with a temperature maximum $t_{\max} \geq 25.0$ °C)

num_heat Number of hot days (days with a temperature maximum $t_{\max} \geq 30.0$ °C)

p Daily mean air pressure in hPa (mean of all measurements at 7 a.m., 2 p.m., 7 p.m. CET; before 1971 9 p.m. instead of 7 p.m.)

p_max Maximum air pressure in hPa (maximum of all measurements 7 am, 2 pm, 7 pm CET; before 1971 9 pm instead of 7 pm)

p_min Minimum air pressure in hPa (minimum of all measurements 7 am, 2 pm, 7 pm CET; before 1971 9 pm instead of 7 pm)

sun_h Monthly total sunshine duration in hours
num_clear Number of clear days (daily mean cloudiness < 20/100)
num_cloud Number of cloudy days (daily mean cloudiness > 80/100)
rel_hum Daily mean relative humidity in percent (2 x RH7 mean + RH14 mean + RH19 mean)/4;
before 1971 9 p.m. instead of 7 p.m.)
rel_hum_max Relative humidity maximum in percent
rel_hum_min Relative humidity minimum in percent
wind_v Monthly average wind speed in km/h
num_wind_v60 Number of days with wind peaks >= 60 km/h
wind_v_max Maximum wind speed in km/h
precip_sum Monthly total precipitation in mm
num_precp_01 Number of days with precipitation >= 0.1 mm

Source

The data were downloaded from <https://www.data.gv.at/katalog/dataset/wetter-seit-1872-hohe-warte-wien> in September 2022.

References

Stadt Wien (2022). “Monthly data from the weather station Hohe Warte since April 1872 - Vienna.”
<https://www.data.gv.at/katalog/dataset/wetter-seit-1872-hohe-warte-wien>.

Examples

```
data("WeatherVienna")  
summary(WeatherVienna)
```

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