

Package ‘alphaOutlier’

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Title Obtain Alpha-Outlier Regions for Well-Known Probability Distributions

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Maintainer Andre Rehage <andre.rehage@gmx.de>

Description Given the parameters of a distribution, the package uses the concept of alpha-outliers by Davies and Gather (1993) to flag outliers in a data set. See Davies, L.; Gather, U. (1993): The identification of multiple outliers, JASA, 88 423, 782-792, <doi:10.1080/01621459.1993.10476339> for details.

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Author Andre Rehage [aut, cre],
Sonja Kuhnt [aut]

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Contents

alphaOutlier-package	2
aout.binom	3
aout.cg	4
aout.chisq	5
aout.conntab	6
aout.exp	7
aout.gandh	8
aout.hyper	9
aout.kernel	10
aout.laplace	11
aout.logis	12

aout.mvnorm	13
aout.nbinom	14
aout.norm	15
aout.pareto	16
aout.pois	17
aout.weibull	18
citiesData	19
createDesMat	19
daysabs	20

Index	21
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alphaOutlier-package *Obtain α -outlier regions for well-known probability distributions*

Description

Given the parameters of a distribution, the package uses the concept of α -outliers by Davies and Gather (1993) to flag outliers in a data set.

Details

The structure of the package is as follows: `aout.[Distribution]` is the name of the function which returns the α -outlier region of a random variable following `[Distribution]`. The names of the distributions are abbreviated as in the `d`, `p`, `q`, `r` functions. Use pre-specified or robustly estimated parameters from your data to obtain reasonable results. The sample size should be taken into account when choosing alpha, for example Gather et al. (2003) propose $\alpha_N = 1 - (1 - \alpha)^{1/N}$.

Author(s)

A. Rehage, S. Kuhnt

References

Davies, L.; Gather, U. (1993) The identification of multiple outliers, *Journal of the American Statistical Association*, **88** 423, 782-792.

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

See Also

[nleqslv](#), [solnp](#), [rq.fit.fnc](#)

Examples

```
iris.setosa <- iris[1:51, 4]
aout.norm(data = iris.setosa, param = c(mean(iris.setosa), sd(iris.setosa)), alpha = 0.01)
aout.pois(data = warpbreaks[,1], param = mean(warpbreaks[,1]), alpha = 0.01,
          hide.outliers = TRUE)
```

aout.binom	<i>Find α-outliers in Binomial data</i>
------------	---

Description

Given the parameters of a Binomial distribution, aout.binom identifies α -outliers in a given data set.

Usage

```
aout.binom(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

data	a vector. The data set to be examined.
param	a vector. Contains the parameters of the Binomial distribution, N and p .
alpha	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
hide.outliers	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named is.outlier that flags the outliers with TRUE. If hide.outliers is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

See Also

[dbinom](#)

Examples

```
data(uis)
medbeck <- median(uis$BECK)
aout.binom(data = uis$BECK, param = c(54, medbeck/54), alpha = 0.001)
```



```
aout.cg(ratweight,
        list(p = c(1/3, 1/3, 1/3), mu = c(7, 7, 14), sigma = c(1.6, 1.4, 3.3)))
```

aout.chisq *Find α -outliers in χ^2 data*

Description

Given the parameters of a χ^2 distribution, `aout.chisq` identifies α -outliers in a given data set.

Usage

```
aout.chisq(data, param, alpha = 0.1, hide.outliers = FALSE, ncp = 0, lower = auto.l,
           upper = auto.u, method.in = "Newton", global.in = "gline",
           control.in = list(sigma = 0.1, maxit = 1000, xtol = 1e-12,
                             ftol = 1e-12, btol = 1e-04))
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	an atomic vector. Contains the degrees of freedom of the χ^2 distribution.
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.
<code>ncp</code>	an atomic vector. Determines the non-centrality parameter of the χ^2 distribution. Defaults to 0.
<code>lower</code>	an atomic vector. First element of <code>x</code> from nleqslv .
<code>upper</code>	an atomic vector. Second element of <code>x</code> from nleqslv .
<code>method.in</code>	See <code>method</code> in nleqslv .
<code>global.in</code>	See <code>global</code> in nleqslv .
<code>control.in</code>	See <code>control</code> in nleqslv .

Details

The α -outlier region of a χ^2 distribution is generally not available in closed form or via the tails, such that a non-linear equation system has to be solved.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

See Also[dchisq](#)**Examples**

```
aout.chisq(chisq.test(occupationalStatus)$statistic, 49)
```

```
aout.conttab
```

```
Find  $\alpha$ -outliers in two-way contingency tables
```

Description

This is a wrapper function for [aout.pois](#). We assume that each entry of a contingency table can be seen as a realization of a Poisson random variable. The parameter λ of each cell can either be set by the user or estimated. Given the parameters, `aout.conttab` identifies α -outliers in a given contingency table.

Usage

```
aout.conttab(data, param, alpha = 0.1, hide.outliers = FALSE, show.estimates = FALSE)
```

Arguments

<code>data</code>	a matrix or data.frame. The contingency table to be examined.
<code>param</code>	a character string from <code>c("ML", "L1", "MP")</code> or a vector containing the parameters of each cell of the Poisson distribution: λ . "ML" yields the maximum likelihood estimate from the log-linear Poisson model using a suitable design matrix. "L1" yields the L1-estimate from <code>rq.fit.fnc</code> . "MP" yields the Median Polish estimate. If the parameter vector is given by the user, it is necessary that the contingency table was filled by <code>byrow = FALSE</code> .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.
<code>show.estimates</code>	boolean. Returns $\hat{\lambda}$ for each cell if set to TRUE. Defaults to FALSE.

Value

Data frame of the vectorized input data and, if desired, an index named `is.outlier` that flags the outliers with TRUE and a vector named `param` containing the estimated lambdas.

Author(s)

A. Rehage

References

Kuhnt, S. (2000) Ausreisseridentifikation im Loglinearen Poissonmodell fuer Kontingenztafeln unter Einbeziehung robuster Schaetzer. Ph.D. Thesis. Universitaet Dortmund, Dortmund. Fachbereich Statistik.

Kuhnt, S.; Rapallo, F.; Rehage, A. (2014) Outlier detection in contingency tables based on minimal patterns. *Statistics and Computing* 24 (3), 481-491.

See Also

[rq.fit.fnc](#), [aout.pois](#)

Examples

```
aout.conttab(data = HairEyeColor[, ,1], param = "L1", alpha = 0.01, show.estimates = TRUE)
aout.conttab(data = HairEyeColor[, ,1], param = "ML", alpha = 0.01, show.estimates = TRUE)
```

aout.exp

Find α -outliers in exponentially distributed data

Description

Given the parameters of an exponential distribution, `aout.exp` identifies α -outliers in a given data set.

Usage

```
aout.exp(data, param, alpha = 0.1, hide.outliers = FALSE, theta = 0)
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	an atomic vector. Contains the parameter of the exponential distribution.
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.
<code>theta</code>	an atomic vector. Determines the lower bound of the support of the exponential distribution. Defaults to 0.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

See Also

[dexp](#)

Examples

```
aout.exp(attenu[,5], median(attenu[,5]), alpha = 0.05)
```

```
aout.gandh
```

Find α -outliers in data from the family of g -and- h distributions

Description

Given the parameters of a g -and- h distribution, `aout.gandh` identifies α -outliers in a given data set.

Usage

```
aout.gandh(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	a vector. Contains the parameters of the g -and- h distribution: median, scale, g , h .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Details

The concept of α -outliers is based on the p.d.f. of the random variable. Since for g -and- h distributions this does not exist in closed form, the computation of the outlier region is based on an optimization of the quantile function with side conditions.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Note

Makes use of [solnp](#).

Author(s)

A. Rehage

References

Xu, Y.; Iglewicz, B.; Chervoneva, I. (2014) Robust estimation of the parameters of g-and-h distributions, with applications to outlier detection. *Computational Statistics and Data Analysis* 75, 66-80.

Examples

```
durations <- faithful$eruptions
aout.gandh(durations, c(4.25, 1.14, 0.05, 0.05), alpha = 0.1)
```

`aout.hyper`*Find α -outliers in hypergeometric data*

Description

Given the parameters of a hypergeometric distribution, `aout.hyper` identifies α -outliers in a given data set.

Usage

```
aout.hyper(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	a vector. Contains the parameters of the hypergeometric distribution: m, n, k .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

See Also[Hypergeometric](#)

Examples

```
set.seed(1)
lotto6aus49 <- rhyper(100, 6, 43, 6)
aout.hyper(lotto6aus49, c(6, 43, 6), 0.1)
```

aout.kernel	<i>Find α-outliers in arbitrary univariate data using kernel density estimation</i>
-------------	---

Description

Given the arguments of the density, aout.kernel identifies α -outliers in a given data set.

Usage

```
aout.kernel(data, alpha, plot = TRUE, plottitle = "", kernel = "gaussian",
nkernel = 1024, kern.bw = "SJ", kern.adj = 1,
xlim = NA, ylim = NA, outints = FALSE, w = NA, ...)
```

Arguments

data	a vector. The data set to be examined.
alpha	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain.
plot	boolean. If TRUE, a plot of the data and estimated density with shaded outlier region is printed.
plottitle	character string. Title of the plot.
kernel	See kernel in density .
nkernel	See n in density .
kern.bw	See bw in density .
kern.adj	See adjust in density .
xlim	a vector. Specify if you want to change the x-limits of the plot.
ylim	a vector. Specify if you want to change the y-limits of the plot.
outints	boolean. If TRUE, then the bounds of the inlier-regions and the chosen bandwidth are shown.
w	a vector. See weights in density .
...	Further arguments for density and plot.

Value

If `outints = TRUE`, a list of

Results A data frame containing one row for each observation. The observations are labelled whether they are outlying, the value of the estimated density at the observation is shown and the bound of the outlier identifier.

Bounds.of.Inlier.Regions The bounds of the inlier region(s).

KDE.Chosen.Bandwidth The bandwidth that was chosen by density.

Author(s)

A. Rehage

Examples

```
set.seed(23)
tempx <- rnorm(1000, 0, 1)
tempx[1] <- -2.5
aout.kernel(tempx[1:10], alpha = 0.1, kern.adj = 1, xlim = c(-3,3), outints = TRUE)
# not run:
# aout.kernel(tempx[1:200], alpha = 0.1, kern.adj = 1, xlim = c(-3,3))
```

aout.laplace *Find α -outliers in Laplace / double exponential data*

Description

Given the parameters of a Laplace distribution, `aout.laplace` identifies α -outliers in a given data set.

Usage

```
aout.laplace(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

data a vector. The data set to be examined.

param a vector. Contains the parameters of the Laplace distribution: μ, σ .

alpha an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.

hide.outliers boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Dumonceaux, R.; Antle, C. E. (1973) Discrimination between the log-normal and the Weibull distributions. *Technometrics*, 15 (4), 923-926.

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

Examples

```
# Using the flood data from Dumonceaux and Antle (1973):
temp <- c(0.265, 0.269, 0.297, 0.315, 0.3225, 0.338, 0.379, 0.380, 0.392, 0.402,
         0.412, 0.416, 0.418, 0.423, 0.449, 0.484, 0.494, 0.613, 0.654, 0.74)
aout.laplace(temp, c(median(temp), median(abs(temp - median(temp))))), 0.05)
```

aout.logis

Find α -outliers in logistic data

Description

Given the parameters of a logistic distribution, aout.logis identifies α -outliers in a given data set.

Usage

```
aout.logis(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

data	a vector. The data set to be examined.
param	a vector. Contains the parameters of the logistic distribution: μ, σ .
alpha	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
hide.outliers	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named is.outlier that flags the outliers with TRUE. If hide.outliers is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Balakrishnan, N. (1992) Maximum likelihood estimation based on complete and type II censored samples. In N. Balakrishnan (Ed.): *Handbook of the Logistic Distribution*. Dekker, New York, 49-78.

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

See Also

[dlogis](#)

Examples

```
# Data example from Balakrishnan (1967)
lifetime <- c(785, 855, 905, 918, 919, 920, 929, 936, 948, 950)
aout.logis(lifetime, c(949.9, 63.44))
```

aout.mvnorm

Find α -outliers in multivariate normal data

Description

Given the parameters of a multivariate normal distribution, `aout.mvnorm` identifies α -outliers in a given data set.

Usage

```
aout.mvnorm(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

<code>data</code>	a data.frame or matrix. The data set to be examined.
<code>param</code>	a list. Contains the parameters of the normal distribution: the mean vector μ and the covariance matrix σ .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a data frame of the outlier-free data.

Author(s)

A. Rehage

References

Kuhnt, S.; Rehage, A. (2013) The concept of α -outliers in structured data situations. In C. Becker, R. Fried, S. Kuhnt (Eds.): *Robustness and Complex Data Structures. Festschrift in Honour of Ursula Gather*. Berlin: Springer, 91-108.

See Also

[dnorm](#)

Examples

```
temp <- iris[1:51,-5]
temp.xq <- apply(FUN = median, MARGIN = 2, temp)
aout.mvnorm(as.matrix(temp), param = list(temp.xq, cov(temp)), alpha = 0.001)
```

aout.nbinom

Find α -outliers in negative Binomial data

Description

Given the parameters of a negative Binomial distribution, `aout.nbinom` identifies α -outliers in a given data set.

Usage

```
aout.nbinom(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

`data` a vector. The data set to be examined.

`param` a vector. Contains the parameters of the negative Binomial distribution: N, p .

`alpha` an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.

`hide.outliers` boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

See Also

[dnbinom](#), [daysabs](#)

Examples

```
data(daysabs)
aout.nbinom(daysabs, c(8, 0.6), 0.05)
```

aout.norm	<i>Find α-outliers in normal data</i>
-----------	---

Description

Given the parameters of a normal distribution, `aout.norm` identifies α -outliers in a given data set.

Usage

```
aout.norm(data, param = c(0, 1), alpha = 0.1, hide.outliers = FALSE)
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	a vector. Contains the parameters of the normal distribution: μ, σ .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

See Also

[dnorm](#)

Examples

```
iris.setosa <- iris[1:51, 4]
# implosion breakdown point:
aout.norm(data = iris.setosa, param = c(median(iris.setosa), mad(iris.setosa)),
          alpha = 0.01)
# better:
aout.norm(data = iris.setosa, param = c(median(iris.setosa), sd(iris.setosa)),
          alpha = 0.01)
```

aout.pareto

Find α -outliers in Pareto data

Description

Given the parameters of a Pareto distribution, `aout.pareto` identifies α -outliers in a given data set.

Usage

```
aout.pareto(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

`data` a vector. The data set to be examined.

`param` a vector. Contains the parameters of the Pareto distribution: λ, θ .

`alpha` an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.

`hide.outliers` boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Details

We use the Pareto distribution with Lebesgue-density $f(x) = \frac{\lambda\theta^\lambda}{x^{\lambda+1}}$.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Gather, U.; Kuhnt, S.; Pawlitschko, J. (2003) Concepts of outlyingness for various data structures. In J. C. Misra (Ed.): *Industrial Mathematics and Statistics*. New Delhi: Narosa Publishing House, 545-585.

See Also[citiesData](#)**Examples**

```
data(citiesData)
aout.pareto(citiesData[[1]], c(1.31, 14815), alpha = 0.01)
```

`aout.pois`*Find α -outliers in Poisson count data*

Description

Given the parameters of a Poisson distribution, `aout.pois` identifies α -outliers in a given data set.

Usage

```
aout.pois(data, param, alpha = 0.1, hide.outliers = FALSE)
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	a vector. Contains the parameter of the Poisson distribution: λ .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

See Also[dpois](#)**Examples**

```
aout.pois(data = c(discoveries), param = median(discoveries), alpha = 0.01)
```

aout.weibull

*Find α -outliers in Weibull data***Description**

Given the parameters of a Weibull distribution, `aout.weibull` identifies α -outliers in a given data set.

Usage

```
aout.weibull(data, param, alpha = 0.1, hide.outliers = FALSE, lower = auto.l,
             upper = auto.u, method.in = "Broyden", global.in = "qline",
             control.in = list(sigma = 0.1, maxit = 1000, xtol = 1e-12,
                               ftol = 1e-12, btol = 1e-04))
```

Arguments

<code>data</code>	a vector. The data set to be examined.
<code>param</code>	a vector. Contains the parameters of the Weibull distribution: β, λ .
<code>alpha</code>	an atomic vector. Determines the maximum amount of probability mass the outlier region may contain. Defaults to 0.1.
<code>hide.outliers</code>	boolean. Returns the outlier-free data if set to TRUE. Defaults to FALSE.
<code>lower</code>	an atomic vector. First element of <code>x</code> from <code>nleqslv</code> .
<code>upper</code>	an atomic vector. Second element of <code>x</code> from <code>nleqslv</code> .
<code>method.in</code>	See method in <code>nleqslv</code>
<code>global.in</code>	See global in <code>nleqslv</code>
<code>control.in</code>	See control in <code>nleqslv</code>

Details

The α -outlier region of a Weibull distribution is generally not available in closed form or via the tails, such that a non-linear equation system has to be solved.

Value

Data frame of the input data and an index named `is.outlier` that flags the outliers with TRUE. If `hide.outliers` is set to TRUE, a simple vector of the outlier-free data.

Author(s)

A. Rehage

References

Dodson, B. (2006) *The Weibull Analysis Handbook*. American Society for Quality, 2nd edition.

See Also[dweibull](#), [nleqslv](#)**Examples**

```
# lifetime data example taken from Table 2.2, Dodson (2006)
temp <- c(12.5, 24.4, 58.2, 68.0, 69.1, 95.5, 96.6, 97.0,
          114.2, 123.2, 125.6, 152.7)
aout.weibull(temp, c(2.25, 97), 0.1)
```

`citiesData`*Population of the 999 largest German cities*

Description

Population of the 999 largest German cities as a real life example for Pareto distributed data

Usage

```
data(citiesData)
```

Format

List with one element

References

<http://bevoelkerungsstatistik.de>

`createDesMat`*Create design matrix for log-linear models of contingency tables*

Description

This function creates a design matrix for contingency tables and is particularly useful for log-linear Poisson models. It uses effect coding of the variables: First the rows of the contingency table from top to bottom, then the columns from left to right.

Usage

```
createDesMat(n, p)
```

Arguments

`n` Number of rows of the corresponding contingency table.
`p` Number of columns of the corresponding contingency table.

Value

A $(n+p-1)$ times $(n*p)$ design matrix.

Author(s)

A. Rehage

References

Kuhnt, S.; Rapallo, F.; Rehage, A. (2014) Outlier detection in contingency tables based on minimal patterns. *Statistics and Computing* 24 (3), 481-491.

Examples

```
createDesMat(3, 5)
```

daysabs

Number of absence days of students

Description

Number of absence days of students

Usage

```
data(daysabs)
```

Format

Vector with 314 elements

References

<http://www.ats.ucla.edu/stat/r/dae/nbreg.htm>

Index

- * **datasets**
 - citiesData, 19
 - daysabs, 20
 - * **design**
 - createDesMat, 19
 - * **distribution**
 - alphaOutlier-package, 2
 - aout.binom, 3
 - aout.cg, 4
 - aout.chisq, 5
 - aout.conttab, 6
 - aout.exp, 7
 - aout.gandh, 8
 - aout.hyper, 9
 - aout.laplace, 11
 - aout.logis, 12
 - aout.mvnorm, 13
 - aout.nbinom, 14
 - aout.norm, 15
 - aout.pareto, 16
 - aout.pois, 17
 - aout.weibull, 18
 - * **nonparametric**
 - aout.kernel, 10
 - * **robust statistics**
 - alphaOutlier-package, 2
- alphaOutlier (alphaOutlier-package), 2
- alphaOutlier-package, 2
- aout.binom, 3
- aout.cg, 4
- aout.chisq, 5
- aout.conttab, 6
- aout.exp, 7
- aout.gandh, 8
- aout.hyper, 9
- aout.kernel, 10
- aout.laplace, 11
- aout.logis, 12
- aout.mvnorm, 13
- aout.nbinom, 14
- aout.norm, 15
- aout.pareto, 16
- aout.pois, 6, 7, 17
- aout.weibull, 18
- citiesData, 17, 19
- createDesMat, 19
- daysabs, 14, 20
- dbinom, 3
- dchisq, 6
- density, 10
- dexp, 8
- dlogis, 13
- dnbinom, 14
- dnorm, 14, 15
- dpois, 17
- dweibull, 19
- Hypergeometric, 9
- nleqslv, 2, 5, 18, 19
- rq.fit.fnc, 2, 6, 7
- solnp, 2, 8