Package 'cvmdisc'

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Type Package

Title Cramer von Mises Tests for Discrete or Grouped Distributions
Version 0.1.0
Depends R (>= 3.5), stats4, CompQuadForm
Implements Cramer-von Mises Statistics for testing fit to (1) fully specified discrete distributions as described in Choulakian, Lockhart and Stephens (1994) <doi:10.2307 3315828=""> (2) discrete distributions with unknown parameters that must be estimated from the sample data, see Spinelli & Stephens (1997) <doi:10.2307 3315735=""> and Lockhart, Spinelli and Stephens (2007) <doi:10.1002 cjs.5550350111=""> (3) grouped continuous distributions with Unknown Parameters, see Spinelli (2001) <doi:10.2307 3316040="">. Maximum likelihood estimation (MLE) is used to estimate the parameters. The package computes the Cramervon Mises Statistics, Anderson-Darling Statistics and the Watson-Stephens Statistics and their p-values.</doi:10.2307></doi:10.1002></doi:10.2307></doi:10.2307>
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ped Data

Description

cvmdisc is a package for fitting data to binomial, poisson, and grouped continuous distributions as well as testing their Goodness of Fit using Cramer-von-Mises and Anderson-Darling statistics

Details

The cymdisc package includes a function that fits and tests grouped/discrete data to multiple distributions, as well as multiple auxillary functions called by the main function.

Distributions

The Binomial, Poisson, Exponential, Gamma, Log-normal, Normal, Uniform and Weibull distributions are currently supported in the cymdisc package.

See Also

```
groupFit: Fit data and test GoF
distrFit: Fit data
cvmTest: Compute test statistics
cvmPval: Compute p values
```

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cvmPval	cvmPval

Description

Calculate P values of CVM statistics using their asymptotic distributions

Usage

```
cvmPval(statistic, Msig, imhof = TRUE)
```

Arguments

statistic	test statistic
Msig	Matrix used to produce eigenvalues to estimate the asymptotic distributions of the test statistic
imhof	Logical. Set to be FALSE if Imhof's method is NOT used to approximate the null. The package "CompQuadForm" is required if "imhof=TRUE"

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Details

cvmPval is used by groupFit to calculate test statistics for fit distributions.

Value

P-value for test statistic

Author(s)

Shaun Zheng Sun and Dillon Duncan

See Also

```
groupFit: Data fitting function
```

Examples

```
# A_squared and MsigA derived from
#(Choulakian, Lockhart and Stephens(1994) Example 3, p8)
A_squared <- 1.172932
MsigA \leftarrow matrix(c(0.05000, 0.03829, 0.02061, 0.00644,
                  0.03829, 0.30000, 0.16153, 0.05050,
                  0.02061, 0.16153, 0.30000, 0.09379,
                  0.00644, 0.05050, 0.09379, 0.30000),
                nrow = 4, ncol = 4, byrow = TRUE)
(U2Pval1 = cvmPval(A_squared, MsigA))
U_squared <- 0
MsigU <- matrix(c(0.16666667, 0.10540926, 0.0745356, 0.05270463, 0.03333333,
                  0.10540926, 0.16666667, 0.1178511, 0.08333333, 0.05270463,
                  0.07453560, 0.11785113, 0.1666667, 0.11785113, 0.07453560,
                  0.05270463, 0.08333333, 0.1178511, 0.16666667, 0.10540926,
                  0.033333333, 0.05270463, 0.0745356, 0.10540926, 0.16666667),
                nrow = 5, ncol = 5, byrow = TRUE)
(U2Pval2 = cvmPval(U_squared, MsigU, imhof = FALSE))
```

cvmTest

cvmTest

Description

Calculate test statistics for grouped data's counts and probabilities

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Usage

```
cvmTest(counts, p, pave = FALSE)
```

Arguments

counts vector containing the frequency of the counts in each group

p vector of probabilities for counts, often called p-hat

pave Logical. Set to be FALSE if the probabilities in groups are used; set to TRUE if

the average of probabilities of groups j and j+1 are used

Details

cvmTest is used by groupFit to calculate test statistics for the fitted distributions.

Value

A list with the components:

Asq Anderson-Darling test statistic
Wsq Cramer-Von-Mises test statistic

Usq Watson's test statistic

Chisq Pearson's Chi-squared test statistic

Hj Estimated cumulative probability

Author(s)

Shaun Zheng Sun and Dillon Duncan

See Also

```
groupFit: Data fitting function
```

Examples

```
#Choulakian, Lockhart and Stephens (1994)
counts <- c(10, 19, 18, 15, 11, 13, 7, 10, 13, 23, 15, 22)
phat <- rep(1/12, 12)

(stats1 <- cvmTest(counts, phat))

#Choulakian, Lockhart and Stephens (1994)
counts <- c(1, 4, 11, 4, 0)
phat <- c(0.05, 0.3, 0.3, 0.3, 0.05)

(stat2 <- cvmTest(counts, phat))

#Utilizing Benford's Law

#Setting pave to TRUE</pre>
```

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```
#genomic data, Lesperance et al (2016)
genomic<- c(48, 14, 12, 6, 18, 5, 7, 8, 9)
phat<- log10(1+1/1:9)
(stat3 <- cvmTest(genomic, phat, pave = TRUE))</pre>
```

distrFit

distrFit

Description

Finds the Maximum Likelihood Estimates of the parameters in a requested distribution.

Usage

```
distrFit(breaks, counts, distr, initials)
```

Arguments

breaks	Vector defining the breaks in each group
counts	Vector containing the frequency of counts in each group
distr	Character; the name of the distribution users want to fit the data to
	distrFit supports all of the continuous distributions supported in groupFit:
initials	Vector of initial values for the maximum likelihood estimates.

Details

distFit uses Maximum Likelihood Estimates to optimize the parameters for a requested distribution.

Value

distFit returns a vector containing the MLEs.

Author(s)

Shaun Zheng Sun and Dillon Duncan

See Also

groupFit for fitting data and providing GoF statistics.

Examples

```
#fitting exponential data without initial values (Spinelli 2001)
breaks <- c(0, 2, 6, 10, 14, 18, 22, 26)
counts <- c(21, 9, 5, 2, 1, 1, 0)

(mle1 <- distrFit(breaks, counts, distr = "exp"))
#fitting generated data with initial values
breaks <- seq(0, 40, 2)
counts <- table(cut(rweibull(200, 0.5, 3), breaks))

(mle2 <- distrFit(breaks, counts, distr = "weibull", initials = c(0.5, 3)))
#fitting generated data to a different distribution
breaks <- seq(-100, 100, 5)
counts <- table(cut(rcauchy(500, -20, 10), breaks))

(mle3 <- distrFit(breaks, counts, distr = "norm"))</pre>
```

groupFit

groupFit

Description

Fits grouped continuous data or discrete data to a distribution and computes Cramer-Von Mises Goodness of Fit statistics and p-values.

Usage

```
groupFit(breaks, counts, data, discrete, distr, N, params, initials,
   pfixed, bootstrap = FALSE, numLoops = 5000, known = FALSE,
   pave = FALSE, imhof = TRUE)
```

Arguments

breaks	Vector defining the breaks in each group.
counts	Vector containing the frequency of counts in each group.
data	Vector containing values of discrete random variables.
discrete	Logical. Is the distribution discrete?
distr	Character; the name of the distribution users want to fit the data to.
	Included continuous distributions are: "exp", "gamma", "lnorm", "norm", "unif" and "weibull".
	Included discrete distributions are: "binom" and "pois".

User defined distributions are supported as "user".

Short-hand or full spelling of these distributions will be recognized, case insen-

sitive.

N Number of trials, used only for the binomial distribution.

params Vector of distribution parameters. This is only required when known == TRUE.

groupFit will estimate the parameters if known == FALSE.

initials Vector of distribution parameters to use as starting points for calculating MLEs.

Vector of known probabilities for corresponding counts vector. pfixed must be

provided when distr = "user".

bootstrap Logical. Should p-values be calculated via bootstrapping? numLoops Number of Bootstrap iterations. Set to be 5000 by default.

known Logical. Set to be TRUE if the parameters are known and do not need to be

estimated.

pave Logical. Set to be FALSE if the probabilities in groups are used; set to TRUE if

the average of probabilities of groups j and j + 1 are used. See page 2 of Spinelli

(2001) for more details.

imhof Logical. Set to be TRUE if Imhof's method from the package "CompQuad-

Form" is used to approximate the null distributions.

Details

For grouped continuous data: call groupFit with arguments breaks, counts, and distr to fit the data.

For discrete data call groupFit with data and distr to fit the data. If distr = "binom", then be sure to call groupFit with N as well.

Provide initials to suggest starting points for parameter estimation.

Provide params and set known = TRUE to test goodness of fit when parameters are known.

Set bootstrap = TRUE to use bootstrapping to estimate p-values rather than using asymptotic distributions.

Set imhof = FALSE when $a + bX_p^2$ is used to approximate the null distributions. See page 5 of Spinelli (2001) for details.

groupFit can test the fit of user defined discrete distributions. To do so, set distr to "user", and provide the vector pfixed, where each cell contains the probability corresponding to that same cell in counts.

Value

List containing the components:

estimates The estimated parameters of the distribution

stats Data frame containing the goodness of fit statistics

pvals Data frame containing p-values for the goodness of fit statistics

Author(s)

Shaun Zheng Sun and Dillon Duncan

References

V. Choulakian, R. A. Lockhart & M. A. Stephens (1994). Cramer-vonMises statistics for discrete distributions. The Canadian Journal of Statistics, 2 2,125-137.

J.J.Spinelli (2001). Testing fit for the grouped exponential distribution. The Canadian Journal of Statistics,29,451-458.

- J.J. Spinelli (1994). Cramer-vonMises statistics for discrete distributions. Phd thesis. Simon Fraser University, Burnaby, Canada.
- S. Z. Sun, J.J. Spinelli & M. A. Stephens (2012). Testing fit for the grouped gamma and weibull distributions. Technical report. Simon Fraser University, Burnaby, Canada.

See Also

distrFit: Parameter estimation function

Examples

```
#Poisson Example (Spinelli 1994) p36
counts < c(9, 22, 6, 2, 1, 0, 0, 0, 0)
vals <- 0:8
data <- rep(vals, counts)</pre>
groupFit(data = data, distr = "pois")
# When the parameters are unknown
#(Spinelli 1994) p56
counts <- c(57, 203, 383, 525, 532, 408, 273, 139, 45, 27, 10, 4, 0, 1, 1)
vals <- 0:14
data <- rep(vals, counts)</pre>
(pois_fit <- groupFit(data = data, distr = "pois"))</pre>
#Binomial example when the parameter is unknown
#Spinelli (1994) P92.
N = 12
counts= c(185, 1149, 3265, 5475, 6114, 5194,
          3067, 1331, 403, 105, 14, 4, 0)
vals <- 0:12
data <- rep(vals, counts)</pre>
(binom_fit <- groupFit(data = data, N = N, distr = "binom"))</pre>
#When the parameter is assumed known and is equal to 1/3
(binom_fit <- groupFit(data = data, N = N, distr = "binom", params = 1/3, known = TRUE))
#uniform example (Choulakian, Lockhart and Stephens(1994) Example 2, p8)
```

```
counts <- c(10, 19, 18, 15, 11, 13, 7, 10, 13, 23, 15, 22)
(uni_fit <- groupFit(0:12, counts, distr = "unif"))</pre>
#uniform example (Choulakian, Lockhart and Stephens(1994) Example 3, p8)
counts <- c(1, 4, 11, 4, 0)
probability \leftarrow c(0.05, 0.3, 0.3, 0.3, 0.05)
breaks <- c(0, cumsum(probability))</pre>
#with bootstrapping
(uni_fit1 <- groupFit(breaks, counts, distr = "unif", bootstrap = TRUE, numLoops = 500))</pre>
#without bootstrapping
(uni_fit2 <- groupFit(breaks, counts, distr = "unif", bootstrap = FALSE))</pre>
#exponential example (Spinelli 2001)
breaks <- c(0, 2, 6, 10, 14, 18, 22, 26)
counts \leftarrow c(21, 9, 5, 2, 1, 1, 0)
(exp_fit <- groupFit(breaks, counts, distr = "exp", pave = TRUE))</pre>
#Example Sun, Stephens & Spinelli (2012) set 2.
breaks <- c(0, 2, 6, 10, 14, 18, 22, 26)
counts <- c(21, 9, 5, 2, 1, 1, 0)
breaks[1] <- 1e-6
breaks[8] <- Inf</pre>
(weibull_fit <- groupFit(breaks, counts, distr = "Weibull"))</pre>
#Example Sun, Stephens & Spinelli (2012) set 3.
breaks <- c(0, seq(0.5, 6.5, 1))
counts <- c(32, 12, 3, 6, 0, 0, 1)
breaks[1] <- 1e-6
breaks[8] <- Inf</pre>
(weibull_fit <- groupFit(breaks, counts, distr = "Weibull"))</pre>
#Example Sun, Stephens & Spinelli (2012) set 3.
breaks <- c(0, 2, 6, 10, 14, 18, 22, 26)
counts \leftarrow c(21, 9, 5, 2, 1, 1, 0)
breaks[1] <- 1e-6
breaks[8] <- Inf</pre>
(gamma_fit <- groupFit(breaks, counts, distr = "exp"))</pre>
#Example Sun, Stephens & Spinelli (2012) set 3.
```

```
breaks <- c(0, seq(0.5, 6.5, 1))
counts <- c(32, 12, 3, 6, 0, 0, 1)
breaks[1] <- 1e-6
breaks[8] <- Inf</pre>
(gamma_fit <- groupFit(breaks, counts, distr = "gamma"))</pre>
#More examples
breaks <- c(0, seq(0.5, 6.5, 1))
counts <- table(cut(rgamma(100, 3, 1/3), breaks))</pre>
breaks[8] <- Inf
#setting pave to true
(exp_fit <- groupFit(breaks, counts, distr = "exp", initials = 0.2, pave = TRUE))</pre>
#Setting known to true, with params
(gamma_fit <- groupFit(breaks, counts, distr = "gamma",</pre>
                       params = c(3, 1/3), known = TRUE))
#with bootstrapping, specifying the number of loops.
(lnorm_fit <- groupFit(breaks, counts, distr = "lnorm",</pre>
                       bootstrap = TRUE, numLoops = 1000))
#fitting with both pave and imhof set to false
#by setting imhof to false, we use a+bX^2_p to approximate
#the distribution of the goodness-of-fit Statistics
(weibull_fit <- groupFit(breaks, counts, distr = "weibull",</pre>
                         pave = TRUE, imhof = FALSE))
#Using the user defined distribution to test for Benford's law
#genomic data, Lesperance et al (2016)
genomic <- c(48, 14, 12, 6, 18, 5, 7, 8, 9)
phat <- log10(1+1/1:9)</pre>
(fit <- groupFit(counts = genomic, distr = "user", pfixed = phat, imhof = FALSE, pave = TRUE))</pre>
```

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