Package ‘KSgeneral’

March 20, 2022

Type Package
Version 1.1.1
Title Computing P-Values of the K-S Test for (Dis)Continuous Null Distribution
Author Dimitrina S. Dimitrova <D.Dimitrova@city.ac.uk>, Vladimir K. Kaishev <Vladimir.Kaishev.1@city.ac.uk> and Senren Tan <raymondtsrtsr@outlook.com>
Maintainer Senren Tan <raymondtsrtsr@outlook.com>
Depends R (>= 3.3.0)
SystemRequirements fftw3 (>=3.3.4), C++11
Copyright Copyright holders of FFTW3: Copyright (c) 2003, 2007-11 Matteo Frigo; Copyright (c) 2003, 2007-11 Massachusetts Institute of Technology
Description Computes a p-value of the one-sample two-sided (or one-sided, as a special case) Kolmogorov-Smirnov (KS) statistic, for any fixed critical level, and an arbitrary, possibly large sample size for a pre-specified purely discrete, mixed or continuous cumulative distribution function (cdf) under the null hypothesis. If a data sample is supplied, 'KSgeneral' computes the p-value corresponding to the value of the KS test statistic computed based on the user provided data sample. The package 'KSgeneral' implements a novel, accurate and efficient method named Exact-KS-FFT, expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using Fast Fourier Transform (FFT). The package can also be used to compute and plot the complementary cdf of the KS statistic which is known to depend on the hypothesized distribution when the latter is discontinuous (i.e. purely discrete or mixed). To cite this package in publication use: Dimitrina S. Dimitrova, Vladimir K. Kaishev, and Senren Tan. Computing the Kolmogorov-Smirnov Distribution When the Underlying CDF is Purely Discrete, Mixed, or Continuous. Journal of Statistical Software. 2020; 95(10): 1–42. <doi:10.18637/jss.v095.i10>.
License GPL (>= 2.0)
URL https://github.com/raymondtsrtsr/KSgeneral
Encoding UTF-8
LazyData true
Imports Rcpp (>= 0.12.12), MASS, dgof
KSgeneral-package

Computing P-Values of the K-S Test for (Dis)Continuous Null Distribution

Description

The one-sample two-sided Kolmogorov-Smirnov (KS) statistic is one of the most popular goodness-of-fit test statistics that is used to measure how well the distribution of a random sample agrees with a prespecified theoretical distribution. Given a random sample \{X_1, \ldots, X_n\} of size \(n\) with an empirical cdf \(F_n(x)\), the two-sided KS statistic is defined as \(D_n = \sup |F_n(x) - F(x)|\), where \(F(x)\) is the cdf of the prespecified theoretical distribution under the null hypothesis \(H_0\), that \{X_1, \ldots, X_n\} comes from \(F(x)\).

The package KSgeneral implements a novel, accurate and efficient Fast Fourier Transform (FFT)-based method, referred as Exact-KS-FFT method to compute the complementary cdf, \(P(D_n \geq q)\), at a fixed \(q \in [0, 1]\) for a given (hypothesized) purely discrete, mixed or continuous underlying cdf \(F(x)\), and arbitrary, possibly large sample size \(n\). A plot of the complementary cdf \(P(D_n \geq q)\), 0 \(\leq q \leq 1\), can also be produced.

In other words, the package computes the p-value, \(P(D_n \geq q)\) for any fixed critical level \(q \in [0, 1]\). If a data sample, \{\(x_1, \ldots, x_n\}\) is supplied, KSgeneral computes the p-value \(P(D_n \geq d_n)\), where \(d_n\) is the value of the KS test statistic computed based on \{\(x_1, \ldots, x_n\}\).

Remark: The description of the package and its functions are primarily tailored to computing the (complementary) cdf of the two-sided KS statistic, \(D_n\). It should be noted however that one can compute the (complementary) cdf for the one-sided KS statistics \(D_{n-}^+\) or \(D_{n+}^+\) (cf., Dimitrova, Kai-shev, Tan (2020)) by appropriately specifying correspondingly \(A_i = 0\) for all \(i\) or \(B_i = 1\) for all \(i\), in the function ks_c_cdf_Rcpp.
Details

The Exact-KS-FFT method underlying \texttt{KSgeneral} is based on expressing the p-value $P(D_n \geq q)$ in terms of an appropriate rectangle probability with respect to the uniform order statistics, as noted by Gleser (1985) for $P(D_n > q)$. The latter representation is used to express $P(D_n \geq q)$ via a double-boundary non-crossing probability for a homogeneous Poisson process, with intensity $n$, which is then efficiently computed using FFT, ensuring total run-time of order $O(n^2 \log(n))$ (see Dimitrova, Kaishev, Tan (2020) and also Moscovich and Nadler (2017) for the special case when $F(x)$ is continuous).

\texttt{KSgeneral} represents an R wrapper of the original C++ code due to Dimitrova, Kaishev, Tan (2020) and based on the C++ code developed by Moscovich and Nadler (2017). The package includes the functions \texttt{disc\_ks\_c\_cdf}, \texttt{mixed\_ks\_c\_cdf} and \texttt{cont\_ks\_c\_cdf} that compute the complementary cdf $P(D_n \geq q)$, for a fixed $q$, $0 \leq q \leq 1$, when $F(x)$ is purely discrete, mixed or continuous, respectively. \texttt{KSgeneral} includes also the functions \texttt{disc\_ks\_test}, \texttt{mixed\_ks\_test} and \texttt{cont\_ks\_test} that compute the p-value $P(D_n \geq d_n)$, where $d_n$ is the value of the KS test statistic computed based on a user provided data sample $\{x_1, ..., x_n\}$, when $F(x)$ is purely discrete, mixed or continuous, respectively.

The functions \texttt{disc\_ks\_test} and \texttt{cont\_ks\_test} represent accurate and fast (run time $O(n^2 \log(n))$) alternatives to the functions \texttt{ks.test} from the package \texttt{dgof} and the function \texttt{ks.test} from the package \texttt{stat}, which compute p-values of $P(D_n \geq d_n)$, assuming $F(x)$ is purely discrete or continuous, respectively.

The package also includes the function \texttt{ks\_c\_cdf\_Rcpp} which gives the flexibility to compute the complementary cdf (p-value) for the one-sided KS test statistics $D_n^-$ or $D_n^+$. It also allows for faster computation time and possibly higher accuracy in computing $P(D_n \geq q)$.

Author(s)

Dimitrina S. Dimitrova <D.Dimitrova@city.ac.uk>, Vladimir K. Kaishev <Vladimir.Kaishev.1@city.ac.uk> and Senren Tan <raymondtsrtsr@outlook.com>

Maintainer: Senren Tan <raymondtsrtsr@outlook.com>

References


<table>
<thead>
<tr>
<th>Function</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>cont_ks_cdf</td>
<td>Computes the cumulative distribution function of the two-sided Kolmogorov-Smirnov statistic when the cdf under the null hypothesis is continuous.</td>
</tr>
</tbody>
</table>

**Description**

Computes the cdf $P(D_n \leq q) \equiv P(D_n < q)$ at a fixed $q$, $q \in [0, 1]$, for the one-sample two-sided Kolmogorov-Smirnov statistic, $D_n$, for a given sample size $n$, when the cdf $F(x)$ under the null hypothesis is continuous.

**Usage**

```r
cont_ks_cdf(q, n)
```

**Arguments**

- `q` numeric value between 0 and 1, at which the cdf $P(D_n \leq q)$ is computed
- `n` the sample size

**Details**

Given a random sample $\{X_1, ..., X_n\}$ of size $n$ with an empirical cdf $F_n(x)$, the Kolmogorov-Smirnov goodness-of-fit statistic is defined as $D_n = \sup |F_n(x) - F(x)|$, where $F(x)$ is the cdf of a prespecified theoretical distribution under the null hypothesis $H_0$, that $\{X_1, ..., X_n\}$ comes from $F(x)$.

The function `cont_ks_cdf` implements the FFT-based algorithm proposed by Moscovich and Nadler (2017) to compute the cdf $P(D_n \leq q)$ at a value $q$, when $F(x)$ is continuous. This algorithm ensures a total worst-case run-time of order $O(n^2 \log(n))$ which makes it more efficient and numerically stable than the algorithm proposed by Marsaglia et al. (2003). The latter is used by many existing packages computing the cdf of $D_n$, e.g., the function `ks.test` in the package `stats` and the function `ks.test` in the package `dgof`. More precisely, in these packages, the exact p-value, $P(D_n \geq q)$ is computed only in the case when $q = d_n$, where $d_n$ is the value of the KS statistic computed based on a user provided sample $\{x_1, ..., x_n\}$. Another limitation of the functions `ks.test` is that the sample size should be less than 100, and the computation time is $O(n^3)$. In contrast, the function `cont_ks_cdf` provides results with at least 10 correct digits after the decimal point for sample sizes $n$ up to 100000 and computation time of 16 seconds on a machine with an 2.5GHz Intel Core i5 processor with 4GB RAM, running MacOS X Yosemite. For $n > 100000$, accurate results can still be computed with similar accuracy, but at a higher computation time. See Dimitrova, Kaishev, Tan (2020), Appendix B for further details and examples.

**Value**

Numeric value corresponding to $P(D_n \leq q)$. 


**Source**

Based on the C++ code available at https://github.com/mosco/crossing-probability developed by Mosovich and Nadler (2017). See also Dimitrova, Kaishev, Tan (2020) for more details.

**References**


**Examples**

```r
## Compute the value for P(D_{100} \leq 0.05)
KSgeneral::cont_ks_cdf(0.05, 100)

## Compute P(D_{n} \leq q)
## for n = 100, q = 1/500, 2/500, ..., 500/500
## and then plot the corresponding values against q
n<-100
q<-1:500/500
plot(q, sapply(q, function(x) KSgeneral::cont_ks_cdf(x, n)), type='l')

## Compute P(D_{n} \leq q) for n = 40, q^{2}(2) = 0.76 as shown
## in Table 9 of Dimitrova, Kaishev, Tan (2020)
KSgeneral::cont_ks_cdf(sqrt(0.76/40), 40)
```

**Description**

Computes the complementary cumulative distribution function of the two-sided Kolmogorov-Smirnov statistic when the cdf under the null hypothesis is continuous.

**Usage**

```r
cont_ks_c_cdf(q, n)
```
Arguments

q numeric value between 0 and 1, at which the complementary cdf \( P(D_n \geq q) \) is computed

n the sample size

Details

Given a random sample \( \{X_1, ..., X_n\} \) of size \( n \) with an empirical cdf \( F_n(x) \), the two-sided Kolmogorov-Smirnov goodness-of-fit statistic is defined as \( D_n = \sup |F_n(x) - F(x)| \), where \( F(x) \) is the cdf of a prespecified theoretical distribution under the null hypothesis \( H_0 \), that \( \{X_1, ..., X_n\} \) comes from \( F(x) \).

The function \texttt{cont_ks_c_cdf} implements the FFT-based algorithm proposed by Moscovich and Nadler (2017) to compute the complementary cdf, \( P(D_n \geq q) \) at a value \( q \), when \( F(x) \) is continuous. This algorithm ensures a total worst-case run-time of order \( O(n^2 \log(n)) \) which makes it more efficient and numerically stable than the algorithm proposed by Marsaglia et al. (2003). The latter is used by many existing packages computing the cdf of \( D_n \), e.g., the function \texttt{ks.test} in the package \texttt{stats} and the function \texttt{ks.test} in the package \texttt{dgof}. More precisely, in these packages, the exact p-value, \( P(D_n \geq q) \) is computed only in the case when \( q = d_n \), where \( d_n \) is the value of the KS test statistic computed based on a user provided sample \( \{x_1, ..., x_n\} \). Another limitation of the functions \texttt{ks.test} is that the sample size should be less than 100, and the computation time is \( O(n^3) \). In contrast, the function \texttt{cont_ks_c_cdf} provides results with at least 10 correct digits after the decimal point for sample sizes \( n \) up to 100000 and computation time of 16 seconds on a machine with an 2.5GHz Intel Core i5 processor with 4GB RAM, running MacOS X Yosemite. For \( n > 100000 \), accurate results can still be computed with similar accuracy, but at a higher computation time. See Dimitrova, Kaishev, Tan (2020), Appendix C for further details and examples.

Value

Numeric value corresponding to \( P(D_n \geq q) \).

Source

Based on the C++ code available at \url{https://github.com/mosco/crossing-probability} developed by Moscovich and Nadler (2017). See also Dimitrova, Kaishev, Tan (2020) for more details.

References


cont_ks_test

Examples

```r
## Compute the value for P(D_{100} \geq 0.05)
KSgeneral::cont_ks_c_cdf(0.05, 100)

## Compute P(D_{n} \geq q)
## for n = 100, q = 1/500, 2/500, \ldots, 500/500
## and then plot the corresponding values against q
n <- 100
q <- 1:500/500
plot(q, sapply(q, function(x) KSgeneral::cont_ks_c_cdf(x, n)), type='l')

## Compute P(D_{n} \geq q) for n = 141, nq^{2} = 2.1 as shown
## in Table 18 of Dimitrova, Kaishev, Tan (2020)
KSgeneral::cont_ks_c_cdf(sqrt(2.1/141), 141)
```

---

cont_ks_test

*Computes the p-value for a one-sample two-sided Kolmogorov-Smirnov test when the cdf under the null hypothesis is continuous*

---

Description

Computes the p-value $P(D_n \geq d_n) \equiv P(D_n > d_n)$, where $d_n$ is the value of the KS test statistic computed based on a data sample $\{x_1, \ldots, x_n\}$, when $F(x)$ is continuous.

Usage

```r
cont_ks_test(x, y, ...)
```

Arguments

- `x` a numeric vector of data sample values $\{x_1, \ldots, x_n\}$.
- `y` a pre-specified continuous cdf, $F(x)$ under the null hypothesis. Note that `y` should be a character string naming a continuous cumulative distribution function such as `pexp`, `pnorm`, etc. Only continuous cdfs are valid!
- `...` values of the parameters of the cdf, $F(x)$ specified (as a character string) by `y`.

Details

Given a random sample $\{X_1, \ldots, X_n\}$ of size $n$ with an empirical cdf $F_n(x)$, the two-sided Kolmogorov-Smirnov goodness-of-fit statistic is defined as $D_n = \sup |F_n(x) - F(x)|$, where $F(x)$ is the cdf of a prespecified theoretical distribution under the null hypothesis $H_0$, that $\{X_1, \ldots, X_n\}$ comes from $F(x)$. 
The function `cont_ks_test` implements the FFT-based algorithm proposed by Moscovich and Nadler (2017) to compute the p-value \( P(D_n \geq d_n) \), where \( d_n \) is the value of the KS test statistic computed based on a user provided data sample \( \{x_1, \ldots, x_n\} \), assuming \( F(x) \) is continuous. This algorithm ensures a total worst-case run-time of order \( O(n^2 \log(n)) \) which makes it more efficient and numerically stable than the algorithm proposed by Marsaglia et al. (2003). The latter is used by many existing packages computing the cdf of \( D_n \), e.g., the function `ks.test` in the package `stats` and the function `ks.test` in the package `dgof`. A limitation of the functions `ks.test` is that the sample size should be less than 100, and the computation time is \( O(n^3) \). In contrast, the function `cont_ks_test` provides results with at least 10 correct digits after the decimal point for sample sizes \( n \) up to 100000 and computation time of 16 seconds on a machine with an 2.5GHz Intel Core i5 processor with 4GB RAM, running MacOS X Yosemite. For \( n > 100000 \), accurate results can still be computed with similar accuracy, but at a higher computation time. See Dimitrova, Kaishev, Tan (2020), Appendix C for further details and examples.

**Value**

A list with class "htest" containing the following components:

- **statistic**: the value of the statistic.
- **p.value**: the p-value of the test.
- **alternative**: "two-sided".
- **data.name**: a character string giving the name of the data.

**Source**

Based on the C++ code available at [https://github.com/mosco/crossing-probability](https://github.com/mosco/crossing-probability) developed by Moscovich and Nadler (2017). See also Dimitrova, Kaishev, Tan (2020) for more details.

**References**


**Examples**

```r
## Comparing the p-values obtained by stat::ks.test
## and KSgeneral::cont_ks_test

x <- abs(rnorm(100))
p.kt <- ks.test(x, "pexp", exact = TRUE)$p
p.kt_fft <- KSgeneral::cont_ks_test(x, "pexp")$p
abs(p.kt - p.kt_fft)
```
**disc_ks_c_cdf**

*Computes the complementary cumulative distribution function of the two-sided Komogorov-Smirnov statistic when the cdf under the null hypothesis is purely discrete*

---

**Description**

Computes the complementary cdf, \( P(D_n \geq q) \) at a fixed \( q, q \in [0, 1] \), of the one-sample two-sided Kolmogorov-Smirnov (KS) statistic, when the cdf \( F(x) \) under the null hypothesis is purely discrete, using the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)). Moreover, for comparison purposes, \( disc_ks_c_cdf \) gives, as an option, the possibility to compute (an approximate value for) the asymptotic \( P(D_n \geq q) \) using the simulation-based algorithm of Wood and Altavela (1978).

**Usage**

\[
disc_ks_c_cdf(q, n, y, ..., exact = \text{NULL}, tol = 1e^{-08}, \text{sim.size} = 1e+06, \text{num.sim} = 10)
\]

**Arguments**

- **q**: numeric value between 0 and 1, at which the complementary cdf \( P(D_n \geq q) \) is computed
- **n**: the sample size
- **y**: a pre-specified discrete cdf, \( F(x) \) under the null hypothesis. Note that \( y \) should be a step function within the class: `stepfun`, of which `ecdf` is a subclass!
- **...**: values of the parameters of the cdf, \( F(x) \), specified (as a character string) by \( y \).
- **exact**: logical variable specifying whether one wants to compute exact p-value \( P(D_n \geq q) \) using the Exact-KS-FFT method, in which case \( exact = \text{TRUE} \) or wants to compute an approximate p-value \( P(D_n \geq q) \) using the simulation-based algorithm of Wood and Altavela (1978), in which case \( exact = \text{FALSE} \). When \( exact = \text{NULL} \) and \( n \leq 100000 \), the exact \( P(D_n \geq q) \) will be computed using the Exact-KS-FFT method. Otherwise, the asymptotic complementary cdf is computed based on Wood and Altavela (1978). By default, \( exact = \text{NULL} \).
- **tol**: the value of \( \epsilon \) that is used to compute the values of \( A_i \) and \( B_i, i = 1, ..., n \), as detailed in Step 1 of Section 2.1 in Dimitrova, Kaishev and Tan (2020) (see also (ii) in the Procedure Exact-KS-FFT therein). By default, \( tol = 1e^{-08} \). Note that a value of NA or 0 will lead to an error!
- **sim.size**: the required number of simulated trajectories in order to produce one Monte Carlo estimate (one MC run) of the asymptotic complementary cdf using the algorithm of Wood and Altavela (1978). By default, \( sim.size = 1e+06 \).
- **num.sim**: the number of MC runs, each producing one estimate (based on \( sim.size \) number of trajectories), which are then averaged in order to produce the final estimate for the asymptotic complementary cdf. This is done in order to reduce the variance of the final estimate. By default, \( num.sim = 10 \).
Details

Given a random sample \( \{X_1, \ldots, X_n\} \) of size \( n \) with an empirical cdf \( F_n(x) \), the two-sided Kolmogorov-Smirnov goodness-of-fit statistic is defined as \( D_n = \sup |F_n(x) - F(x)| \), where \( F(x) \) is the cdf of a prespecified theoretical distribution under the null hypothesis \( H_0 \), that \( \{X_1, \ldots, X_n\} \) comes from \( F(x) \).

The function \texttt{disc ks c cdf} implements the Exact-KS-FFT method, proposed by Dimitrova, Kaishev, Tan (2020) to compute the complementary cdf \( P(D_n \geq q) \) at a value \( q \), when \( F(x) \) is purely discrete. This algorithm ensures a total worst-case run-time of order \( O(n^2 \log(n)) \) which makes it more efficient and numerically stable than the only alternative algorithm developed by Arnold and Emerson (2011) and implemented as the function \texttt{ks.test} in the package \texttt{dgof}. The latter only computes a p-value \( P(D_n \geq d_n) \), corresponding to the value of the KS test statistic \( d_n \) computed based on a user provided sample \( \{x_1, \ldots, x_n\} \). More precisely, in the package \texttt{dgof} (function \texttt{ks.test}), the p-value for a one-sample two-sided KS test is calculated by combining the approaches of Gleser (1985) and Niederhausen (1981). However, the function \texttt{ks.test} only provides exact p-values for \( n \leq 30 \), since as noted by the authors (see Arnold and Emerson (2011)), when \( n \) is large, numerical instabilities may occur. In the latter case, \texttt{ks.test} uses simulation to approximate p-values, which may be rather slow and inaccurate (see Table 6 of Dimitrova, Kaishev, Tan (2020)).

Thus, making use of the Exact-KS-FFT method, the function \texttt{disc ks c cdf} provides an exact and highly computationally efficient (alternative) way of computing \( P(D_n \geq q) \) at a value \( q \), when \( F(x) \) is purely discrete.

Lastly, incorporated into the function \texttt{disc ks c cdf} is the MC simulation-based method of Wood and Altavela (1978) for estimating the asymptotic complementary cdf of \( D_n \). The latter method is the default method behind \texttt{disc ks c cdf} when the sample size \( n \) is \( n \geq 100000 \).

Value

Numeric value corresponding to \( P(D_n \geq q) \).

References


See Also

\texttt{ks.test}
Examples

## Example to compute the exact complementary cdf for $D_n$
## when the underlying cdf $F(x)$ is a binomial(3, 0.5) distribution,
## as shown in Example 3.4 of Dimitrova, Kaishev, Tan (2020)

```r
binom_3 <- stepfun(c(0:3), c(0, pbinom(0:3, 3, 0.5)))
KSgeneral::disc_ks_c_cdf(0.05, 400, binom_3)
```

## Not run:
## Compute $P(D_n \geq q)$ for $n = 100$,
## $q = 1/5000$, $2/5000$, ..., $5000/5000$, when
## the underlying cdf $F(x)$ is a binomial(3, 0.5) distribution,
## as shown in Example 3.4 of Dimitrova, Kaishev, Tan (2020),
## and then plot the corresponding values against $q$,
## i.e. plot the resulting complementary cdf of $D_n$

```r
n <- 100
q <- 1:5000
binom_3 <- stepfun(c(0:3), c(0, pbinom(0:3, 3, 0.5)))
plot(q, sapply(q, function(x) KSgeneral::disc_ks_c_cdf(x, n, binom_3)), type='l')
```

## End(Not run)

## Not run:
## Example to compute the asymptotic complementary cdf for $D_n$
## based on Wood and Altavela (1978),
## when the underlying cdf $F(x)$ is a binomial(3, 0.5) distribution,
## as shown in Example 3.4 of Dimitrova, Kaishev, Tan (2020)

```r
binom_3 <- stepfun(c(0:3), c(0, pbinom(0:3, 3, 0.5)))
KSgeneral::disc_ks_c_cdf(0.05, 400, binom_3, exact = FALSE, tol = 1e-08,
sim.size = 1e+06, num.sim = 10)
```

## End(Not run)

---

**disc_ks_test**

*Computes the p-value for a one-sample two-sided Kolmogorov-Smirnov test when the cdf under the null hypothesis is purely discrete*

### Description

Computes the p-value $P(D_n \geq d_n)$, where $d_n$ is the value of the KS test statistic computed based on a data sample $\{x_1, ..., x_n\}$, when $F(x)$ is purely discrete, using the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)).

### Usage

```r
disc_ks_test(x, y, ..., exact = NULL, tol = 1e-08, sim.size = 1e+06, num.sim = 10)
```
Arguments

- **x**: a numeric vector of data sample values \( \{x_1, ..., x_n\} \).
- **y**: a pre-specified discrete cdf, \( F(x) \), under the null hypothesis. Note that \( y \) should be a step function within the class: \texttt{stepfun}, of which \texttt{ecdf} is a subclass!
- **exact**: values of the parameters of the cdf, \( F(x) \), specified (as a character string) by \( y \).
- **tol**: the value of \( \epsilon \) that is used to compute the values of \( A_i \) and \( B_i \), \( i = 1, ..., n \), as detailed in Step 1 of Section 2.1 in Dimitrova, Kaishev and Tan (2020) (see also (ii) in the Procedure Exact-KS-FFT therein). By default, \( \text{tol} = 1e^{-08} \). Note that a value of \( \text{NA} \) or \( 0 \) will lead to an error!
- **sim.size**: the required number of simulated trajectories in order to produce one Monte Carlo estimate (one MC run) of the asymptotic p-value using the algorithm of Wood and Altavela (1978). By default, \( \text{sim.size} = 1e+06 \).
- **num.sim**: the number of MC runs, each producing one estimate (based on \( \text{sim.size} \) number of trajectories), which are then averaged in order to produce the final estimate for the asymptotic p-value. This is done in order to reduce the variance of the final estimate. By default, \( \text{num.sim} = 10 \).

Details

Given a random sample \( \{X_1, ..., X_n\} \) of size \( n \) with an empirical cdf \( F_n(x) \), the two-sided Kolmogorov-Smirnov goodness-of-fit statistic is defined as \( D_n = \sup |F_n(x) - F(x)| \), where \( F(x) \) is the cdf of a pre-specified theoretical distribution under the null hypothesis \( H_0 \), that \( \{X_1, ..., X_n\} \) comes from \( F(x) \).

The function \texttt{disc_ks_test} implements the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)). It represents an accurate and fast (run time \( O(n^2 \log(n)) \)) alternative to the function \texttt{ks.test} from the package \texttt{dgof}, which computes a p-value \( P(D_n \geq d_n) \), where \( d_n \) is the value of the KS test statistic computed based on a user provided data sample \( \{x_1, ..., x_n\} \), assuming \( F(x) \) is purely discrete.

In the function \texttt{ks.test}, the p-value for a one-sample two-sided KS test is calculated by combining the approaches of Gleser (1985) and Niederhausen (1981). However, the function \texttt{ks.test} due to Arnold and Emerson (2011) only provides exact p-values for \( n \leq 30 \), since as noted by the authors, when \( n \) is large, numerical instabilities may occur. In the latter case, \texttt{ks.test} uses simulation to approximate p-values, which may be rather slow and inaccurate (see Table 6 of Dimitrova, Kaishev, Tan (2020)).

Thus, making use of the Exact-KS-FFT method, the function \texttt{disc_ks_test} provides an exact and highly computationally efficient (alternative) way of computing the p-value \( P(D_n \geq d_n) \), when \( F(x) \) is purely discrete.
Lastly, incorporated into the function `disc ks test` is the MC simulation-based method of Wood and Altavela (1978) for estimating the asymptotic p-value of $D_n$. The latter method is the default method behind `disc ks test` when the sample size $n$ is $n \geq 100000$.

**Value**

A list with class "htest" containing the following components:

- `statistic` the value of the statistic.
- `p.value` the p-value of the test.
- `alternative` "two-sided".
- `data.name` a character string giving the name of the data.

**References**


**See Also**

`ks.test`

**Examples**

# Comparison of results obtained from dgof::ks.test
# and KSgeneral::disc ks test, when F(x) follows the discrete
# Uniform[1, 10] distribution as in Example 3.5 of
# Dimitrova, Kaishev, Tan (2020)

# When the sample size is larger than 100, the
# function dgof::ks.test will be numerically
# unstable

x3 <- sample(1:10, 25, replace = TRUE)
KSgeneral::disc ks_test(x3, ecdf(1:10), exact = TRUE)
dgof::ks.test(x3, ecdf(1:10), exact = TRUE)
KSgeneral::disc ks_test(x3, ecdf(1:10), exact = TRUE)$p -
dgof::ks.test(x3, ecdf(1:10), exact = TRUE)$p

x4 <- sample(1:10, 500, replace = TRUE)
ks_c_cdf_Rcpp

R function calling directly the C++ routines that compute the complementary cumulative distribution function of the two-sided (or one-sided, as a special case) Kolmogorov-Smirnov statistic, when the cdf under the null hypothesis is arbitrary (i.e., purely discrete, mixed or continuous)

Description

Function calling directly the C++ routines that compute the complementary cdf for the one-sample two-sided Kolmogorov-Smirnov statistic, given the sample size $n$ and the file "Boundary_Crossing_Time.txt" in the working directory. The latter file contains $A_i$ and $B_i$, $i = 1, ..., n$, specified in Steps 1 and 2 of the Exact-KS-FFT method (see Equation (5) in Section 2 of Dimitrova, Kaishev, Tan (2020)). The latter values form the $n$-dimensional rectangular region for the uniform order statistics (see Equations (3), (5) and (6) in Dimitrova, Kaishev, Tan (2020)), namely $P(D_n \geq q) = 1 - P(A_i \leq U(t_i) \leq B_i, 1 \leq i \leq n) = 1 - P(g(t) \leq nU_n(t) \leq h(t), 0 \leq t \leq 1)$, where the upper and lower boundary functions $h(t), g(t)$ are defined as $h(t) = \sum_{i=1}^{n} 1_{(A_i \leq t)}$, $g(t) = \sum_{i=1}^{n} 1_{(B_i \leq t)}$, or equivalently, noting that $h(t)$ and $g(t)$ are correspondingly left and right continuous functions, we have $\sup\{t \in [0, 1] : h(t) < i\} = A_i$ and $\inf\{t \in [0, 1] : g(t) > i - 1\} = B_i$.

Note that one can also compute the (complementary) cdf for the one-sided KS statistics $D_n^-$ or $D_n^+$ (cf., Dimitrova, Kaishev, Tan (2020)) by appropriately specifying correspondingly $A_i = 0$ for all $i$ or $B_i = 1$ for all $i$, in the function `ks_c_cdf_Rcpp`.

Usage

`ks_c_cdf_Rcpp(n)`

Arguments

- `n` the sample size

Details

Note that all calculations here are done directly in C++ and output in R. That leads to faster computation time, as well as in some cases, possibly higher accuracy (depending on the accuracy of the pre-computed values $A_i$ and $B_i$, $i = 1, ..., n$, provided in the file "Boundary_Crossing_Time.txt") compared to the functions `cont_ks_c_cdf, disc_ks_c_cdf, mixed_ks_c_cdf`.
Given a random sample \( \{X_1, \ldots, X_n\} \) of size \( n \) with an empirical cdf \( F_n(x) \), the two-sided Kolmogorov-Smirnov goodness-of-fit statistic is defined as \( D_n = \sup \{|F_n(x) - F(x)|\} \), where \( F(x) \) is the cdf of a prespecified theoretical distribution under the null hypothesis \( H_0 \), that \( \{X_1, \ldots, X_n\} \) comes from \( F(x) \). The one-sided KS test statistics are correspondingly defined as \( D_n^+ = \sup_x (F_n(x) - F(x)) \) and \( D_n^- = \sup_x (F(x) - F_n(x)) \).

The function \texttt{ks_c_cdf_Rcpp} implements the Exact-KS-FFT method, proposed by Dimitrova, Kaishev, Tan (2020), to compute the complementary cdf, \( P(D_n \geq q) \) at a value \( q \), when \( F(x) \) is arbitrary (i.e. purely discrete, mixed or continuous). It is based on expressing the complementary cdf as

\[
P(D_n \geq q) = 1 - P(A_i \leq U(i) \leq B_i, 1 \leq i \leq n),
\]

where \( A_i \) and \( B_i \) are defined as in Step 1 of Dimitrova, Kaishev, Tan (2020).

The complementary cdf is then re-expressed in terms of the conditional probability that a homogeneous Poisson process, \( \xi_n(t) \) with intensity \( n \) will not cross an upper boundary \( h(t) \) and a lower boundary \( g(t) \), given that \( \xi_n(1) = n \) (see Steps 2 and 3 in Section 2.1 of Dimitrova, Kaishev, Tan (2020)). This conditional probability is evaluated using FFT in Step 4 of the method in order to obtain the value of the complementary cdf \( P(D_n \geq q) \). This algorithm ensures a total worst-case run-time of order \( O(n^2 \log(n)) \) which makes it highly computationally efficient compared to other known algorithms developed for the special cases of continuous or purely discrete \( F(x) \).

The values \( A_i \) and \( B_i \), \( i = 1, \ldots, n \), specified in Steps 1 and 2 of the Exact-KS-FFT method (see Dimitrova, Kaishev, Tan (2020), Section 2) must be pre-computed (in R or, if needed, using alternative softwares offering high accuracy, e.g. Mathematica) and saved in a file with the name "Boundary_Crossing_Time.txt" (in the current working directory).

The function \texttt{ks_c_cdf_Rcpp} is called in R and it first reads the file "Boundary_Crossing_Time.txt" and then computes the value for the complementary cdf \( P(D_n \geq q) = 1 - P(A_i \leq U(i) \leq B_i, 1 \leq i \leq n) = 1 - P(g(t) \leq nU_n(t) \leq h(t), 0 \leq t \leq 1) \) in C++ and output in R (or as noted above, as a special case, computes the value of the complementary cdf \( P(D_n^+ \geq q) = 1 - P(A_i \leq U(i) \leq 1, 1 \leq i \leq n) \) or \( P(D_n^- \geq q) = 1 - P(0 \leq U(i) \leq B_i, 1 \leq i \leq n) \)).

**Value**

Numeric value corresponding to \( P(D_n \geq q) = 1 - P(A_i \leq U(i) \leq B_i, 1 \leq i \leq n) = 1 - P(g(t) \leq \eta_n(t) \leq h(t), 0 \leq t \leq 1) \) (or, as a special case, to \( P(D_n^+ \geq q) \) or \( P(D_n^- \geq q) \)), given a sample size \( n \) and the file "Boundary_Crossing_Time.txt" containing \( A_i \) and \( B_i \), \( i = 1, \ldots, n \), specified in Steps 1 and 2 of the Exact-KS-FFT method (see Dimitrova, Kaishev, Tan (2020), Section 2).

**References**


**Examples**

```r
## Computing the complementary cdf P(D_\geq(n) >= q)
```
## for n = 10 and q = 0.1, when F(x) is continuous,
## In this case,
## B_i = (i-1)/n + q
## A_i = i/n - q

n <- 10
g <- 0.1
up_rec <- ((1:n)-1)/n + g
low_rec <- (1:n)/n - g
df <- data.frame(rbind(up_rec, low_rec))
write.table(df, "Boundary_Crossing_Time.txt", sep = " ", 
row.names = FALSE, col.names = FALSE)
ks_c_cdf_Rcpp(n)

mixed_ks_c_cdf

Computes the complementary cumulative distribution function of the
two-sided Kolmogorov-Smirnov statistic when the cdf under the null
hypothesis is mixed

Description

Computes the complementary cdf, $P(D_n \geq q)$ at a fixed $q$, $q \in [0,1]$, of the one-sample two-sided Kolmogorov-Smirnov statistic, when the cdf $F(x)$ under the null hypothesis is mixed, using the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)).

Usage

mixed_ks_c_cdf(q, n, jump_points, Mixed_dist, ..., tol = 1e-10)

Arguments

- **q**: numeric value between 0 and 1, at which the complementary cdf $P(D_n \geq q)$ is computed
- **n**: the sample size
- **jump_points**: a numeric vector containing the points of (jump) discontinuity, i.e. where the underlying cdf $F(x)$ has jump(s)
- **Mixed_dist**: a pre-specified (user-defined) mixed cdf, $F(x)$, under the null hypothesis.
- **...**: values of the parameters of the cdf, $F(x)$ specified (as a character string) by Mixed_dist.
- **tol**: the value of $\epsilon$ that is used to compute the values of $A_i$ and $B_i$, $i = 1, \ldots, n$, as detailed in Step 1 of Section 2.1 in Dimitrova, Kaishev and Tan (2020) (see also (ii) in the Procedure Exact-KS-FFT therein). By default, tol = 1e-10. Note that a value of NA or 0 will lead to an error!
Details

Given a random sample \( \{X_1, ..., X_n\} \) of size \( n \) with an empirical cdf \( F_n(x) \), the Kolmogorov-Smirnov goodness-of-fit statistic is defined as \( D_n = \sup |F_n(x) - F(x)| \), where \( F(x) \) is the cdf of a prespecified theoretical distribution under the null hypothesis \( H_0 \), that \( \{X_1, ..., X_n\} \) comes from \( F(x) \).

The function \texttt{mixed\_ks\_c\_cdf} implements the Exact-KS-FFT method, proposed by Dimitrova, Kaishev, Tan (2020) to compute the complementary cdf \( P(D_n \geq q) \) at a value \( q \), when \( F(x) \) is mixed. This algorithm ensures a total worst-case run-time of order \( O(n^2 \log(n)) \).

We have not been able to identify alternative, fast and accurate, method (software) that has been developed/implemented when the hypothesized \( F(x) \) is mixed.

Value

Numeric value corresponding to \( P(D_n \geq q) \).

References


Examples

```r
# Compute the complementary cdf of D_{n}\n# when the underlying distribution is a mixed distribution
# with two jumps at 0 and \log(2.5),
# as in Example 3.1 of Dimitrova, Kaishev, Tan (2020)

## Defining the mixed distribution

Mixed_cdf_example <- function(x)
{
  result <- 0
  if (x < 0){
    result <- 0
  }
  else if (x == 0){
    result <- 0.5
  }
  else if (x < log(2.5)){
    result <- 1 - 0.5 * exp(-x)
  }
  else{
    result <- 1
  }
  return (result)
}

KSgeneral::mixed ks c cdf(0.1, 25, c(0, log(2.5)), Mixed_cdf_example)
```
## Not run:
## Compute \( \text{P}(D_n \geq q) \) for \( n = 5 \),
## \( q = 1/5000, 2/5000, \ldots, 5000/5000 \)
## when the underlying distribution is a mixed distribution
## with four jumps at 0, 0.2, 0.8, 1.0,
## as in Example 2.8 of Dimitrova, Kaishev, Tan (2020)

\[
\begin{align*}
n & \leftarrow 5 \\
q & \leftarrow 1:5000/5000
\end{align*}
\]

\[
\text{Mixed_cdf_example} \leftarrow \text{function(x)}
\]
\[
\{ \\
\quad \text{result} \leftarrow 0 \\
\quad \text{if} \ (x < 0) \{ \\
\quad\quad \text{result} \leftarrow 0 \\
\quad \} \\
\quad \text{else if} \ (x == 0) \{ \\
\quad\quad \text{result} \leftarrow 0.2 \\
\quad \} \\
\quad \text{else if} \ (x < 0.2) \{ \\
\quad\quad \text{result} \leftarrow 0.2 + x \\
\quad \} \\
\quad \text{else if} \ (x < 0.8) \{ \\
\quad\quad \text{result} \leftarrow 0.5 \\
\quad \} \\
\quad \text{else if} \ (x < 1) \{ \\
\quad\quad \text{result} \leftarrow x - 0.1 \\
\quad \} \\
\quad \text{else} \{ \\
\quad\quad \text{result} \leftarrow 1 \\
\quad \} \\
\quad \text{return (result)}
\}
\]

\[
\text{plot(q, sapply(q, \text{function(x) KSgeneral::mixed_ks_c_cdf(x, n, c(0, 0.2, 0.8, 1.0), Mixed_cdf_example)), type='1'})}
\]

## End(Not run)

---

**mixed_ks_test**

Computes the p-value for a one-sample two-sided Kolmogorov-
Smirnov test when the cdf under the null hypothesis is mixed
mixed_ks_test

Description
Computes the p-value $P(D_n \geq d_n)$, where $d_n$ is the value of the KS test statistic computed based on a data sample $\{x_1, ..., x_n\}$, when $F(x)$ is mixed, using the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)).

Usage
mixed_ks_test(x, jump_points, Mixed_dist, ..., tol = 1e-10)

Arguments
- **x**: a numeric vector of data sample values $\{x_1, ..., x_n\}$.
- **jump_points**: a numeric vector containing the points of (jump) discontinuity, i.e. where the underlying cdf $F(x)$ has jump(s).
- **Mixed_dist**: a pre-specified (user-defined) mixed cdf, $F(x)$, under the null hypothesis.
- **...**: values of the parameters of the cdf, $F(x)$ specified (as a character string) by Mixed_dist.
- **tol**: the value of $\epsilon$ that is used to compute the values of $A_i$ and $B_i$, $i = 1, ..., n$, as detailed in Step 1 of Section 2.1 in Dimitrova, Kaishev and Tan (2020) (see also (ii) in the Procedure Exact-KS-FFT therein). By default, tol = 1e-10. Note that a value of NA or 0 will lead to an error!

Details
Given a random sample $\{X_1, ..., X_n\}$ of size $n$ with an empirical cdf $F_n(x)$, the Kolmogorov-Smirnov goodness-of-fit statistic is defined as $D_n = \sup |F_n(x) - F(x)|$, where $F(x)$ is the cdf of a prespecified theoretical distribution under the null hypothesis $H_0$, that $\{X_1, ..., X_n\}$ comes from $F(x)$.

The function mixed_ks_test implements the Exact-KS-FFT method expressing the p-value as a double-boundary non-crossing probability for a homogeneous Poisson process, which is then efficiently computed using FFT (see Dimitrova, Kaishev, Tan (2020)). This algorithm ensures a total worst-case run-time of order $O(n^2 log(n))$.

The function mixed_ks_test computes the p-value $P(D_n \geq d_n)$, where $d_n$ is the value of the KS test statistic computed based on a user-provided data sample $\{x_1, ..., x_n\}$, when $F(x)$ is mixed.

We have not been able to identify alternative, fast and accurate, method (software) that has been developed/implemented when the hypothesized $F(x)$ is mixed.

Value
A list with class "htest" containing the following components:
- **statistic**: the value of the statistic.
- **p.value**: the p-value of the test.
- **alternative**: "two-sided".
- **data.name**: a character string giving the name of the data.
References


Examples

# Example to compute the p-value of the one-sample two-sided KS test,
# when the underlying distribution is a mixed distribution
# with two jumps at 0 and log(2.5),
# as in Example 3.1 of Dimitrova, Kaishev, Tan (2020)

# Defining the mixed distribution

Mixed_cdf_example <- function(x)
{
  result <- 0
  if (x < 0)
  {
    result <- 0
  }
  else if (x == 0)
  {
    result <- 0.5
  }
  else if (x < log(2.5))
  {
    result <- 1 - 0.5 * exp(-x)
  }
  else{
    result <- 1
  }
  return (result)
}
test_data <- c(0,0,0,0,0,0.1,0.2,0.3,0.4,
               0.5,0.6,0.7,0.8,log(2.5),log(2.5),
               log(2.5),log(2.5),log(2.5),log(2.5))
KSgeneral::mixed_ks_test(test_data, c(0, log(2.5)),
                              Mixed_cdf_example)

## Compute the p-value of a two-sided K-S test
## when F(x) follows a zero-and-one-inflated
## beta distribution, as in Example 3.3
## of Dimitrova, Kaishev, Tan (2020)

## The data set is the proportion of inhabitants
## living within a 200 kilometer wide costal strip
## in 232 countries in the year 2010

data("Population_Data")
mu <- 0.6189
phi <- 0.6615
a <- mu * phi
Population_Data

\[ b \leftarrow (1 - \mu) \cdot \phi \]

\[
\text{Mixed_cdf_example} \leftarrow \text{function}\(x\) \{
\begin{align*}
\text{result} & \leftarrow 0 \\
\text{if} \ (x < 0)\{ \\
& \quad \text{result} \leftarrow 0 \\
\} \\
\text{else if} \ (x == 0)\{ \\
& \quad \text{result} \leftarrow 0.1141 \\
\} \\
\text{else if} \ (x < 1)\{ \\
& \quad \text{result} \leftarrow 0.1141 + 0.4795 \cdot \text{pbeta}(x, a, b) \\
\} \\
\text{else} \{ \\
& \quad \text{result} \leftarrow 1 \\
\}
\}
\]

\text{return (result)}
\}

\text{KSgeneral::mixed_ks_test(Population_Data, c(0, 1), Mixed_cdf_example)}

<table>
<thead>
<tr>
<th>Population_Data</th>
<th>\text{The proportion of inhabitants living within a 200 kilometer wide costal strip in 232 countries in the year 2010}</th>
</tr>
</thead>
</table>

\textbf{Description}

This data set contains the proportion of inhabitants living within a 200 kilometer wide costal strip in 232 countries in the year 2010. In Example 3.3 of Dimitrova, Kaishev, Tan (2020), the data set is modelled using a zero-and-one-inflated beta distribution in the null hypothesis and a one-sample two-sided Kolmogorov-Smirnov test is performed to test whether the proposed distribution fits the data well enough.

\textbf{Usage}

\text{data("Population_Data")}

\textbf{Format}

A data frame with 232 observations on the proportion of inhabitants living within a 200 kilometer wide costal strip in 2010.

\textbf{Source}

References

Index

* datasets
  Population_Data, 21

  cont_ks_c_cdf, 3, 5, 6, 14
  cont_ks_cdf, 4, 4
  cont_ks_test, 3, 7, 8

  disc_ks_c_cdf, 3, 9, 9, 10, 14
  disc_ks_test, 3, 11, 12, 13

  ecdf, 9, 12

  ks.test, 3, 4, 6, 8, 10, 12, 13
  ks_c_cdf_Rcpp, 2, 3, 14, 14, 15
  KSgeneral-package, 2

  mixed_ks_c_cdf, 3, 14, 16, 17
  mixed_ks_test, 3, 18, 19

  pexp, 7
  pnorm, 7
  Population_Data, 21

  stepfun, 9, 12