# Package 'serieslcb'

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Title Lower Confidence Bounds for Binomial Series System

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<b>Description</b> Calculate and compare lower confidence bounds for binomial series system reliability. The R 'shiny' application, launched by the function launch_app(), weaves together a workflow of customized simulations and delta coverage calculations to output recommended lower confidence bound methods.
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bayes Bayesian method

## Description

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Calculate a binomial series lower confidence bound using Bayes' method with a Beta prior distribution.

## Usage

```
bayes(s, n, alpha, MonteCarlo, beta.a, beta.b, ...)
```

## Arguments

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
MonteCarlo	Number of samples to draw from the posterior distribution for the Monte Carlo estimate.
beta.a	Shape1 parameter for the Beta prior distribution.
beta.b	Shape2 parameter for the Beta prior distribution.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
bayes(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000, beta.a=1, beta.b=1)
```

bayes\_jeffreys 3

bayes_jeffreys	Bayesian method (Jeffrey's prior)

## Description

Calculate a binomial series lower confidence bound using Bayes' method with Jeffrey's prior.

## Usage

```
bayes_jeffreys(s, n, alpha, MonteCarlo, ...)
```

#### **Arguments**

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
MonteCarlo	Number of samples to draw from the posterior distribution for the Monte Carlo estimate.
•••	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

## **Examples**

```
bayes_jeffreys(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000)
```

bayes_nlg	Bayesian method (Negative Log Gamma Prior)

## Description

Caclulate a binomal series lower confidence bound using Bayes' method with negative log gamma priors on the components, defined such that the prior on the system is a uniform distribution.

```
bayes_nlg(s, n, alpha, MonteCarlo, ...)
```

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#### **Arguments**

s Vector of successes.n Vector of sample sizes.

alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

MonteCarlo Number of samples to draw from the posterior distribution for the Monte Carlo

estimate.

... Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

#### **Examples**

```
bayes_nlg(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000)
```

bayes\_uniform

Bayesian method (Uniform prior)

#### **Description**

Calculate a binomial series lower confidence bound using Bayes' method with a uniform prior distribution.

#### Usage

```
bayes_uniform(s, n, alpha, MonteCarlo, ...)
```

#### **Arguments**

s Vector of successes.n Vector of sample sizes.

alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

MonteCarlo Number of samples to draw from the posterior distribution for the Monte Carlo

estimate.

... Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
bayes_uniform(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000)
```

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chao_huwang	Chao-Huwang method	

## Description

Calculate a binomial series lower confidence bound using Chao and Huwang's (1987) method.

## Usage

```
chao_huwang(s, n, alpha, MonteCarlo, ...)
```

## Arguments

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
MonteCarlo	Number of samples to draw from the posterior distribution for the Monte Carlo estimate.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

## **Examples**

```
chao_huwang(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000)
```

(	coit	Coit's method	

## Description

Calculate a binomial series lower confidence bound using Coit's (1997) method.

```
coit(s, n, alpha, use.backup = FALSE, backup.method, ...)
```

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## Arguments

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
use.backup	If TRUE, then a backup.method in the will be used for the methods with calculate $LCB = 1$ in the case of no failures across all components. If FALSE (default), no backup.method is used.
backup.method	The backup method which is used for the methods which calculate $LCB = 1$ in the case of zero failures. Use function name.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

## **Examples**

```
coit(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

d	
---	--

## Description

Calculate a binomial series lower confidence bound using Easterling's (1972) method.

#### Usage

```
easterling(s, n, alpha, ...)
```

## Arguments

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
easterling(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

launch\_app 7

launch_app	Launch Shiny App
------------	------------------

#### **Description**

Launches an instance of an R Shiny App, which runs locally on the user's computer.

#### Usage

```
launch_app(MonteCarlo = 1000, use.backup = TRUE,
backup.method = lindstrom_madden_AC, sample.omega = "corners",
number = 50)
```

#### **Arguments**

MonteCarlo	The number of Monte Carlo samples to take. E.g. In a Bayesian method, how many samples to take from a posterior distribution to estimate the lower $\alpha$ -th quantile. The default value is 1000.
use.backup	If TRUE (default), then a backup.method in the will be used for the methods with calculate LCB = 1 in the case of no failures across all components. If FALSE, no backup.method is used.
backup.method	The backup method which is used for the methods which calculate $LCB = 1$ in the case of zero failures. The default is lindstrom_madden_AC.
sample.omega	The method used to define component reliabilities. Can be only one of "corners" (default), "random", or "both". See Details below.
number	The number of component reliability vectors sampled if sample.omega = "random" or "both". Default is 50.

## **Details**

If the "Download Histograms" button does not work, it can be fixed by launching the Shiny App on your local browser. This can be done by clicking on "Open in Browser" located at the top of your Shiny App. This seems to be an issue with the Download Handler that Shiny uses.

Define

$$\Omega = \{(p_1, p_2, \dots, p_m) : \prod_{i=1}^m p_i \in [R_L, R_U]\}$$

and

$$\Omega' = \{(p_1, p_2, \dots, p_m) : p_i = R_L^{1/m} or R_U^{1/m} \forall i\}$$

. If sample.omega = "corners" (the default), then the elements of

 $\Omega'$ 

are used for component reliabilities, of which there are

combinations. If sample.omega = "random", then each component reliability is sampled uniformly from the interval

$$[R_L^m, R_U^m]$$

. If sample.omega = "both", then the results of "corners" and "random" are appended together and both are used.

lindstrom\_madden

Lindstrom and Madden's method

#### **Description**

Calculate a binomial series lower confidence bound using Lindstrom and Madden's (1962) method.

#### Usage

```
lindstrom_madden(s, n, alpha, ...)
```

## **Arguments**

s Vector of successes.n Vector of sample sizes.

alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

. . . Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

#### **Examples**

```
lindstrom_madden(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

lindstrom\_madden\_AC

Lindstrom and Madden's method with Agresti-Coull

## Description

Calculate a binomial series lower confidence bound using Agresti-Coull (1998) lower confidence bound calculation in the Lindstrom and Madden's (1962) method.

```
lindstrom_madden_AC(s, n, alpha, ...)
```

madansky 9

## Arguments

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

## **Examples**

```
lindstrom_madden_AC(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

|--|

## Description

Calculate a binomial series lower confidence bound using Madansky's (1965) method.

#### Usage

```
madansky(s, n, alpha, use.backup = FALSE, backup.method, ...)
```

#### **Arguments**

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
use.backup	If TRUE, then a backup.method in the will be used for the methods with calculate $LCB = 1$ in the case of no failures across all components. If FALSE (default), no backup.method is used.
backup.method	The backup method which is used for the methods which calculate $LCB = 1$ in the case of zero failures. Use function name.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound. Note that if there are zero observed failures across all components, the output is LCB = 0.

```
madansky(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

mann\_grubbs

madansky.fun Lagrange multiplier in Madansky's method	madansky.fun	Lagrange multiplier in Madansky's method	
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#### **Description**

This function is called in the madansky() function to solve for the Lagrange multipliers.

#### Usage

```
madansky.fun(lam, s, n, alpha)
```

## Arguments

s Vector of successes.n Vector of sample sizes.

alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

mann_grubbs	Mann and Grubb's method

## Description

Calculate a binomial series lower confidence bound using Mann and Grubb's (1974) method.

#### Usage

```
mann_grubbs(s, n, alpha, ...)
```

## Arguments

s Vector of successes. n Vector of sample sizes. alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound. . . . Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
mann_grubbs(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

mann\_grubbs\_calc 11

		-
mann	grubbs	calc

Function to calculate the LCB in the Mann-Grubbs method.

## Description

Calculate the LCB in the Mann-Grubbs method.

#### Usage

```
mann_grubbs_calc(s, n, A, alpha)
```

#### **Arguments**

s Vector of successes.n Vector of sample sizes.

A The restricted sum, as calculated by the mann\_grubbs\_sum() function. alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

#### Value

The LCB for the Mann-Grubbs method.

mann\_grubbs\_sum

Function to calculate the restricted sum in the Mann-Grubbs method.

## Description

Calculate the restricted sum in the Mann-Grubbs method.

#### Usage

```
mann_grubbs_sum(s, n)
```

## Arguments

s Vector of successes.

n Vector of sample sizes.

#### Value

The restricted sum.

myhre\_rennie1

mr.fun	Function of $\beta$ in the Myhre-Rennie 2 method
--------	--

#### **Description**

This function is called in myhre\_rennie2() function to solve for the  $\beta$  value.

#### Usage

```
mr.fun(beta, s, n)
```

#### **Arguments**

beta	The value of $\beta$ .	
S	Vector of successes.	
n	Vector of sample sizes.	

#### **Description**

Calculate a binomial series lower confidence bound using the Myhre-Rennie (modified ML) method (1986).

#### Usage

```
myhre_rennie1(s, n, alpha, use.backup = FALSE, backup.method, ...)
```

## **Arguments**

s Vector of successes.n Vector of sample sizes.

alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound.

use.backup 
If TRUE, then a backup.method in the will be used for the methods with cal-

culate LCB = 1 in the case of no failures across all components. If FALSE

(default), no backup.method is used.

backup.method The backup method which is used for the methods which calculate LCB = 1 in

the case of zero failures. Use function name.

. . . Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

myhre\_rennie2

## **Examples**

```
myhre_rennie1(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

myhre\_rennie2

Myhre and Rennie (reliability invariant) method

## Description

Calculate a binomial series lower confidence bound using the Myhre-Rennie (reliability invariant) method (1986).

## Usage

```
myhre_rennie2(s, n, alpha, use.backup = FALSE, backup.method, ...)
```

#### **Arguments**

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
use.backup	If TRUE, then a backup.method in the will be used for the methods with calculate $LCB = 1$ in the case of no failures across all components. If FALSE (default), no backup.method is used.
backup.method	The backup method which is used for the methods which calculate $LCB = 1$ in the case of zero failures. Use function name.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
myhre_rennie2(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

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nishime	Nishime's method
---------	------------------

#### **Description**

Calculate a binomial series lower confidence bound using Nishime's (1959) method.

#### Usage

```
nishime(s, n, alpha, ...)
```

## Arguments

s Vector of successes. 
n Vector of sample sizes. 
alpha The significance level; to calculate a  $100(1-\alpha)\%$  lower confidence bound. 
... Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

#### **Examples**

```
nishime(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

nlg.post.sample	Sampling from Posterior of Negative Log Gamma prior and Binomial
	data.

## Description

Randomly sample from the posterior distribution resulting from a NLG prior and Binomial data.

#### Usage

```
nlg.post.sample(sample.size, shape, scale, s, n)
```

#### **Arguments**

sample.size	The number of draws from the posterior distribution.
shape	The shape parameter for the NLG prior.
scale	The scale parameter for the NLG prior.
s	The number of successes for the binomial data (should be a scalar).
n	The number of tests for the binomial data (should be a scalar).

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## **Examples**

```
nlg.post.sample(sample.size=50, shape=.2, scale=1, s=29, n=30)
```

normal\_approximation Normal approximation method

## Description

Calculate a binomial series lower confidence bound using a normal approximation with MLE estimates.

## Usage

```
normal_approximation(s, n, alpha, use.backup = FALSE, backup.method, ...)
```

#### **Arguments**

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
use.backup	If TRUE, then a backup.method in the will be used for the methods with calculate $LCB = 1$ in the case of no failures across all components. If FALSE (default), no backup.method is used.
backup.method	The backup method which is used for the methods which calculate $LCB = 1$ in the case of zero failures. Use function name.
	Additional arguments to be ignored.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

```
normal_approximation(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10)
```

pm.random

pm

Matrix of p-vector combinations

#### **Description**

Calculate a matrix of *p*-vector combinations (component reliabilities) which lie in the specified interval of system reliability. Rows correspond to *p*-vectors and columns correspond to components.

#### Usage

```
pm(Rs.int, m)
```

#### **Arguments**

Rs.int Interval (or single number) of total system reliability.

m Number of components.

#### **Details**

Denote Rs.int =  $(R_L, R_U)$ . This function calculates all elements of the set

$$\Omega' = \{(p_1, p_2, \dots, p_m) : p_i = R_L^{1/m} or R_U^{1/m} \forall i\}$$

.

#### Value

The  $2^m$  by m matrix of p-vector combinations.

#### **Examples**

```
pm(Rs.int = c(.9, .95), m=3)
```

pm.random

*Matrix of* p-vector combinations sampled randomly.

## Description

Randomly sample to build a matrix of *p*-vector combinations (component reliabilities) which lie in the specified interval of system reliability. Rows correspond to *p*-vectors and columns correspond to components.

```
pm.random(Rs.int, m, number)
```

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## Arguments

Rs.int Interval (or single number) of total system reliability.

m Number of components.

number The number of random samples to draw.

## **Examples**

```
pm.random(Rs.int=c(.9, .95), m=3, number=100)
```

rice\_moore

Rice and Moore's method

## Description

Calculate a binomial series lower confidence bound using Rice and Moore's (1983) method.

#### Usage

```
rice_moore(s, n, alpha, MonteCarlo, f.star = 1.5 - min(n) + 0.5 * sqrt((3 - 2 * min(n))^2 - 4 * (min(n) - 1) * log(alpha) * qchisq(p = alpha, df = 2)), ...)
```

#### **Arguments**

S	Vector of successes.
n	Vector of sample sizes.
alpha	The significance level; to calculate a $100(1-\alpha)\%$ lower confidence bound.
MonteCarlo	Number of samples to draw from the posterior distribution for the Monte Carlo estimate.
f.star	The number of psuedo-failures to use for a component that exhibits zero observed failures. The default value is from the log-gamma procedure proposed by Gatliffe (1976), and is the value used by Rice and Moore.

#### Value

The  $100(1-\alpha)\%$  lower confidence bound.

#### **Examples**

```
rice_moore(s=c(35, 97, 59), n=c(35, 100, 60), alpha=.10, MonteCarlo=1000)
```

Additional arguments to be ignored.

18 rmse.LCB

 ${\sf rmse.LCB}$ 

Root Mean Square Error

## Description

Calculate the root mean squared errors of the LCB's from the true system reliability. A measure of spread.

## Usage

```
rmse.LCB(LCB, R)
```

## Arguments

LCB Vector of LCB's.

R The true system reliability .

#### Value

The root mean squared error of the LCB's from the true system reliability.

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