

Package ‘pg’

July 22, 2023

Title Polya Gamma Distribution Sampler

Version 0.2.4

Description Provides access to a high performant random distribution sampler for the Polya Gamma Distribution using either 'C++' headers for 'Rcpp' or 'RcppArmadillo' and 'R'.

URL <https://tmsalab.github.io/pg/>, <https://github.com/tmsalab/pg>

BugReports <https://github.com/tmsalab/pg/issues>

License GPL (>= 3)

LinkingTo Rcpp, RcppArmadillo

Imports Rcpp

Encoding UTF-8

RoxygenNote 7.2.3

Suggests testthat (>= 2.1.0)

NeedsCompilation yes

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Repository CRAN

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pg_mean

Theoretical Polya Gamma Distribution's Mean and Variance

Description

Compute the theoretical mean and variance for a Polya Gamma variable.

Usage

```
pg_mean(h, z)
```

```
pg_var(h, z)
```

Arguments

h A single integer value corresponding to the "shape" parameter.
z A single numeric value corresponding to the "scale" parameter.

Value

Either the theoretical mean or theoretical variance for a Polya Gamma distribution.

Examples

```
# Fixed parameter distribution simulation ----  
  
## Parameters ----  
h = 1; z = .5  
## Attempt distribution recovery ----  
vector_of_pg_samples = rpg_vector(1e6, h, z)  
  
head(vector_of_pg_samples)  
length(vector_of_pg_samples)  
  
## Obtain the empirical results ----  
empirical_mean = mean(vector_of_pg_samples)  
empirical_var = var(vector_of_pg_samples)  
  
## Take the theoretical values ----  
theoretical_mean = pg_mean(h, z)  
theoretical_var = pg_var(h, z)  
  
## Form a comparison table ----  
  
# empirically sampled vs. theoretical values  
rbind(c(empirical_mean, theoretical_mean),  
      c(empirical_var, theoretical_var))
```

`rpg_scalar`*Sample from the Polya Gamma distribution $PG(h, z)$*

Description

Chooses the most efficient implemented method to sample from a Polya Gamma distribution. Details on algorithm selection presented below.

Usage`rpg_scalar(h, z)``rpg_vector(n, h, z)``rpg_hybrid(h, z)``rpg_gamma(h, z, trunc = 1000L)``rpg_devroye(h, z)``rpg_sp(h, z)``rpg_normal(h, z)`**Arguments**

<code>h</code>	integer values corresponding to the "shape" parameter.
<code>z</code>	numeric values corresponding to the "scale" parameter.
<code>n</code>	The number of samples to taken from a $PG(h, z)$. Used only by the vector sampler.
<code>trunc</code>	Truncation cut-off. Only used by the gamma sampler.

Details

The following sampling cases are enabled:

- $h > 170$: Normal approximation method
- $h > 13$: Saddlepoint approximation method
- $h = 1$ or $h = 2$: Devroye method
- $h > 0$: Sum of Gammas method.
- $h < 0$: Result is automatically set to zero.

Value

A single numeric value.

Examples

```
# Fixed parameter distribution simulation ----

## Parameters ----
h = 1; z = .5

## Sample only one value ----
single_value = rpg_scalar(h, z)
single_value

## Attempt distribution recovery ----
vector_of_pg_samples = rpg_vector(1e6, h, z)

head(vector_of_pg_samples)
length(vector_of_pg_samples)

## Obtain the empirical results ----
empirical_mean = mean(vector_of_pg_samples)
empirical_var = var(vector_of_pg_samples)

## Take the theoretical values ----
theoretical_mean = pg_mean(h, z)
theoretical_var = pg_var(h, z)

## Form a comparison table ----

# empirically sampled vs. theoretical values
rbind(c(empirical_mean, theoretical_mean),
      c(empirical_var, theoretical_var))

# Varying distribution parameters ----

## Generate varying parameters ----
u_h = 20:100
u_z = 0.5*u_h

## Sample from varying parameters ----
x = rpg_hybrid(u_h, u_z)
```

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