# Package 'UnivRNG'

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UnivRNG-package Univariate Pseudo-Random Number Generation

#### Description

This package implements the algorithms described in Demirtas (2005) for pseudo-random number generation of 17 univariate distributions. The following distributions are available: Left Truncated Gamma, Laplace, Inverse Gaussian, Von Mises, Zeta (Zipf), Logarithmic, Beta-Binomial, Rayleigh, Pareto, Non-central t, Non-central Chi-squared, Doubly non-central F, Standard t, Weibull, Gamma with  $\alpha$ <1, Gamma with  $\alpha$ <1, and Beta with  $\alpha$ <1 and  $\beta$ <1. For some distributions, functions that have similar capabilities exist in the base package; the functions herein should be regarded as complementary tools.

The methodology for each random-number generation procedure varies and each distribution has its own function. draw.left.truncated.gamma, draw.von.mises, draw.inverse.gaussian, draw.zeta, draw.gamma.alpha.less.than.one, and draw.beta.alphabeta.less.than.one are based on acceptance/rejection region techniques. draw.rayleigh, draw.pareto, and draw.weibull utilize the inverse CDF method. The chop-down method is used for draw.logarithmic. In draw.laplace, a sample from an exponential distribution with mean  $1/\lambda$  is generated and subsequently the sign is changed with probability 0.5 and all variables are shifted by  $\alpha$ . For the Beta-Binomial distribution in draw.beta.binomial,  $\pi$  is generated as the appropriate  $\beta$  and used as the success probability for the binomial portion. draw.noncentral.t utilizes on arithmetic functions of normal and chi-squared random variables. draw.noncentral.chisquared is based on the sum of squared random normal variables, and draw.noncentral.F is a ratio of chi-squared random variables generated via draw.noncentral.chisquared. draw.t employs a rejection polar method developed by Bailey (1994). draw.gamma.alpha.greater.than.one uses a ratio of uniforms method by Cheng and Feast (1979).

#### **Details**

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#### Author(s)

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#### References

Bailey, R. W. (1994). Polar generation of random variates with the t-distribution. *Mathematics of Computation*, **62**, 779-781.

Cheng, R. C. H., & Feast, G. M. (1979). Some simple gamma variate generation. *Applied Statistics*, **28**, 290-295.

Demirtas, H. (2005). Pseudo-random number generation in R for some univariate distributions. *Journal of Modern Applied Statistical Methods*, **4(1)**, 300-311.

draw.beta.alphabeta.less.than.one

*Generates variates from Beta distribution with max* $(\alpha, \beta) < 1$ 

#### **Description**

This function implements pseudo-random number generation for a Beta distribution for  $\max(\alpha, \beta) < 1$  with pdf

$$f(x|\alpha,\beta) = \frac{1}{B(\alpha,\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

for  $0 \le x \le 1, 0 < \alpha < 1$ , and  $0 < \beta < 1$  where  $\alpha$  and  $\beta$  are the shape parameters and  $B(\alpha, \beta)$  is the complete beta function.

#### Usage

draw.beta.alphabeta.less.than.one(nrep,alpha,beta)

# **Arguments**

nrep Number of data points to generate.

alpha First shape parameter. Must be less than 1. beta Second shape parameter. Must be less than 1.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### References

Jhonk, M. D. (1964). Erzeugung von betaverteilter und gammaverteilter zufallszahlen. *Metrika*, **8**, 5-15.

#### **Examples**

draw.beta.alphabeta.less.than.one(nrep=100000,alpha=0.7,beta=0.4)

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draw.beta.binomial

Generates variates from Beta-binomial distribution

# Description

This function implements pseudo-random number generation for a Beta-binomial distribution with pmf

$$f(x|n,\alpha,\beta) = \frac{n!}{x!(n-x)!B(\alpha,\beta)} \int_0^1 \pi^{\alpha-1+x} (1-\pi)^{n+\beta-1-x} d\pi$$

for  $x=0,1,2,...,\alpha>0$ , and  $\beta>0$ , where n is the sample size,  $\alpha$  and  $\beta$  are the shape parameters and  $B(\alpha,\beta)$  is the complete beta function.

# Usage

draw.beta.binomial(nrep,alpha,beta,n)

#### **Arguments**

nrep Number of data points to generate.

alpha First shape parameter.

beta Second shape parameter.

n Number of trials.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### **Examples**

```
draw.beta.binomial(nrep=100000,alpha=0.2,beta=0.25,n=10)
```

draw.beta.binomial(nrep=100000,alpha=2,beta=3,n=10)

draw.beta.binomial(nrep=100000,alpha=600,beta=400,n=20)

draw.gamma.alpha.greater.than.one

Generates variation from Gamma distribution with  $\alpha > 1$ 

#### **Description**

This function implements pseudo-random number generation for a Gamma distribution for  $\alpha>1$  with pdf

$$f(x|\alpha,\beta) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}} x^{\alpha-1} e^{-x/\beta}$$

for  $0 \le x < \infty$  and  $\min(\alpha, \beta) > 0$  where  $\alpha$  and  $\beta$  are the shape and scale parameters, respectively.

## Usage

draw.gamma.alpha.greater.than.one(nrep,alpha,beta)

## **Arguments**

nrep Number of data points to generate.

alpha Shape parameter for desired gamma distribution. Must be greater than 1.

beta Scale parameter for desired gamma distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

# References

Cheng, R. C. H., & Feast, G. M. (1979). Some simple gamma variate generation. *Applied Statistics*, **28**, 290-295.

```
draw.gamma.alpha.greater.than.one(nrep=100000,alpha=2,beta=2)
draw.gamma.alpha.greater.than.one(nrep=100000,alpha=3,beta=0.4)
```

draw.gamma.alpha.less.than.one

Generates variation from Gamma distribution with  $\alpha < 1$ 

# **Description**

This function implements pseudo-random number generation for a gamma distribution for  $\alpha<1$  with pdf

$$f(x|\alpha,\beta) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}} x^{\alpha-1} e^{-x/\beta}$$

for  $0 \le x < \infty$  and  $\min(\alpha, \beta) > 0$  where  $\alpha$  and  $\beta$  are the shape and scale parameters, respectively.

# Usage

draw.gamma.alpha.less.than.one(nrep,alpha,beta)

## **Arguments**

nrep Number of data points to generate.

alpha Shape parameter for desired gamma distribution. Must be less than 1.

beta Scale parameter for desired gamma distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

# References

Ahrens, J. H., & Dieter, U. (1974). Computer methods for sampling from gamma, beta, poisson and binomial distributions. *Computing*, **1**, 223-246.

#### **Examples**

draw.gamma.alpha.less.than.one(nrep=100000,alpha=0.5,beta=2)

draw.inverse.gaussian 7

draw.inverse.gaussian Generates variation from inverse Gaussian distribution

#### **Description**

This function implements pseudo-random number generation for an inverse Gaussian distribution with pdf

$$f(x|\mu,\lambda) = (\frac{\lambda}{2\pi})^{1/2} x^{-3/2} e^{-\frac{\lambda(x-\mu)^2}{2\mu^2 x}}$$

for x > 0,  $\mu > 0$ , and  $\lambda > 0$  where  $\mu$  and  $\lambda$  are the location and scale parameters, respectively.

## Usage

draw.inverse.gaussian(nrep,mu,lambda)

#### **Arguments**

nrep Number of data points to generate.

mu Location parameter for the desired inverse Gaussian distribution.

lambda Scale parameter for the desired inverse Gaussian distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### References

Michael, J. R., William, R. S., & Haas, R. W. (1976). Generating random variates using transformations with multiple roots. *The American Statistician*, **30**, 88-90.

```
draw.inverse.gaussian(nrep=100000, mu=1,lambda=1)
draw.inverse.gaussian(nrep=100000, mu=3,lambda=1)
```

draw.laplace

Generates variates from Laplace distribution

# Description

This function implements pseudo-random number generation for a Laplace (double exponential) distribution with pdf

$$f(x|\lambda,\alpha) = \frac{\lambda}{2}e^{-\lambda|x-\alpha|}$$

for  $\lambda$ >0 where  $\alpha$  and  $\lambda$  are the location and scale parameters, respectively.

# Usage

draw.laplace(nrep, alpha, lambda)

## **Arguments**

nrep Number of data points to generate.

alpha Location parameter for the desired Laplace distribution.

Scale parameter for the desired Laplace distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### **Examples**

```
draw.laplace(nrep=100000, alpha=4, lambda=2)
draw.laplace(nrep=100000, alpha=-5, lambda=4)
```

draw.left.truncated.gamma

Generates variates from left truncated Gamma distribution

#### **Description**

This function implements pseudo-random number generation for a left-truncated gamma distribution with pdf

$$f(x|\alpha,\beta) = \frac{1}{(\Gamma(\alpha) - \Gamma_{\tau/\beta}(\alpha))\beta^{\alpha}} x^{\alpha-1} e^{-x/\beta}$$

for  $0<\tau\leq x$ , and  $\min(\tau,\beta)>0$  where  $\alpha$  and  $\beta$  are the shape and scale parameters, respectively,  $\tau$  is the cutoff point at which truncation occurs, and  $\Gamma_{\tau/\beta}$  is the incomplete gamma function.

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## Usage

```
draw.left.truncated.gamma(nrep,alpha,beta,tau)
```

#### **Arguments**

nrep Number of data points to generate.

alpha Shape parameter for the desired gamma distribution. beta Scale parameter fot the desired gamma distribution.

tau Point of left truncation.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### References

Dagpunar, J. S. (1978). Sampling of variates from a truncated gamma distribution. *Journal of Statistical Computation and Simulation*, **8**, 59-64.

#### **Examples**

```
draw.left.truncated.gamma(nrep=100000,alpha=5,beta=1,tau=0.5)
draw.left.truncated.gamma(nrep=100000,alpha=2,beta=2,tau=0.1)
```

draw.logarithmic

Generates variates from logarithmic distribution

#### **Description**

This function implements pseudo-random number generation for a logarithmic distribution with pmf

$$f(x|\theta) = -\frac{\theta^x}{x\log(1-\theta)}$$

for  $x = 1, 2, 3, \dots$  and  $0 < \theta < 1$ .

## Usage

draw.logarithmic(nrep,theta)

#### **Arguments**

nrep Number of data points to generate.

theta Rate parameter of the desired logarithmic distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

#### References

Kemp, A. W. Efficient generation of logarithmically distributed pseudo-random variables. *Applied Statistics*, **30**, 249-253.

## **Examples**

```
draw.logarithmic(nrep=100000,theta=0.33)
draw.logarithmic(nrep=100000,theta=0.66)
```

draw.noncentral.chisquared

Generates variates from non-central chi-squared distribution

# **Description**

This function implements pseudo-random number generation for a non-central chi-squared distribution with pdf

$$f(x|\lambda,\nu) = \frac{e^{-(x+\lambda)/2}x^{\nu/2-1}}{2^{\nu/2}} \sum_{k=0}^{\infty} \frac{(\lambda x)^k}{4^k k! \Gamma(k+\nu/2)}$$

for  $0 \le x < \infty$ ,  $\lambda > 0$ , and  $\nu > 1$ , where  $\lambda$  is the non-centrality parameter and  $\nu$  is the degrees of freedom.

# Usage

draw.noncentral.chisquared(nrep,dof,ncp)

#### **Arguments**

nrep Number of data points to generate.

dof Degrees of freedom of the desired non-central chi-squared distribution.

ncp Non-centrality parameter of the desired non-central chi-squared distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

draw.noncentral.F

## **Examples**

```
draw.noncentral.chisquared(nrep=100000,dof=2,ncp=1)
draw.noncentral.chisquared(nrep=100000,dof=5,ncp=2)
```

draw.noncentral.F

Generates variates from doubly non-central F distribution

#### **Description**

This function implements pseudo-random number generation for a doubly non-central F distribution

$$F = \frac{X_1^2/n}{X_2^2/m}$$

where  $X_1^2 \sim \chi^2(n, \lambda_1)$ ,  $X_2^2 \sim \chi^2(m, \lambda_2)$ , n and m are numerator and denominator degrees of freedom, respectively, and  $\lambda_1$  and  $\lambda_2$  are the numerator and denominator non-centrality parameters, respectively. It includes central and singly non-central F distributions as a special case.

# Usage

```
draw.noncentral.F(nrep,dof1,dof2,ncp1,ncp2)
```

# **Arguments**

nrep	Number of data points to generate.
dof1	Numerator degress of freedom.
dof2	Denominator degrees of freedom.
ncp1	Numerator non-centrality parameter.
ncp2	Denominator non-centrality parameter.

#### Value

A vector containing generated data.

# See Also

```
draw.noncentral.chisquared
```

```
draw.noncentral.F(nrep=100000, dof1=2, dof2=4, ncp1=2, ncp2=4)
```

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draw.noncentral.t

Generates variates from doubly non-central t distribution

# Description

This function implements pseudo-random number generation for a non-central t distribution

$$\frac{Y}{\sqrt{U/\nu}}$$

where U is a central chi-square random variable with  $\nu$  degrees of freedom and Y is an independent, normally distributed random variable with variance 1 and mean  $\lambda$ .

#### Usage

```
draw.noncentral.t(nrep,nu,lambda)
```

#### **Arguments**

nrep Number of data points to generate.

nu Degrees of freedom of the desired non-central t distribution.

Non-centrality parameter of the desired non-central t distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

## **Examples**

```
draw.noncentral.t(nrep=100000,nu=4,lambda=2)
draw.noncentral.t(nrep=100000,nu=5,lambda=1)
```

draw.pareto

Generates variates from Pareto distribution

## Description

This function implements pseudo-random number generation for a Pareto distribution with pdf

$$f(x|\alpha,\beta) = \frac{ab^a}{x^{a+1}}$$

for  $0 < b \le x < \infty$  and a > 0 where a and b are the shape and location parameters, respectively.

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## Usage

```
draw.pareto(nrep, shape, location)
```

#### **Arguments**

nrep Number of data points to generate.

shape Shape parameter of the desired Pareto distribution.

location Location parameter of the desired Pareto distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

# **Examples**

```
draw.pareto(nrep=100000, shape=11, location=11)
draw.pareto(nrep=100000, shape=8, location=10)
```

draw.rayleigh

Generates variates from Rayleigh distribution

# Description

This function implements pseudo-random number generation for a Rayleigh distribution with pdf

$$f(x|\sigma) = \frac{x}{\sigma^2} e^{-x^2/2\sigma^2}$$

for  $x \ge 0$  and  $\sigma > 0$  where  $\sigma$  is the scale parameter.

# Usage

```
draw.rayleigh(nrep,sigma)
```

#### **Arguments**

nrep Number of data points to generate.

sigma Scale parameter of the desired Rayleigh distribution.

# Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

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# **Examples**

```
draw.rayleigh(nrep=100000, sigma=0.5)
draw.rayleigh(nrep=100000, sigma=3)
```

draw.t

Generates variates from standard t distribution

# **Description**

This function implements pseudo-random number generation for a standard-t distribution with pdf

$$f(x|\nu) = \frac{\Gamma(\frac{\nu+1}{2})}{\Gamma(\frac{\nu}{2})\sqrt{\nu\pi}} (1 + \frac{x^2}{\nu})^{-(\nu+1)/2}$$

for  $-\infty < x < \infty$  where  $\nu$  is the degrees of freedom.

# Usage

draw.t(nrep,dof)

#### **Arguments**

nrep Number of data points to generate.

dof Degrees of freedom of the desired t distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

## References

Bailey, R. W. (1994). Polar generation of random variates with the t-distribution. *Mathematics of Computation*, **62**, 779-781.

# **Examples**

```
draw.t(nrep=100000,dof=2)
```

draw.t(nrep=100000,dof=6)

draw.von.mises 15

draw.von.mises

Generates variates from Von Mises distribution

# **Description**

This function implements pseudo-random number generation for a Von Mises distribution with pdf

$$f(x|K) = \frac{1}{2\pi I_0(K)} e^{K\cos(x)}$$

for  $-\pi \le x \le \pi$  and K > 0 where  $I_0(K)$  is a modified Bessel function of the first kind of order 0.

# Usage

```
draw.von.mises(nrep,K)
```

# **Arguments**

nrep

Number of data points to generate.

Κ

Parameter of the desired von Mises distribution.

#### Value

A list of length three containing generated data, the theoretical mean, and the empirical mean with names y, theo.mean, and emp.mean, respectively.

# References

Best, D. J., & Fisher, N. I. (1979). Efficient simulation of the von mises distribution. *Applied Statistics*, **28**, 152-157.

```
draw.von.mises(nrep=100000,K=10)
draw.von.mises(nrep=100000,K=0.5)
```

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draw.weibull

Generates variates from Weibull distribution

# **Description**

This function implements pseudo-random number generation for a Weibull distribution with pdf

$$f(x|\alpha,\beta) = \frac{\alpha}{\beta^{\alpha}} x^{\alpha-1} e^{-(x/\beta)^{\alpha}}$$

for  $0 \le x < \infty$  and  $\min(\alpha, \beta) > 0$  where  $\alpha$  and  $\beta$  are the shape and scale parameters, respectively.

# Usage

```
draw.weibull(nrep, alpha, beta)
```

# **Arguments**

nrep Number of data points to generate.

alpha Shape parameter of the desired Weibull distribution. beta Scale parameter of the desired Weibull distribution.

# Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

# **Examples**

```
draw.weibull(nrep=100000, alpha=0.5, beta=1)
draw.weibull(nrep=100000, alpha=5, beta=1)
```

draw.zeta

Generates variates from Zeta (Zipf) distribution

# **Description**

This function implements pseudo-random number generation for a Zeta (Zipf) distribution with pmf

$$f(x|\alpha) = \frac{1}{\zeta(\alpha)x^{\alpha}}$$

for x=1,2,3,... and  $\alpha>1$  where  $\zeta(\alpha)=\sum_{x=1}^{\infty}x^{-\alpha}.$ 

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# Usage

```
draw.zeta(nrep, alpha)
```

# **Arguments**

nrep Number of data points to generate.

alpha Parameter of the desired zeta distribution.

#### Value

A list of length five containing generated data, the theoretical mean, the empirical mean, the theoretical variance, and the empirical variance with names y, theo.mean, emp.mean, theo.var, and emp.var, respectively.

# References

Devroye, L. (1986). Non-Uniform random variate generation. New York: Springer-Verlag.

```
draw.zeta(nrep=100000,alpha=4)
```

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