

# Package: UnivRNG (via r-universe)

September 17, 2024

**Type** Package

**Title** Univariate Pseudo-Random Number Generation

**Version** 1.2.3

**Date** 2021-03-05

**Author** Hakan Demirtas, Rawan Allozi, Ran Gao

**Maintainer** Ran Gao <rgao8@uic.edu>

**Description** Pseudo-random number generation of 17 univariate distributions proposed by Demirtas. (2005)  
<[DOI:10.22237/jmasm/1114907220](https://doi.org/10.22237/jmasm/1114907220)>.

**License** GPL-2 | GPL-3

**NeedsCompilation** no

**Date/Publication** 2021-03-05 18:10:02 UTC

**Repository** <https://bernice0321.r-universe.dev>

**RemoteUrl** <https://github.com/cran/UnivRNG>

**RemoteRef** HEAD

**RemoteSha** 5ca292d93a183c59162040215062081659aa3bae

## Contents

UnivRNG-package . . . . .	2
draw.beta.alphabeta.less.than.one . . . . .	3
draw.beta.binomial . . . . .	4
draw.gamma.alpha.greater.than.one . . . . .	5
draw.gamma.alpha.less.than.one . . . . .	6
draw.inverse.gaussian . . . . .	7
draw.laplace . . . . .	8
draw.left.truncated.gamma . . . . .	8
draw.logarithmic . . . . .	9
draw.noncentral.chisquared . . . . .	10
draw.noncentral.F . . . . .	11
draw.noncentral.t . . . . .	12

draw.pareto . . . . .	12
draw.rayleigh . . . . .	13
draw.t . . . . .	14
draw.von.mises . . . . .	15
draw.weibull . . . . .	16
draw.zeta . . . . .	16

<b>Index</b>	<b>18</b>
--------------	-----------

---

 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

Package: UnivRNG  
 Type: Package  
 Version: 1.2.3  
 Date: 2021-03-05  
 License: GPL-2 | GPL-3

**Author(s)**

Hakan Demirtas, Rawan Allozi, Ran Gao

Maintainer: Ran Gao <rgao8@uic.edu>

**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  $0 \leq x \leq 1$ ,  $0 < \alpha < 1$ , 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)
```

---

```
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  $x = 0, 1, 2, \dots$ ,  $\alpha > 0$ , 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  $0 \leq x < \infty$  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.

### Examples

```
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  $0 \leq x < \infty$  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 *Generates variation from inverse Gaussian distribution*

---

### Description

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

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

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.

### Examples

```
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.
lambda	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.



**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 for 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	<i>Generates variates from logarithmic distribution</i>
------------------	---

---

**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  $0 \leq x < \infty$ ,  $\lambda > 0$ , 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.

**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 degree 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](#)

**Examples**

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

---

`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

<code>nrep</code>	Number of data points to generate.
<code>nu</code>	Degrees of freedom of the desired non-central t distribution.
<code>lambda</code>	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  $0 < b \leq x < \infty$  and  $a > 0$  where  $a$  and  $b$  are the shape and location parameters, respectively.

**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	<i>Generates variates from Rayleigh distribution</i>
---------------	--

---

**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  $x \geq 0$  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.

**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

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

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	<i>Generates variates from Von Mises distribution</i>
----------------	---

---

### 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  $-\pi \leq x \leq \pi$  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.
K	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.

### Examples

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

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

---

draw.weibull	<i>Generates variates from Weibull distribution</i>
--------------	---

---

### 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  $0 \leq x < \infty$  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	<i>Generates variates from Zeta (Zipf) distribution</i>
-----------	---

---

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



**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.

**Examples**

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

# Index

`draw.beta.alphabeta.less.than.one`, 3  
`draw.beta.binomial`, 4  
`draw.gamma.alpha.greater.than.one`, 5  
`draw.gamma.alpha.less.than.one`, 6  
`draw.inverse.gaussian`, 7  
`draw.laplace`, 8  
`draw.left.truncated.gamma`, 8  
`draw.logarithmic`, 9  
`draw.noncentral.chisquared`, 10, 11  
`draw.noncentral.F`, 11  
`draw.noncentral.t`, 12  
`draw.pareto`, 12  
`draw.rayleigh`, 13  
`draw.t`, 14  
`draw.von.mises`, 15  
`draw.weibull`, 16  
`draw.zeta`, 16

UnivRNG (UnivRNG-package), 2  
UnivRNG-package, 2