RANDOM NUMBER GENERATOR FOR THE COMMAND LINE RG 0.4.4 BETA

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1. PROPER USE OF THE GENERATOR TO PRESERVE ITS INTERNAL INGEGRITY

When executed repeatedly from the command line or a script, rg preserves the integrity of its internal random number generator via the option -I (capital i). Generating N random numbers using the -N option is thus the same as calling rg N times with the -I option.

The generator returns a single line of output. It consists of a space-separated list of random numbers, followed by ':' and a comma-separated list of shuffle table values that are to be used with the -I option during the next call.

```
#!/bin/bash
# The initialization option -1 is a capital i, not a small ell.
# Initialization of rg
nextInit = 'rg -S 0 | awk -F' :' '{ print $2}''
# Random values in a single variable
out='rg -l $nextInit -N 6 -D Exponential -m 5'
randomValues='echo $out | awk -F' :' '{ print $1 }''
nextInit='echo $out | awk -F' :' '{ print $2}''
# do something with the random numbers
echo Random values in a single variable
for value in $randomValues; do
 echo $value
done
# Random values in an array - note the () for random Values
out='rg -l $nextInit -N 5 -D Pareto -k 83456 -b 1.1'
randomValues = ( 'echo $out | awk -F' :' '{ print $1 } ' ')
nextInit = 'echo $out | awk -F' :' '{ print $2}'
# do something with the random numbers
numberOfValues=${#randomValues[@]}
echo ""; echo Random values in an array
for ((i=0; i < \text{SnumberOfValues}; i++)); do
 echo ${randomValues[$i]}
done
```

Usage of rg

Usage:

```
rg -S seed | -I initialization_vector [-N sample_size]
[-D distribution_name] [-m mu] [-l lambda] ...
[-E | -F | -A] [-O output_format] [-V] [-Q]
```

rg generates a sample of pseudo random numbers. It supports the following distributions: Uniform on (0,1), Exponential, Erlang, Gamma, Normal, Lognormal, Weibull, Pareto, Geometric, and Poisson.

When executed repeatedly from the command line, rg preserves the integrity of its internal generator via the option -I. Generating N random numbers using the -N option is thus the same as calling rg N times with the -I option.

Parameters

- -S seed of the generator. A non-negative integer value.
- -I initialization (capital I) of the generator. The argument is a comma-separated list of shuffle table values that were returned with random numbers generated during the previous call.
- -N the size of the generated random sample
- -D the desired probability distribution; one of: Uniform, Exponential, Erlang, Gamma, Normal, Lognormal, Weibull, Pareto, Poisson
- -m mu (it is the mean for the Exponential, Poisson, and Normal distributions)
- -I the rate parameter lambda (small ell) of the Exponential and Erlang distribution
- -s sigma squared the variance of the (underlying) Normal distribution
- -a the minimum value a of the Uniform distribution, the shape parameter alpha of the Gamma and Weibull distribution
- -b the maximum value b of the Uniform distribution, the scale parameter beta of the Gamma and Weibull distribution, the shape parameter beta of the Pareto distribution
- -k the scale parameter (minimum) of the Pareto distribution, or the shape parameter of the Erlang distribution
- -p the probability of success for the Geometric distribution
- -E engineering (fixed-point) output format, with C's default number of significant digits (%e). For continuous distributions only.
- -F floating-point output format, with C's default number of significant digits (%f). For continuous distributions only.
- -A automatic output format preserve as many significant digits as possible, using the '%e' or '%f' format as necessary (%.53g). For continuous distributions only.
- -O specify the output format as for printf in C. E.g. '%.6g'. For continuous distributions only.
- -V verbose ouput to stderr, including the version of the generator
- -Q quiet ouput (suppress output of initialization vector)

Returns

Space-separated list of random numbers, followed by ':' and a comma-separated list of shuffle table values that are to be used with the -I option during the next call.

Version

0.4.4 beta

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2. SUPPORTED DISCRETE DISTRIBUTIONS AND THEIR PARAMETERS

In the following sections X denotes a random variable with the discussed distribution.

2.1. Geometric Distribution.

 $X \sim \text{Geom}(p)$ if $P(X = k) = p(1-p)^k$ for k = 1, 2, 3, ...

where 0 . The expectation (mean) and variance of the random variable X are, respectively,

$$E X = \frac{1}{p} - 1$$
, $Var X = \frac{1 - p}{p^2}$.

The variable X can be viewed as the index of the first successful trial in a sequence of independent Bernoulli trials with probability of success equal to p. We often think of waiting for the first Head in repeated independent coin tosses with P(Head) = p. Then X is the toss number of the first Head.

Geometric distribution is closely related to the exponential distribution, which is often used to model a continuous random waiting time for the next customer. Geometric distribution can be viewed as a discrete version of such waiting time. Both these distributions share many properties, for example they are both memoryless.

2.2. Poisson Distribution.

$$X \sim \text{Poisson}(\mu)$$
 if $P(X = k) = \frac{e^{-\mu}\mu^k}{k!}$ for $k = 0, 1, 2, 3, ...$

where $\mu > 0$ is the mean of X, i.e. $EX = \mu$. The expectation and variance of the random variable X are, respectively,

$$E X = \mu$$
, $Var X = \mu$.

rg -S <seed> -D Poisson -m <mu>

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3. SUPPORTED CONTINUOUS DISTRIBUTIONS AND THEIR PARAMETERS

In the following sections X denotes a random variable with the discussed distribution, and f(x) represents its probability density function (pdf).

3.1. Continuous Uniform Distribution. The distribution is uniform over an interval [a, b]. The expectation and variance of the random variable X are, respectively,

$$E X = \frac{a+b}{2}$$
, $Var X = \frac{(b-a)^2}{12}$.

rg -S <seed> -D Uniform -a <minimum> -b <maximum>

Uniform(0,1) is the default distribution:

rg -S <seed> rg -S <seed> -D Uniform -a 0 -b 1

3.2. Exponential Distribution.

$$X \sim \mathsf{Exp}(\lambda)$$
 if its pdf is $f(x) = \begin{cases} \lambda \ e^{-\lambda x} & x \ge 0\\ 0 & \text{otherwise,} \end{cases}$

where $\lambda > 0$ is the rate parameter. The expectation and variance are

$$\mathsf{E} X = rac{1}{\lambda}, \qquad \qquad \mathsf{Var} X = rac{1}{\lambda^2}.$$

Exponential distribution is often used to model a continuous random waiting time for the next customer. A related distribution is the geometric distribution. It can be viewed as a discrete version of such waiting time. Both these distributions share many properties, for example they are both memoryless.

rg -S <seed> -D Exponential -I <lambda> rg -S <seed> -D Exponential -m <mean>

3.3. Erlang Distribution.

$$X \sim \text{Erlang}(k, \lambda) \quad \text{if its pdf is} \quad f(x) = \begin{cases} \frac{1}{(k-1)!} \lambda^k x^{k-1} e^{-\lambda x} & x \ge 0\\ 0 & \text{otherwise} \end{cases}$$

where $k \in \mathbb{N}^+$ (a positive integer) is the shape parameter, and $\lambda > 0$ is the rate parameter. The sum of k independent $\text{Exp}(\lambda)$ random variables has the $\text{Erlang}(k,\lambda)$ distribution. The expectation and variance of the random variable X are

$$\mathsf{E}\,X = \frac{\mathsf{k}}{\lambda}, \qquad \mathsf{Var}\,X = \frac{\mathsf{k}}{\lambda^2}.$$
rg -S -D Erlang -k -I

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3.4. Gamma Distribution.

$$X \sim \text{Gamma}(\alpha, \beta) \quad \text{if its pdf is} \quad f(x) = \begin{cases} \frac{1}{\Gamma(\alpha)} \beta^{-\alpha} x^{\alpha-1} e^{-x/\beta} & x \ge 0\\ 0 & \text{otherwise,} \end{cases}$$

where $\alpha > 0$ is the shape parameter, and $\beta > 0$ is the scale parameter. The distribution may also be denoted by $\Gamma(\alpha, \beta)$. The expectation and variance of the random variable X are

$$\mathsf{E} X = \alpha \, \beta,$$
 $\operatorname{Var} X = \alpha \, \beta^2.$

The function Γ in the pdf is the Euler gamma function

(1)
$$\Gamma(\alpha) = \int_0^\infty t^{\alpha-1} e^{-t} dt.$$

If α is a positive integer then $\Gamma(\alpha) = (\alpha - 1)!$ and hence the Gamma(α, β) distribution is the same as the Erlang($k = \alpha, \lambda = 1/\beta$) distribution. Notice the inverse of the scale parameter.

Caution, sometimes the Gamma distribution is defined with the scale parameter $1/\beta$ as in the Erlang distribution. Make sure that you use the correct scale parameter! If you cannot verify the form of the pdf, you can use the expectation $EX = \alpha\beta$ if its value is known. Then the scale parameter to use with the rg generator is $\beta = EX/\alpha$.

rg
$$-S$$
 $-D$ Gamma $-a$ $-b$

3.5. Normal Distribution (Gaussian Distribution).

$$X \sim N(\mu, \sigma^2)$$
 if its pdf is $f(x) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

where μ is the mean, and $\sigma^2 > 0$ is the variance of the distribution. Caution, some books use the standard deviation σ instead of the variance σ^2 in the normal distribution notation. Make sure that you use the variance σ^2 with rg!

rg -S <seed> -D Normal -m <mu> -s <sigma_squared>

3.6. Lognormal Distribution.

$$X \sim \ln N(\mu, \sigma^2)$$
 if $\ln(X) \sim N(\mu, \sigma^2)$.

The parameters μ and $\sigma^2 > 0$ are the mean and variance, respectively, of the *underlying normal* distribution. Caution, some books use the standard deviation σ instead of the variance σ^2 in the normal distribution notation. Make sure that you use the variance σ^2 with rg! The expectation and variance of X are

$$E X = e^{\mu + \sigma^2/2}$$
, $Var X = (e^{\sigma^2} - 1) e^{2\mu + \sigma^2}$.

rg -S <seed> -D Lognormal -m <mu> -s <sigma_squared>

3.7. Weibull Distribution.

$$X \sim \text{Weibull}(\alpha, \beta) \quad \text{if its pdf is} \quad f(x) = \begin{cases} \frac{\alpha}{\beta} \left(\frac{x}{\beta}\right)^{\alpha-1} e^{-(x/\beta)^{\alpha}} & x \ge 0\\ 0 & \text{otherwise,} \end{cases}$$

where $\alpha > 0$ is the shape parameter, and $\beta > 0$ is the scale parameter. The expectation and variance are

$$\mathsf{E} X = \beta \, \Gamma \left(1 + \frac{1}{\alpha} \right), \qquad \qquad \mathsf{Var} \, X = \beta^2 \, \Gamma \left(1 + \frac{2}{\alpha} \right) - (\mathsf{E} \, X)^2,$$

where Γ is the Euler gamma function as defined in (1).

If E X and Var X are known then the parameters can be found e.g. numerically:

(2)
$$\beta = \frac{\mathsf{E} X}{\Gamma\left(1 + \frac{1}{\alpha}\right)}$$

(3)
$$\operatorname{Var} X = (\mathsf{E} X)^2 \frac{\Gamma \left(1 + \frac{2}{\alpha}\right)}{\left(\Gamma \left(1 + \frac{1}{\alpha}\right)\right)^2} - (\mathsf{E} X)^2$$

Equation (3) can be solved for α numerically using Mathematica, Matlab, R, or a similar system, and equation (2) directly yields β .

3.8. Pareto Distribution.

 $X \sim \text{Pareto}(k, \beta) \quad \text{if its pdf is} \quad f(x) = \begin{cases} \beta k^{\beta} x^{-(\beta+1)} & x \geq k \\ 0 & \text{otherwise,} \end{cases}$

where k > 0 is the scale parameter, and $\beta > 0$ is the shape parameter. The parameter k is the minimal possible value of X. The parameter β is sometimes called the tail index. The expectation and variance are

$$\mathsf{E} X = \begin{cases} \frac{k\beta}{\beta-1} & \beta > 1\\ \infty & 0 < \beta \le 1, \end{cases} \qquad \qquad \mathsf{Var} \, X = \begin{cases} \frac{k^2\beta}{(\beta-2)(\beta-1)^2} & \beta > 2\\ \infty & 1 < \beta \le 2. \end{cases}$$

rg -S <seed> -D Pareto -k <minimal_value> -b <beta>

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