Package 'ars'

April 5, 2024

Арт 3, 2024
Title Adaptive Rejection Sampling
Version 0.7
Date 2024-04-04
Author Paulino Perez Rodriguez <perpdgo@colpos.mx> original C++ code from Arnost Komarek based on ars.f written by P. Wild and W. R. Gilks</perpdgo@colpos.mx>
Maintainer Paulino Perez Rodriguez <perpdgo@colpos.mx></perpdgo@colpos.mx>
Depends R (>= $3.1.2$)
Description Adaptive Rejection Sampling, Original version.
License GPL (>= 2)
NeedsCompilation yes
Repository CRAN
Date/Publication 2024-04-05 16:03:06 UTC
R topics documented: ars
ars Adaptive Rejection Sampling
Description Adaptive Rejection Sampling from log-concave density functions
Usage
ars(n=1,f,fprima,x=c(-4,1,4),ns=100,m=3,emax=64,lb=FALSE,ub=FALSE,xlb=0,xub=0,)

2 ars

Arguments

n	sample size
f	function that computes $\log(f(u,))$, for given u, where $f(u)$ is proportional to the density we want to sample from
fprima	d/du log(f(u,))
x	some starting points in wich log(f(u,) is defined
ns	maximum number of points defining the hulls
m	number of starting points
emax	large value for which it is possible to compute an exponential
lb	boolean indicating if there is a lower bound to the domain
xlb	value of the lower bound
ub	boolean indicating if there is a upper bound to the domain
xub	value of the upper bound bound

arguments to be passed to f and fprima

Details

. . .

ifault codes, subroutine initial

0: successful initialisation

1: not enough starting points

2: ns is less than m

3: no abscissae to left of mode (if lb = false)

4: no abscissae to right of mode (if ub = false)

5: non-log-concavity detect

ifault codes, subroutine sample

0: successful sampling

5: non-concavity detected

6: random number generator generated zero

7: numerical instability

Value

a sampled value from density

Author(s)

Paulino Perez Rodriguez, original C++ code from Arnost Komarek based on ars.f written by P. Wild and W. R. Gilks

ars 3

References

Gilks, W.R., P. Wild. (1992) Adaptive Rejection Sampling for Gibbs Sampling, *Applied Statistics* **41**:337–348.

Examples

```
library(ars)
 #Example 1: sample 20 values from the normal distribution N(2,3)
 f<-function(x,mu=0,sigma=1)\{-1/(2*sigma^2)*(x-mu)^2\}
 fprima<-function(x,mu=0,sigma=1){-1/sigma^2*(x-mu)}</pre>
 mysample<-ars(20,f,fprima,mu=2,sigma=3)</pre>
 mysample
hist(mysample)
 #Example 2: sample 20 values from a gamma(2,0.5)
 f1 < -function(x, shape, scale=1){(shape-1)*log(x)-x/scale}
 f1prima < -function(x, shape, scale=1) {(shape-1)/x-1/scale}
 \label{lem:mysample1} \verb| mysample1| <- ars(20, f1, f1prima, x=4.5, m=1, lb=TRUE, xlb=0, shape=2, scale=0.5) |
 mysample1
hist(mysample1)
 #Example 3: sample 20 values from a beta(1.3,2.7) distribution
 f2 < -function(x,a,b)\{(a-1)*log(x)+(b-1)*log(1-x)\}
 f2prima < -function(x,a,b)\{(a-1)/x-(b-1)/(1-x)\}
  \label{eq:mysample2} \\ \text{mysample2} \\ -\text{ars}(20, f2, f2 \\ \text{prima}, \\ \\ \text{x=c}(0.3, 0.6), \\ \text{m=2,lb=TRUE}, \\ \text{xlb=0,ub=TRUE}, \\ \text{xub=1,a=1.3,b=2.7}) \\ \\ \text{mysample2} \\ -\text{ars}(20, f2, f2 \\ \text{prima}, \\ \\ \text{x=c}(0.3, 0.6), \\ \text{m=2,lb=TRUE}, \\ \text{xlb=0,ub=TRUE}, \\ \text{xub=1,a=1.3,b=2.7}) \\ \\ \text{mysample2} \\ -\text{ars}(20, f2, f2 \\ \text{prima}, \\ \\ \text{x=c}(0.3, 0.6), \\ \text{m=2,lb=TRUE}, \\ \text{xlb=0,ub=TRUE}, \\ \text{xub=1,a=1.3,b=2.7}) \\ \\ \text{mysample2} \\ -\text{mysample3} \\ \text{mysample3} \\ -\text{mysample3} \\ \text{mysample3} \\ -\text{mysample3} \\ \text{mysample3} \\ -\text{mysample3} \\ -\text{mys
 mysample2
 hist(mysample2)
```

Index

```
* distribution
ars, 1
```