# Package 'FAmle'

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R topics documented:
FAmle-package
boot.mle
ColesData
FAmle-internal
floodsNB
metropolis
mle
plot.metropolis
plot.mle
printineuropons

2 boot.mle

FAmle	e-package	um Like Distribu	and Bayesiar	Estimation of U	nivariate Proba-	
Index					1	18
	yarns	 	 		1	17
	station01AJ010 .					
	Q.conf.int	 	 		1	16
	Q.boot.ci	 	 		1	15
	print.mle	 	 		1	14

# Description

This package contains a series of functions that might be useful in carrying out maximum likelihood and Bayesian estimations of univariate probability distributions.

# Author(s)

Francois Aucoin (author and original maintainer), Thomas Petzoldt (actual maintainer, applied formal changes to pass CRANs package check). Many thanks to the original author for his agreement.

boot.mle Bootstrap Distribution for Fitted Model

# **Description**

This function allows the user to obtain draws from the (parametric) bootstrap distribution of the fitted model's parameters.

#### Usage

```
boot.mle(model, B = 200, seed = NULL, start = NULL,
  method = "Nelder-Mead")
```

# Arguments

model	mle object corresponding to the fitted model.
В	Requested number of bootstrap samples.
seed	A seed may be specified (see set.seed)
start	Starting values for the optimization algorithm (if is.null(start)==TRUE, the fitted model's parameters are used as starting values).
method	The optimization method to be used (see optim and mle).

ColesData 3

# **Details**

Parametric bootstrap – see References.

#### Value

model	mle object corresponding to the fitted model.
В	Requested number of bootstrap samples.
seed	The specified seed (see set.seed)
par.star	Array containing realized values from the bootstrap distribution of the maximum likelihood parameter estimators.
gof	The bootstrap distributions of two goodness-of-fit statistics: Anderson-Darling statistic and Pearson's correlation coefficient for the pair ("observed quantiles", "fitted quantiles").
p.value	Bootstrap p-values for the two goodness-of-fit statistics.
failure.rate	The proportion of bootstrap samples for which optimization failed using the specified starting values.
total.time	The total amount of time required to generate B bootstrap samples.

# References

Davison, A.C., and Hinkley, D.V. (1997). Bootstrap methods and their application. Cambridge University Press.

# See Also

```
mle, Q.conf.int, Q.boot.ci
```

# Examples

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'weibull',c(.1,.1))
boot.x <- boot.mle(fit.x,B=10)
boot.x$par.star
boot.x$p.value</pre>
```

ColesData

Annual Maximum Sea Levels at Port Pirie, South Australia

# Description

This dataset is taken from Coles (2001) (also see references therein), and consists of 64 sea level (in meters) yearly maxima for the time period 1923-1987.

4 distr

# Usage

```
data(ColesData)
```

#### **Format**

A data.frame containing two columns named year and sea.level (in meters).

#### **Source**

Coles (2001), page 4 (also see references therein).

#### References

Coles, S. (2001). An introduction to statistical modeling of extreme values. Springer.

distr

Distribution functions 4-in-1

# Description

This function can be used to call any of the 4 functions specific to a given probability distribution available in R.

#### Usage

```
distr(x, dist, param, type = "d", model = NULL, ...)
```

# Arguments

х	Vector (or array) of quantiles, vector (or array) of probabilities, or number of observations.
dist	Distribution name.
param	Vector (or array) of parameters.
type	Type of function to be called ('d', 'p', 'q', or 'r').
model	Object from the class mle - may be specified instead of param and dist.
	Additional arguments log, lower.tail, and log.p, depending on type.

# **Details**

For each distribution available in R, 4 functions can be called. For example, for the normal distribution, the following 4 functions are available: dnorm, pnorm, qnorm, and rnorm. For the normal distribution, based on the argument type, distr may be used to call any one of the previous four functions.

# Value

Returns the density, the distribution function, the quantile function, or random variates.

FAmle-internal 5

# Note

Most functions in FAmle rely upon distr.

#### **Examples**

```
## Example 1
dnorm(-4:4,0,1,log=TRUE)
distr(-4:4,'norm',c(0,1),type='d',log=TRUE)

## Example 2
mu.vec <- c(1,100,100)
sigma.vec <- c(1,11,111)
n <- 3
set.seed(123)
rnorm(n,mu.vec,sigma.vec)
set.seed(123)
distr(n,'norm',cbind(mu.vec,sigma.vec),'r')

## Example 3
qnorm(.9,mu.vec,sigma.vec)
distr(.9,'norm',cbind(mu.vec,sigma.vec),'q')</pre>
```

FAmle-internal

Internal Functions in the FAmle Package

#### **Description**

Internal functions in the FAmle package.

# Usage

```
cdf.plot(z)
delta.Q(p, model, ln = FALSE)
delta.QQ(model, alpha = 0.1, ln = FALSE)
Diff.1(x, f, h = 1e-04)
Diff.2(k, i, model, p, ln = FALSE)
Diff.3(i, model, p, ln = FALSE)
## S3 method for class 'metropolis'
hist(x, density = TRUE, ...)
## S3 method for class 'plot'
hist(x,...)
Plot.post.pred(x, ...)
post.pred(z, fun = NULL)
Quantile.plot(z, ci = FALSE, alpha = 0.05)
Return.plot(model, ci = FALSE, alpha = 0.05)
Carlin(x)
```

6 floodsNB

Arguments	
z	A mle object.
р	A vector of probabilities.
model	A mle object.
ln	Whether or not (TRUE or FALSE) computations should be carried out on the natural logarithmic scale.
alpha	The significance level.
х	Value at which the numerical derivative should be evaluated. For the Carlin function (see References for metropolis), this x corresponds to an object from the class metropolis.
f	A function to be differentiated.
h	Small number representing a small change in x.
k	Parameter value at which the first derivative should be evaluated.
i	Position of the parameter, within a vector of parameters, with respect to which differentiation should be carried out.
density	Whether or not (TRUE or FALSE) a Kernel density should be added to the histogram - see density.
	Additional arguments pertaining to hist.
fun	optional argument that may be used to modify the scale on which the histogram will be plotted.
ci	Whether or not (TRUE or FALSE) approximated $100*(1-\text{alpha})$ confidence intervals should be added to the plot (either Quantile.plot or Return.plot).

floodsNB	New Brunswick (Canada) Flood Dataset	
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# Description

floodsNB is a list object containing the hydrometric stations considered for analysis... Each element from the list corresponds to an hydrometric station located in the Canadian province of New Brunswick, for which the flow is unregulated. For each station, the following information is available:

- data: Maximum annual daily mean discharge (in  $m^3/s$ );
- peak: Maximum annual daily peak discharge (in  $m^3/s$ );
- In.drain: Natural logarithm of the drainage area (in  $km^2$ );
- coor: Coordinates (in latitude and longitude) of the hydrometric station;
- status: Station's status Active or Inactive;
- Aucoin. 2001: Whether or not (TRUE or FALSE) the station is retained for analysis in ....

metropolis 7

#### Usage

```
data(floodsNB)
```

#### **Format**

A list object whose elements correspond to distinct hydrometric stations.

#### Source

HYDAT database.

#### References

Environment and Climate Change Canada Historical Hydrometric Data web site, https://wateroffice.ec.gc.ca/mainmenu/historical\_data\_index\_e.html

metropolis

Bayesian Estimation of Univariate Probability Distributions

#### **Description**

For a given dataset, this function serves to approximate (using a Metropolis algorithm) the posterior distribution of the parameters for some specified parametric probability distribution.

# Usage

```
metropolis(model, iter = 1000, tun = 2, trans.list = NULL,
start = NULL, variance = NULL, prior = NULL, burn = 0,
uniroot.interval = c(-100, 100),pass.down.to.C=FALSE)
```

#### **Arguments**

model mle object corresponding to the fitted (by maximum likelihood) model.

A list(x=dataset, dist=distribution) object may also be provided, but the user will then have to make sure to specify the arguments start and variance. Moreover, the latter two arguments will have to be specified on their transformed scales (see transplaint)

scales (see trans.list).

iter The requested number of iterations - the Markov Chain's length.

tun A tuning constant; value by which the covariance matrix of the multivariate

normal proposal will be multiplied - see References.

trans.list A list object containing a function for each parameter that is to be estimated.

For each parameter, the function must correspond to the inverse transformation that will determine the parametrization for which the simulation will be carried

out (see Example and Details).

8 metropolis

start A vector of starting values for the algorithm. If NULL, the maximum likelihood

parameter estimates will be used as starting values for the Markov Chain. If model is not an object from the class mle, this argument will have to be specified, along with the argument variance. Moreover, as already stated above, the user will have to make sure that both start and variance are those for the

transformed parameters (see trans.list).

variance Covariance matrix of the multivariate normal proposal distribution. If NULL, the

observed Fisher's information will be used and multiplied by the specified tun. As for start, this argument needs to be specified if model is not from the class

mle.

prior A function that corresponds to the joint prior distribution (see Example). Note

that the prior distribution will be evaluated on the transformed parameter space(s).

burn Burn-in period (see References).

uniroot.interval

Default is c(-100,100). This interval is used by R's function uniroot to search

for the inverse of each element in trans.list.

pass.down.to.C If TRUE, the iterative task is passed down to a C program for faster implementa-

tion of the MCMC algorithm.

#### **Details**

This function uses a single block Metropolis algorithm with multivariate normal proposal. For this function to work properly, all parameters should be defined on the real line - parameter transformation(s) might be required. If trans.list is not specified, the function will assume that the parameter distributions are all defined on the real line (i.e., function(x) x will be used for each parameter). If no prior distribution is provided, an improper prior distribution - uniform on the interval )-Inf,+Inf( - will be used for all parameters (i.e., prior distribution proportional to 1 - function(x) 1).

In order to minimize the number of arguments for metropolis, the function automatically computes the inverse of trans.list: this suppresses the need for the user to provide both the "inverse transformation" and the "transformation". However, problems may occur, and it is why the user is allowed to alter uniroot.interval. Depending on the number of errors reported, future versions of this package may end up requesting that a list for both the "inverse transformation" and the "transformation" be provided by the user.

A nice list of references is provided below for more information on topics such as: MCMC algorithms, tuning of Metropolis-Hastings algorithms, MCMC convergence diagnostics, the Bayesian paradigm ...

#### Value

rate MCMC acceptance rate. This value is computed before applying the burn-in;

i.e., it is computed for sims.all.

total.time Total computation time.

sims.all Array containing all iterations.

sims Array containing iterations after burn-in.

input Inputted mle object.

metropolis 9

iter	Number of iterations.
prior	Prior distribution.
burn	Integer corresponding to the number of iterations to be discarded - burn-in period.
М	Parameter vector whose elements correspond to the parameter values (on the scales specified by trans.list) obtained at the last iteration of the Metropolis sampler; i.e. sims[iter,].
V	Covariance matrix computed using, after removing the burn-in period, the joint posterior distribution of the parameters (on the scales specified by trans.list). This matrix might be used to tune the MCMC algorithm.

#### References

Gelman, A., Carlin, J.B., Stern, H.S., and Rubin, D.B. (2004). Bayesian data analysis, 2nd edition, Chapman & Hall/CRC.

Carlin, B.P, and Louis, T.A. (2009). Bayesian methods for data analysis. Chapman & Hall/CRC.

Gamerman, D., and Lopes H.F. (2006). Markov Chain Monte Carlo: Stochastic simulation for Bayesian inference. 2nd edition, Chapman & Hall/CRC.

Gilks, W.R., Richardson, S., and Spiegelhalter, D.J. (1996). Markov Chain Monte Carlo in Practice. Chapman & Hall.

#### See Also

```
plot.metropolis, mle
```

```
### These examples should be re-run with, e.g., iter > 2000.
data(yarns)
x <- yarns$x
fit.x \leftarrow mle(x, 'gamma', c(.1, .1))
bayes.x.no.prior <- metropolis(model=fit.x,iter=150,</pre>
trans.list=list(function(x) x,function(x) exp(x)))
plot(bayes.x.no.prior)
# examples of prior distributions (note that these prior distribution
# are specified for the transformated parameters;
# i.e., in this case, 'meanlog' -> 'meanlog' and 'sdlog' -> 'ln.sdlog')
# for the scale parameter only
prior.1 <- function(x) dnorm(x[2],.8,.1)
# for both parameters (joint but independent in this case)
prior.2 <- function(x) dunif(x[1], 3.4, 3.6)*dnorm(x[2], 1, 1)
bayes.x.prior.2 <- metropolis(model=fit.x,iter=150,</pre>
trans.list=list(function(x) x, function(x) exp(x)), prior=prior.2)
plot(bayes.x.prior.2)
# Example where 'model' is not from the class 'mle'; i.e.
```

10 mle

mle

Maximum Likelihood Estimation of Univariate Probability Distributions

# **Description**

For a given dataset, this function serves to find maximum likelihood parameter estimates for some specified parametric probability distribution.

### Usage

```
mle(x, dist, start = NULL, method = "Nelder-Mead")
```

#### **Arguments**

x A univariate dataset (a vector).

dist Distribution to be fitted to x.

start Starting parameter values for the optimization algorithm (see optim).

method The optimization method to be used (see optim).

# Value

fit optim output (see optim). x.info Array that contains the following columns: i: (1:length(x)), x: (original dataset), z: (sorted dataset), Fx: (CDF of x evaluated at the estimated parameter values), Fz: (sorted values of Fx), Emp: (i/(length(x)+1)), zF: (distr(Emp, 'dist', par.hat, 'q') evaluated at estimated parameter values (par.hat)), fx: (PDF of x evaluated at the estimated parameter values), fz: (PDF of z evaluated at the estimated parameter values) dist Distribution fitted to x.

plot.metropolis 11

par.hat	Vector of estiamted parameters.
cov.hat	Observed Fisher's information matrix.
k	Number of parameters
n	Number of observations (i.e., $length(x)$ ).
log.like	Log-likelihood value evaluated at the estimated parameter (i.e. par.hat).
aic	Akaike information criterion computed as 2*k - 2*log.like.
ad	Anderson Darling statistic evaluated at the estimated parameter values.
data.name	Name for x.
rho	Pearson's correlation coefficient computed as cor(x.info[,'z'],x.info[,'zF']).

# See Also

```
optim, distr, boot.mle, metropolis, Q.conf.int
```

# **Examples**

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'weibull',c(.1,.1))
fit.x
names(fit.x)
#plot(fit.x)
#plot(fit.x,TRUE,alpha=.01)
p <- c(.9,.95,.99)
distr(p,model=fit.x,type='q')
Q.conf.int(p,fit.x,.01)
Q.conf.int(p,fit.x,.01,TRUE)</pre>
```

plot.metropolis

A Function to Plot metropolis objects

# **Description**

This function allows to user to call different plots for visual assessment of the posterior distribution(s).

# Usage

```
## S3 method for class 'metropolis'
plot(x, plot.type = "carlin", pos = 1:x$iter, ...)
```

12 plot.mle

# **Arguments**

mle object corresponding to the fitted model. plot.type The user may choose betweew: carlin returns the same plot as in Carlin and Louis (2009) (see References); ts returns plot.ts; pairs returns a pairs; hist returns an hist for each marginal posterior distribution; post.pred returns an histogram of the data's posterior predictive distribution. May be used by the user to plot a subset (i.e. a random subset, sample)) of the pos posterior distribution when pairs is called. This avoids using too much memory while building the plot.

Additional arguments pertaining to function plot.default.

# References

See list of references for metropolis.

#### See Also

```
metropolis
```

# **Examples**

```
data(yarns)
x <- yarns$x
fit.x \leftarrow mle(x, 'gamma', c(.1, .1))
bayes.x <- metropolis(model=fit.x,iter=100,</pre>
trans.list=list(function(x) exp(x), function(x) exp(x)))
plot(bayes.x)
plot(bayes.x,'hist',col='cyan')
plot(bayes.x,'pairs',cex=.1,pch=19)
plot(bayes.x,'pairs',pos=sample(1:bayes.x$iter,20),col='red')
plot(bayes.x,'post.pred',col='green')
```

plot.mle

Diagnostic Plots for the Fitted Model

#### **Description**

This function returns diagnotic plots for a mle object.

## Usage

```
## S3 method for class 'mle'
plot(x, ci = FALSE, alpha = 0.05,...)
```

print.metropolis 13

# Arguments

X	mle object corresponding to the fitted model.
ci	Whether or not approximate confidence intervals should be added to the return period and quantile plots.
alpha	1-alpha is the requested coverage probability for the confidence interval.
	none

#### See Also

```
mle, Q.conf.int
```

# **Examples**

```
data(yarns)
x <- yarns$x
fit.1 <- mle(x,'weibull',c(.1,.1))
fit.2 <- mle(x,'logis',c(.1,.1))
plot(fit.1,TRUE,.05)
dev.new();plot(fit.2,TRUE,.05)</pre>
```

print.metropolis

Bayesian Estimation of Univariate Probability Distributions

# **Description**

```
See metropolis.
```

#### Usage

```
## S3 method for class 'metropolis'
print(x, stats.fun = NULL,...)
```

# Arguments

```
    x metropolis object corresponding to the fitted model.
    stats.fun An optional function that may be provided by the user in order to obtain a posterior summary (see Example).
    ... none...
```

#### See Also

```
metropolis, print
```

14 print.mle

# **Examples**

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'gamma',c(.1,.1))
bayes.x <- metropolis(fit.x,50,trans.list=
list(function(x) exp(x), function(x) exp(x)))
print(bayes.x)
print(bayes.x,stats.fun=function(x) c(mean=mean(x),CV=sd(x)/mean(x)))</pre>
```

print.mle

Maximum Likelihood Estimation of Univariate Probability Distributions

# **Description**

See mle.

# Usage

```
## S3 method for class 'mle'
print(x,...)
```

# Arguments

x mle object corresponding to the fitted model.

... none...

# See Also

```
mle, print
```

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'gamma',c(.1,.1))
print(fit.x)
print.mle(fit.x)</pre>
```

Q.boot.ci

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Parametric Bootstrap Confidence Intervals for p-th Quantile

# **Description**

This function can be used to derive parametric bootstrap confidence intervals for the p-th quantile of the fitted distribution (see mle).

## Usage

```
Q.boot.ci(p,boot,alpha=.1)
```

#### **Arguments**

p Vector of probabilities.

boot An object obtained using boot.mle.

alpha 1-alpha is the interval's coverage probability.

#### Value

This functions returns two types of bootstrap confidence intervals for the p-th quantile - one is based on the "percentile" method, while the other corresponds to the basis bootstrap interval or "reflexion" (see References).

#### Note

See References for other means of deriving bootstrap intervals.

#### References

Davison, A.C., and Hinkley, D.V. (1997) Bootstrap methods and their application. Cambridge University Press.

#### See Also

```
boot.mle, mle, Q.conf.int
```

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'gamma',c(.1,.1))
Q.conf.int(p=c(.5,.9,.95,.99),model=fit.x,alpha=.01,ln=FALSE)
# should be run again with B = 1000, for example...
boot.x <- boot.mle(model=fit.x,B=50)
Q.boot.ci(p=c(.5,.9,.95,.99),boot=boot.x,alpha=.01)</pre>
```

Q.conf.int

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Approximate Confidence Intervals for p-th Quantile

## Description

This function can be used to derive approximate confidence intervals for the p-th quantile of the fitted distribution (see mle).

#### Usage

```
Q.conf.int(p, model, alpha = 0.1, ln = FALSE)
```

#### **Arguments**

model mle object corresponding to the fitted model.
alpha 1-alpha is the interval's coverage probability.

In whether or not the confidence interval of the p-th quantile should be computed

on the natural logarithmic scale (see Details).

#### **Details**

The p-th quantile confidence interval is derived using the observed Fisher's information matrix in conjuction with the well-known delta method. Here, Q. conf. int allows the user to chose between two types of confidence intervals: one that is computed on the original scale and one that is computed on the quantile's natural logarithmic scale.

#### Value

The function returns a 3-by-length(p) array containing, for each value of p, the confidence interval's lower and upper bounds, as well as the quantile point estimate (maximum likelihood).

#### References

Rice, J.A. (2006) Mathematical statistics and data analysis. Duxbury Press, 3rd edition (regarding the Delta method).

## See Also

```
plot.mle
```

```
data(yarns)
x <- yarns$x
fit.x <- mle(x,'gamma',c(.1,.1))
Q.conf.int(p=c(.5,.9,.95,.99),model=fit.x,alpha=.01,ln=FALSE)
Q.conf.int(p=c(.5,.9,.95,.99),model=fit.x,alpha=.01,ln=TRUE)</pre>
```

station01AJ010 17

station01AJ010

Annual Maximum Daily Mean Flow Data (NB, Canada)

# **Description**

This dataset is taken from the HYDAT database, and corresponds to realized values of annual maximum daily mean flows (in  $m^3/s$ ).

# Usage

```
data(station01AJ010)
```

#### **Format**

A vector of observations.

#### Source

Hydrometric station 01AJ010

#### References

Environment Canada: https://wateroffice.ec.gc.ca/

yarns

Yarns Failure Data

# **Description**

This dataset is taken from Gamerman and Lopes (2006) (also see references therein), and consists of 100 cycles-to-failure times for airplane yarns.

# Usage

```
data(yarns)
```

#### **Format**

A data. frame object - one column of 100 observations.

#### Source

Gamerman and Lopes (2006), page 255 (also see references therein).

# References

Gamerman, D., and Lopes H.F. (2006). Markov Chain Monte Carlo: Stochastic simulation for Bayesian inference. 2nd edition, Chapman & Hall/CRC.

# **Index**

* aplot plot.metropolis, 11 plot.mle, 12  * datasets ColesData, 3 floodsNB, 6 station01AJ010, 17	<pre>delta.QQ(FAmle-internal), 5 density, 6 Diff.1(FAmle-internal), 5 Diff.2(FAmle-internal), 5 Diff.3(FAmle-internal), 5 distr, 4, 11 dnorm, 4</pre>
yarns, 17 * distribution	FAmle, 5
distribution distr, 4	FAmle (FAmle-package), 2
* models	FAmle-internal, 5
boot.mle, 2	FAmle-package, 2
metropolis, 7	floodsNB, 6
mle, 10	h: -1 6 12
plot.metropolis, 11	hist, 6, 12 hist.metropolis(FAmle-internal), 5
plot.mle, 12	hist.plot (FAmle-internal), 5
print.metropolis, 13	mist.piot (railie internal), 3
print.mle, 14	list, 7
Q.boot.ci, 15	,
Q.conf.int, 16	metropolis, 6, 7, 11–13
* optimize	mle, 2-4, 6-9, 10, 12-16
boot.mle, 2 mle, 10	
* package	optim, 2, 10, 11
FAmle-internal, 5	pairs, <i>12</i>
FAmle-package, 2	plot.default, 12
* print	plot.metropolis, 9, 11
print.metropolis, 13	plot.mle, 12, 16
print.mle, 14	Plot.post.pred (FAmle-internal), 5
* robust	plot.ts, <i>12</i>
boot.mle, 2	pnorm, 4
	<pre>post.pred(FAmle-internal), 5</pre>
boot.mle, 2, 11, 15	print, <i>13</i> , <i>14</i>
Carlin (FAmle-internal), 5 cdf.plot (FAmle-internal), 5	print.metropolis, 13 print.mle, 14
ColesData, 3	Q.boot.ci, 3, 15 Q.conf.int, 3, 11, 13, 15, 16
data.frame, <i>4</i> , <i>17</i>	qnorm, 4
delta.Q(FAmle-internal),5	Quantile.plot(FAmle-internal), 5

INDEX 19

```
Return.plot (FAmle-internal), 5 rnorm, 4 set.seed, 2, 3 station01AJ010, 17 uniroot, 8 yarns, 17
```