# The BRugs Package

February 15, 2006

2 BRugs

write.datafile	
summary	
setValues	
samplesStats	
samplesSize	
samplesSetting	
samplesSet	
samplesSample	
samplesMonitors	
samplesHistory	
samplesGet	
samplesDensity	
samplesCorrel	
samplesCoda	
samplesClear	
samplesBgr	
samplesAutoC	
rats	
ranks	
plotHistory	
plotDensity	
plotBgr	
plotAutoC	
modelUpdate	
modelSetAP	
modelSeed	
modelSaveState	
modelPrecision	
modelNames	
modelModules	
modelIteration	
modelInits	
modelGenInits	
modelData	
modelCompile	
modelCheck	
modelAdaptivePhase	

BRugs

Introduction to BRugs

# Description

This manual describes how to use the BRugs software

# Usage

```
help.BRugs(browser = getOption("browser"))
```

BRugs 3

#### **Arguments**

browser

the name of the program to be used as hypertext browser. It should be in the PATH, or a full path specified.

#### **Details**

BRugs is a collection of R functions that allow users to analys graphical models using MCMC techniques. Most of the R functions in BRugs provide a interface to the BRugs dynamic link library (shared object file). The BRugs dynamic link library is able to make use of many of the WinBUGS components, in particular those components concerned with graphical models and MCMC simulation. BRugs lacks the GUI interface of WinBUGS but is able to use R to create graphical displays of the MCMC simulation. BRugs uses the same model specification language as WinBUGS and the same format for data and initial values. However BRugs always uses plain text files for input inplace of WinBUGS compound documents. The BRugs functions can be split into two groups: those associated with setting up and simulating the graphical model and those associated with making statistical inference. In general the R functions in BRugs correspond to the command buttons and and text entry fields in the menus of WinBUGS. Each WinBUGS text entry field splits into two R functions, one to set the quantity and the other to get the value of the quantity.

#### **Permission and Disclaimer**

BRugs is released under the GNU GENERAL PUBLIC LICENSE. For details see <a href="http://mathstat.helsinki.fi/openbugs/ortype">http://mathstat.helsinki.fi/openbugs/ortype</a> help.BRugs ().

More informally, potential users are reminded to be extremely careful if using this program for serious statistical analysis. We have tested the program on quite a wide set of examples, but be particularly careful with types of model that are currently not featured. If there is a problem, BRugs might just crash, which is not very good, but it might well carry on and produce answers that are wrong, which is even worse. Please let us know of any successes or failures.

#### See Also

help.WinBUGS and the meta function BRugsFit

# Examples

```
###
       Step by step example:
                                ###
library(BRugs) # loading BRugs
## Now setting the working directory to the examples' one:
oldwd <- getwd()
setwd(system.file("OpenBUGS", "Examples", package="BRugs"))
## some usual steps (like clicking in WinBUGS):
modelCheck("ratsmodel.txt")
                                    # check model file
modelData("ratsdata.txt")
                                    # read data file
modelCompile(numChains=2)
                                    # compile model with 2 chains
modelInits(rep("ratsinits.txt", 2)) # read init data file
                                     # burn in
modelUpdate(1000)
samplesSet(c("alpha0", "alpha"))
                                    # alpha0 and alpha should be monitored
modelUpdate(1000)
                                     # 1000 more iterations ....
samplesStats("*")
                                     # the summarized results
## some plots
```

4 BRugsFit

```
samplesHistory("*", mfrow = c(4, 2)) # plot the chain,
samplesDensity("alpha") # plot the densities,
samplesBgr("alpha[1:6]") # plot the bgr statistics, and
samplesAutoC("alpha[1:6]", 1) # plot autocorrelations of 1st chain

## switch back to the previous working directory:
setwd(oldwd)
## Not run:
# Getting more (online-)help:
help.BRugs()
## End(Not run)
```

BRugsFit

BRugs' meta function

## **Description**

This function takes model, data and starting values as input and automatically runs a simulation in BRugs.

# Usage

```
BRugsFit (modelFile, data, inits, numChains = 3, parametersToSave, nBurnin = 1000, nIter = 1000, nThin = 1, DIC = TRUE, working.directory = NULL, digits = 5)
```

# Arguments

DIC

modelFile	File containing the model written in OpenBUGS code.
data	Either a named list (names corresponding to variable names in the modelFile) of the data for the OpenBUGS model, <i>or</i> a vector or list of the names of the data objects used by the model. In these cases data are written into a file 'data.txt' into the working directory.
	If a filename of an existing file is given, data are read from that file.
inits	A list with numChains elements; each element of the list is itself a list of starting values for the OpenBUGS model, <i>or</i> a function creating (possibly random) initial values. In these cases inits are written into files 'inits1.txt',, 'initsN.txt' into the working directory.
	If a vector of filenames of existing files is given, inits are read from those files. Alternatively, if inits is not specified, initial values are generated by Open-BUGS.
numChains	Number of Markov chains (default: 3).
parametersTo	Save
	Character vector of the names of the parameters to save which should be monitored.
nBurnin	Length of burn in (before nIter iterations start).
nIter	Number of iterations (without burn in).
nThin	Every nThin-th iteration of each chain is stored.

Logical, whether to calculate and return the DIC.

bgrPoint 5

```
working.directory
```

Sets working directory during execution of this function; data, inits and other files are written to / read from this directory if no other directory is explicitly given in those arguments. If NULL, the current working directory is chosen.

digits Number of significant digits used for OpenBUGS input, see formatC.

#### Value

## A list containg components

Stats A data frame containing sample statistics. See samplesStats.

DIC The DIC statistics, if DIC=TRUE, else NULL. See dicStats.

#### See Also

```
BRugs, help.WinBUGS
```

#### **Examples**

bgrPoint

Internal functions (to support plotting the Gelman-Rubin convergence statistic)

#### **Description**

These functions are for internal use only. They support samplesBgr and plotBgr.

#### Usage

```
bgrGrid(node, bins = 50)
bgrPoint(node, iteration)
```

## **Arguments**

node Character vector of length 1, name of a variable in the model.

bins Blocksize

iteration Calculated by bgrGrid

## Note

Intended for internal use only.

```
samplesBgr, BRugs, help.WinBUGS
```

6 bugsData

buffer

Reading OpenBUGS buffer file

## **Description**

Reads OpenBUGS buffer file, internally used for intefacing to OpenBUGS.

# Usage

```
buffer()
```

#### Value

Prints the buffer, returns nothing.

#### See Also

```
BRugs, help.WinBUGS
```

bugsData

Writing input for OpenBUGS

## **Description**

Write data file for OpenBUGS.

# Usage

```
bugsData(data, fileName = file.path(getwd(), "data.txt"), digits = 5)
```

#### **Arguments**

data either a named list (names corresponding to variable names in the model file) of

the data for the OpenBUGS model, or a vector or list of the names of the data

objects used by the model

fileName the filename, defaults to 'data.txt' in the current working directory

digits number of significant digits used for OpenBUGS input, see formatC

## Value

Invisibly returns the fileName.

# See Also

BRugs

bugsInits 7

Writing input for OpenBUC
---------------------------

#### **Description**

Write files containing inits.

#### Usage

```
bugsInits(inits, numChains = 1, fileName, digits = 5)
```

#### **Arguments**

a list with n.chains elements; each element of the list is itself a list of starting values for the OpenBUGS model, or a function creating (possibly random) initial values

numChains
number of Markov chains
fileName the filename(s), one for each chain. Defaults to 'inits1.txt', ..., 'initsN.txt' in the

current working directory.

digits number of significant digits used for OpenBUGS input, see formatC

## Value

Invisibly returns the fileName(s).

#### See Also

BRugs

buildMCMC	Generating mcmc.list objects for package coda	
	g to the same of t	

#### **Description**

This functions reads samples from OpenBUGS and converts the results into an object of class mcmc.list that can directly be used by package coda for further analysis.

#### Usage

```
buildMCMC(node, beg = samplesGetBeg(), end = samplesGetEnd(),
    firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin())
```

# **Arguments**

node Character vector of length 1, name of a variable in the model.

beg, end Arguments to select a slice of monitored values corresponding to iterations

beg:end.
firstChain, lastChain

Arguments to select a sub group of chains.

thin To only use every thin-th value of the stored sample.

8 dic

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

# Value

An object of class meme.list which is a list containing meme objects.

#### See Also

```
mcmc.list, mcmc, BRugs, help.WinBUGS
```

currentValues

Last sampled values

# Description

This function returns the current (last sampled) values of a variable.

## Usage

```
currentValues (nodeLabel)
```

# **Arguments**

nodeLabel Character vector of length 1, name of a variable in the model.

#### Value

Vector of the current (last sampled) values of a variable.

#### See Also

```
setValues, BRugs, help.WinBUGS
```

dic

DIC

## **Description**

These functions are used to evaluate the Deviance Information Criterion.

## Usage

```
dicSet()
dicStats()
dicClear()
```

dic 9

#### **Details**

These functions are used to evaluate the Deviance Information Criterion (DIC; Spiegelhalter et al., 2002) and related statistics - these can be used to assess model complexity and compare different models. Most of the examples packaged with OpenBUGS contain an example of their usage.

It is important to note that DIC assumes the posterior mean to be a good estimate of the stochastic parameters. If this is not so, say because of extreme skewness or even bimodality, then DIC may not be appropriate. There are also circumstances, such as with mixture models, in which Open-BUGS will not permit the calculation of DIC and so the menu option is greyed out. Please see help.WinBUGS for restrictions.

#### Value

dicStats returns a data frame with columns:

Dbar	The posterior mean of the deviance, which is exactly the same as if the node 'deviance' had been monitored. This deviance is defined as $-2 * \log(\text{likelihood})$ : 'likelihood' is defined as $p(y \mid \text{theta})$ , where y comprises all stochastic nodes given values (i.e. data), and theta comprises the stochastic parents of y - 'stochastic parents' are the stochastic nodes upon which the distribution of y depends, when collapsing over all logical relationships.
Dhat	A point estimate of the deviance (-2 * log(likelihood)) obtained by substituting in the posterior means theta.bar of theta: thus Dhat = -2 * log(p(y   theta.bar)).
pD	The effective number of parameters is given by $pD = Dbar - Dhat$ . Thus $pD$ is the posterior mean of the deviance minus the deviance of the posterior means.
DIC	The Deviance Information Criterion is given by $DIC = Dbar + pD = Dhat + 2 * pD$ . The model with the smallest DIC is estimated to be the model that would best predict a replicate dataset of the same structure as that currently observed.

## Note

Users should ensure their simulation has converged before using these functions. If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

#### References

Spiegelhalter, D.J., Best, N.G., Carlin B.P., and van der Linde, A. (2002): Bayesian measures of model complexity and fit (with discussion). *J. Roy. Statist. Soc. B.* 64, 583-640.

## See Also

BRugs, help.WinBUGS

10 getObj

dimensions

Dimension of BUGS variables

# **Description**

This function is intended for internal use only.

#### Usage

```
dimensions (node)
```

#### **Arguments**

node

Character vector of length 1, name of a variable in the model.

#### Value

Dimension of BUGS variable specified by node, if it is a non-scalar one, else NULL.

#### See Also

```
BRugs, help.WinBUGS
```

get0bj

Expert functions

## **Description**

Getting class names of Component Pascal object

# Usage

```
getGraphObj(node)
getUpdaterObj(node)
```

## **Arguments**

node

Character vector of length 1, name of a variable in the model.

#### **Details**

OpenBUGS creates Component Pascal objects to represent each component of a name in the graphial model.

#### Value

getGraphObj returns a data frame of the class names of the Component Pascal object associated with each component.

getUpdaterObj returns a data frame of the class names of the Component Pascal object for each component of a variable that needs updating.

getChain 11

#### See Also

```
BRugs, help.WinBUGS
```

getChain

Current chain to be initialized

# Description

This function is intended for internal use only.

## Usage

```
getChain()
```

## Value

Number of the chain to be initialized next.

# See Also

```
BRugs, help.WinBUGS
```

getNumChains

Number of chains

# Description

This function returns the number of chains being simulated.

# Usage

```
getNumChains()
```

#### Value

Returns the number of chains from the current simulation.

```
BRugs, help.WinBUGS
```

12 loadModule

help.WinBUGS

WinBUGS documentation

## **Description**

Function that open the html version of the WinBUGS manual

# Usage

```
help.WinBUGS(browser = getOption("browser"))
```

# Arguments

browser

the name of the program to be used as hypertext browser. It should be in the PATH, or a full path specified.

#### See Also

```
help.BRugs
```

# **Examples**

```
## Not run:
help.WinBUGS()
## End(Not run)
```

loadModule

Load a module

# **Description**

This function loads a module.

# Usage

```
loadModule(module)
```

# **Arguments**

module

character, name of the module.

## See Also

See modelModules for currently loaded modules. BRugs, help.WinBUGS

modelAdaptivePhase 13

modelAdaptivePhase Getting length of adaptive phase

## **Description**

This function returns the length of the adaptive phase of the simulation.

## Usage

```
modelAdaptivePhase()
```

#### Value

This function returns the length of the adaptive phase of the simulation. This is only known after the simulation has finished adapting. If this function is called while the simulation is still adapting MAX (INTEGER) is returned. If the simulation does not have an adaptive phase then zero is returned.

#### Note

This function can be executed once the model has been compiled and initialized.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

#### See Also

```
BRugs, help.WinBUGS
```

modelCheck

Checking the model file

#### **Description**

This function parses a BUGS language description of the statistical model.

# Usage

```
modelCheck(fileName)
```

#### **Arguments**

fileName

file containing the BUGS language description of the statistical model.

#### Value

If a syntax error is detected the position of the error and a description of the error is printed, otherwise the 'model is syntaxicaly correct' message is displayed.

14 modelCompile

#### Note

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

#### See Also

```
BRugs, help.WinBUGS
```

modelCompile

Compiling the model

## **Description**

This function builds the data structures needed to carry out MCMC sampling.

# Usage

```
modelCompile(numChains = 1)
```

# **Arguments**

numChains

Simulation is carried out for numChains chains.

#### **Details**

The model is checked for completeness and consistency with the data. A node called 'deviance' is automatically created which calculates minus twice the log-likelihood at each iteration, up to a constant. This node can be used like any other node in the graphical model.

## Value

When the model has been successfully compiled, 'model compiled' message should be printed.

#### Note

This command becomes active once the model has been successfully checked (see modelCheck).

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

```
BRugs, help.WinBUGS
```

modelData 15

modelData

Loading the data

#### **Description**

This function loads data into the statistical model.

#### Usage

```
modelData(fileName = "data.txt")
```

#### **Arguments**

fileName

Filename(s) of file(s) containing the data in OpenBUGS format.

#### Value

If any syntax errors or data inconsistencies are detected an error message is displayed. Corrections can be made to the data without returning to the 'check model' stage. When the data have been loaded successfully the message 'data loaded' should appear.

#### Note

This function can be executed once a model has been successfully checked (see modelCheck), it can no longer be executed once the model has been successfully compiled.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

#### See Also

```
BRugs, help.WinBUGS
```

modelGenInits

Generating initial values

# Description

This function attempts to generate initial values by sampling either from the prior or from an approximation to the prior.

# Usage

```
modelGenInits()
```

# **Details**

In the case of discrete variables a check is made that a configuration of zero probability is not generated. This function will generate extreme values if any of the priors are very vague.

16 modelInits

#### Value

If the function is successful the message 'initial values generated: model initialized' is displayed otherwise the message 'could not generate initial values' is displayed.

#### Note

This function can be executed once the model has been successfully compiled (modelCompile), and can no longer be executed once the model has been initialized.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

#### See Also

```
BRugs, help.WinBUGS
```

modelInits

Loading initial values

# **Description**

This function loads initial values for the MCMC simulation.

#### Usage

```
modelInits(fileName, chainNum = NULL)
```

# **Arguments**

fileName

Character vector of filenames containing the initial values in OpenBUGS format.

chainNum

The initial values will be loaded for the chain number chainNum. By default chainNum is one the first time modelInits is executed and incremented by one after each call modula the number of chains numChains being simulated (and restarts at 1 after that). If fileName is a vector, chainNum is increased automatically by default after processing each file. If there is more than one file containing initial values for one chain, either set chainNum explicitly, or wait until cycle restarts at chain 1.

#### **Details**

This function checks that initial values are in the form of an appropriate R object or rectangular array and that they are consistent with any previously loaded data. If some of the elements in an array are known (say because they are constraints in a parameterisation), those elements should be specified as missing (NA) in the initial values file.

Generally it is recommended to load initial values for all fixed effect nodes (founder nodes with no parents) for all chains, initial values for random effects can be generated using the modelGenInits function.

modelIteration 17

#### Value

Any syntax errors or inconsistencies in the initial value are displayed. If, after loading the initial values, the model is fully initialized this will be reported by displaying the message 'model initialized'. Otherwise the message 'initial values loaded but this or another chain contain uninitialized variables' will be displayed. The second message can have several meanings:

- a) If only one chain is simulated it means that the chain contains some nodes that have not been initialized yet.
- b) If several chains are to be simulated it could mean (a) or that no initial values have been loaded for one of the chains.

In either case further initial values can be loaded, or modelGenInits can be executed to try and generate initial values for all the uninitialized nodes in all the simulated chains.

#### Note

This function can be executed once the model has been successfully compiled. It can still be executed once MCMC sampling has been started having the effect of starting the sampler out on a new trajectory.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

#### See Also

```
BRugs, help.WinBUGS
```

modelIteration

Returns number of iterations

## **Description**

This function returns the total number of iterations carried out divided by thin.

# Usage

```
modelIteration()
```

#### Value

This function returns the total number of iterations carried out divided by thin.

#### Note

This function can be executed once the model has been compiled and initialized.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

```
BRugs, help.WinBUGS
```

modelNames

modelModules

Loaded modules

# Description

Displays all the modules (dynamic link libraries) in use.

## Usage

```
modelModules()
```

## Value

Dataframe containing information on all the modules (dynamic link libraries) in use.

# See Also

```
BRugs, help.WinBUGS
```

modelNames

Get variable names in model

# **Description**

This function returns the names of variables contained in the current model.

# Usage

```
modelNames()
```

# Value

Character vector of names of variables contained in the current model.

```
BRugs, help.WinBUGS
```

modelPrecision 19

modelPrecision

Setting precision for prec figures

# Description

This function sets the precision to which results are displayed to prec figures.

## Usage

```
modelPrecision(prec)
```

# **Arguments**

prec

#### **Details**

It does not affect the precision of any calculations!

#### See Also

```
BRugs, help.WinBUGS
```

modelSaveState

Save the model's current state

# Description

This function saves the sate of each chain in OpenBUGS model

#### Usage

```
modelSaveState(stem)
```

#### **Arguments**

stem

?????.

## Value

#### Note

This function can be executed once a model has been successfully checked ?????? (see modelCheck).

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed. ??????

```
BRugs, help.WinBUGS
```

20 modelSetAP

	- 7 ~ -	1
mode	2 I S 6	$\sim \sim \sim$

Seed of Random Number Generator

#### **Description**

These functions set/return the seed of the random number generator.

## Usage

```
modelSetSeed(newSeed)
modelGetSeed(i = 1)
```

#### **Arguments**

newSeed a positive, non zero (vector of) integer(s). More than one integer if the chosen

random number generator requires more seed components.

i indicates which component of the seed should be returned.

#### See Also

```
BRugs, help.WinBUGS
```

modelSetAP

Changing settings of updating algorithms

# Description

These functions change adaptivePhase, iterations, and overRelaxation settings.

## Usage

```
modelSetAP(factoryName, adaptivePhase)
modelSetIts(factoryName, iterations)
modelSetOR(factoryName, overRelaxation)
```

# **Arguments**

factoryName

String defining which particular MCMC updating algorithm is to be tuned. Technically this string is the type name of the factory object used to create the updater, for example 'UpdaterMetnormal.Factory' for the random walk metropolis sampler.

adaptivePhase

length of the updater's adaptive phase

iterations number of times an iterative algorithm is run before a failure is reported

overRelaxation

amount of over relaxation the updater uses

modelUpdate 21

#### **Details**

Once a model has been compiled, the various updating algorithms required in order to perform the MCMC simulation may be 'tuned' somewhat via these three functions.

#### See Also

```
BRugs, help. WinBUGS
```

modelUpdate

Updating the model

#### **Description**

This function updates the model.

## Usage

```
modelUpdate(numUpdates, thin = 1, overRelax = FALSE)
```

#### **Arguments**

individuales in this function updates the model by carrying out cliff * individuales inclin	numUpdates	This function updates the model by carrying	goutthin *	numUpdates MCMC
---	------------	---	------------	-----------------

iterations for each chain.

thin The samples from every kth iteration will be used for inference, where k is the

value of thin. Setting thin > 1 can help to reduce the autocorrelation in the sample, but there is no real advantage in thinning except to reduce storage

requirements.

overRelax If overRelax is TRUE an over-relaxed form of MCMC (Neal, 1998) which

will be executed where possible. This generates multiple samples at each iteration and then selects one that is negatively correlated with the current value. The time per iteration will be increased, but the within-chain correlations should be reduced and hence fewer iterations may be necessary. However, this method is not always effective and should be used with caution. The auto-correlation function may be used to check whether the mixing of the chain is improved.

# Note

This function can be executed once the model has been compiled and initialized.

If an attempt is made to execute this function in an inappropriate context the generic error message 'command is not allowed (greyed out)' is displayed.

# References

Neal, R. (1998): Suppressing random walks in Markov chain Monte Carlo using ordered over-relaxation. In M.I. Jordan (Ed.): *Learning in Graphical Models*, Kluwer Academic Publishers, Dordrecht, 205-230. http://www.cs.utoronto.ca/~radford/publications.html

```
BRugs, help.WinBUGS
```

22 plotBgr

ρl	$\sim$ $\pm$	7\	+ 4	$\sim$
$\nu$	OL.	ΑIJ		. )(.

Plot autocorrelation function for a scalar variable

## **Description**

This function plots the autocorrelation function of a scalar variable.

## Usage

```
plotAutoC(node, plot = TRUE,
    colour = c("red", "blue", "green", "yellow", "black"),
    lwd = 5, main = NULL, ...)
```

# **Arguments**

node	Character, name of a scalar variable in the model.
plot	Logical, whether to plot the ACF or only return the values. If ${\tt TRUE},$ values are returned invisibly.
colour	Colours used to represent different chains.
lwd, main	graphical parameters, see plot.default
	Further graphical parameters as in par.

#### **Details**

Acts on a scalar variable. See the wrapper function samplesAutoC for more details.

# Value

```
An acf object. See acf for details.
```

#### See Also

```
samplesAutoC, acf, BRugs, help.WinBUGS
```

plotBgr

Plot the Gelman-Rubin convergence statistic for a scalar variable

## **Description**

This function calculates and plots the Gelman-Rubin convergence statistic for a scalar variable, as modified by Brooks and Gelman (1998).

## Usage

```
plotBgr(node, plot = TRUE, main = NULL, xlab = "iteration",
    ylab = "bgr", col = c("red", "blue", "green"), bins = 50,
    ...)
```

plotDensity 23

# **Arguments**

node		Character, name of a scalar variable in the model.
plot		Logical, whether to plot the BGR statistics or only return the values. If TRUE, values are returned invisibly.
main,	xlab,	ylab
		annotation, see plot.default
col		Colours, see Details Section in samplesBgr.
bins		Number of blocks
		Further graphical parameters as in par.

# **Details**

Acts on a scalar variable. See the wrapper function samplesBgr for more details.

#### Value

## Data frame with elements

```
Iteration end iteration of corresponding bin

pooledChain80pct)
80pct interval (normalized) of pooled chains

withinChain80pct
80pct interval (normalized) of mean within chain

bgrRatio BGR ratio
```

## See Also

```
samplesBgr, BRugs, help.WinBUGS
```

plotDensity

Plot density estimate or histogram of a scalar variable

## **Description**

This function plots a smoothed kernel density estimate for a scalar variable if it is continuous or a histogram if it is discrete.

## Usage

```
plotDensity(node, main = NULL, xlab = "" , ylab = "", col = "red", ...)
```

# Arguments

```
node Character, name of a scalar variable in the model.

main, xlab, ylab, col
graphical parameters, see plot.default
... Further graphical parameters as in par.
```

24 plotHistory

#### **Details**

Acts on a scalar variable. See the wrapper function samplesDensity for more details.

#### See Also

```
samplesDensity, BRugs, help.WinBUGS
```

plotHistory

Trace of a scalar variable

#### **Description**

This function returns and plots a complete trace for a scalar variable.

# Usage

```
plotHistory(node, plot = TRUE,
    colour = c("red", "blue", "green", "yellow", "black"),
    main = NULL, xlab = "iteration", ylab = "", ...)
```

## **Arguments**

```
node Character, name of a scalar variable in the model.

plot Logical, whether to plot the trace or only return the values. If TRUE, values are returned invisibly.

colour Colours used to represent different chains.

main, xlab, ylab graphical parameters, see plot.default

... Further graphical parameters as in par.
```

#### **Details**

Acts on a scalar variable. See the wrapper function samplesHistory for more details.

#### Value

A matrix containing samples of node, each row corresponds to one chain.

```
samplesHistory, BRugs, help.WinBUGS
```

ranks 25

#### **Description**

These functions are used to calculate ranks of vector valued quantities in the model.

# Usage

```
ranksSet (node)
ranksStats (node)
ranksClear (node)
```

#### **Arguments**

node

Character, name of a vector (one dimensional array) variable in the model.

#### **Details**

ranksSet creates a monitor that starts building running histograms to represent the rank of each component of node. An amount of storage proportional to the square of the number of components of node is allocated. Even for large numbers of components this can require less storage than calculating the ranks explicitly in the model specification and storing their samples, and it is also much quicker.

ranksStats displays summarises of the distribution of the ranks of each component of node.

ranksClear removes the monitor calculating running histograms for node.

#### Value

ranksStats returns a data frame with columns:

```
val2.5pc 0.025 quantiles median medians val97.5pc 0.975 quantiles
```

## Note

Users should ensure their simulation has converged before using these functions. Note that if the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

```
BRugs, help.WinBUGS
```

26 samplesAutoC

rats	ratsdata	exam	ole

# Description

ratsdata example

## Usage

```
data(ratsdata)
data(ratsinits)
```

#### **Format**

The list ratsdata contains data originally taken from section 6 of Gelfand and Smith (1990).

#### **Source**

A. Gelfand and A. Smith (1990): Sampling-based Approaches to Calculating Marginal Densities. *Journal of the American Statistical Association*, 85, 398-409.

samplesAutoC Plot autocorrelation function

# Description

This function calculates and plots the autocorrelation function of a variable.

# Usage

```
samplesAutoC(node, chain, beg = samplesGetBeg(),
   end = samplesGetEnd(), thin = samplesGetThin(), plot = TRUE,
   mfrow = c(3, 2), ask = NULL, ann = TRUE, ...)
```

## **Arguments**

node	Character vector of length 1, name of a variable in the model.
chain	Selects a chain to plot autocorrelation function for.
beg, end	Arguments to select a slice of monitored values corresponding to iterations beg:end.
thin	To only use every thin-th value of the stored sample for statistics.
plot	Logical, whether to plot the ACF or only return the values. If TRUE, values are returned invisibly.
mfrow, ask,	ann
	Graphical parameters, see par for details. ask defaults to TRUE unless it is plotting into an already opened non-interactive device.
•••	Further graphical parameters as in par may also be passed as arguments to plotAutoC.

samplesBgr 27

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

#### Value

A list containing acf objects - one for each scalar variable contained in argument node. See acf for details on the list elements.

#### Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

#### See Also

```
plotAutoC, acf, BRugs, help.WinBUGS
```

samplesBgr

Plot the Gelman-Rubin convergence statistic

#### **Description**

This function calculates and plots the Gelman-Rubin convergence statistic, as modified by Brooks and Gelman (1998).

## Usage

```
samplesBgr(node, beg = samplesGetBeg(), end = samplesGetEnd(),
    firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin(),
    bins = 50, plot = TRUE, mfrow = c(3, 2), ask = NULL,
    ann = TRUE, ...)
```

## **Arguments**

node	Character vector of length 1, name of a variable in the model.
beg, end	Arguments to select a slice of monitored values corresponding to iterations
	beg:end.
firstChain,	lastChain
	Arguments to select a sub group of chains to calculate the Gelman-Rubin con-
	vergence statistics for. Number of chains must be larger than one.
thin	Only use every thin-th value of the stored sample for statistics.
bins	Number of blocks
plot	Logical, whether to plot the BGR statistics or only return the values. If TRUE, values are returned invisibly.
mfrow, ask,	ann
	Graphical parameters, see par for details. ask defaults to TRUE unless it is plotting into an already opened non-interactive device.
• • •	Further graphical parameters as in par may also be passed as arguments to plotBgr.

28 samplesClear

#### **Details**

The width of the central 80% interval of the pooled runs is green, the average width of the 80% intervals within the individual runs is blue, and their ratio R(=pooled/within) is red. For plotting purposes the pooled and within interval widths are normalised to have an overall maximum of one. The statistics are calculated in bins of length 50: R would generally be expected to be greater than 1 if the starting values are suitably over-dispersed. Brooks and Gelman (1998) emphasise that one should be concerned both with convergence of R to 1, and with convergence of both the pooled and within interval widths to stability.

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

#### Value

A list containing data frames - one for each scalar variable contained in argument node. Each data frames contains elements

```
Iteration end iteration of corresponding bin

pooledChain80pct)

80pct interval (normalized) of pooled chains

withinChain80pct

80pct interval (normalized) of mean within chain

bgrRatio

BGR ratio
```

#### Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

#### References

Brooks, S.P. and Gelman A. (1998): Alternative Methods for Monitoring Convergence of Iterative Simulations. *Journal of Computational and Graphical Statistics*, 7, 434-455.

# See Also

```
plotBgr, BRugs, help.WinBUGS
```

samplesClear

Clear recorded values

#### **Description**

This function is used to remove the stored values of a variable.

## Usage

```
samplesClear(node)
```

samplesCoda 29

#### **Arguments**

node

Character vector of length 1, name of a variable in the model.

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

#### See Also

```
BRugs, help.WinBUGS
```

samplesCoda

Writing files in CODA format

## **Description**

This function writes files in CODA format to be processed or imported, e.g, by some other software.

#### Usage

```
samplesCoda(node, stem, beg = samplesGetBeg(),
    end = samplesGetEnd(), firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin())
```

# **Arguments**

node Character vector of length 1, name of a variable in the model.

stem The filestem of the CODA files to be generated. See details.

beg, end Arguments to select a slice of monitored values corresponding to iterations beg:end.

firstChain, lastChain

Arguments to select a sub group of chains.

thin to only use every thin-th value of the stored sample.

#### **Details**

Example for argument stem: If stem = "c:/myFolder/foo", the resulting files are called 'fooCODAchain1.txt', ..., 'fooCODAchainN.txt', and 'fooCODAindex.txt'. They are written into the tempdir() and copied to the path '"c:/myFolder".

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...].

If the arguments are left at their defaults the whole sample for all chains will be used for output.

## Value

Prints 'CODA files written'.

30 samplesCorrel

#### Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

#### See Also

```
BRugs, help.WinBUGS
```

samplesCorrel

Correlation

# **Description**

This function calculates the correlation matrix between two vectors of variables.

## Usage

```
samplesCorrel(node0, node1, beg = samplesGetBeg(),
   end = samplesGetEnd(), firstChain = samplesGetFirstChain(),
   lastChain = samplesGetLastChain(), thin = samplesGetThin())
```

#### **Arguments**

```
node0, node1 Character vectors of length 1, name of variables in the model.

beg, end Arguments to select a slice of monitored values corresponding to iterations beg:end.

firstChain, lastChain
Arguments to select a sub group of chains to calculate correlation(s) for.

thin to only use every thin-th value of the stored sample for statistics.
```

# Details

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...].

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

#### Value

Correlation matrix.

## Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

```
BRugs, help.WinBUGS
```

samplesDensity 31

samplesDensity Plot density estimate or histogram

#### **Description**

This function plots a smoothed kernel density estimate for a variable if it is continuous or a histogram if it is discrete.

#### **Usage**

```
samplesDensity(node, beg = samplesGetBeg(), end = samplesGetEnd(),
    firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin(),
    mfrow = c(3, 2), ask = NULL, ann = TRUE, ...)
```

## **Arguments**

node Character vector of length 1, name of a variable in the model.

beg, end Arguments to select a slice of monitored values corresponding to iterations beg:end.

firstChain, lastChain
Arguments to select a sub group of chains to plot density estimate or histogram for.

thin to only use every thin-th value of the stored sample for statistics.

mfrow, ask, ann
Graphical parameters, see par for details. ask defaults to TRUE unless it is plotting into an already opened non-interactive device.

... Further graphical parameters as in par may also be passed as arguments to plotDensity.

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

# Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

```
BRugs, help.WinBUGS
```

32 samplesHistory

samplesGet

Get settings used for calculations

# Description

These low level functions can be used to get information on settings of begin, end, and thinning of chains, as well as the number of the first/last chain of the stored sample.

## Usage

```
samplesGetBeg()
samplesGetEnd()
samplesGetThin()
samplesGetFirstChain()
samplesGetLastChain()
```

#### Value

samplesGetBeg returns the first iteration of the stored sample used for calculating statistics.

 ${\tt samplesGetEnd}$  returns the last iteration of the stored sample used for calculating statistics to end.

samplesGetThin returns the thin parameter, see samplesSetThin.

samplesGetFirstChain returns the number of the first chain of the stored sample used for calculating statistics.

samplesGetLastChain returns the number of the last chain of the stored sample used for calculating statistics.

#### See Also

```
samplesSetBeg, BRugs, help.WinBUGS
```

samplesHistory

Trace of a variable

## **Description**

This function returns and plots a complete trace for a variable.

## Usage

```
samplesHistory(node, beg = samplesGetBeg(), end = samplesGetEnd(),
    firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin(),
    plot = TRUE, mfrow = c(3, 1), ask = NULL, ann = TRUE, ...)
```

samplesMonitors 33

Character and the office of 1 and 1

#### **Arguments**

node	Character vector of length 1, name of a variable in the model.
beg, end	Arguments to select a slice of monitored values corresponding to iterations
	beg:end.
firstChain,	lastChain
	Arguments to select a sub group of chains to plot the trace for.
thin	to only use every thin-th value of the stored sample for statistics.
plot	Logical, whether to plot the trace or only return the values. If TRUE, values are returned invisibly.
mfrow, ask,	ann
	Graphical parameters, see par for details. ask defaults to TRUE unless it is plotting into an already opened non-interactive device.
•••	Further graphical parameters as in par may also be passed as arguments to plotHistory.

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

#### Value

A list containing matrices - one for each scalar variable contained in argument node. Each row of a matrix corresponds to one chain.

## See Also

```
plotHistory, BRugs, help.WinBUGS
```

samplesMonitors

Names of monitored scalar variables

## **Description**

This function returns names of monitored scalar variables.

# Usage

```
samplesMonitors(node)
```

# Arguments

node Character vector of length 1, name of a variable in the model, or simply '\*'.

node can be a vector quantity with sub ranges given to indices (e.g. samplesMonitors ("node[

# Value

A list of names that are monitored. If sampling a vector of parameters of node, all elements are printed, e.g.: "node[beg]", ..., "node[end]".

34 samplesSet

#### See Also

```
BRugs, help.WinBUGS
```

samplesSample

Stored values

# Description

This function returns an array of stored values.

## Usage

```
samplesSample(node)
```

# Arguments

node

Character vector of length 1, name of a variable in the model.

## Value

Values of the stored sample.

## Note

If sampling a vector of parameters, the function must be called for each parameter separately such as samplesSample (node[1]).

## See Also

```
BRugs, help.WinBUGS
```

samplesSet

Start recording

# Description

This function is used to start recording a chain of values for particular variables.

# Usage

```
samplesSet (node)
```

## **Arguments**

node

Character vector of names of variables in the model.

# **Details**

WinBUGS generally automatically sets up a logical node to measure a quantity known as deviance; this may be accessed, in the same way as any other variable of interest, by typing its name, i.e. "deviance"

samplesSetting 35

#### See Also

```
BRugs, help.WinBUGS
```

samplesSetting

Change settings used for calculations

#### **Description**

These low level functions can be used to set begin, end, and thinning of chains as well as the first/last chain of the stored sample.

## Usage

```
samplesSetBeg(begIt)
samplesSetEnd(endIt)
samplesSetThin(thin)
samplesSetFirstChain(first)
samplesSetLastChain(last)
```

#### **Arguments**

begIt First iteration of the stored sample used for calculating statistics.

endIt Last iteration of the stored sample used for calculating statistics.

thin Every thin-th iteration of each chain is used to contribute to the statistics being calculated.

first, last First/last chain of the stored sample used for calculating statistics.

#### **Details**

samplesSetBeg sets the first iteration of the stored sample used for calculating statistics to begIt.

samplesSetEnd sets the last iteration of the stored sample used for calculating statistics to endIt.

samplesSetThin sets the numerical field used to select every thin-th iteration of each chain to contribute to the statistics being calculated.

samplesSetFirstChain is used to set the first chain of the stored sample used for calculating statistics to be first.

samplesSetLastChain is used to set the last chain of the stored sample used for calculating statistics to be last.

#### Note

Note the difference between this and the thinning facility of the update function: when thinning via the update function we are permanently discarding samples as the MCMC simulation runs, whereas here we have already generated (and stored) a suitable number of (posterior) samples and may wish to discard some of them only temporarily. Thus, setting  $\verb|thin| > 1$  here will not have any impact on the storage (memory) requirements; if you wish to reduce the number of samples actually stored (to free-up memory) you should thin via the update function.

36 samplesStats

#### See Also

```
BRugs, help.WinBUGS
```

samplesSize

Size of the stored sample

# Description

This function returns the size of the stored sample.

#### Usage

```
samplesSize(node)
```

## **Arguments**

node

Character vector of length 1, name of a variable in the model.

#### Value

Size of the stored sample. If no samples exist, -1 will be returned.

#### Note

If sampling a vector of parameters, the function must be called for each parameter separately such as samplesSize(node[1]).

#### See Also

```
BRugs, help.WinBUGS
```

samplesStats

Calculate summary statistics

# Description

This function produces summary statistics for a variable, pooling over the chains selected.

## Usage

```
samplesStats(node, beg = samplesGetBeg(), end = samplesGetEnd(),
    firstChain = samplesGetFirstChain(),
    lastChain = samplesGetLastChain(), thin = samplesGetThin())
```

samplesStats 37

#### **Arguments**

node Character vector containing names of variables in the model.

beg, end Arguments to select a slice of monitored values corresponding to iterations

beg:end.

firstChain, lastChain

Arguments to select a sub group of chains to calculate summary statistics for.

thin to only use every thin-th value of the stored sample for statistics.

#### **Details**

If the variable of interest is an array, slices of the array can be selected using the notation variable [lower0:upper0 lower1:upper1, ...]. A star '\*' can be entered as shorthand for all the stored samples.

If the arguments are left at their defaults the whole sample for all chains will be used for calculation.

#### Value

samples.stats returns a data frame with columns:

mean means

sd standard deviations

MC\_error Estimate of  $s/\sqrt(N)$ , the Monte Carlo standard error of the mean. The batch

means method outlined by Roberts (1996; p.50) is used to estimate s.

val2.5pc 0.025 quantiles

median medians

val 97.5pc 0.975 quantiles

 $\begin{array}{ccc} \text{start} & \text{beg} + 1 \\ \text{sample} & \text{sample sizes} \end{array}$ 

#### Note

If the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

#### References

Roberts, G.O. (1996): Markov Chain Concepts Related to Sampling Algorithms. In: W.R. Gilks, S. Richardson and D.J. Spiegelhalter (Eds.): *Markov Chain Monte Carlo in Practice*. Chapman and Hall, London, UK.

#### See Also

BRugs, help.WinBUGS

38 summary

setValues

Setting current values

# Description

This function sets current values for a variable for future iterations.

# Usage

```
setValues(nodeLabel, values)
```

## **Arguments**

nodeLabel Character vector of length 1, name of a variable in the model.

values The values to be set, generated, e.g., by currentValues.

## **Details**

currentValues of a model can be stored in order to be used as initial values.

#### Value

The number of values set.

#### See Also

```
currentValues, BRugs, help.WinBUGS
```

summary

Summary of MCMC simulation

## **Description**

These functions are used to calculate running means, standard deviations and quantiles.

# Usage

```
summarySet(node)
summaryStats(node)
summaryClear(node)
```

#### **Arguments**

node

Character vector containing names of a variables in the model.

write.datafile 39

#### **Details**

summarySet creates monitor(s) that starts recording the running totals for node.

summaryStats displays the running means, standard deviations, and 2.5%, 50% (median) and 97.5% quantiles for node. Note that these running quantiles are calculated via an approximate algorithm and should therefore be used with caution.

summaryClear removes the monitor(s) calculating running totals for node.

These functions are less powerful and general than the samples functions (e.g., see samplesSet), but they also require much less storage (an important consideration when many variables and/or long runs are of interest).

#### Value

summaryStats returns a data frame with columns:

mean means

sd standard deviations

val2.5pc 0.025 quantiles

median medians

val97.5pc 0.975 quantiles

sample sample sizes

#### Note

Users should ensure their simulation has converged before using these functions. Note that if the MCMC simulation has an adaptive phase it will not be possible to make inference using values sampled before the end of this phase.

# See Also

```
BRugs, help.WinBUGS
```

write.datafile

Write data for OpenBUGS - intended for internal use only

# Description

Write data in files that can be read by OpenBUGS - intended for internal use only

# Usage

```
write.datafile(datalist, towhere, fill = TRUE)
formatdata(datalist)
```

# **Arguments**

```
datalist a list to be written into an appropriate structure
towhere the name of the file which the data are to be written to
fill see cat, defaults to TRUE
```

40 write.datafile

# Value

datalist.tofile

A structure appropriate to be read in by OpenBUGS.

# See Also

The main functions to be called by the user are  ${\tt bugsData}$  and  ${\tt bugsInits}$ .

# Index

Tonia IO	madal Cand 10
*Topic IO	modelSeed, 19
samplesCoda, 28	modelSetAP, 19
*Topic datasets	modelUpdate, 20
rats, 25	plotAutoC, 21
*Topic documentation	plotBgr, 21
BRugs, 1	plotDensity, 22
help.WinBUGS, 11	plotHistory, 23
*Topic <b>file</b>	ranks, 24
bugsData, 5	samplesAutoC, 25
bugsInits,6	samplesBgr, 26
samplesCoda, 28	samplesClear, 27
*Topic <b>hplot</b>	samplesCoda, 28
plotAutoC, 21	samplesCorrel, 29
plotBgr, <mark>21</mark>	samplesDensity, 30
plotDensity, 22	samplesGet, 31
plotHistory, 23	samplesHistory, 31
samplesAutoC, 25	samplesMonitors, 32
samplesBgr, 26	samplesSample, 33
samplesDensity, $30$	samplesSet, 33
samplesHistory, 31	samplesSetting, 34
*Topic <b>interface</b>	samplesSize, 35
BRugs, 1	samplesStats, 35
BRugsFit, 3	setValues, 37
buffer, 5	summary, 37
buildMCMC, $6$	*Topic internal
currentValues,7	bgrPoint,4
dic,7	buffer, 5
dimensions, 9	dimensions, 9
getChain, 10	getChain, 10
getNumChains, 10	write.datafile,38
getObj,9	*Topic <b>univar</b>
loadModule, 11	samplesCorrel, 29
modelAdaptivePhase, 12	samplesStats, 35
modelCheck, 12	acf, 21, 26
modelCompile, 13	ac1, 21, 20
modelData, 14	bgrGrid(bgrPoint),4
modelGenInits, 14	bgrPoint, 4
modelInits, 15	BRugs, 1, 4–24, 26–38
modelIteration, 16	BRugsFit, 2, 3
modelModules, 17	buffer, 5
modelNames, 17	buffer, 5 bugsData, 5, 39
modelPrecision, 18	bugsInits, 6, 39
modelSaveState, 18	bugsinits, 0, 39 buildMCMC, 6
moderbavestate, 10	Dullardic, U

INDEX

cat, 38	ranks, 24
currentValues, 7, 37	ranksClear(ranks), 24
, ,	ranksSet (ranks), 24
dic,7	ranksStats (ranks), 24
dicClear(dic), 7	rats, <b>25</b>
dicSet (dic), 7	ratsdata(rats), 25
dicStats,4	ratsinits (rats), 25
dicStats (dic), 7	· //
dimensions, 9	samplesAutoC, 21, 25
	samplesBgr, 4, 22, 26
formatC, 4-6	samplesClear, 27
formatdata(write.datafile), 38	samplesCoda, 28
, , , , , , , , , , , , , , , , , , , ,	samplesCorrel, 29
getChain, 10	samplesDensity, $23,30$
getGraphObj(getObj),9	samplesGet, 31
getNumChains, 10	samplesGetBeg(samplesGet), 31
getObj,9	<pre>samplesGetEnd(samplesGet), 31</pre>
getUpdaterObj(getObj),9	samplesGetFirstChain
	(samplesGet), 31
help.BRugs, 11	samplesGetLastChain(samplesGet),
help.BRugs (BRugs), 1	31
help.WinBUGS, 2, 4, 5, 7-10, 11, 11-24,	samplesGetThin(samplesGet), 31
26–38	samplesHistory, 23, 31
	samplesMonitors, 32
loadModule, 11	samplesSample, 33
,	samplesSet, 33, 38
mcmc, 7	samplesSetBeg, 31
mcmc.list,7	samplesSetBeg(samplesSetting), 34
modelAdaptivePhase, 12	samplesSetEnd(samplesSetting), 34
modelCheck, 12, 13, 14, 18	samplesSetFirstChain
modelCompile, 13, 15	(samplesSetting), 34
modelData, 14	samplesSetLastChain
modelGenInits, 14, 15, 16	(samplesSetting), 34
modelGetSeed (modelSeed), 19	samplesSetThin, 31
modelInits, 15	samplesSetThin(samplesSetting),
modelIteration, 16	34
modelModules, 11, 17	samplesSetting, 34
modelNames, 17	samplesSize, 35
modelPrecision, 18	samplesStats, 4, 35
modelSaveState, 18	setValues, 7, 37
modelSeed, 19	summary, 37
modelSetAP, 19	summaryClear(summary), 37
modelSetIts (modelSetAP), 19	summarySet (summary), 37
modelSetOR (modelSetAP), 19	summaryStats (summary), 37
modelSetSeed (modelSeed), 19	2
modelUpdate, 20	write.datafile,38
par, 21–23, 25, 26, 30, 32	
plot.default, 21-23	
plotAutoC, 21, 25, 26	
plotBgr, 4, 21, 26, 27	
plotDensity, 22, 30	
plotHistory, 23, 32	
<u>_</u> _ / · · · · · · ·	