

Markov chain Monte Carlo Simulation Using the DREAM Software Package: Erratum

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Abstract

In a recent paper published in *Environmental Modeling & Software* (Vrugt, 2016), I introduced a MATLAB toolbox of the Differential Evolution Adaptive Metropolis (DREAM) algorithm developed by Vrugt *et al.* (2008a, 2009a). This algorithm is used widely to solve Bayesian inference problems in fields ranging from physics, chemistry and engineering, to ecology, hydrology, and geophysics. This manuscript serves as an epilogue to the original DREAM toolbox paper published by Vrugt (2016) and documents the changes that have been made to the original MATLAB code. Most of these modifications are in response to users that request new options, whereas others changes involve removal of certain built-in variables, as their use has become obsolete.

Keywords: Bayesian inference, Markov chain Monte Carlo (MCMC) simulation, Differential Evolution Adaptive Metropolis (DREAM) algorithm

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1. ERRATUM

2 All the changes that are documented herein are in reference to the source
code of DREAM described in *Vrugt* (2016). I use two different subsections to
4 detail the changes that have been made. The first sub section introduces new
options that have been added to the source code. The second subsection de-
6 tails the options that have become obsolete and therefore have been removed.
Note, the changes discussed herein have been made to the source codes of
8 all the different algorithms of the DREAM family discussed in Section 7 of
Vrugt (2016).

10 1.1. New options

Upon request of a user of the DREAM_(ZS) algorithm, a new option has
12 been added in December 2015 to the field `initial` of `Par_info`. This field
specifies with a string enclosed between quotes how to sample the initial
14 state of each of the N different Markov chains. The existing four options (1)
'uniform' (2) 'latin' (3) 'normal' and (4) 'prior' are now augmented with (5)
16 'user'. This new option allows the user to explicate the initial state of each
Markov chain. If this new option is exercised then the user has to define in
18 a $N \times d$ matrix the initial state (per row) of each of the N Markov chains.
This matrix is called `x0` and should be stored as field of structure `Par_info`,
20 for instance

$$\text{Par_info.x0} = \text{rand}(N,d) \quad (1)$$

22 where N denotes the number of Markov chains that is used by DREAM to
solve for the target distribution, and d equals the number of parameters of
24 this distribution. The values of these two variables are stored in field `N` and
`d` of structure `DREAMPar`, respectively. The field `x0` is thus a new field of
26 `Par_info` and necessary if `Par_info.initial = 'user'`.

The postprocessor (function `POSTPROC_DREAM`) has been integrated
28 into the main function `DREAM`. This has a few practical advantages. The
user can deactivate screen writing of tables and figures by setting the field

30 `print` of structure `options` to 'no'. This field has been added to the recent
 version of DREAM and allows the user to control the postprocessor. The
 32 default setting of `print` is 'yes'. This setting will be assumed if the user does
 not specify the field `print` (or its content) of input argument `options`.

34 The latest release of DREAM also uses the latex text interpreter in all
 figures printed to the screen. I believe this looks better, but this might
 36 be subjective. What is more, the postprocessor (figures and tables) now
 use the labels x_1, \dots, x_d to indicate each of the dimensions of the posterior
 38 distribution.

A new input argument has been added to the main function of the
 40 DREAM toolbox, namely `DREAM` which allows the user to port other
 information to the plugin model/function. Thus the new function call to
 42 DREAM now becomes

```
44 [chain,output,fx] = DREAM(Func_name,DREAMPar,Par_info,  

46                               Meas_info,options,plugin) (2)
```

where `Func_name` (string), `DREAMPar` (structure array), and `Par_info` (struc-
 48 ture array) are input arguments defined by the user, and `chain` (matrix),
`output` (structure array) and `fx` (matrix) are output variables computed
 50 by DREAM and returned to the user. The last three input arguments of
 the DREAM subroutine (function), `Meas_info`, `options` and `plugin` are
 52 optional.

The content of the different input variables has been defined in the main
 54 manual of DREAM - which has been published in *Vrugt* (2016). Interested
 readers are referred to this publication for further details. The new input
 56 argument `plugin` is optional and allows user to pass to their model (stored
 as `Func_name`) additional information as follows

```
58 Y = model(x,plugin), (3)
```

where Y signifies the likelihood (`DREAMPar.lik = 1`), the log-likelihood (`DREAMPar.lik = 2`), model simulation (`(DREAMPar.lik = {11-17,31-34})`) or vector of summary statistics (`DREAMPar.lik = {21-23}`). The content of `plugin` can be determined by the user. It can be declared a scalar, vector, matrix, structure, string or cell-array, whatever is deemed appropriate for the model plugin.

The revised DREAM codes also allows the user to define the length of the burn-in period that is used to monitor convergence via the scale-reduction diagnostics of *Gelman and Rubin* (1992) and *Brooks and Gelman* (1998). The field `burnin` of structure `options` stipulates the burn-in percentage. The default value of this field is 50, that means that only the second half of the sampled chains is used to compute both scale-reduction convergence diagnostics. A value of `options.burnin = 80` would use only the last 20% of the chain to calculate the \hat{R}_j -diagnostic of *Gelman and Rubin* (1992), where $j = \{1, \dots, d\}$ and the \hat{R}^d -statistic of *Brooks and Gelman* (1998). Furthermore, a new field `logprior` of structure `DREAMPar` enables the user to work directly with a log-density prior rather than the density of `DREAMPar.prior`. The content of `DREAMPar.logprior` is identical to `DREAMPar.prior` except that `logprior` expects the output of the prior distribution to be a log-density rather than density. The use of a log-density is encouraged, to avoid numerical underflow (prior goes to zero) in large-dimensional search spaces.

I also made two recent changes to DREAM_D (discrete estimation) that were not documented in the original published DREAM manual (*Vrugt*, 2016). First, a new field called `combinatorial` has been added to the structure `DREAMPar` and allows the user to specify with 'yes' whether the inference involves combinatorial estimation or not. The default setting of this field is 'no'. The sudoku puzzle in case study 1 of *Vrugt et al.* (2011) constitutes an example of a combinatorial parameter estimation problem, and is part of the DREAM_D software toolbox (see example 24). Another field called `sort` has been added to `DREAMPar` and allows the user to solve discrete problems for which the order of the parameter values does not matter as long as the

"right" values are included in the parameter vector. Measurement selection
 90 is such an example (see case study 2 of *Vrugt et al.* (2011); included as ex-
 92 ample 25 in MATLAB toolbox of DREAM_D) as the order in which the water
 94 retention measurements are selected will not matter for the results of the
 inference (only the measurement set counts). Note, that detailed balance
 cannot be proven if the parameter vector is sorted in ascending order (low
 to high) after a proposal is created. Nevertheless, case studies demonstrate
 96 that the DREAM_D code converges properly.

Finally, and if so desired, the users can request a script for postprocessing
 98 outside the main DREAM programs when `options.print = 'no'`. This script
 will take the output of the program and generate automatically the same
 100 figures as when `options.print = 'yes'`.

1.2. *Obsolete options*

102 The new source code of DREAM automatically detects which hardware
 system is used (PC, Mac or Linux) and automatically configures the compu-
 104 tational environment in MATLAB. Hence, the field `linux` of structure `options`
 has become obsolete and has been removed.

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