

bmp_description

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1 Using bayesmixedlogit and bayesmixedlogitwtp in Stata

This document presents an overview of the bayesmixedlogit and bayesmixedlogitwtp Stata packages. It mirrors closely the helpfile obtainable in stata (i.e., through `help bayesmixedlogit` or `help bayesmixedlogitwtp`). Further background for the packages can be found in [Baker\(2014\)](#).

1.0.1 Description

bayesmixedlogit can be used to fit mixed logit models using Bayesian methods – more precisely, bayesmixedlogit produces draws from the posterior parameter distribution and then presents summary and other statistics describing the results of the drawing. Detailed analysis of the draws is left to the discretion of the user.

Implementation of bayesmixedlogit follows [Train \(2009, chap. 12\)](#), and details of how the algorithm works are described in [Baker\(2014\)](#). A diffuse prior for the mean values of the random coefficients is assumed, and the prior distribution on the covariance matrix of random coefficients is taken to be an identity inverse Wishart. bayesmixedlogit uses the Mata routines `amcmc()` (if not installed, search online; from the Stata command-line prompt, type `ssc install amcmc`) for adaptive Markov chain Monte Carlo sampling from the posterior distribution of individual level coefficients and fixed coefficients. The data setup for bayesmixedlogit is the same as for `clogit` (`ssc install clogit`). Much of the syntax follows that used by [Hole \(2007\)](#) in development of the command `mixlogit`.

1.0.2 Options

`group(varname)` specifies a numeric identifier variable for choice occasions. `group()` is required.

`identifier(varname)` identifies coefficient sets (those observations for which a set of coefficients apply). Thus, when a person is observed making choices over multiple occasions, one would use `group(varname)` to specify the choice occasions, while `identifier(varname)` would identify the person. `identifier()` is required.

`rand(varlist)` specifies independent variables with random coefficients. The variables immediately following the dependent variable in the syntax are considered to have fixed coefficients (see the examples below). While a model can be run without any independent variables with fixed coefficients, at least one random-coefficient independent variable is required for bayesmixedlogit to work. `rand()` is required.

`draws(#)` specifies the number of draws that are to be taken from the posterior distribution of the parameters. The default is `draws(1000)`.

`drawsrandom(#)` is an advanced option. The drawing algorithm treats each set of random coefficients as a Gibbs step in sampling from the joint posterior distribution of parameters. In difficult, large-dimensional problems, it might be desirable to let individual Gibbs steps run for more than one draw to achieve better mixing and convergence of the algorithm.

`drawsfixed(#)` is a more advanced option. The drawing algorithm treats fixed coefficients as a Gibbs step in sampling from the joint posterior distribution of parameters. In difficult, large-dimensional problems, it might be desirable to let this step in Gibbs sampling run for more than a single draw. The default is `drawsfixed(1)`.

`burn(#)` specifies the length of the burn-in period; the first # draws are discarded upon completion.

`thin(#)` specifies that only every #th draw is to be retained, so if `thin(3)` is specified, only every third draw is retained. `burn` and `thin` can be applied together, which randomly mixes draws. Both options may be applied.

`araterandom(#)` specifies the desired acceptance rate for random coefficients and should be a number between 0 and 1. If the desired acceptance rate is too low, the user has some control over adaptation of the algorithm to the problem.

`aratefixed(#)` specifies the desired acceptance rate for fixed coefficients and works in the same way as `araterandom`.

`samplerrandom(string)` specifies the type of sampler that is to be used when random parameters are drawn. The options are `global` (the default), `mwg` (an acronym for "Metropolis within Gibbs"), and `gibbs`. If `mwg` is specified, the Gibbs step is used when random parameters are drawn all at once. If `mwg` is specified, the Gibbs step is used when random parameters are drawn all at once. The default is `samplerrandom(global)`, but `mwg` might be useful in situations in which the acceptance rate is low.

`samplerfixed(string)` specifies the type of sampler that is used when fixed parameters are drawn. The options are `global` (the default), `mwg` (an acronym for "Metropolis within Gibbs"), and `gibbs`.

`dampparmfixed(#)` works exactly as option `dampparmrandom(#)` but is applied to drawing fixed parameters.

`dampparmrandom(#)` is a parameter that controls how aggressively the proposal distributions for random parameters are updated. The default is 1.0. If the parameter is set to 0.5, the proposal distributions are updated less aggressively. If the parameter is set to 0.1, the proposal distributions are updated very aggressively.

`from(rowvector)` specifies a row vector of starting values for all parameters in order. If these values are not specified, the algorithm will use default starting values.

`fromvariance(matrix)` specifies a matrix of starting values for the random parameters. The matrix should be a square matrix with the same number of rows as the number of random parameters.

`jumble` specifies to randomly mix draws.

`noisy` specifies that a dot be produced every time a complete pass through the algorithm is finished. This is useful for monitoring the progress of the algorithm. While `ln_fc(p)` is not an objective function, it is useful for monitoring the progress of the algorithm. While `ln_fc(p)` is not an objective function, it is useful for monitoring the progress of the algorithm. While `ln_fc(p)` is not an objective function, it is useful for monitoring the progress of the algorithm.

`saving(filename)` specifies a location to store the draws from the distribution. The file will be created if it does not exist.

`replace` specifies that an existing file is to be overwritten.

`append` specifies that an existing file is to be appended, which might be useful if multiple runs are being saved.

`indsave(filename)` specifies a file to which individual-level random parameters are to be saved. If the number of individuals is large, specifying this option can cause memory problems. Users should be cautious when using this option.

`indkeep(#)` is for use with `indsave` and specifies that only the last `#` draws of the individual-level

`indwide` is for use with `indsave` and affords the user a degree of control over how individual-level data is saved in a row, where draws are marked by the group identifier. If instead the user would like to save with the first entry of the row being the group identifier. By analogy with `reshape`, by default

`replaceind` functions in the same way as `replace`, but in reference to the file specified in `indsave`

`appendind` functions in the same way as `append`, but in reference to the file specified in `indsave`

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