

# Package ‘BayMor’

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**Type** Package

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**Title** Bayesian Multivariate Ordinal Regression

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## Description

This package has been created to perform a multivariate ordinal regression using a Bayesian approach. In cases that the response variables of a system are ordinal in nature, (e.g. Likert items), it returns the Bayesian posterior density distributions of the coefficients of explanatory variables in a multivariate regression. After obtaining the distributions in R, a user can simply calculate 95% Highest Density Interval (HDI) for each of the coefficients and if the interval does not include zero, the respective explanatory variable is statistically significant. There are two functions available in this package as follows: “mor” and “rtruncnorm”.

**Repository** CRAN

**License** GPL-2

**Needs Compilation** yes

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<code>mor</code>	Bayesian Multivariate Ordinal Regression
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## Description

## Returns

## Usage

`mor(Y, l, X, nx, Nit)`

## Arguments

- Y*       $N \times m$  matrix of responses
- l*      Number of levels responses in each dimension
- X*      List of explanatory variables of each dimension
- nx*     Number of explanatory variables in each dimension
- Nit*    Number of iteration of the MCMC

## Guideline

Consider the following multivariate regression example in which a user wants to investigate the effect of the explanatory variables on the ordinal response variables. These are the steps that he/she should take to use “mor” function of the package.

$$\hat{y}_1 = b_{01} + b_{11}x_{11} + b_{21}x_{21} + \cdots + b_{r1}x_{r1} \quad (1)$$

$$\hat{y}_2 = b_{02} + b_{12}x_{12} + b_{22}x_{22} + \cdots + b_{r2}x_{r2} \quad (2)$$

$$\vdots$$

$$\hat{y}_m = b_{0m} + b_{1m}x_{1m} + b_{2m}x_{2m} + \cdots + b_{rm}x_{rm} \quad (m)$$

*Step 1:* Create a  $N \times m$  matrix of responses, *Y*, where *N* is the number of observations and *m* is the number of dimension. Please note that the first column of the matrix must correspond to the first set of observations, equation 1, and the last column to the last set of observations, equation m.

*Step 2:* Create a vector “ $l$ ” for the number of levels response variables in equations 1 through  $m$ . The first element of the resulting vector must correspond to equation 1 and the last element to equation  $m$ .

*Step 3:* For explanatory variables of each set of observations, create a matrix,  $X_1, X_2, \dots, X_m$ .  $X_1$  refers to the matrix of explanatory variables for the first equation and  $X_m$  is the matrix of explanatory variables for the  $m$  equation.

*Step 4:* Create a list with  $X_1, X_2, \dots, X_m$  as  $X = \text{list}(X_1, X_2, \dots, X_m)$ .

*Step 5:* Create a vector for number of explanatory variables in each of the equations 1 through  $m$ . (i.e.  $nx = c(r1, r2, \dots, rm)$ ).

*Step 6:* Determine  $Nit$ , the number of iteration of the MCMC.

*Step 7:* run the function, determine the burn-in steps, and then either find HDI, or plot the density

### **Example**

```
> Mybeta <- mor(Y, l, X, nx, 5000)
> plot(density(Mybeta[500:5000,]))
```

OR

```
> hid(Mybeta[500:5000,])           #Requires HDInterval package
```

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rtruncnorm

Drawing random samples from a truncated normal distribution

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This function generate random numbers from a normal distribution with mean  $\mu$  and standard deviation  $\sigma$  that is truncated to the lower bound  $l$  and the upper bound  $up$ .

### Usage

`rtruncnorm( $\mu, \sigma, l, up$ )`

### Arguments

$\mu$  The mean of the normal distribution

$\sigma$  The standard deviation of the normal distribution

$l$  The lower bound

$up$  The upper bound

### Example

```
> rtruncnorm (1,1,0,1)
```