

# Package ‘saeHB.ZIB’

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**Type** Package

**Title** Small Area Estimation under ZIB Distribution using Hierarchical Bayesian

**Version** 0.1.0

**Author** Rizqina Rahmati [aut, cre], Azka Ubaidillah [aut]

**Maintainer** Rizqina Rahmati <221810583@stis.ac.id>

**Description** Provides function for area level of small area estimation using hierarchical Bayesian (HB) method with Zero-Inflated Binomial (ZIB) distribution for variables of interest. Some dataset produced by a data generation are also provided. The 'rjags' package is employed to obtain parameter estimates. Model-based estimators involves the HB estimators which include the mean and the variation of mean. For references, see Wiley (2013) <[doi:10.1111/cdoe.12044](https://doi.org/10.1111/cdoe.12044)>.

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.1.2

**URL** <https://github.com/rizqinaR/saeHB.ZIB>

**BugReports** <https://github.com/rizqinaR/saeHB.ZIB/issues>

**Imports** stringr, coda, rjags, stats, grDevices, graphics

**Suggests** rmarkdown, knitr

**VignetteBuilder** knitr

**Depends** R (>= 2.10)

**NeedsCompilation** no

**Repository** CRAN

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## R topics documented:

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| dataZIB | <i>Sample Data for Small Area Estimation using Hierarchical Bayesian Method under Zero-Inflated Binomial Distribution</i> |
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### Description

Dataset to simulate Small Area Estimation using Hierarchical Bayesian Method under Zero-Inflated Binomial distribution

This data is generated by these following steps:

1. Generate sampling random area effect  $u.Z$  and  $u.nZ$  with  $(u.Z \sim N(0, 1))$  and  $(u.nZ \sim N(0, 1))$ .  
The auxiliary variabls are generated by Uniform distribution with  $(x1 \sim U(0, 1))$  and  $(x2 \sim U(0, 1))$ .  
The coefficient parameters  $\alpha_0, \alpha_1, \alpha_2, \beta_0, \beta_1, \beta_2$  are set as 0.
2. Calculate  $logit(p) = \alpha_0 + \alpha_1 * x1 + \alpha_2 * x2 + u.Z$  and  $logit(\pi) = \beta_0 + \beta_1 * x1 + \beta_2 * x2 + u.nZ$
3. Generate number of sample with  $n.samp \sim U(10, 30)$
4. Generate  $delta \sim bernoulli(p)$  and  $y_s, tar \sim binomial(s, \pi)$
5. calculate  $y = delta * y_s, tar$
6. Calculate variance of direct estimates (vardir) with  $var(y) = (1-p) * s * pi * (1-p * (1-p * s))$
7. Auxiliary variables  $x1, x2$ , direct estimation ( $y$ ), vardir, and  $s$  are combined in a dataframe called dataZIB

### Usage

```
data(dataZIB)
```

### Format

A data frame with 64 observations on the following 4 variables:

**y** Direct Estimation of y

**x1** Auxiliary variable of x1

**x2** Auxiliary variable of x2

**vardir** sampling variance of y

**n.samp** number of sample

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|            |   |
|------------|---|
| dataZIB.ns | <i>Sample Data for Small Area Estimation using Hierarchical Bayesian Method under Zero-Inflated Binomial Distribution</i> |
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### Description

Dataset to simulate Small Area Estimation using Hierarchical Bayesian Method under Zero-Inflated Binomial distribution with non-sampled areas

This data contains NA values that indicates no sampled at one or more small areas. It uses the dataZIB.ns with the direct estimates and the related variances in 3 small areas are missing.

### Usage

```
data(dataZIB.ns)
```

### Format

A data frame with 30 rows and 4 variables :

**y** Direct Estimation of y

**x1** Auxiliary variable of x1

**x2** Auxiliary variable of x2

**vardir** sampling variance of y

**n.samp** number of sample

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|            |  |
|------------|--|
| ziBinomial | <i>Small Area Estimation using Hierarchical Bayesian under Zero Inflated Binomial Distribution</i> |
|------------|--|

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### Description

This function is implemented to variable of interest ( $y$ ) that assumed to be a Zero Inflated Binomial Distribution. The range of data is ( $0 < y < \infty$ ). This model can be used to handle overdispersion and excess zero in data.

### Usage

```
ziBinomial(  
  formula,  
  n.samp,  
  iter.update = 3,  
  iter.mcmc = 10000,  
  coef.nonzero,  
  var.coef.nonzero,
```

```

coef.zero,
var.coef.zero,
thin = 2,
burn.in = 2000,
tau.u.nZ = 1,
tau.u.Z = 1,
data
)

```

### Arguments

|                  |  |
|------------------|--|
| formula          | Formula that describe the fitted model   |
| n.samp           | Number of sample in each area  |
| iter.update      | Number of updates with default 3   |
| iter.mcmc        | Number of total iterations per chain with default 2000   |
| coef.nonzero     | Optional argument for mean on coefficient's prior distribution or $\beta$ 's prior distribution which value is non-zero  |
| var.coef.nonzero | Optional argument for the variances of the prior distribution of the model coefficients ( $\beta$ )                      |
| coef.zero        | Optional argument for mean on coefficient's prior distribution or $\alpha$ 's prior distribution which value is non-zero |
| var.coef.zero    | Optional argument for the variances of the prior distribution of the model coefficients ( $\alpha$ )                     |
| thin             | Thinning rate, must be a positive integer with default 1   |
| burn.in          | Number of iterations to discard at the beginning with default 1000   |
| tau.u.nZ         | Variance of random effect area for non-zero of variable interest ( $y$ ) with default 1                                  |
| tau.u.Z          | Variance of random effect area for zero of variable interest ( $y$ ) with default 1                                      |
| data             | The data frame   |

### Value

This function returns a list of the following objects:

|             |  |
|-------------|--|
| Est         | A vector with the values of Small Area mean Estimates using Hierarchical bayesian method |
| refVar      | Estimated random effect variances  |
| coefficient | A dataframe with the estimated model coefficient   |
| plot_alpha  | Trace, Density, Autocorrelation Function Plot of MCMC samples                            |
| plot_beta   | Trace, Density, Autocorrelation Function Plot of MCMC samples                            |

**Examples**

```

#Compute Fitted Model
y ~ x1 +x2

# For data without any nonsampled area
# Load Dataset
data(dataZIB)
saeHB.ZIB <- ziBinomial(formula = y~x1+x2, iter.update=3, iter.mcmc = 1000,
                        burn.in = 200,data = dataZIB)
#the setting of iter.update, iter.mcmc, and burn.in in this example
#is considered to make the example execution time be faster.
#Result
saeHB.ZIB$Est                                #Small Area mean Estimates
saeHB.ZIB$sd                                  #Standard deviation of Small Area Mean Estimates
saeHB.ZIB$refVar                              #refVar
saeHB.ZIB$coefficient                         #coefficient
#Load Library 'coda' to execute the plot
#autocorr.plot(saeHB.ZIB$plot[[3]]) is used to #ACF Plot for alpha
#autocorr.plot(saeHB.ZIB$plot[[3]]) is used to #ACF Plot for alpha
#plot(saeHB.ZIB$plot_alpha[[3]]) is used to   #Dencity and trace plot for alpha
#plot(saeHB.ZIB$plot_beta[[3]]) is used to   #Dencity and trace plot for beta

```

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