

# Package ‘ZIBR’

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**Type** Package

**Title** A Zero-Inflated Beta Random Effect Model

**Version** 1.0.2

**Date** 2023-10-16

**Description** A two-part zero-inflated Beta regression model with random effects (ZIBR) for testing the association between microbial abundance and clinical covariates for longitudinal microbiome data. Eric Z. Chen and Hongzhe Li (2016) <[doi:10.1093/bioinformatics/btw308](https://doi.org/10.1093/bioinformatics/btw308)>.

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**LazyData** TRUE

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**Suggests** betareg, dplyr, lme4 (>= 1.1-34), nlme, knitr, rmarkdown, testthat (>= 3.0.0)

**Config/testthat/edition** 3

**Depends** R (>= 2.10), statmod

**VignetteBuilder** knitr

**URL** <https://github.com/PennChopMicrobiomeProgram/ZIBR>

**BugReports** <https://github.com/PennChopMicrobiomeProgram/ZIBR/issues>

**NeedsCompilation** no

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fit\_beta\_random\_effect

*Fit beta random effect*

---

### Description

Fit beta random effect

### Usage

```
fit_beta_random_effect(
  Z = Z,
  Y = Y,
  subject.ind = subject.ind,
  time.ind = time.ind,
  quad.n = 30,
  verbose = FALSE
)
```

### Arguments

Z	FILL
Y	FILL
subject.ind	the subject index
time.ind	the time index
quad.n	number of points in gaussian quadrature
verbose	a boolean to enable more output

### Value

a named list

- est.table
- s2.est
- v.est

---

fit\_logistic\_random\_effect  
*Fit logistic random effect*

---

**Description**

Fit logisitic random effect

**Usage**

```
fit_logistic_random_effect(  
  X = X,  
  Y = Y,  
  subject.ind = subject.ind,  
  time.ind = time.ind,  
  quad.n = 30,  
  verbose = FALSE  
)
```

**Arguments**

X	FILL
Y	FILL
subject.ind	the subject index
time.ind	the time index
quad.n	number of points in gaussian quadrature
verbose	a boolean to enable more output

**Value**

a named list

- est.table
- s1.est

---

fit\_zero\_inflated\_beta\_random\_effect  
*Fit zero inflated beta random effect*

---

**Description**

Fit zero inflated beta random effect

**Usage**

```
fit_zero_inflated_beta_random_effect(  
  X = X,  
  Z = Z,  
  Y = Y,  
  subject_ind = subject_ind,  
  time_ind = time_ind,  
  component_wise_test = TRUE,  
  joint_test = TRUE,  
  quad_n = 30,  
  verbose = FALSE  
)
```

**Arguments**

X	FILL
Z	FILL
Y	FILL
subject_ind	the subject index
time_ind	the time index
component_wise_test	boolean to run component-wise test
joint_test	boolean to run joint test
quad_n	number of points in gaussian quadrature
verbose	a boolean to enable more output

**Value**

a named list

- logistic\_est\_table
- logistic\_s1\_est
- beta\_est\_table
- beta\_s2\_est
- beta\_v\_est
- loglikelihood
- joint\_p

---

ibd

*Longitudinal human microbiome data*

---

### Description

A dataset containing the bacterial abundance and clinical information from a longitudinal human microbiome study

### Usage

ibd

### Format

A data frame with 236 rows and 5 variables:

**Sample** Sample IDs

**Subject** Subject IDs

**Time** Time points

**Treatment** Treatment, 0 for antiTNF, 1 for EEN

**Abundance** Abundance for Eubacterium ...

### References

Lewis and Chen et al. (2016) Cell Host & Microbe 18 (4), 489-500

---

simulate\_beta\_random\_effect\_data

*Simulate beta data*

---

### Description

Simulate beta data

### Usage

```
simulate_beta_random_effect_data(  
  subject_n = 50,  
  time_n = 5,  
  v = 2,  
  beta = as.matrix(c(-0.5, -0.5, 0.5)),  
  Z = NA,  
  s2 = 1,  
  sim_seed = 100  
)
```

**Arguments**

subject_n	the number of subjects
time_n	the number of time points
v	FILL
beta	FILL
Z	FILL
s2	FILL
sim_seed	the random seed with which to simulate the data

**Value**

a named list

- Y
- Z
- c
- u
- v
- beta
- s2
- subject\_ind
- time\_ind

---

simulate\_logistic\_data

*Simulate logistic data*

---

**Description**

Simulate logistic data

**Usage**

```
simulate_logistic_data(  
  subject_n = 50,  
  time_n = 5,  
  alpha = as.matrix(c(0, 0.5, -1)),  
  s1 = 0.5,  
  sim_seed = 100  
)
```

**Arguments**

subject_n	the number of subjects
time_n	the number of time points
alpha	FILL
s1	FILL
sim_seed	the random seed with which to simulate the data

**Value**

a named list

- X
- Y
- b
- subject\_ind
- time\_ind

---

simulate\_zero\_inflated\_beta\_random\_effect\_data

*Simulate data according to zero-inflated beta random effects model*

---

**Description**

Simulate data according to zero-inflated beta random effects model

**Usage**

```
simulate_zero_inflated_beta_random_effect_data(  
  subject_n = 50,  
  time_n = 5,  
  v = 2,  
  alpha = as.matrix(c(0, 0.5, -1)),  
  beta = as.matrix(c(-0.5, -0.5, 0.5)),  
  X = NA,  
  Z = NA,  
  s1 = 0.2,  
  s2 = 0.2,  
  sim_seed = 100  
)
```

**Arguments**

subject_n	number of subjects
time_n	number of time points for each subject
v	the dispersion parameter in beta component
alpha	the coefficients in logistic component
beta	the coefficients in beta component
X	the covariates in logistic component
Z	the covariates in beta component
s1	the standard deviation of random effect in logistic component
s2	the standard deviation of random effect in beta component
sim_seed	the random seed

**Value**

a named list

- Y the bacterial abundance generated from the model
- X the covariates in logistic component
- Z the covariates in beta component
- alpha the coefficients in logistic component
- beta the coefficients in beta component
- s1 the standard deviation of random effect in logistic component
- s2 the standard deviation of random effect in beta component
- subject\_ind the IDs for each subject
- time\_ind time points

**Examples**

```
simulate_zero_inflated_beta_random_effect_data(  
  subject_n = 100, time_n = 5,  
  X = as.matrix(c(rep(0, 50 * 5), rep(1, 50 * 5))),  
  alpha = as.matrix(c(-0.5, 1)),  
  beta = as.matrix(c(-0.5, 0.5)),  
  s1 = 1, s2 = 0.8,  
  v = 5,  
  sim_seed = 100  
)
```

---

zibr *Fit zero-inflated beta regression with random effects*

---

### Description

Fit zero-inflated beta regression with random effects

### Usage

```
zibr(
  logistic_cov,
  beta_cov,
  Y,
  subject_ind,
  time_ind,
  component_wise_test = TRUE,
  quad_n = 30,
  verbose = FALSE
)
```

### Arguments

logistic_cov	the covariates in logistic component
beta_cov	the covariates in beta component
Y	the response variable in the regression model
subject_ind	the variable for subject IDs
time_ind	the variable for time points
component_wise_test	whether to perform component wise test. If true, ZIBR will calculate p-values for logistic and beta component respectively.
quad_n	Gaussian quadrature points
verbose	print the fitting process

### Value

a named list

- logistic\_est\_table - the estimated coefficients for logistic component.
- logistic\_s1\_est - the estimated standard deviation for the random effect in the logistic component.
- beta\_est\_table - the estimated coefficients for logistic component.
- beta\_s2\_est - the estimated standard deviation for the random effect in the beta component.
- beta\_v\_est - the estimated dispersion parameter in the beta component.
- loglikelihood - the log likelihood of fitting ZIBR model on the data.
- joint\_p - the p-values for jointly testing each covariate in both logistic and beta component.

**Examples**

```
## simulate some data
sim <- simulate_zero_inflated_beta_random_effect_data(
  subject_n = 100, time_n = 5,
  X = as.matrix(c(rep(0, 50 * 5), rep(1, 50 * 5))),
  Z = as.matrix(c(rep(0, 50 * 5), rep(1, 50 * 5))),
  alpha = as.matrix(c(-0.5, 1)),
  beta = as.matrix(c(-0.5, 0.5)),
  s1 = 1, s2 = 0.8,
  v = 5,
  sim_seed = 100
)

## run zibr on the simulated data
zibr_fit <- zibr(
  logistic_cov = sim$X, beta_cov = sim$Z, Y = sim$Y,
  subject_ind = sim$subject_ind, time_ind = sim$time_ind
)

zibr_fit
```

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