## Package 'ZIBR'

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Type Package

Title A Zero-Inflated Beta Random Effect Model

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**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>.

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**Encoding** UTF-8

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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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## **R** topics documented:

```
fit_beta_random_effect
fit_logistic_random_effect
fit_zero_inflated_beta_random_effect
ibd
simulate_beta_random_effect_data
simulate_logistic_data
simulate_zero_inflated_beta_random_effect_data
zibr
```

11

fit\_beta\_random\_effect

Fit beta random effect

## **Description**

Fit beta random effect

## Usage

Index

```
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

```
\verb|fit_logistic_random_effect|\\
```

Fit logisitic 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

## 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

Χ **FILL** FILL Ζ Υ **FILL** the subject index subject\_ind  ${\tt time\_ind}$ the time index  ${\tt component\_wise\_test}$ boolean to run component-wise test joint\_test boolean to run joint test number of points in gaussian quadrature quad\_n 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 5

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

s1 FILL

s2 FILL

s2 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
Χ	the covariates in logistic component
Z	the covariates in beta component
s1	the stardard deviation of random effect in logistic component
s2	the stardard 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 stardard deviation of random effect in logistic component
- s2 the stardard 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 9

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.

10 zibr

## **Examples**

```
## simulate some data
sim <- simulate_zero_inflated_beta_random_effect_data(</pre>
  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(</pre>
 logistic_cov = sim$X, beta_cov = sim$Z, Y = sim$Y,
  subject_ind = sim$subject_ind, time_ind = sim$time_ind
)
zibr_fit
```

# **Index**