

# Package ‘IFAA’

April 12, 2022

**Title** Robust Inference for Absolute Abundance in Microbiome Analysis

**Version** 1.0.6

**Description** IFAA is a robust approach to make inference on the association of covariates with the absolute abundance (AA) of microbiome in an ecosystem. It can be also directly applied to relative abundance (RA) data to make inference on AA because the ratio of two RA is equal ratio of their AA. This algorithm can estimate and test the associations of interest while adjusting for potential confounders. High-dimensional covariates are handled with regularization. The estimates of this method have easy interpretation like a typical regression analysis. High-dimensional covariates are handled with regularization and it is implemented by parallel computing. False discovery rate is automatically controlled by this approach. Zeros do not need to be imputed by a positive value for the analysis. The IFAA package also offers the 'MZILN' function for estimating and testing associations of abundance ratios with covariates.

**License** GNU General Public License version 2

**Encoding** UTF-8

**URL** <https://github.com/gitlzg/IFAA>,  
<https://arxiv.org/abs/1909.10101v3>,  
<https://link.springer.com/article/10.1007/s12561-018-9219-2>

**LazyData** true

**RoxygenNote** 7.1.2

**Depends** R (>= 3.6.0),

**Imports** qIcMatrix (>= 0.9.7), methods (>= 3.3.0), mathjaxr (>= 1.0-1),  
expm (>= 0.999-3), foreach (>= 1.4.3),  
Matrix (>= 1.4-0), HDCI (>= 1.0-2), parallel (>= 3.3.0),  
doParallel (>= 1.0.11), future (>= 1.12.0), glmnet, stats

**RdMacros** mathjaxr

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

**NeedsCompilation** no

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**R topics documented:**

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dataC	<i>Sample covariates data</i>
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**Description**

A dataset ontains 3 covariates.

**Usage**

dataC

**Format**

A data frame with 40 rows and 4 variables:

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dataM	<i>Sample microbiome data</i>
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**Description**

A dataset contains 60 taxa with absolute abundances and these are gut microbiome.

**Usage**

dataM

**Format**

A data frame with 40 rows and 61 variables:

IFAA

*Robust association identification and inference for absolute abundance in microbiome analyses***Description**

Make inference on the association of microbiome with covariates

**Usage**

```
IFAA(
  MicrobData,
  CovData,
  linkIDname,
  testCov = NULL,
  ctrlCov = NULL,
  testMany = TRUE,
  ctrlMany = FALSE,
  nRef = 40,
  nRefMaxForEsti = 2,
  refTaxa = NULL,
  adjust_method = "BY",
  fdrRate = 0.15,
  paraJobs = NULL,
  bootB = 500,
  standardize = FALSE,
  sequentialRun = FALSE,
  refReadsThresh = 0.2,
  taxDropThresh = 0,
  SDThresh = 0.05,
  SDquantilThresh = 0,
  balanceCut = 0.2,
  seed = 1
)
```

**Arguments**

MicrobData	Microbiome data matrix containing microbiome absolute abundance or relative abundance with each row per sample and each column per taxon/OTU/ASV (or any other unit). It should contain an id variable to be linked with the id variable in the covariates data: CovData. This argument can take directory path. For example, MicrobData="C://...//microbiomeData.tsv".
CovData	Covariates data matrix containing covariates and confounders with each row per sample and each column per variable. Any categorical variable should be converted into dummy variables in this data matrix unless it can be treated as a continuous variable. It should also contain an id variable to be linked with the id variable in the microbiome data: MicrobData. This argument can take directory path. For example, CovData = "C://...//covariatesData.tsv".
linkIDname	The common variable name of the id variable in both MicrobData and CovData. The two data sets will be merged by this id variable.

testCov	Covariates that are of primary interest for testing and estimating the associations. It corresponds to $X_i$ in the equation. Default is NULL which means all covariates are testCov.
ctrlCov	Potential confounders that will be adjusted in the model. It corresponds to $W_i$ in the equation. Default is NULL which means all covariates except those in testCov are adjusted as confounders.
testMany	This takes logical value TRUE or FALSE. If TRUE, the testCov will contain all the variables in CovData provided testCov is set to be NULL. The default value is TRUE which does not do anything if testCov is not NULL.
ctrlMany	This takes logical value TRUE or FALSE. If TRUE, all variables except testCov are considered as control covariates provided ctrlCov is set to be NULL. The default value is FALSE.
nRef	The number of randomly picked reference taxa used in phase 1. Default number is 40.
nRefMaxForEsti	The maximum number of final reference taxa used in phase 2. The default is 2.
refTaxa	A vector of taxa or OTU or ASV names. These are reference taxa specified by the user to be used in phase 1. If the number of reference taxa is less than 'nRef', the algorithm will randomly pick extra reference taxa to make up 'nRef'. The default is NULL since the algorithm will pick reference taxa randomly.
adjust_method	The adjusting method for p value adjustment. Default is "BY" for dependent FDR adjustment. It can take any adjustment method for p.adjust function in R.
fdrRate	The false discovery rate for identifying taxa/OTU/ASV associated with testCov. Default is 0.15.
paraJobs	If sequentialRun is FALSE, this specifies the number of parallel jobs that will be registered to run the algorithm. If specified as NULL, it will automatically detect the cores to decide the number of parallel jobs. Default is NULL.
bootB	Number of bootstrap samples for obtaining confidence interval of estimates in phase 2 for the high dimensional regression. The default is 500.
standardize	This takes a logical value TRUE or FALSE. If TRUE, the design matrix for X will be standardized in the analyses and the results. Default is FALSE.
sequentialRun	This takes a logical value TRUE or FALSE. Default is FALSE. This argument could be useful for debug.
refReadsThresh	The threshold of proportion of non-zero sequencing reads for choosing the reference taxon in phase 2. The default is 0.2 which means at least 20% non-zero sequencing reads.
taxDropThresh	The threshold of number of non-zero sequencing reads for each taxon to be dropped from the analysis. The default is 0 which means taxon without any sequencing reads will be dropped from the analysis.
SDThresh	The threshold of standard deviations of sequencing reads for been chosen as the reference taxon in phase 2. The default is 0.05 which means the standard deviation of sequencing reads should be at least 0.05 in order to be chosen as reference taxon.
SDquantilThresh	The threshold of the quantile of standard deviation of sequencing reads, above which could be selected as reference taxon. The default is 0.
balanceCut	The threshold of the proportion of non-zero sequencing reads in each group of a binary variable for choosing the final reference taxa in phase 2. The default number is 0.2 which means at least 20% non-zero sequencing reads in each group are needed to be eligible for being chosen as a final reference taxon.

seed Random seed for reproducibility. Default is 1. It can be set to be NULL to remove seeding.

### Details

Most of the time, users just need to feed the first five inputs to the function: MicrobData, CovData, linkIDname, testCov and ctrlCov. All other inputs can just take their default values. To model the association, the following equation is used:

$$\log(\mathcal{Y}_i^k) | \mathcal{Y}_i^k > 0 = \beta^{0k} + X_i^T \beta^k + W_i^T \gamma^k + Z_i^T b_i + \epsilon_i^k, \quad k = 1, \dots, K + 1$$

where

- $\mathcal{Y}_i^k$  is the AA of taxa  $k$  in subject  $i$  in the entire ecosystem.
- $X_i$  is the covariate matrix.
- $W_i$  is the confounder matrix.
- $Z_i$  is the design matrix for random effects.
- $\beta^k$  is the regression coefficients that will be estimated and tested with the IFAA() function.

The challenge in microbiome analysis is that  $\mathcal{Y}_i^k$  can not be observed. What is observed is its small proportion:  $Y_i^k = C_i \mathcal{Y}_i^k$ , where  $C_i$  is an unknown number between 0 and 1 that denote the observed proportion.

The IFAA method can successfully address this challenge. The IFAA() will estimate the parameter  $\beta^k$  and their 95% confidence intervals. High-dimensional  $X_i$  is handled by regularization.

### Value

A list containing the estimation results.

- sig\_results: A list containing estimating results that are statistically significant.
- full\_results: A list containing all estimating results. NA denotes unestimable.
- covariatesData: A dataset containing covariates and confounders used in the analyses.

### References

Li et al.(2021) IFAA: Robust association identification and Inference For Absolute Abundance in microbiome analyses. Journal of the American Statistical Association

Zhang CH (2010) Nearly unbiased variable selection under minimax concave penalty. Annals of Statistics. 38(2):894-942.

Liu et al.(2020) A bootstrap lasso + partial ridge method to construct confidence intervals for parameters in high-dimensional sparse linear models. Statistica Sinica

### Examples

```
data(dataM)
dim(dataM)
dataM[1:5, 1:8]
data(dataC)
dim(dataC)
dataC[1:5, ]

results <- IFAA(MicrobData = dataM,
```

```

CovData = dataC,
linkIDname = "id",
testCov = c("v1", "v2"),
ctrlCov = c("v3"),
fdrRate = 0.15)

```

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MZILN

*Conditional regression for microbiome analysis based on multivariate zero-inflated logistic normal model*


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### Description

For estimating and testing the associations of abundance ratios with covariates.

### Usage

```

MZILN(
  MicrobData,
  CovData,
  linkIDname,
  targetTaxa = NULL,
  refTaxa,
  allCov = NULL,
  adjust_method = "BY",
  fdrRate = 0.15,
  paraJobs = NULL,
  bootB = 500,
  taxDropThresh = 0,
  standardize = FALSE,
  sequentialRun = TRUE,
  seed = 1
)

```

### Arguments

MicrobData	Microbiome data matrix containing microbiome absolute abundance or relative abundance with each row per sample and each column per taxon/OTU/ASV (or any other unit). It should contain an id variable to be linked with the id variable in the covariates data: CovData. This argument can take directory path. For example, MicrobData="C://...//microbiomeData.tsv".
CovData	Covariates data matrix containing covariates and confounders with each row per sample and each column per variable. Any categorical variable should be converted into dummy variables in this data matrix unless it can be treated as a continuous variable. It should also contain an id variable to be linked with the id variable in the microbiome data: MicrobData. This argument can take directory path. For example, CovData="C://...//covariatesData.tsv".

linkIDname	The common variable name of the id variable in both MicrobData and CovData. The two data sets will be merged by this id variable.
targetTaxa	The numerator taxa names specified by users for the targeted ratios. Default is NULL in which case all taxa are numerator taxa (except the taxa in the argument 'refTaxa').
refTaxa	Denominator taxa names specified by the user for the targeted ratios. This could be a vector of names.
allCov	All covariates of interest (including confounders) for estimating and testing their associations with the targeted ratios. Default is 'NULL' meaning that all covariates in covData are of interest.
adjust_method	The adjusting method for p value adjustment. Default is "BY" for dependent FDR adjustment. It can take any adjustment method for p.adjust function in R.
fdrRate	The false discovery rate for identifying taxa/OTU/ASV associated with allCov. Default is 0.15.
paraJobs	If sequentialRun is FALSE, this specifies the number of parallel jobs that will be registered to run the algorithm. If specified as NULL, it will automatically detect the cores to decide the number of parallel jobs. Default is NULL.
bootB	Number of bootstrap samples for obtaining confidence interval of estimates for the high dimensional regression. The default is 500.
taxDropThresh	The threshold of number of non-zero sequencing reads for each taxon to be dropped from the analysis. The default is 0 which means taxon without any sequencing reads will be dropped from the analysis.
standardize	This takes a logical value TRUE or FALSE. If TRUE, the design matrix for X will be standardized in the analyses and the results. Default is FALSE.
sequentialRun	This takes a logical value TRUE or FALSE. Default is TRUE. It can be set to be "FALSE" to increase speed if there are multiple taxa in the argument 'refTaxa'.
seed	Random seed for reproducibility. Default is 1. It can be set to be NULL to remove seeding.

## Details

Most of the time, users just need to feed the first six inputs to the function: MicrobData, CovData, linkIDname, targetTaxa, refTaxa and allCov. All other inputs can just take their default values. The regression model for MZILN() can be expressed as follows:

$$\log\left(\frac{\mathcal{Y}_i^k}{\mathcal{Y}_i^{K+1}}\right) | \mathcal{Y}_i^k > 0, \mathcal{Y}_i^{K+1} > 0 = \alpha^{0k} + \mathcal{X}_i^T \alpha^k + \epsilon_i^k, \quad k = 1, \dots, K$$

where

- $\mathcal{Y}_i^k$  is the AA of taxa  $k$  in subject  $i$  in the entire ecosystem.
- $\mathcal{Y}_i^{K+1}$  is the reference taxon (specified by user).
- $\mathcal{X}_i$  is the covariate matrix for all covariates including confounders.
- $\alpha^k$  is the regression coefficients along with their 95% confidence intervals that will be estimated by the MZILN() function.

High-dimensional  $X_i$  is handled by regularization.

**Value**

A list containing the estimation results.

`##` - `targettaxa_result_list`: A list containing estimating results for the targeted ratios. Only available when `targetTaxa` is non-empty.

- `sig_results`: A list containing estimating results for all significant ratios
- `covariatesData`: A dataset containing all covariates used in the analyses.

**References**

Li et al.(2018) Conditional Regression Based on a Multivariate Zero-Inflated Logistic-Normal Model for Microbiome Relative Abundance Data. *Statistics in Biosciences* 10(3): 587-608

Zhang CH (2010) Nearly unbiased variable selection under minimax concave penalty. *Annals of Statistics*. 38(2):894-942.

Liu et al.(2020) A bootstrap lasso + partial ridge method to construct confidence intervals for parameters in high-dimensional sparse linear models. *Statistica Sinica*

**Examples**

```
data(dataM)
dim(dataM)
dataM[1:5, 1:8]
data(dataC)
dim(dataC)
dataC[1:5, ]

results <- MZILN(MicrobData = dataM,
                 CovData = dataC,
                 linkIDname = "id",
                 targetTaxa = "rawCount6",
                 refTaxa=c("rawCount11"),
                 allCov=c("v1", "v2", "v3"),
                 fdrRate=0.15)
```

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