# Package 'MTA'

## December 24, 2019

<b>Title</b> Longitudinal microbiome data analysis: microbial trend analysis, group comparison and classification
Version 0.1.0
<b>Date</b> 2019-12-24
Author Chan Wang, Jiyuan Hu, Martin J. Blaser, Huilin Li.
Maintainer Chan Wang <chan.wang@nyulangone.org> and Huilin Li <huilin.li@nyulangone.org></huilin.li@nyulangone.org></chan.wang@nyulangone.org>
Description  Microbial trend analysis (MTA) framework is proposed for the high-dimensional and phylogenetic longitudinal microbiome data. The MTA framework can describe microbial dynamics, differential group difference, extract key taxa driving the microbial temporal trend, and classify the subjects, is a highly in-time and attractive analytical tool to study the dynamics of microbiome composition and identify key bacterial species that affect susceptibility to complex diseases.
<b>Depends</b> adephylo, fda, ggplot2, reshape2, toOrdinal, matlib, R (>= 3.6.1)
License GPL (>= 2)
Encoding UTF-8
LazyData true
<pre>URL https://github.com/chanw0/MTA</pre>
<pre>BugReports http://github.com/chanw0/MTA/issues</pre>
RoxygenNote 6.1.1
R topics documented:
MTA-package
Index

MTA-package	Longitudinal microbiome data analysis: microbial trend analysis, group comparison and classification

### Description

Microbial trend analysis (MTA) framework is an attractive and effective tool in investigating the dynamic microbial pattern in the longitudinal microbiome study. In particular, MTA can perform three tasks: 1) MTA can capture the common and major community level microbial dynamic trends for a given group of subjects and identify the dominant taxa; 2) MTA can examine whether the microbial overall dynamic trends are significantly different or not in groups; 3) MTA can classify a given subject based on its longitudinal microbial profiling. Both simulations and real data applications showed that MTA framework is robust and powerful in hypothesis testing, taxon identification, and subject classification.

#### **Details**

Package: MTA
Type: Package
Version: 1.0

Date: 2019-12-24 License: GPL (>= 2)

#### Author(s)

Chan Wang, Jiyuan Hu, Martin J. Blaser, Huilin Li.

Maintainer: Chan Wang < Chan. Wang @nyulangone.org>, Huilin Li < Huilin.Li@nyulangone.org>

#### References

Wang C, Hu J, Blaser M J, Li H (2019). Longitudinal microbiome data analysis: microbial trend analysis, group comparison and classification.

MTA The overall function in MTA framework

#### **Description**

This overall function performs three tasks: 1) capturing the common and major community level microbial dynamic trends for a given group of subjects and identifying the dominant taxa; 2) examining whether the microbial overall dynamic trends are significantly different or not in groups; 3) classifying a given subject based on its longitudinal microbial profiling.

#### **Usage**

### Arguments Y1

Y\_new

phy.tree

timevec

transf

Y1 can be an N\*P\*T numeric array containing compositional microbiome data for N subjects having P taxa across T time points, or a list, in which each element represents a group of subjects having P taxa across T time points. The number of groups is equal to or larger than 2. The list's name can be assigned as the group labels, otherwise, it is 1:length(Y1) by default. If Y1 is an N\*P\*T numeric array, MTA gives the microbial dynamics for Y1 and identifies the key taxa. If Y1 is a list, MTA provides the group comparison or classification.

An optional N\*P\*T numeric array containing compositional microbiome data for N subjects having P taxa across T time points. Y\_new represents the subjects who need to be classified based on the training set Y1. Default=NULL, no calculation for the classification part.

The phylogenetic tree (phylo-class) object for P taxa. Default=NULL, no Laplacian penalty dealing with the phylogenetical correlation among taxa.

An optional numeric vector of the continuous time point with length = T.

A character value, which represents the transformation technique one wants to take. "None" represents MTA analyzes the microbial relative abundance directly; "Log", "ALR", and "CLR" represent the corresponding log, additive log-ratio transformation and centered log-ratio transformation respectively. Defended by "None".

fault="None".

M The number of top common trends one wants to select. Default=NULL.

proportion.explained

A numeric value used to determine the number of top common trends as an alternative. Default=0.85.

. .

k A numeric value represents k-fold cross-validation to determine the tuning parameters. Default=5.

lambda1.set A numeric vector containing the candidate tunning parameters for the smoothing technique to deal with the smoothness of the extracted trends. Default=c(0.01,0.1,1,10).

lambda2.set A numeric vector containing the candidate tunning parameters for the Lasso penalty to deal with the high dimensionality of taxa. Default=seq(0,1,0.1).

lambda3.set A numeric vector containing the candidate tunning parameters for the Laplacian penalty to deal with the phylogenetical correlation. Default=c(0).

An integer value, the number of resampling for the group comparison. Default=10.

alpha An numeric value, the signficance level for the group comparison. Default=0.05.

#### Value

num.sam

When Y1 is an N\*P\*T numeric array, MTA performs the first task: capturing the common and major community level microbial dynamic trends and identifying the dominant taxa based on the Y1. A list which contains two elements:

**Trend Plot** The extracted common trends extracted from all subjects Y1

**Factor Score** The estimated factor scores for all key taxa which contribute to the extracted common trends.

When Y1 is a list involving two or more than two groups, but Y\_new is NULL, MTA performs the second task: examining whether the microbial overall dynamic trends are significantly different or not in groups and provides a list which contains comparing results between any two groups. The comparing results between any two groups contains five elements:

**P value** P value for the common trends representing the difference in the microbial dynamic pattern between these two groups

**Trend Plot** The plot for the extracted common trends among R resamplings

**Discover rate** The discover rates for all taxa among R resamplings. This item is provided when the common trend is significant, otherwise, on this item.

**Factor Score** The estimated factor scores for all taxa which contribute to the common trends. This item is provided when the common trend is significant, otherwise, on this item.

**Standard error** The standard error of the estimated factor scores for all taxa among R resamplings. This item is provided when the common trend is significant, otherwise, on this item.

When Y1 is a list involving two or more than two groups, and Y\_new is an N\*P\*T numeric array, MTA performs the third task: classifying the subjects in Y\_new based on the training set Y1. The result contains an N\*2 matrix, in which the first column is the subject's ID and the second column is its group label accordingly.

#### Author(s)

Chan Wang, Jiyuan Hu, Martin J. Blaser, Huilin Li.

Maintainer: Chan Wang <a href="mailto:Chan.Wang@nyulangone.org">Chan.Wang@nyulangone.org</a>, Huilin Li <a href="mailto:Huilin.Li@nyulangone.org">Huilin Li <a href="mailto:Huilin.Li <a href="mailto:Hu

#### References

Wang C, Hu J, Blaser M J, Li H (2019). Longitudinal microbiome data analysis: microbial trend analysis, group comparison and classification.

#### **Examples**

```
# ########## generation data
# library(dirmult)
# ## sample size
# N=20;
# ## time point
# T=10;
# ##number of taxa
# P=30;
#
# beta0=abs(rnorm(P,10,20))
# ## key taxa
# causal.ind=sample(1:20,3)
#
# #### generating two groups
# group1=array(NA, dim = c(N,P,T))
# for(t in 1:T) group1[, ,t] = rdirichlet(N,(beta0))
```

# Index

```
*Topic MTA
MTA, 2
*Topic package
MTA-package, 2

MTA, 2
MTA-package, 2
```