# Package 'MTA'

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Title Microbial Trend Analysis for Common Dynamic Trend, Group

Comparison, and Classification in Longitudinal Microbiome Study

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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.
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R topics documented:
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MTA-package	Microbial Trend Analysis for Common Dynamic Trend, Group Com-
, 0	parison, and Classification in Longitudinal Microbiome Study

# 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

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#### References

Wang C, Hu J, Blaser M J, Li H (2021). Microbial Trend Analysis for Common Dynamic Trend, Group Comparison, and Classification in Longitudinal Microbiome Study.

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.

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#### 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. De-

fault="None".

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

proportion.explained

lambda3.set

num.sam

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(0,1,0.001).

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. De-

fault=10.

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

#### Value

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:

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**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

**Trends** 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.

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#### References

Wang C, Hu J, Blaser M J, Li H (2021). Microbial Trend Analysis for Common Dynamic Trend, Group Comparison, and Classification in Longitudinal Microbiome Study.

#### **Examples**

```
########### generation data
library(dirmult)
## sample size
N=20;
## time point
T=10;
##number of taxa
P=20;

beta0=abs(rnorm(P,10,20))
## key taxa
causal.ind=sample(1:P,3)

#### generating two groups
group1=array(NA, dim = c(N,P,T))
for(t in 1:T) group1[, ,t] = rdirichlet(N,beta0)
```

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Toydata

Toy data used for illustration

#### **Description**

Toydata is a list with the length=3.

### Usage

Toydata

#### **Format**

A list with the length=3 with names Case, Control, and phy.tree:

Case A 15 \* 35 \* 4 Array of relative abundances for 15 Cases

**Control** A 15 \* 35 \* 4 Array of relative abundances for 15 Controls

phy.tree The phylogenetic tree (phylo-class) object for 35 taxa

## **Examples**

```
####### Toy data for illustration
library(ggpubr)
library(toOrdinal)
library(reshape2)

Case=Toydata[[1]]
Control=Toydata[[2]]
phy.tree=Toydata[[3]]
```

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```
### Group comparison
aa=MTA(list(Control,Case),phy.tree=phy.tree)$`Comparison between group 1 and group 2`
aa1=aa[[4]]
aa2=aa[[5]]
for(i in 1:nrow(aa2)) {
 bb1=aa1[i,-1];bb2=aa2[i,-1]
 bb=order(bb1^2/sum(bb1^2),decreasing =TRUE)
bb1[-c(bb[1:which(cumsum(cumsum(sort(bb1^2/sum(bb1^2),decreasing =TRUE))>0.99)==1)])]=0
bb1[((bb1+1.96*bb2)>=0 & (bb1-1.96*bb2)<=0)]=NA
aa1[i,-1]=bb1
}
effect1=aa1
colnames(effect1)=c("trend",unlist(dimnames(Control)[2]))
effect1=data.frame(effect1,check.names =FALSE)
effect1=melt(effect1,id.vars = 1,variable.name = "taxa",value.name = "score")
effect1$trend=paste(sapply(effect1$trend, toOrdinal), "common trend")
effect1=na.omit(effect1)
pp1=ggbarplot(effect1, x = "taxa", y = "score",
             ylab = "Estimated factor score", xlab="", facet.by = "trend",
             add = "mean")+
facet_wrap(~trend,scales="free")+theme(axis.text.x = element_text(angle = 90, hjust = 1))
####### trend figure
trend1=aa[[2]]
pp2=ggline(trend1, x = "time",y = "Escore",
          ylab = "Common trend", xlab="Week", facet.by = "trend",
          add = "mean")+
 facet_wrap(~trend, scales="free_y")
ggarrange(pp2,pp1,labels = c("(A)", "(B)"), ncol = 2)
#### Classification
aa=MTA(list(Control,Case),Case, phy.tree=phy.tree)$`Comparison between group 1 and group 2`
### Subject ID
aa$`New subject`
### Group label 1: the first group (Control); 2: the second group (Case)
aa$`Group label`
```

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