16S Microbiome Data Visualization

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Introduction to this script

This script is for the Vinatzer lab's need of 16S microbiome data analysis. Taking the output BIOM file and a couple of intermediate files from QIIME as well as the mapping file, this script aims to analyze and visualize the data with the R package of phyloseq from BioConductor and ggplot2, analyses will include but not limitted to abundance analysis, alpha diversity, beta diversity, and hierarchical clustering of the samples.

Set working directory to where the input files are/will be

```
setwd("/Users/longtian/Desktop/VinatzerLab/Dianthus")
```

Library/package needed

```
source('http://bioconductor.org/biocLite.R')
## Warning: 'BiocInstaller' and 'biocLite()' are deprecated, use the 'BiocManager'
     CRAN package instead.
## Bioconductor version 3.8 (BiocInstaller 1.32.1), ?biocLite for help
biocLite('phyloseq')
## Warning: 'biocLite' is deprecated.
## Use 'BiocManager::install' instead.
## See help("Deprecated")
## BioC_mirror: https://bioconductor.org
## Using Bioconductor 3.8 (BiocInstaller 1.32.1), R 3.5.2 (2018-12-20).
## Installing package(s) 'phyloseq'
##
## The downloaded binary packages are in
   /var/folders/wf/clzm6h595g1f40xhj4x9b2vm0000gn/T//Rtmpo5cGmq/downloaded_packages
## Old packages: 'dplyr', 'igraph', 'readxl'
library("phyloseq")
library("scales")
library("ggplot2")
library("Rcpp")
library("dplyr")
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
## filter, lag
## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
```

Import files

The main files needed are the BIOM file and the mapping file. Only to make sure there should be 8 columns in the mapping file, not sure why is that

Abundance analysis

Filtering the data

There would be a lot of taxon whose percentage is very low, and it's not really necessary to show them all. So the first step is to keep only those taxa which each contributes to at least 2% or 5% of the whole microbial community of the sample.

```
run_genus <- run %%

tax_glom(taxrank = "Genus") %>%

transform_sample_counts(function(x){x/sum(x)}) %>%

psmelt() %>%

filter(Abundance > 0.02) %>%

arrange(Genus)
```

Prepare palette

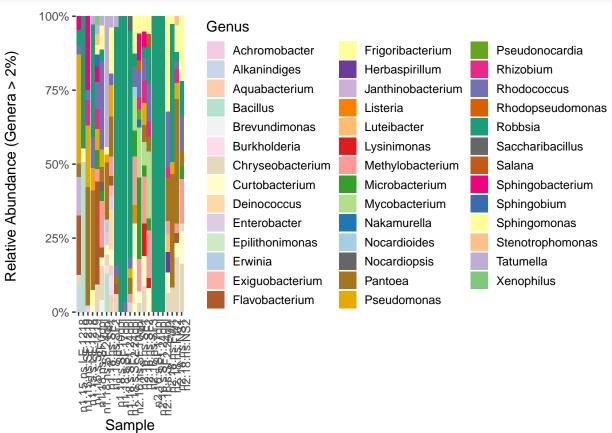
```
library(RColorBrewer)
n <- dim(run_genus)[1]
qual_col_pals = brewer.pal.info[brewer.pal.info$category == 'qual',]
col_vector = unlist(mapply(brewer.pal, qual_col_pals$maxcolors, rownames(qual_col_pals)))</pre>
```

Barplot

A normal abundance barplot could be easily generated in this way

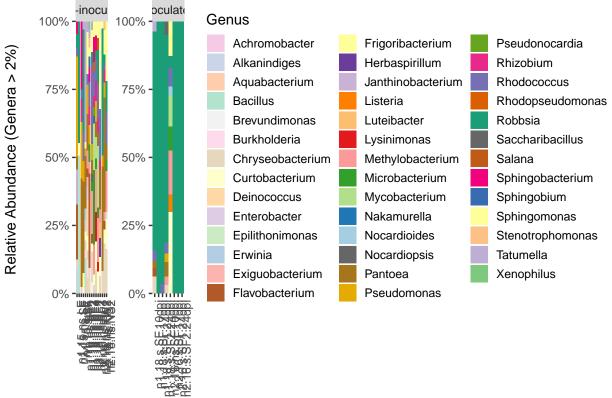
```
run_genus$Genus <- factor(run_genus$Genus, levels = rev(levels(run_genus$Genus)))
ggplot(run_genus,aes(x=Sample,y=Abundance,fill=Genus)) +
   geom_bar(position="fill",stat="identity") +
   scale_fill_manual(values = col_vector) +
   guides(fill=guide_legend(reverse=T,keywidth = 1,keyheight = 1)) +
   ylab("Relative Abundance (Genera > 2%) \n") +
```

```
xlab("Sample") +
# ggtitle("Genus Composition")+
scale_y_continuous(labels=percent_format(),expand=c(0,0))+
theme(axis.text.x = element_text(angle=90,hjust=1), panel.background = element_blank())
```



If there the experiment consists of different groups/treatments, the barplot can be divided into groups of choice

```
# run_genus$Genus <- factor(run_genus$Genus, levels = rev(levels(run_genus$Genus)))
# Add grouping info to the table
run_genus$Condition_rev <- factor(run_genus$Condition,level=rev(levels(map$Condition)))</pre>
levels(run_genus$Condition_rev) <- c('Non-inoculated','Inoculated')</pre>
library(plyr)
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
##
##
## Attaching package: 'plyr'
## The following objects are masked from 'package:dplyr':
##
       arrange, count, desc, failwith, id, mutate, rename, summarise,
##
##
       summarize
```



Alpha diversity

Calculate rarefaction

Sample

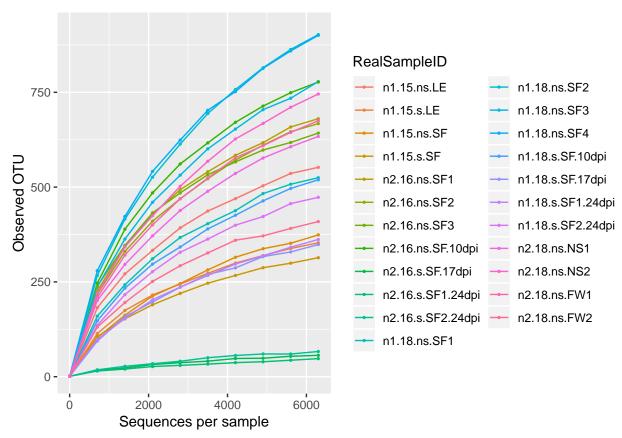
First we create a function that can iterate the BIOM file, i.e. rarefaction

```
calculate_rarefaction_curves <- function(psdata, measures, depths) {
  require('plyr') # ldply
  require('reshape2') # melt
  estimate_rarified_richness <- function(psdata, measures, depth) {
    if(max(sample_sums(psdata)) < depth) return()
    psdata <- prune_samples(sample_sums(psdata) >= depth, psdata)
```

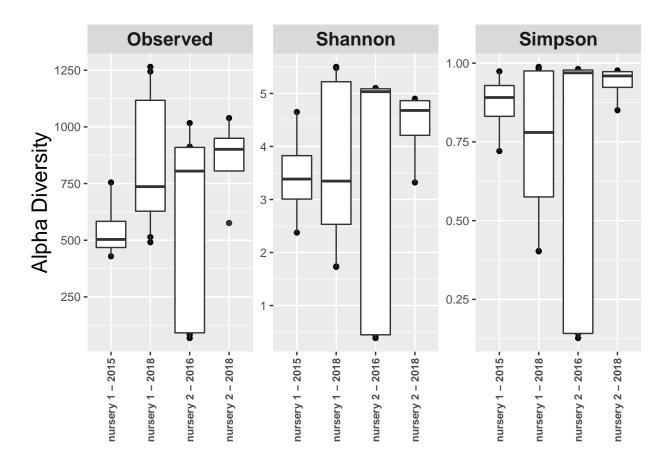
```
rarified_psdata <- rarefy_even_depth(psdata, depth, verbose = FALSE)
alpha_diversity <- estimate_richness(rarified_psdata, measures = measures) # as.matrix forces the u
molten_alpha_diversity <- melt(as.matrix(alpha_diversity), varnames = c('Sample', 'Measure'), value
molten_alpha_diversity
}
names(depths) <- depths # this enables automatic addition of the Depth to the output by ldply
rarefaction_curve_data <- ldply(depths, estimate_rarified_richness, psdata = psdata, measures = measurarefaction_curve_data$Depth <- as.numeric(levels(rarefaction_curve_data$Depth))[rarefaction_curve_data
}</pre>
```

One thing worth to note is the choice of depth. For the most of the time, rarefaction curve is used to compare the richness of each sample, so the depth is the minimal number of OTU of the samples. However, if you want to see the raw sample size, use the maximum.

```
depth = 7000
step = 700
rarefaction_curve_data <- calculate_rarefaction_curves(run, c('Observed'), rep(seq(1,depth,by=step), ea
## Loading required package: reshape2
rarefaction_curve_data_summary <- ddply(rarefaction_curve_data, c('Depth', 'Sample', 'Measure'), summar
rarefaction_curve_data_summary_verbose <- merge(rarefaction_curve_data_summary, data.frame(sample_data(
rarefaction_curve_data_summary_verbose$RealSampleID <- factor(rarefaction_curve_data_summary_verbose$Re
n <- dim(rarefaction_curve_data_summary_verbose)[1]</pre>
qual_col_pals = brewer.pal.info[brewer.pal.info$category == 'qual',]
col_vector = unlist(mapply(brewer.pal, qual_col_pals$maxcolors, rownames(qual_col_pals)))
ggplot( data = rarefaction_curve_data_summary_verbose,
        mapping = aes( x = Depth, y = Alpha_diversity_mean,
                       ymin = Alpha_diversity_mean - Alpha_diversity_sd,
                       ymax = Alpha_diversity_mean + Alpha_diversity_sd,
                       colour = RealSampleID,
                       group = RealSampleID)
        ) +
  scale fill manual(values=col vector) +
  geom_line( ) +
  geom_point(size=0.5 ) +
  guides(fill=guide_legend(title="Sample")) +
  xlab("Sequences per sample") +
  ylab("Observed OTU")
```



Also, boxplot with other alpha diversity indices can be generated. If there is such a column in the mapping file representing the desired grouping condition, say FigureTwo here.



Beta diversity

Beta diviersity indicates the relatedness between samples.

First, pairwise distances between samples are calculated. The two measurement used are weighted-Unifrac and unweighted-Unifrac here.

```
dm_weighted_unifrac <- distance(run, method = "wUniFrac")

## Warning in UniFrac(physeq, weighted = TRUE, ...): Randomly assigning root

## as -- JF703554.1.1369 -- in the phylogenetic tree in the data you provided.

## Warning in matrix(tree$edge[order(tree$edge[, 1]), ][, 2], byrow = TRUE, :

## data length [11107] is not a sub-multiple or multiple of the number of rows

## [5554]

dm_unweighted_unifrac <- distance(run, method='Unifrac')

## Warning in UniFrac(physeq, ...): Randomly assigning root as --

## FN421553.1.1379 -- in the phylogenetic tree in the data you provided.

## Warning in UniFrac(physeq, ...): data length [11107] is not a sub-multiple

## or multiple of the number of rows [5554]

Treatment <- map$Treatment

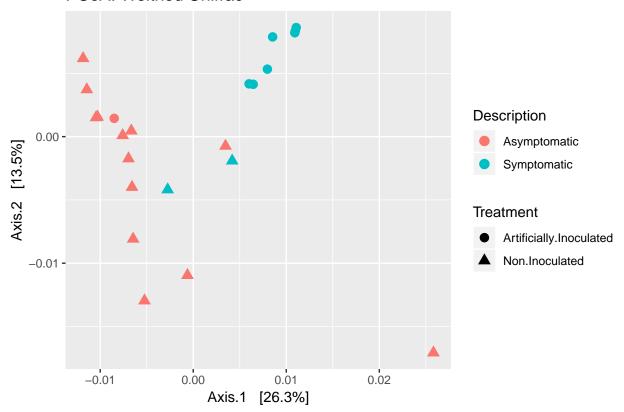
Symptom <- map$Description

ordWU <- ordinate(run, method='PCoA', distance=dm_weighted_unifrac)

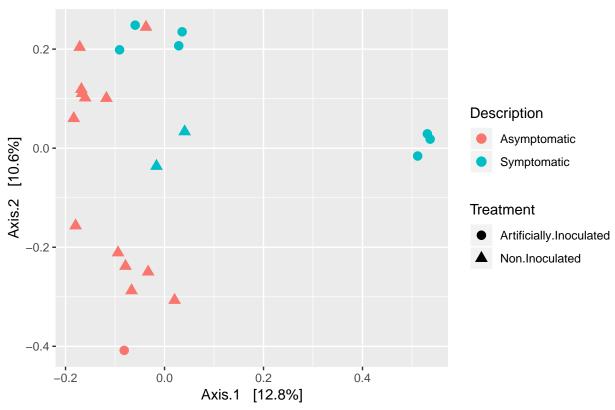
plot_ordination(run, ordWU,,color = 'Description', shape='Treatment') +</pre>
```

```
geom_point(size=3) +
# geom_text(aes(label=map$RealSampleID),hjust=0,vjust=0) +
# theme(axis.title = element_text(size=20),
# axis.text = element_text(size=16),
# legend.text = element_text(size=16)) +
ggtitle("PCoA: Weithed Unifrac")
```

PCoA: Weithed Unifrac







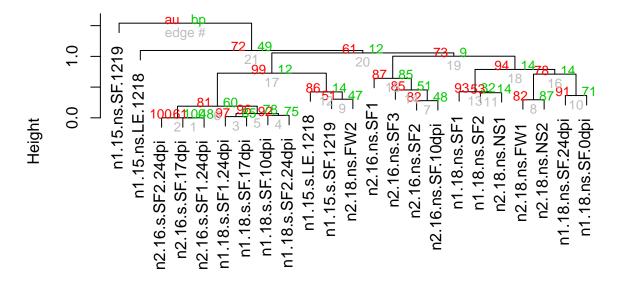
Phylogenetics tree of the samples

Bootstrap values can be generated from pvclust, and the topology is the same with what calculated from regular hclust function

```
library(pvclust)
hc_pv <- pvclust(as.matrix(dm_weighted_unifrac),method.hclust = "complete")

## Bootstrap (r = 0.48)... Done.
## Bootstrap (r = 0.57)... Done.
## Bootstrap (r = 0.7)... Done.
## Bootstrap (r = 0.78)... Done.
## Bootstrap (r = 0.87)... Done.
## Bootstrap (r = 1.0)... Done.
## Bootstrap (r = 1.0)... Done.
## Bootstrap (r = 1.17)... Done.
## Bootstrap (r = 1.17)... Done.
## Bootstrap (r = 1.26)... Done.
## Bootstrap (r = 1.39)... Done.</pre>
```

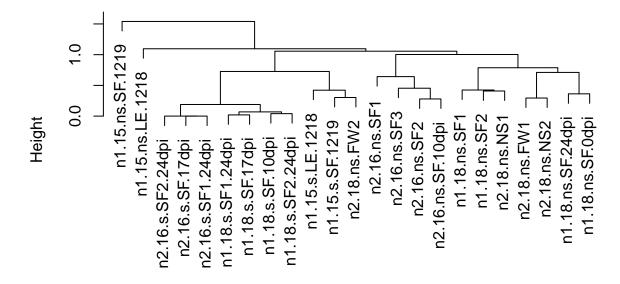
Cluster dendrogram with AU/BP values (%)



Distance: correlation Cluster method: complete

hc <- hclust(as.dist(1-cor(as.matrix(dm_weighted_unifrac))),method="complete")
plot(hc)</pre>

Cluster Dendrogram

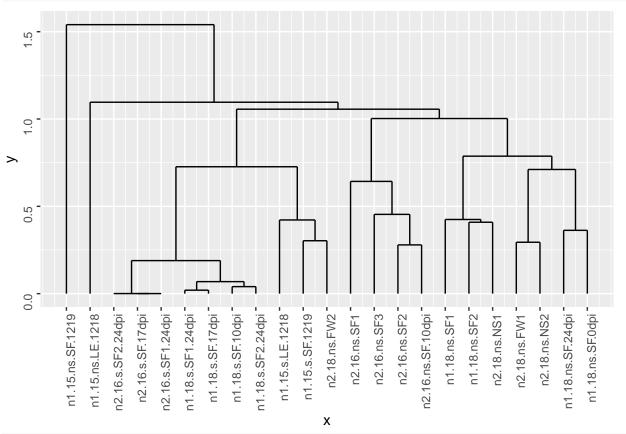


as.dist(1 - cor(as.matrix(dm_weighted_unifrac))) hclust (*, "complete")

The den-

drogram can be plotted through ggplot2 with the help from ggdendro.

```
library(ggdendro)
ggdendrogram(hc,rotate=F, size=4, theme_dendro = F)
```



```
ddata <- dendro_data(as.dendrogram(hc))</pre>
mapping <- map
label <- as.character(ddata$labels$label)</pre>
rownames(mapping) <- mapping$RealSampleID</pre>
Treatment <- mapping$Condition</pre>
labeled_data <- cbind(label(ddata), Treatment)</pre>
ggplot(segment(ddata)) +
  geom_segment(aes(x=x,y=y,xend=xend,yend=yend)) +
  coord_flip() + scale_y_reverse(expand=c(0.5,0)) +
  geom_text(data = labeled_data, aes(x=x,y=y,label=label,hjust=0,color=Treatment),size=3) +
  ylab("Weighted Unifrac distance") +
  theme(axis.line.y=element_blank(),
        axis.ticks.y = element_blank(),
        axis.text.y = element_blank(),
        axis.title.y=element_blank(),
        plot.title = element_text(hjust=0.5),
        legend.text = element_text(size=16),
        axis.text.x = element_text(size=10),
        axis.title.x = element_text(size=20))
```

