

# Script R9: Plotting Microbial Taxonomy from MEGAN

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## Abstract

This protocol outlines the analysis used to plot MEGAN taxonomic assignments. Based on the methods from the following publication:

Hannigan, Geoffrey D., et al. "The Human Skin Double-Stranded DNA Virome: Topographical and Temporal Diversity, Genetic Enrichment, and Dynamic Associations with the Host Microbiome." *mBio* 6.5 (2015): e01578-15.

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## Guidelines

sessionInfo()

```
## R version 3.2.0 (2015-04-16)
## Platform: x86_64-apple-darwin13.4.0 (64-bit)
## Running under: OS X 10.10.4 (Yosemite)
## ## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats graphics grDevices utils datasets methods base
##
## loaded via a namespace (and not attached):
## [1] magrittr_1.5   formatR_1.2   tools_3.2.0   htmltools_0.2.6
## [5] yaml_2.1.13   stringi_0.4-1 rmarkdown_0.7 knitr_1.10.5
## [9] stringr_1.0.0 digest_0.6.8  evaluate_0.7
```

## Before start

Supplemental information available at:

[https://figshare.com/articles/The\\_Human\\_Skin\\_dsDNA\\_Virome\\_Topographical\\_and\\_Temporal\\_Diversity\\_Genetic\\_Enrichment\\_and\\_Dynamic\\_Associations\\_with\\_the\\_Host\\_Microbiome/1281248](https://figshare.com/articles/The_Human_Skin_dsDNA_Virome_Topographical_and_Temporal_Diversity_Genetic_Enrichment_and_Dynamic_Associations_with_the_Host_Microbiome/1281248)

## Protocol

### Step 1.

First load the libraries necessary for analysis.

```
cmd COMMAND
library(ggplot2)
packageVersion("ggplot2")

library(reshape2)
packageVersion("reshape2")

library(plyr)
packageVersion("plyr")

library(RColorBrewer)
packageVersion("RColorBrewer")
```

### ✓ EXPECTED RESULTS

```
## [1] '1.0.1'
```

```
## [1] '1.4.1'
```

```
## [1] '1.8.2'
```

```
## [1] '1.1.2'
```

### Step 2.

Then read in and format the metadata file.

```
cmd COMMAND
skinmet_metadata<-
read.delim("../IntermediateOutput/Mapping_files/SkinMet_and_Virome_001_metadata.tsv")
skinmet_metadata<-
skinmet_metadata[,c("NexteraXT_SampleID","SubjectID","Site_Symbol","TimePoint")]
skinmet_metadata$NexteraXT_SampleID<-as.character(skinmet_metadata$NexteraXT_SampleID)
colnames(skinmet_metadata)[1]<- "SampleID"
skinmet_metadata<-subset(skinmet_metadata, skinmet_metadata$SampleID != "NA")
skinmet_metadata<-subset(skinmet_metadata, skinmet_metadata$TimePoint != 1)
skinmet_metadata<-subset(skinmet_metadata, !(skinmet_metadata$SubjectID %in% c(2,3,9,11)))
skinmet_metadata<-
subset(skinmet_metadata, !(skinmet_metadata$Site_Symbol %in% c("Neg", "Vf", "Ba", "Ph")))
skinmet_metadata$SubjectID<-NULL
skinmet_metadata$TimePoint<-NULL
```

### Step 3.

Now we are ready to read in the actual data files. In the initial analysis, each sample had its own file. We wrote a function to read in the data for each sample and combine it into one data file. For simplicity, we only put the file with the combined data in the intermediate files, but the function readData shows you how we generated those files.

#### cmd **COMMAND**

```
readData <- function(taxa){

  ## Extract vector of empty files' names
  empty <- taxa[file.info(taxa)[["size"]]==0]
  ## Remove empty files
  unlink(empty, recursive=TRUE, force=FALSE)

  for(i in taxa){
    name<-gsub(x=i,pattern="_megan_.*.txt",replacement="",perl=TRUE)
    tmp<-read.delim(i, header=FALSE)
    colnames(tmp)<-c("Taxa", name)
    if(i==taxa[1]) {data<-tmp} else {data<-merge(data,tmp,"Taxa", all=TRUE)}
  }
  data[is.na(data)]=0

  name<-gsub(x=taxa[1], pattern="MG100.*_megan_", replacement="megan_", perl=TRUE)
  name2<-paste("../IntermediateOutput/MEGAN/", name, sep="")
  write.table(data, name2, row.names=FALSE, quote=FALSE, eol="\r\n", sep="\t")

  return(data)
}
```

#### Step 4.

To make our results more interpretable, we only want to look at a certain number of taxa. We want to write a function that will keep a certain number of taxa and combine the remaining taxa into the category "Other".

#### cmd **COMMAND**

```
topTaxa<- function(data, numTaxa){
  # check to see if you need to condense the taxa list to the specified number
  if(ncol(data)>numTaxa){
    # the data is organized where the taxa are the columns and the samples are the rows
    # add a row containing the total cumulative frequency of the taxon
    data[nrow(data)+1,]<-colSums(data)
    # order the columns by their frequency
    data<-data[,order(-data[nrow(data),])]
    # remove the last row containing the cumulative frequency
    data<-data[-nrow(data),]
    # remove the least frequent taxa
    tmp<-data[,numTaxa:ncol(data)]
    data<-data[, -c(numTaxa:ncol(data))]
    other<-rowSums(tmp)
    data<-cbind(data,other)
    colnames(data)[ncol(data)]<-"Other"
  }
  return(data)
}
```

#### Step 5.

Some of the genus level taxa were not correctly classified at the phylum level (they incorrectly labeled unclassified at phylum level). We need to correct this:

#### cmd **COMMAND**

```
removeDupes <- function(taxa_list, data){
  for (i in 1:length(taxa_list)){
    taxa<-taxa_list[i]
    dup<-grep(pattern=taxa, x=data$Taxa)
    dup_sum<-colSums(data[grep(pattern=taxa, x=data$Taxa),-1])
    data[dup[1],2:ncol(data)]<-dup_sum
    data<-data[-dup[2],]
```

```
}
return(data)
```

```
}
```

## Step 6.

We want to read the data in and generate bar plots to visualize it.

cmd **COMMAND**

```
plotTaxa<- function(level, data){

  # Remove duplicates that were correctly classified at phylum level but also unclassified at phylum level
  if( level=="bacteria"){
    to_remove<-
c("g__Corynebacterium","g__Mycobacterium", "g__Propionibacterium", "g__Staphylococcus")
    data<-removeDups(to_remove, data)
  }

  if( level=="eukaryotes"){
    to_remove<-c("g__Malassezia","g__Ustilago", "g__Canis", "g__Homo")
    data<-removeDups(to_remove, data)
  }

  # format data table
  row.names(data)<-data$Taxa
  data$Taxa<-NULL

  # if the data is for the eukaryotes, we only want to look at the fungal assignments
  if( level=="eukaryotes"){
    fungi_bas<-grep(pattern="p__Basidiomycota", x=row.names(data), fixed=FALSE)
    fungi_asc<-grep(pattern="p__Ascomycota", x=row.names(data), fixed=FALSE)
    fungi<-c(fungi_bas,fungi_asc)
    taxa_level<-data[fungi,]

    # look at fungal vs. nonfungal assignments for use with superkingdom analysis
    # fungi:
    eukaryota_level.t<-t(taxa_level)
    eukaryota_level2<-
merge(eukaryota_level.t,skinmet_metadata,by.x="row.names",by.y="SampleID")

    eukaryota_level_sum<-ddply(eukaryota_level2, c("Site_Symbol"), numcolwise(sum))
    row.names(eukaryota_level_sum)<-eukaryota_level_sum$Site_Symbol
    eukaryota_level_sum$Site_Symbol<-NULL
    eukaryota_level_sum<-eukaryota_level_sum[,order(-colSums(eukaryota_level_sum))]
    eukaryota_fungi<-rowSums(eukaryota_level_sum)

    # non-fungi
    eukaryota_level_other<-data[-fungi,]

    eukaryota_level_other.t<-t(eukaryota_level_other)
    eukaryota_level_other2<-
merge(eukaryota_level_other.t,skinmet_metadata,by.x="row.names",by.y="SampleID")
    eukaryota_level_other_sum<-
ddply(eukaryota_level_other2, c("Site_Symbol"), numcolwise(sum))
    row.names(eukaryota_level_other_sum)<-eukaryota_level_other_sum$Site_Symbol
    eukaryota_level_other_sum$Site_Symbol<-NULL
    eukaryota_non_fungi<-rowSums(eukaryota_level_other_sum)
    eukaryotes<-cbind(eukaryota_non_fungi, eukaryota_fungi)
    colnames(eukaryotes)<-c("sk__Eukaryota:Other", "sk__Eukaryota:Fungi")
  }
}
```

## Step 7.

Write out to table.

```
cmd COMMAND
write.table(eukaryotes, "../IntermediateOutput/MEGAN/eukaryote_counts.txt", row.names=FALSE, quote=FALSE, eol="\r\n", sep="\t")
}
else{
  taxa_level<-data
}
}
```

### Step 8.

Merge the data with the mapping file.

```
cmd COMMAND
taxa_level.t<-t(taxa_level)
taxa_level2<-merge(taxa_level.t,skinmet_metadata,by.x="row.names",by.y="SampleID")
```

### Step 9.

Combine samples from the same body sites.

```
cmd COMMAND
taxa_level_sum<-ddply(taxa_level2, c("Site_Symbol"), numcolwise(sum))
# if doing super kingdom level analysis, break up eukaryotes into fungi vs. non-fungi
if(level=="superkingdom")
{
  sk_eukaryotes<-read.delim("../IntermediateOutput/MEGAN/eukaryote_counts.txt")
  taxa_level_sum<-cbind(taxa_level_sum, sk_eukaryotes)
  taxa_level_sum$sk__Eukaryota<-NULL
  taxa_level_sum$Unassigned<-
rowSums(taxa_level_sum[,c("sk__Not_assigned","sk__other_sequences","sk__Low_complexity", "sk__No_hits")])
  taxa_level_sum$sk__Not_assigned <-NULL
  taxa_level_sum$sk__other_sequences<-NULL
  taxa_level_sum$sk__Low_complexity<-NULL
  taxa_level_sum$sk__No_hits<-NULL
}
row.names(taxa_level_sum)<-taxa_level_sum$Site_Symbol
taxa_level_sum$Site_Symbol<-NULL
```

### Step 10.

Look at the top 10 taxa.

```
cmd COMMAND
taxa_level_sum<-topTaxa(taxa_level_sum,10)
if(ncol(taxa_level_sum) < 10){
  taxa_level_sum<-taxa_level_sum[,order(-colSums(taxa_level_sum))]
}
taxa_order<-as.vector(colnames(taxa_level))
```

### Step 11.

Convert counts into relative abundances.

```
cmd COMMAND
taxa_level_rel_abund<-(taxa_level_sum/rowSums(taxa_level_sum))*100
```

### Step 12.

Format the data frame for use with ggplot2.

```
cmd COMMAND
taxa_level_rel_abund$Site<- row.names(taxa_level_rel_abund)
colnames(taxa_level_rel_abund)<-
gsub(x=colnames(taxa_level_rel_abund), pattern=".*s__",replacement="", perl=TRUE)
colnames(taxa_level_rel_abund)<-
gsub(x=colnames(taxa_level_rel_abund), pattern=".*g__",replacement="", perl=TRUE)
colnames(taxa_level_rel_abund)<-
```

```
gsub(x=colnames(taxa_level_rel_abund), pattern="_<phylum>", replacement="", perl=TRUE)
colnames(taxa_level_rel_abund)<-
gsub(x=colnames(taxa_level_rel_abund), pattern=".*sk__", replacement="", perl=TRUE)
taxa<-melt(taxa_level_rel_abund, id=c("Site"))
taxa_level_rel_abund$Site<-NULL
taxa$variable<-factor(taxa$variable, levels=taxa_order)
```

### Step 13.

Plot.

**cmd** **COMMAND**

```
ggplot(taxa, aes(x=Site, y=value, fill=variable, order=variable))+theme_bw()+geom_bar(stat=
"identity")+guides(fill = guide_legend(reverse = TRUE))+scale_fill_manual(values = c("#e41a
1c", "#377eb8", "#33CCCC", "#4daf4a", "#984ea3", "#ff7f00", "#ffff33", "#a65628", "#f781bf",
"#999999"))
}
```

### Step 14.

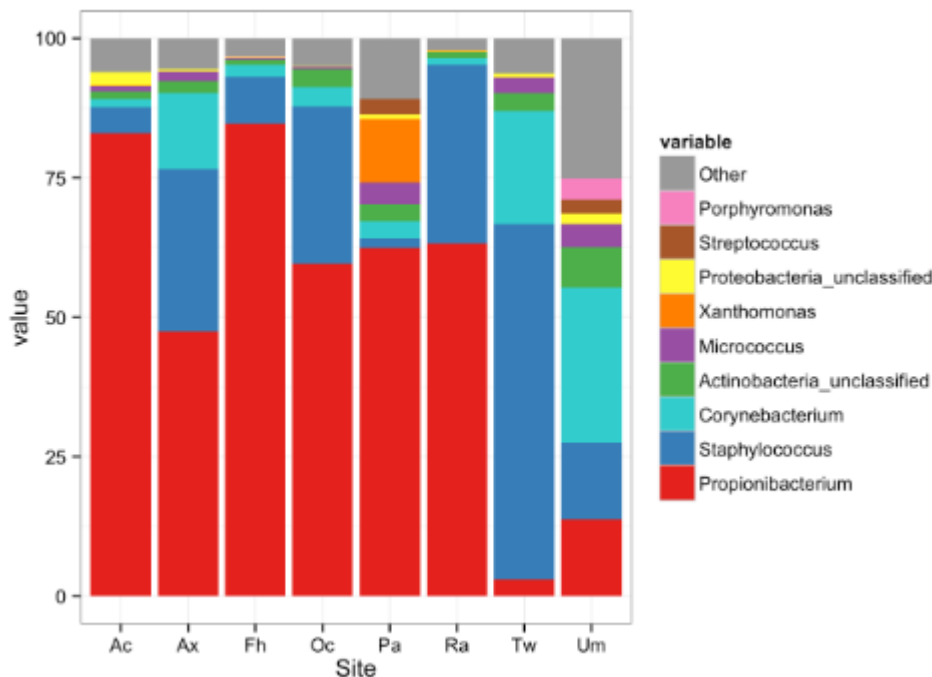
Plot the bacteria data.

**cmd** **COMMAND**

```
#setwd("formatted_output/genera")
#bact_genera=list.files("~/Desktop/SkinMet1/Updated_MEGAN_Figure/formatted_output/genera",
pattern="*_bacteria.txt")
#bact_data<-readData(bact_genera)
bact_data<-read.delim("../IntermediateOutput/MEGAN/megan_genera_bacteria.txt")
plotTaxa("bacteria", bact_data)
```

The commented out lines are how we initially generated the data files that we now just read in.

### ✓ EXPECTED RESULTS



### Step 15.

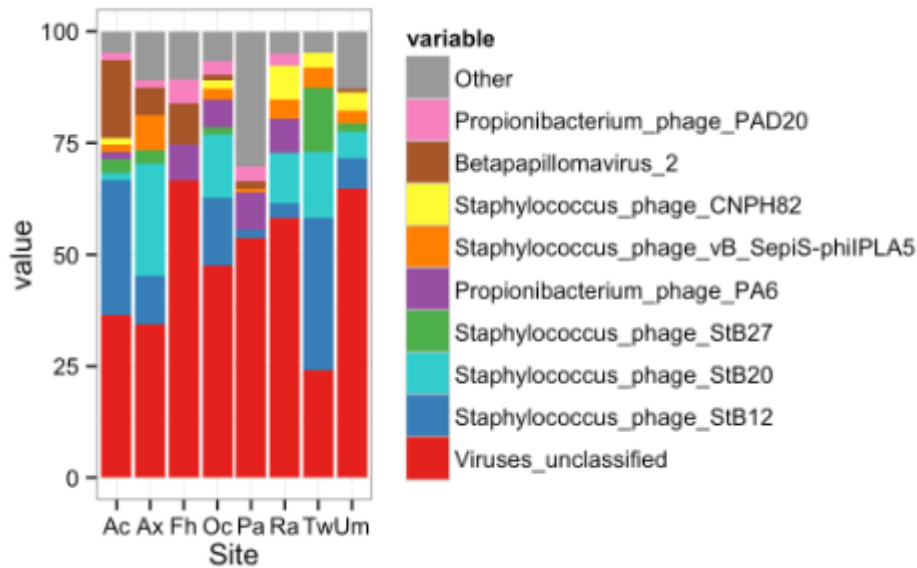
Plot the viruses.

**cmd** **COMMAND**

```
#viral_species=list.files("formatted_output/species", pattern="*_viruses.txt")
#viral_data<-readData(viral_species)
viral_data<-read.delim("../IntermediateOutput/MEGAN/megan_species_viruses.txt")
plotTaxa("viruses", viral_data)
```

The commented out lines are how we initially generated the data files that we now just read in.

## ✓ EXPECTED RESULTS



### Step 16.

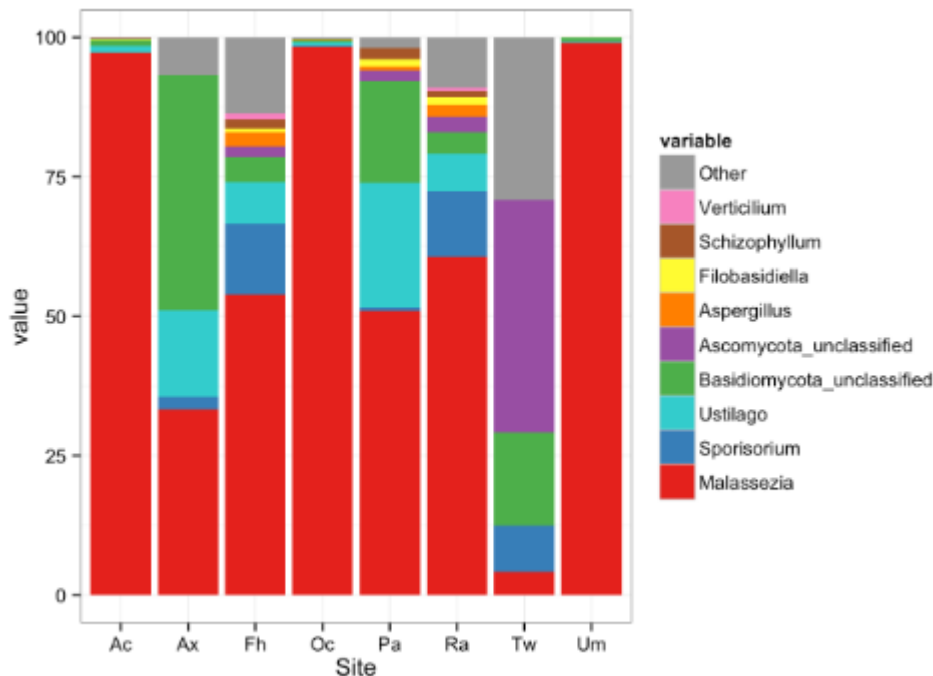
Plot the fungi.

cmd **COMMAND**

```
#eukaryota_genera=list.files("formatted_output/genera", pattern="*_eukaryotes.txt")
#eukaryota_data<- readData(eukaryota_genera)
eukaryota_data<- read.delim("../IntermediateOutput/MEGAN/megan_genera_eukaryotes.txt")
plotTaxa("eukaryotes", eukaryota_data)
```

The commented out lines are how we initially generated the data files that we now just read in.

## ✓ EXPECTED RESULTS



### Step 17.

Finally plot an overview of everything.

cmd **COMMAND**

```
#setwd("formatted_output/super_kingdom/")
#super_kingdom=list.files("formatted_output/super_kingdom/")
```

```
#sk_data<- readData(super_kingdom)
sk_data<-read.delim("../IntermediateOutput/MEGAN/megan_sk.txt")
plotTaxa("superkingdom",sk_data)
```

The commented out lines are how we initially generated the data files that we now just read in.

## 📈 EXPECTED RESULTS

