Phyloseq tutorial

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Load necessary libraries

- phyloseq
- ggplot2
- readxl : necessary to import the data from Excel file
- dplyr : becessary to reformat dataframe

```
library("phyloseq")
library("ggplot2")
library("readxl")
library("dplyr")

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
```

Read the data and create phyloseq objects

intersect, setdiff, setequal, union

Three tables are needed:

filter, lag

* OTU

##

##

##

- * Taxonomy
- * Samples

They are read from a single Excel file where each sheet contains one of the tables

The following objects are masked from 'package:base':

```
otu_mat<- read_excel("CARBOM data.xlsx", sheet = "OTU matrix")
tax_mat<- read_excel("CARBOM data.xlsx", sheet = "Taxonomy table")
samples_df <- read_excel("CARBOM data.xlsx", sheet = "Samples")</pre>
```

Phyloseq objects need to have row.names

• define the row names from the otu column

```
row.names(otu_mat) <- otu_mat$otu
```

- ## Warning: Setting row names on a tibble is deprecated.
 - remove the column of u since it is now used as a row name

```
otu_mat <- otu_mat <mark>%>% select (-</mark>otu)
```

• Idem for the two other matrixes

```
row.names(tax_mat) <- tax_mat$otu</pre>
```

Warning: Setting row names on a tibble is deprecated.

```
tax_mat <- tax_mat %>% select (-otu)
  row.names(samples_df) <- samples_df$sample</pre>
## Warning: Setting row names on a tibble is deprecated.
  samples_df <- samples_df %>% select (-sample)
Transform into matrixes out and tax tables (sample table can be left as data frame)
  otu_mat <- as.matrix(otu_mat)</pre>
  tax_mat <- as.matrix(tax_mat)</pre>
Transform to phyloseq objects
  OTU = otu_table(otu_mat, taxa_are_rows = TRUE)
  TAX = tax_table(tax_mat)
  samples = sample_data(samples_df)
  carbom <- phyloseq(OTU, TAX, samples)</pre>
  carbom
## phyloseq-class experiment-level object
## otu_table()
                OTU Table:
                               [ 20 taxa and 10 samples ]
Visualize data
  sample_names(carbom)
## [1] "X10n" "X10p" "X11n" "X11p" "X13n" "X13p" "X15n" "X15p" "X9n" "X9p"
  rank_names(carbom)
                    "Supergroup" "Division"
## [1] "Domain"
                                              "Class"
                                                           "Order"
## [6] "Family"
                    "Genus"
  sample_variables(carbom)
  [1] "fraction"
                            "Select_18S_nifH"
                                                "total 18S"
## [4] "total 16S"
                            "total nifH"
                                                "sample_number"
## [7] "transect"
                            "station"
                                                "depth"
                                                "picoeuks"
## [10] "latitude"
                            "longitude"
## [13] "nanoeuks"
                                                "level"
                            "bottom depth"
## [16] "transect_distance" "date"
                                                "time"
## [19] "phosphates"
                            "silicates"
                                                "ammonia"
## [22] "nitrates"
                            "nitrites"
                                                "temperature"
## [25] "fluorescence"
                            "salinity"
                                                "sample_label"
Keep only samples to be analyzed
  carbom <- subset_samples(carbom, Select_18S_nifH =="Yes")</pre>
  carbom
## phyloseq-class experiment-level object
## otu_table()
                 OTU Table:
                                   [ 20 taxa and 9 samples ]
## sample_data() Sample Data:
                                    [ 9 samples by 27 sample variables ]
## tax_table()
                 Taxonomy Table:
                                   [ 20 taxa by 7 taxonomic ranks ]
Keep only photosynthetic taxa
```

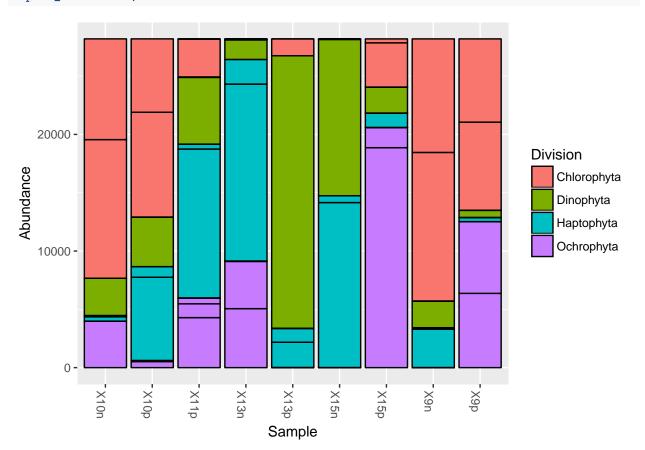
```
total = median(sample_sums(carbom))
standf = function(x, t=total) round(t * (x / sum(x)))
carbom = transform_sample_counts(carbom, standf)
```

The number of reads used for normalization is **28193**.

Bar graphs

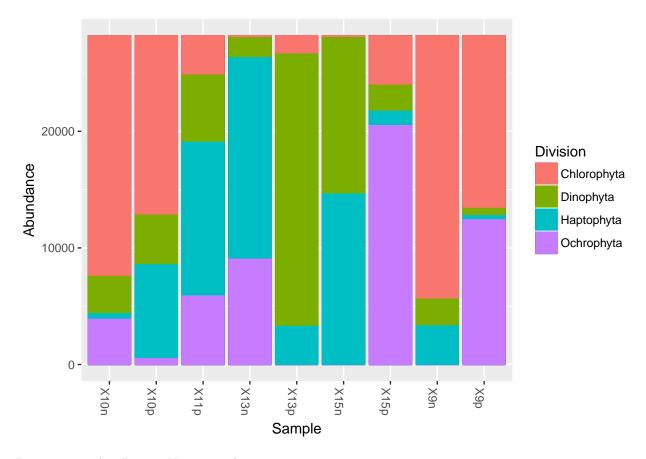
Basic bar graph based on Division

```
plot_bar(carbom, fill = "Division")
```



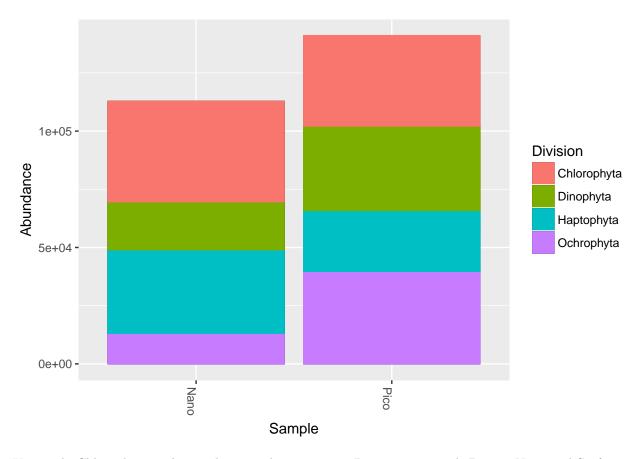
Make the bargraph nicer by removing OTUs boundaries. This is done by adding ggplot2 modifier.

```
plot_bar(carbom, fill = "Division") +
geom_bar(aes(color=Division, fill=Division), stat="identity", position="stack")
```



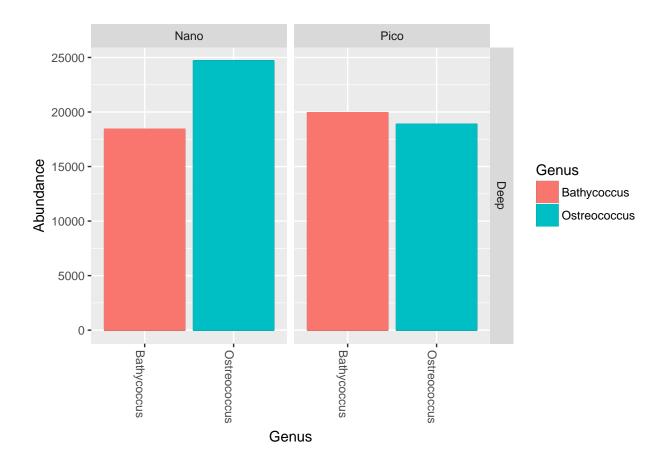
Regroup together Pico vs Nano samples

```
carbom_fraction <- merge_samples(carbom, "fraction")
plot_bar(carbom_fraction, fill = "Division") +
geom_bar(aes(color=Division, fill=Division), stat="identity", position="stack")</pre>
```



Keep only Chlorophyta and use color according to genus. Do separate panels Pico vs Nano and Surface vs Deep samples.

```
carbom_chloro <- subset_taxa(carbom, Division %in% c("Chlorophyta"))
plot_bar(carbom_chloro, x="Genus", fill = "Genus", facet_grid = level~fraction) +
geom_bar(aes(color=Genus, fill=Genus), stat="identity", position="stack")</pre>
```

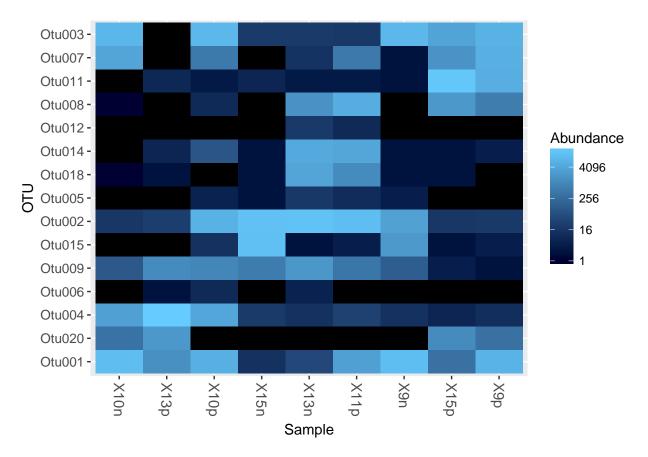


Heatmaps

A basic heatmap using the default parameters.

```
plot_heatmap(carbom, method = "NMDS", distance = "bray")
```

 $\hbox{\tt \#\# Warning: Transformation introduced infinite values in discrete y-axis}$

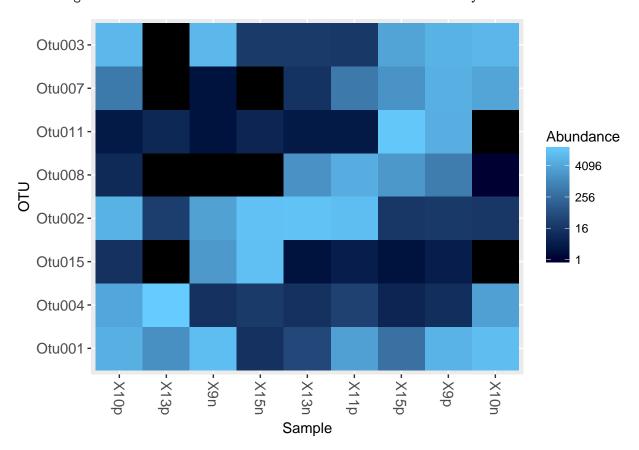


It is very very cluttered. It is better to only consider the most abundant OTUs for heatmaps. For example one can only take OTUs that represent at least 20% of reads in at least one sample. Remember we normalized all the sampples to median number of reads (total). We are left with only 33 OTUS which makes the reading much more easy.

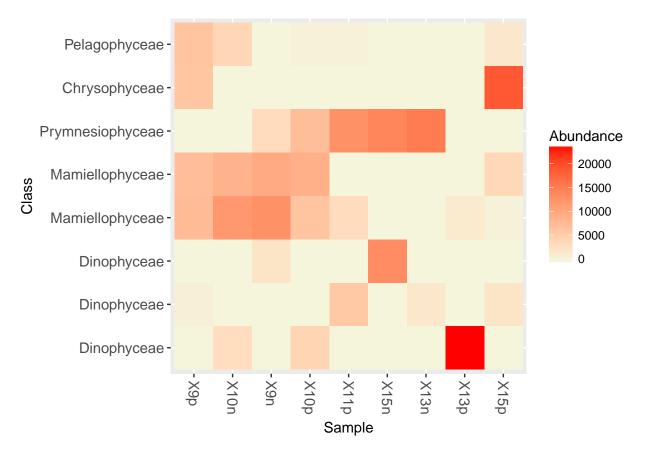
```
carbom_abund <- filter_taxa(carbom, function(x) sum(x > total*0.20) > 0, TRUE)
  carbom abund
## phyloseq-class experiment-level object
## otu table()
                 OTU Table:
                                      [ 8 taxa and 9 samples ]
## sample_data() Sample Data:
                                      [ 9 samples by 27 sample variables ]
## tax_table()
                 Taxonomy Table:
                                      [ 8 taxa by 7 taxonomic ranks ]
  otu_table(carbom_abund)[1:8, 1:5]
## OTU Table:
                        [8 taxa and 5 samples]
##
                         taxa are rows
##
           X10n X10p X11p
                            X13n X13p
## Otu001 11888 6292
                      3245
                               40
                                   1448
             16 7134 12761 15170
                                      24
## Otu002
           8638 8983
## Otu003
                         17
                               19
                                      0
                         29
                               12 23366
## Otu004
           3194 4234
                 518
## Otu007
           3986
                        501
                               13
                                      0
## Otu008
                    8
                       5722
                             1672
                                      0
              1
                    3
                          3
                                      7
## Otu011
              0
                                3
                                2
## Otu015
              0
                   12
                          4
```

```
plot_heatmap(carbom_abund, method = "NMDS", distance = "bray")
```

Warning: Transformation introduced infinite values in discrete y-axis

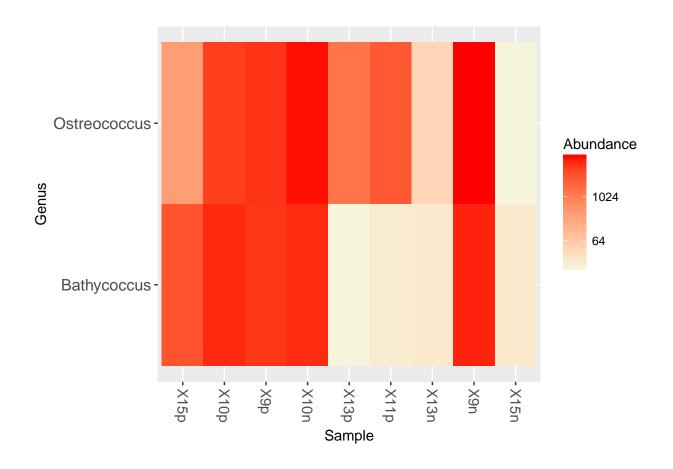


It is possible to use different distances and different multivaraite methods. For example Jaccard distance and MDS and label OTUs with Class, order by Class. We can also change the Palette (the default palette is a bit ugly...).



Another strategy is to do a heatmap for a specific taxonomy group. For example we can taget the Chlorophyta and then label the OTUs using the Genus.

- ## Warning in metaMDS(veganifyOTU(physeq), distance, ...): Stress is (nearly)
 ## zero you may have insufficient data
- ## Warning: Transformation introduced infinite values in discrete y-axis

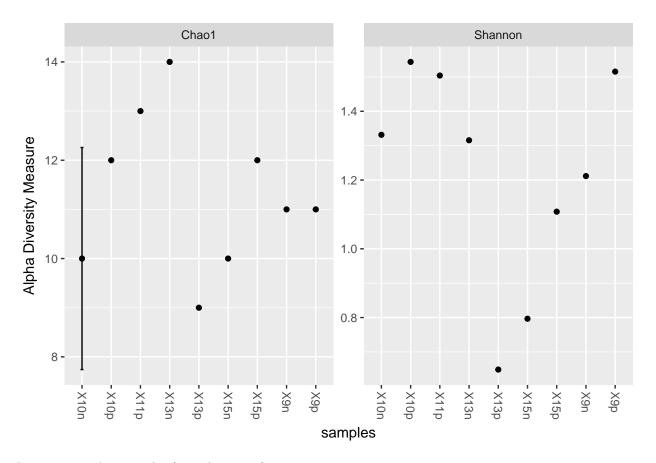


Alpha diversity

Plot Chao1 richness estimator and Shannon diversity estimator.

```
plot_richness(carbom, measures=c("Chao1", "Shannon"))
```

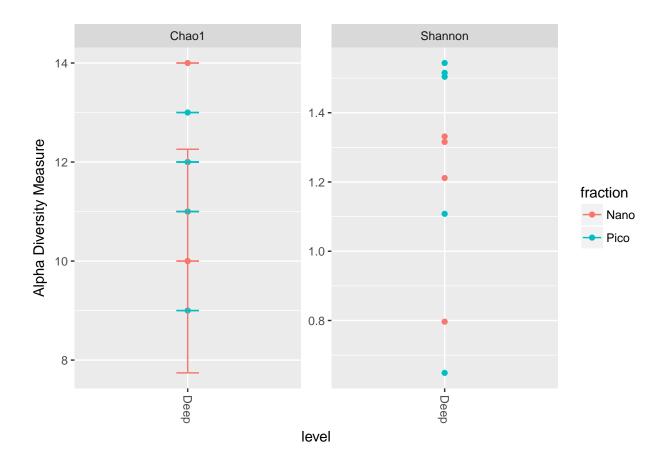
Warning: Removed 9 rows containing missing values (geom_errorbar).



Regroup together samples from the same fraction.

```
plot_richness(carbom, measures=c("Chao1", "Shannon"), x="level", color="fraction")
```

Warning: Removed 9 rows containing missing values (geom_errorbar).



Ordination

Do multivariate analysis based on Bray-Curtis distance and NMDS ordination.

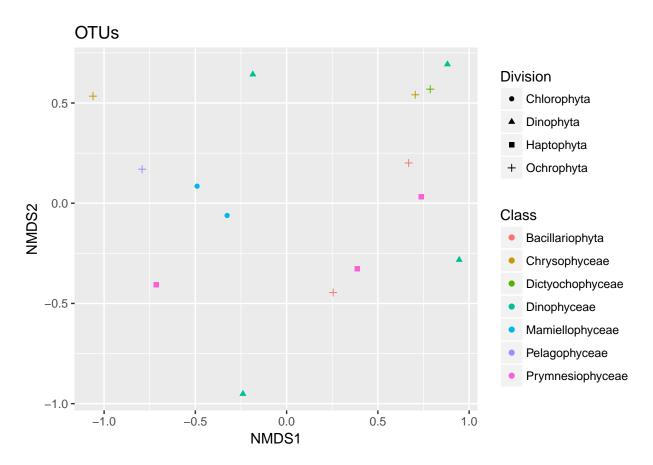
```
carbom.ord <- ordinate(carbom, "NMDS", "bray")
## Square root transformation</pre>
```

```
## Wisconsin double standardization
## Run 0 stress 0.1069238
## Run 1 stress 0.1626382
## Run 2 stress 0.1626382
## Run 3 stress 0.1024627
## ... New best solution
## ... Procrustes: rmse 0.2095694 max resid 0.2941394
## Run 4 stress 0.106924
## Run 5 stress 0.1069241
## Run 6 stress 0.2803014
## Run 7 stress 0.1024627
## ... New best solution
## ... Procrustes: rmse 5.416408e-05 max resid 0.0001053451
## ... Similar to previous best
## Run 8 stress 0.1024628
## ... Procrustes: rmse 7.215393e-05 max resid 0.0001096521
## ... Similar to previous best
## Run 9 stress 0.1175082
## Run 10 stress 0.1069242
```

```
## ... Procrustes: rmse 4.985961e-05 max resid 9.577882e-05
## ... Similar to previous best
## Run 12 stress 0.1024627
## ... Procrustes: rmse 9.026579e-06 max resid 1.719178e-05
## ... Similar to previous best
## Run 13 stress 0.1024627
## ... New best solution
## ... Procrustes: rmse 7.652603e-06 max resid 1.248114e-05
## ... Similar to previous best
## Run 14 stress 0.1069245
## Run 15 stress 0.1069238
## Run 16 stress 0.1024627
## ... New best solution
## ... Procrustes: rmse 4.48323e-05 max resid 9.510291e-05
## ... Similar to previous best
## Run 17 stress 0.1175083
## Run 18 stress 0.117508
## Run 19 stress 0.1024627
## ... New best solution
## ... Procrustes: rmse 8.530685e-06 max resid 1.765247e-05
## ... Similar to previous best
## Run 20 stress 0.1626382
## *** Solution reached
PLot OTUs
 plot_ordination(carbom, carbom.ord, type="taxa", color="Class", shape= "Division",
```

Run 11 stress 0.1024627

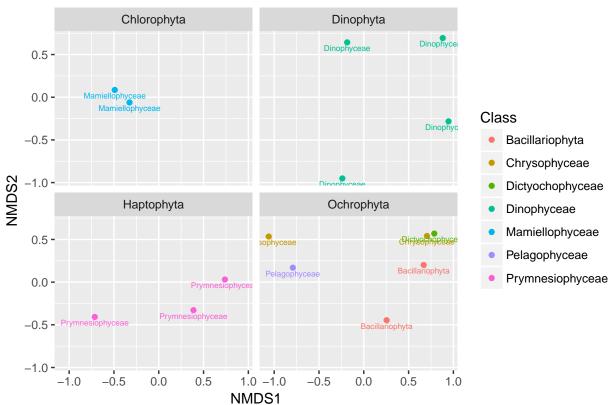
title="OTUs")



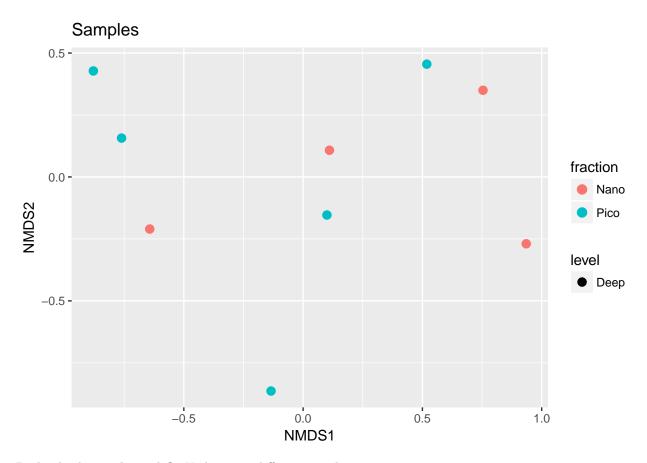
A bit confusing, so make it more easy to visualize by breaking according to taxonomic division.

Warning: Ignoring unknown aesthetics: na.rm





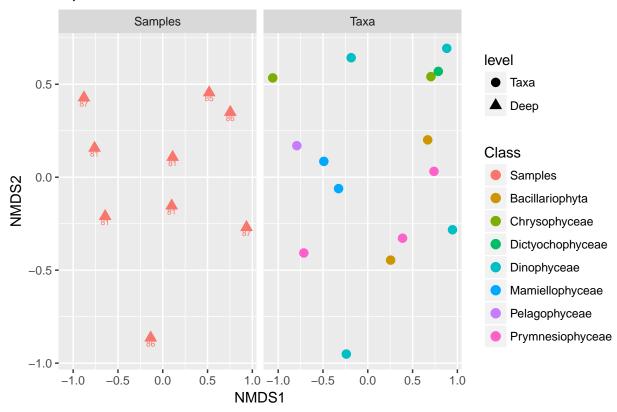
Now display **samples** and enlarge the points to make it more easy to read.



Diplay both samples and OTUs but in 2 different panels.

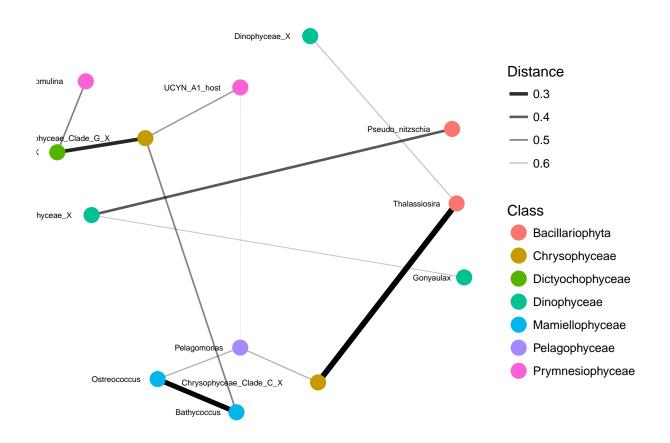
Warning: Ignoring unknown aesthetics: na.rm

biplot



Network analysis

Simple network analysis



This is quite confusing. Let us make it more simple by using only major OTUs

