Chapter 3: ITS data analysis using the R package PhyloSeq

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Load the relevant images, then save this chapter's image as a separate image to retain environment data specific to the ITS processing and analysis workflow.

```
# Set the name for this chapter's image:
chptImage <- "ecobiomics_ITS_analysis.RData"

# Save this chapter's image:
save.image(paste(imageDirPath, chptImage, sep = ""))</pre>
```

When re-starting a session, you can quickly load up the image by running the chunk below:

```
sharedPath <- "/isilon/cfia-ottawa-fallowfield/users/girouxeml/PIRL_working_directory/"
analysis <- "ecobiomics/"
sharedPathAn <- paste(sharedPath, analysis, sep = "")
imageDirPath <- "/home/CFIA-ACIA/girouxeml/GitHub_Repos/r_environments/ecobiomics/"
chptImageA <- "ecobiomics_ITS.RData"
load(paste(imageDirPath, chptImageA, sep = ""))
chptImage <- "ecobiomics_ITS_analysis.RData"
save.image(paste(imageDirPath, chptImage, sep = ""))</pre>
```

Let's get familiar with our phyloseq object created at the end of our sequencing sample processing chapter:

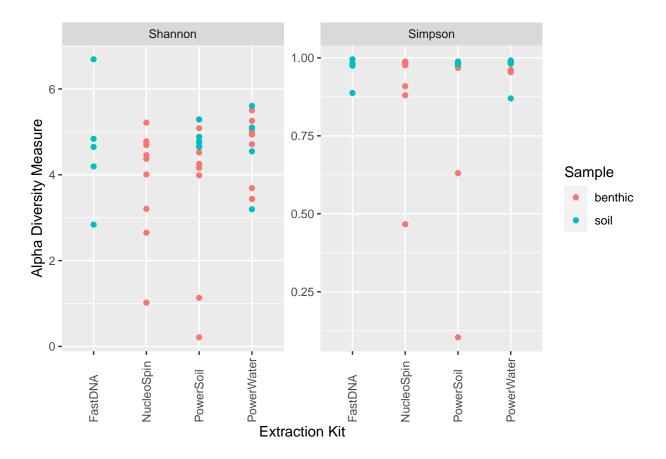
Below I am using the ps, rather than phySeq objects. Recall from the last chunks of Chapter 2, the phySeq object is a phyloseq object without the fitGTRtreeinfo, whilethepsobjectwascreatedleveragingthefitGTR information. The phySeq object can be used instead of the ps object if the optim.pml command wasn't run.

Here is if we filter based on having to know the species:

```
library("phyloseq")
# table(tax_table(ps)[, "Genus"], exclude = NULL)
t <- table(tax_table(ps)[, "Genus"], exclude = NULL)
# head(t[order(-t)])
t[order(-t)][2:10]</pre>
```

```
g_Acidea g_Bannoa g_Mortierella g_Trichoderma g_Venturia
1429 512 197 99 60
g_Malassezia g_Tetracladium g_Myrmecridium g_Alatospora
57 50 46 41
```

Visualize alpha-diversity, phylum:



Prevalence evaluation for species:

```
Species
                    Mean Sum
   s_abeliceae 5.000000
2 s_acerophilum 1.000000
     s__acicola 1.000000
  s__aculeatus 2.666667
   s_acuminata 3.400000 102
5
6
      s__adusta 1.000000
             Species
                         Mean Sum
427
                <NA> 1.422086 8834
124
          s extrema 1.000000 1429
271 s_ogasawarensis 1.000000 512
349
        s_schulzeri 3.465116 149
        s_harzianum 1.359551 121
172
295 s__piceae-abietis 6.562500 105
```

Prevalence evaluation for phyla:

```
colnames(prevalenceTblPhyla) <- c("Phylum", "Mean", "Sum")
prevalenceTblPhyla[order(-prevalenceTblPhyla[,3]),]</pre>
```

```
Phylum
                             Mean Sum
2
           p__Ascomycota 1.700448 9480
4
        p__Basidiomycota 1.189139 2584
14
                    <NA> 1.164004 2186
   p_Mortierellomycota 1.777228
10
      p__Chytridiomycota 1.292857
6
                                   181
7
        p__Glomeromycota 1.111111
                                    30
11
         p_Mucoromycota 1.230769
                                    16
        p_Rozellomycota 1.428571
13
                                    10
1
      p__Aphelidiomycota 1.500000
                                     6
12
        p__Olpidiomycota 1.000000
                                     6
5 p_Blastocladiomycota 1.666667
                                     5
9
  p_Monoblepharomycota 1.000000
                                     4
      p__Kickxellomycota 1.000000
                                     3
    p_Basidiobolomycota 1.000000
3
                                     1
```

From the above calculations, there are a few low-abundance Phylas that appear in less than 10 samples:

Aphelidiomycota

Basidiobolomycota

Blastocladiomycota

Kickxellomycota

Monoble pharomy cota

Olpidiomycota

Rozellomycota

Filter entries with unidentified Phylum, or those phyla that appear in less than 10 samples:

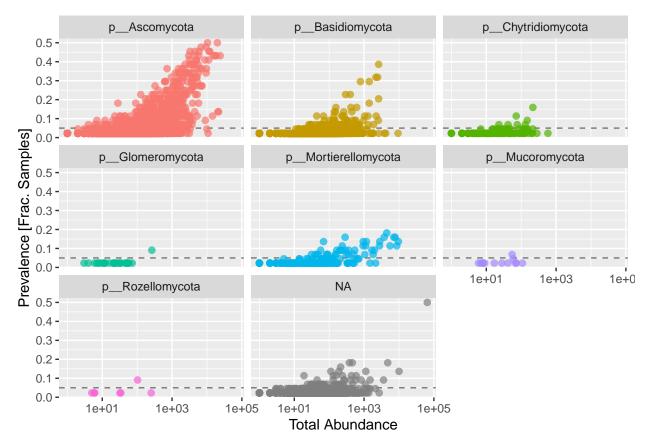
```
[1] "Kingdom" "Phylum" "Class" "Order" "Family" "Genus" "Species"

p__Ascomycota p__Basidiomycota p__Chytridiomycota
5575 2173 140

p__Glomeromycota p__Mortierellomycota p__Mucoromycota
27 202 13
```

```
s__abeliceae s__acerophilum
                                            s__aculeatus
                             s__acicola
                                                           s__acuminata
                                                                     30
                                        2
  s__adusta
          4
    s__extrema s__ogasawarensis
                                   s_harzianum
                                                  s__minutissima
          1429
                           512
 s__schulzeri
                   s__elongata
                                     s__hyalina
                                                    s__acuminata
                            33
                                             33
           43
  s__aquaticus
           30
```

Plot Phylum:

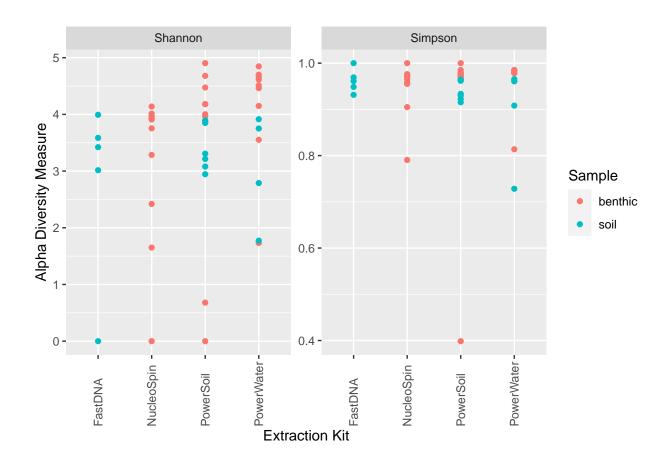


Each point in the above plots is a different taxa, Phylum.

```
prevalenceThreshold = 0.05*nsamples(ps1)
prevalenceThreshold
```

[1] 2.2

The taxa with a prevalence threshold less than the one set in the above chunk are removed using prune_taxa and put into a new phyloseq object, ps2, and we look at the resulting richness plot:



[1] "Kingdom" "Phylum" "Class" "Order" "Family" "Genus" "Species"

Phylum - curiosity:

Curious about the mean and sum prevalence after keeping only taxa passing prevalenceThreshold:

Phylum Mean Sum

```
1
         p__Ascomycota 6.277439 4118
2
      p_Basidiomycota 4.890411
                                 357
3
    p__Chytridiomycota 4.000000
                                  28
4
      p__Glomeromycota 4.000000
                                   4
5 p Mortierellomycota 4.450000
                                 178
       p__Mucoromycota 3.000000
6
                                   3
7
     p Rozellomycota 4.000000
                                   4
8
                  <NA> 4.510638
                                 212
```

Note: I am assuming that the mean is the mean number of times the phylum was seen in a sample for all samples in which it was seen, while the sum is the total times it was seen across all samples. Ascomycota was seen a total of 1,857 times, with about 6 occurances per sample, while Mucoromycota was seen 3 times total ans the mean is simply 3 because when it was seen, it was all three in one sample.

Number of unique phyla, genera and species, across all samples:

```
library("phyloseq")
uniqueClasses <- c("Phylum", "Genus", "Species")
for(i in unique(uniqueClasses))
  cat(cat(i), length(phyloseq::get_taxa_unique(ps2, taxonomic.rank = i)), "\n")

Phylum 8
Genus 139
Species 132</pre>
```

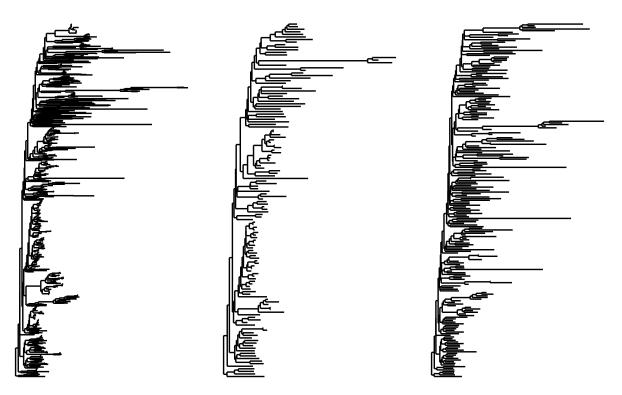
The tax_glom function of phyloseq merges species that have the same taxonomy at certain taxonomic rank, using categorical data. The tip_glom function agglomerates tree tips into a single taxa if they are separated by less than a height specified by h.

```
library("phyloseq")
ps3 <- phyloseq::tax_glom(ps2, "Genus", NArm = TRUE)
h1 = 0.4
ps4 <- phyloseq::tip_glom(ps2, h = h1)</pre>
```

Below we will look at plots of our trees before agglomeration, with agglomeration using tax_glom, and with agglomeration by tip separation using tip_glom:

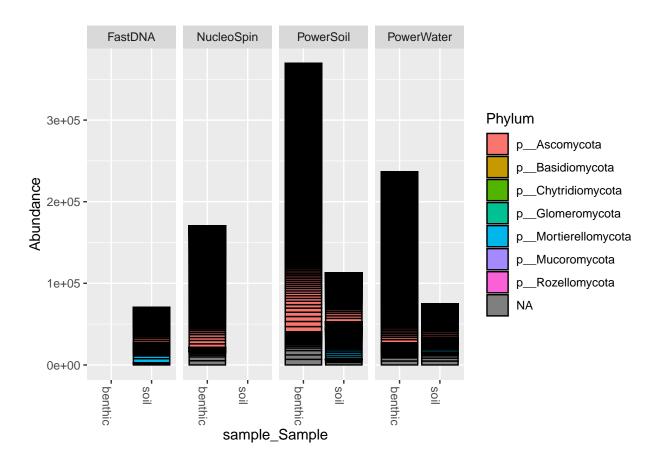
Before Agglomeration By Genus

By Height



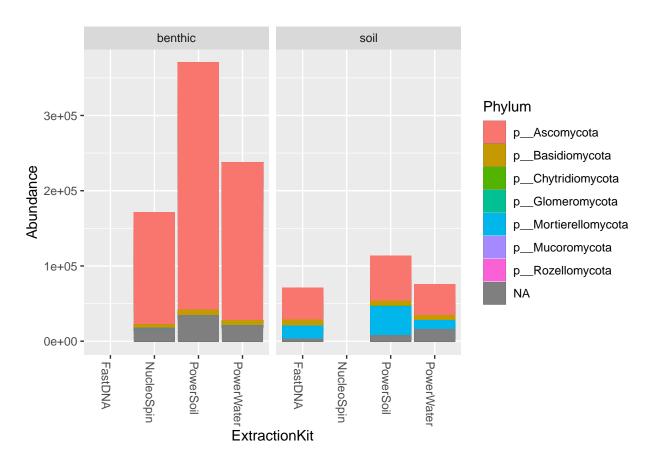
From here on we will continue using the ps2 phyloseq object, that has and the 'NA', low-abundance, and prevalence threshold filters applied.

```
Warning in psmelt(physeq): The sample variables:
Sample
have been renamed to:
sample_Sample
to avoid conflicts with special phyloseq plot attribute names.
```



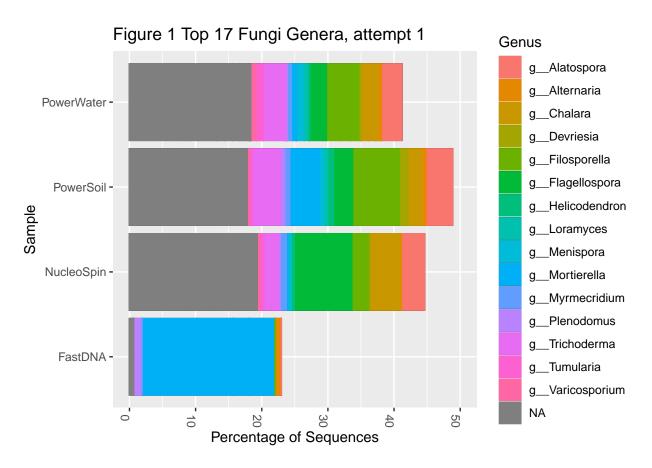
```
Warning in psmelt(physeq): The sample variables:
Sample
have been renamed to:
sample_Sample
to avoid conflicts with special phyloseq plot attribute names.
```

plotPhylum



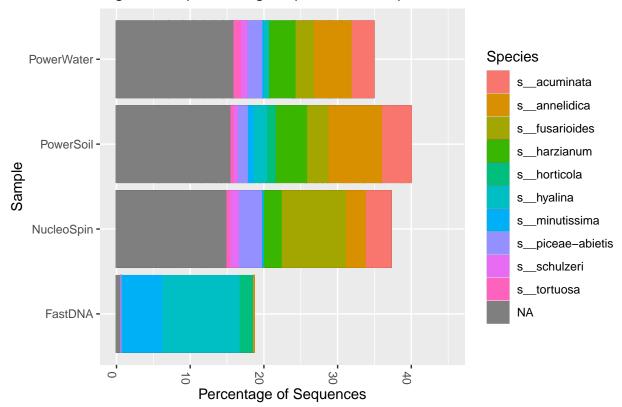
```
library("phyloseq")
library("ggplot2")
            <- names(sort(phyloseq::taxa_sums(ps2), TRUE)[1:41])</pre>
taxTabGenus <- cbind(phyloseq::tax_table(ps2), Genus = NA)
taxTabGenus[topGenus, "Genus"] <- as(tax_table(ps2)[topGenus, "Genus"],</pre>
                                       "character")
tax_table(ps2) <- phyloseq::tax_table(taxTabGenus)</pre>
ps2m <- merge_samples(ps2, "ExtractionKit")</pre>
sample_data(ps2m)$ExtractionKit <- levels(sample_data(ps2)$ExtractionKit)</pre>
ps2m <- phyloseq::transform_sample_counts(ps2m, function(x) 100 * x/sum(x))
ps2mTop = prune_taxa(topGenus, ps2m)
title = "Figure 1 Top 17 Fungi Genera, attempt 1"
plotGenus <- plot_bar(ps2mTop,</pre>
                       \#x = "Sample",
                       fill = "Genus",
                       title = title) +
  coord_flip() +
  ylab("Percentage of Sequences") + ylim(0, 50) +
  geom_bar(aes(color = Genus, fill = Genus),
```

```
stat = "identity", position = "stack")
plotGenus
```

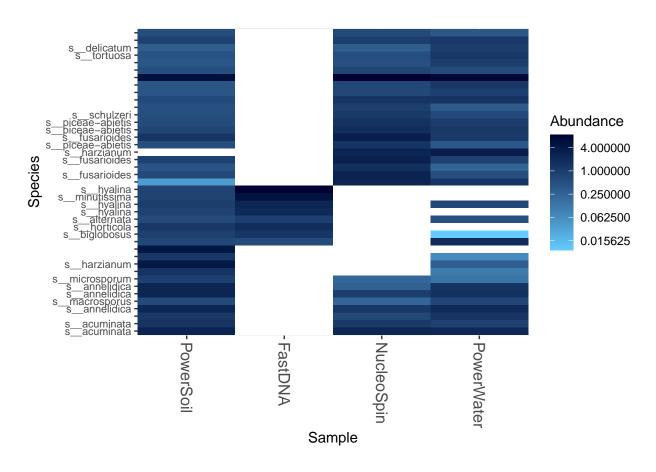


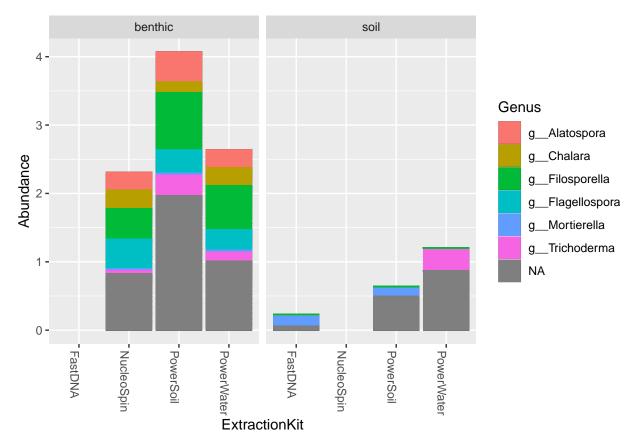
```
library("phyloseq")
library("ggplot2")
topSpecies <- names(sort(taxa_sums(ps2), TRUE)[1:27])</pre>
taxTabSpp <- cbind(phyloseq::tax_table(ps2), Species = NA)</pre>
taxTabSpp[topSpecies, "Species"] <- as(phyloseq::tax_table(ps2)[topSpecies, "Species"],</pre>
                                          "character")
tax_table(ps2) <- phyloseq::tax_table(taxTabSpp)</pre>
ps2mSpp <- phyloseq::merge_samples(ps2, "ExtractionKit")</pre>
sample_data(ps2mSpp)$ExtractionKit <- levels(sample_data(ps2)$ExtractionKit)</pre>
ps2mSpp <- phyloseq::transform_sample_counts(ps2mSpp, function(x) 100 * x/sum(x))
ps2mSppTop = prune_taxa(topSpecies, ps2mSpp)
title = "Figure 2 Top 17 Fungal Species, attempt 1"
plotSpecies <- plot_bar(ps2mSppTop,</pre>
                         x = "Sample",
                         fill = "Species",
                         title = title) +
```

Figure 2 Top 17 Fungal Species, attempt 1



Below I'm testing what a heatmap would look like for taxa abundance across extraction kits:



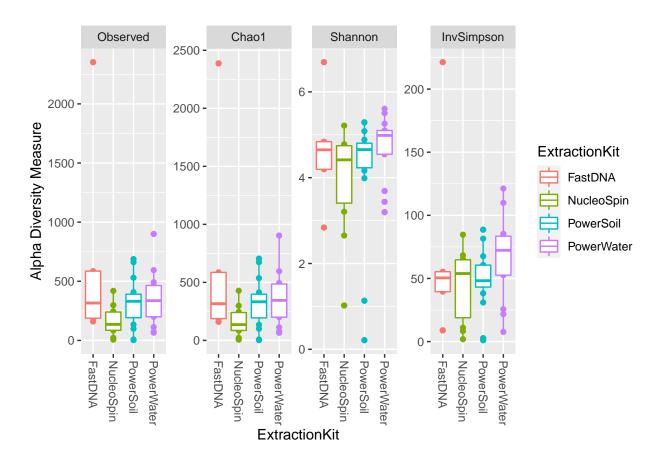


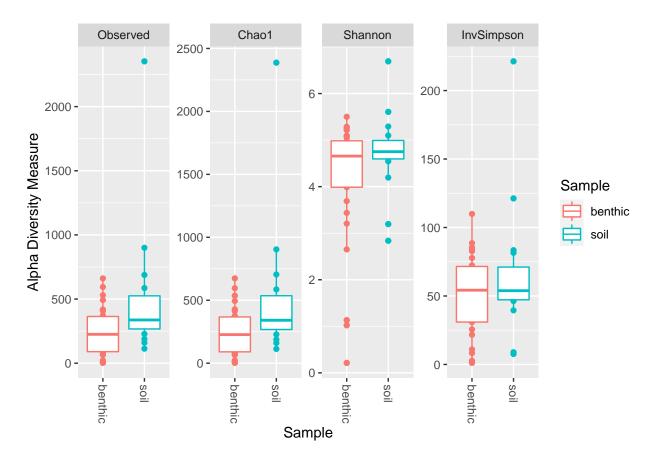
Fo help see:

 $https://www.gdc-docs.ethz.ch/MDA/handouts/MDA20_PhyloseqFormation_Mahendra_Mariadassou.pdf$

Not yet working:

Transform data to proportions as appropriate for Bray-Curtis distances:





Statistical analysis Post hoc comparisons between the four tested methods were made using the Tukey HSD test. OTUs or OTUs pooled at phylum, class, order, family or genera level with different abundances were identified using a generalized linear model where the counts follow an overdispersed Poisson distribution (Kristiansson, Hugenholtz and Dalevi 2009; Jonsson et al.2016). The p-values were corrected for multiple testing using the false discovery rate (FDR) method. The OTU abundance was used for principal component analysis (PCA). Shared OTUs between DNA extraction methods were graphically visualised in Venn diagrams using the corresponding OTU tables exported from QIIME. The hypergeometric distribution was used to test the distribution of gram negatives and gram positives among the taxa identified with the respective four DNA extraction methods. Pearson correlations were used to test for correlations between descriptors of DNA quantity and quality (Table 1), and descriptors of taxonomic diversity (Table 2). The statistical significance for all the analyses was set to P < 0.05 or FDR< 0.05. All statistical analyses were carried out using the R v.3.2.0 software (R Core Team 2013).

Table 2. Detected 16S rRNA richness and biodiversity from marine periphyton biofilm DNA extracted with the four studied methods.

FastDNA Soil PowerPlant PowerBiofilm PlantDNAzol

n 3 2 3 3 P-values No. of OTUs 666 \pm 42 704 \pm 58 809 \pm 11 791 \pm 7 P < 0.05 No. of phyla 17 \pm 1 17 \pm 1 18 \pm 0 ns No. of classes 39 \pm 1 40 \pm 1 40 \pm 2 41 \pm 1 ns No. of orders 68 \pm 3 70 \pm 3 71 \pm 2 72 \pm 1 ns No. of families 91 \pm 2 95 \pm 6 104 \pm 3 106 \pm 2 P < 0.05 No. of genera 141 \pm 4 145 \pm 12 159 \pm 4 162 \pm 3 P < 0.05

Each value represents the arithmetic mean \pm standard error of the mean. n: number of replicates. Statistical significance between extraction methods is denoted as P < 0.05 (ANOVA).

ns: indicates no statistically significant differences between extraction methods.