

MB590-012 Microbiome Analysis

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Load and install R packages

```
library(phyloseq)
library(microbiome)
library(ggplot2)
library(tidyverse)
library(compositions)
library(rmarkdown)
library(knitr)
library(Biostrings)
library(vegan)

#install.packages("RColorBrewer")
library(RColorBrewer)
#install.packages("reshape2")
library(reshape2)
#devtools::install_github("Russel88/MicEco")
library(MicEco)
```

CODING EXERCISES

Please submit as a knitted html or pdf markdown to GitHub due on 3/9

1. Subset to *Vaccinium* unique OTUs and clr transform

- Goal is to retain only fungal OTUs found uniquely associated with *Vaccinium* by removing *Pinus* OTUs
- p1 venn diagram can help you to confirm expected numbers
- use `phylosmith::unique_taxa` to identify taxa associated only with *Pinus* in `ps_nosing`
 - https://schuyler-smith.github.io/phylosmith/analytics.html#unique_taxa
 - `devtools::install_github("schuyler-smith/phylosmith")`
 - `library(phylosmith)`
`library(phylosmith)`
- convert list to vector using `base::unlist`
 - <https://www.rdocumentation.org/packages/base/versions/3.6.2/topics/unlist>
 - note that this gives you unique *Pinus* OTUs + OTUs shared with *Pinus*
- export list of all taxa with `phyloseq::taxa_names` from `ps_nosing`
 - make new `ps` object `ps_vacc` by subsetting the list by removing taxa from *Pinus*
 - hint: look back at code from `lulu`
- in new `ps` object, `ps_vacc`
 - use remaining taxa list to retain only truly unique taxa with `phyloseq::prune_taxa`
 - use `phyloseq::subset_samples` to limit to `Species=="Vaccinium"`
 - check for and remove new singletons
- create `ps_vacc_clr` with clr transformed `otu_table` using `microbiome::transform`
- include new *Vaccinium* venn diagram by `EcoType`
- optional: if you have time and want to practice more, repeat for *Pinus*

```
# Subset the ps object for Vacc
library(phylosmith)
pine_uniq <- phylosmith::unique_taxa(ps_nosing, "Species", subset="Pinus")
pine_uniq <- unlist(pine_uniq) # convert from list to vector
str(pine_uniq) # confirm conversion
```

```
## Named chr [1:493] "OTU4" "OTU5" "OTU1" "OTU3" "OTU16" "OTU2" "OTU7" ...
## - attr(*, "names")= chr [1:493] "Pinus1" "Pinus2" "Pinus3" "Pinus4" ...
```

```
length(pine_uniq) # count num of taxa = 493 (pine + pine shared with vacc)
```

```
## [1] 493
```

```
allTaxa <- phyloseq::taxa_names(ps_nosing) # make vector of taxa names
vaccTaxa <- allTaxa[!(allTaxa %in% pine_uniq)] # remove pine taxa
ps_vacc <- phyloseq::prune_taxa(vaccTaxa, ps_nosing) # retain vacc only taxa in ps obj
ps_vacc # confirm 699 taxa
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 699 taxa and 85 samples ]
## sample_data() Sample Data: [ 85 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 699 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 699 reference sequences ]
```

```
ps_vacc <- phyloseq::subset_samples(ps_vacc, Species == "Vaccinium")
ps_vacc # confirm that # of samples is reduced from 85 to 38
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 699 taxa and 38 samples ]
## sample_data() Sample Data: [ 38 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 699 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 699 reference sequences ]
```

```
phyloseq::sample_names(ps_vacc) #check that these are only "TVxxx" samples
```

```
## [1] "T2V10" "T2V1" "T2V2" "T2V3" "T2V4" "T2V5" "T2V6" "T2V7" "T2V8"
## [10] "T2V9" "T3V10" "T3V1" "T3V2" "T3V3" "T3V4" "T3V5" "T3V6" "T3V7"
## [19] "T3V8" "T3V9" "T4V10" "T4V1" "T4V2" "T4V3" "T4V4" "T4V6" "T4V8"
## [28] "T4V9" "T5V10" "T5V1" "T5V2" "T5V3" "T5V4" "T5V5" "T5V6" "T5V7"
## [37] "T5V8" "T5V9"
```

```
# check for and remove singletons if needed
ps_vacc <- phyloseq::prune_taxa(phyloseq::taxa_sums(ps_vacc) > 1, ps_vacc)
phyloseq::ntaxa(ps_vacc) # no losses
```

```
## [1] 699
```

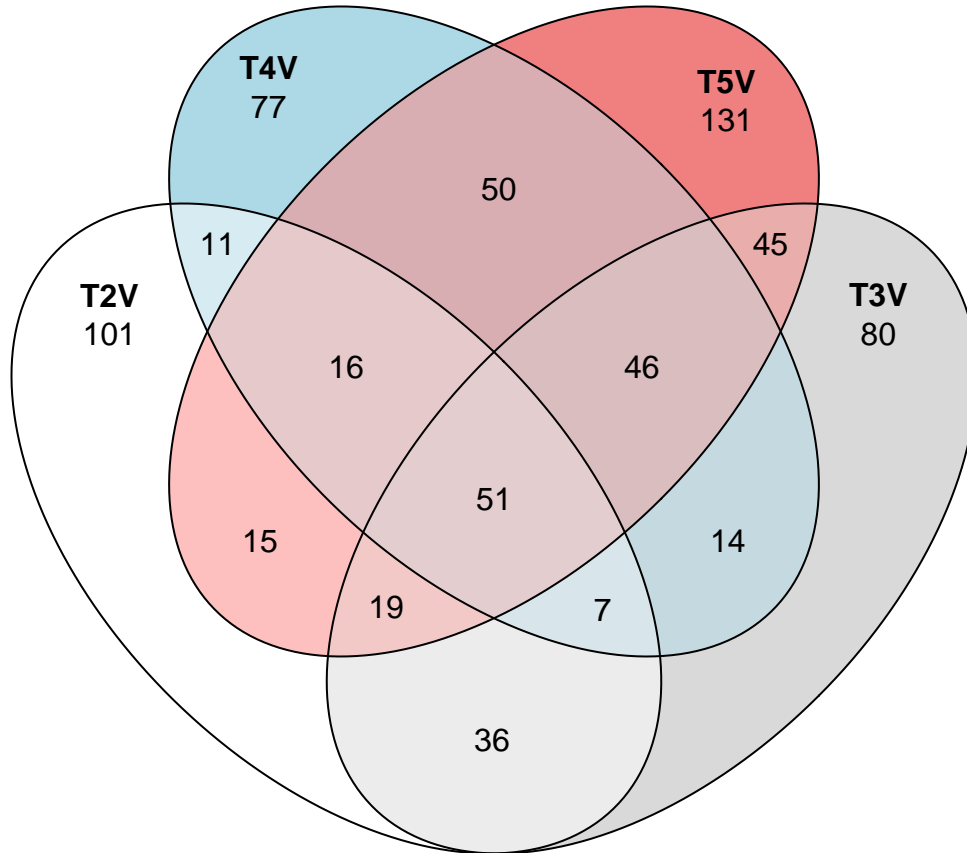
```
# check for and remove samples with zero row sums if needed
phyloseq::nsamples(ps_vacc)
```

```
## [1] 38
```

```
ps_vacc <- phyloseq::prune_samples(phyloseq::sample_sums(ps_vacc)>0, ps_vacc)
phyloseq::nsamples(ps_vacc) # no change 38 samples
```

```
## [1] 38
```

```
# Venn diagram
MicEco::ps_venn(ps_vacc, "EcoType", fraction=0, weight=FALSE, type="counts", relative=FALSE, plot=TRUE)
```



```
# 51 taxa found in all EcoTypes
```

```
# transforms
```

```
ps_vacc_ra <- microbiome::transform(ps_vacc, "compositional")
ps_vacc_clr <- microbiome::transform(ps_vacc, transform="clr")
```

```
# Optional
```

```
# Subset the ps object for Pine
```

```
vacc_uniq <- phylosmith::unique_taxa(ps_nosing, "Species", subset="Vaccinium")
vacc_uniq <- unlist(vacc_uniq)
str(vacc_uniq)
```

```
## Named chr [1:1052] "OTU4" "OTU5" "OTU1" "OTU3" "OTU16" "OTU2" "OTU7" ...
```

```
## - attr(*, "names")= chr [1:1052] "Vaccinium1" "Vaccinium2" "Vaccinium3" "Vaccinium4" ...
```

```
length(vacc_uniq) #1052 (vacc + vacc shared with pine)
```

```
## [1] 1052
```

```
pineTaxa <- allTaxa[!(allTaxa %in% vacc_uniq)] # remove vacc taxa
ps_pine <- phyloseq::prune_taxa(pineTaxa, ps_nosing) # retain pine only taxa in ps obj
ps_pine # confirm 140 taxa
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 140 taxa and 85 samples ]
## sample_data() Sample Data: [ 85 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 140 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 140 reference sequences ]
```

```
ps_pine <- phyloseq::subset_samples(ps_pine, Species == "Pinus")
ps_pine # confirm that # of samples is reduced from 85 to 47
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 140 taxa and 47 samples ]
## sample_data() Sample Data: [ 47 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 140 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 140 reference sequences ]
```

```
phyloseq::sample_names(ps_pine) #check that "TVxxx" samples are gone
```

```
## [1] "T1P10" "T1P1" "T1P2" "T1P3" "T1P4" "T1P5" "T1P6" "T1P7" "T1P8"
## [10] "T1P9" "T2P10" "T2P1" "T2P2" "T2P3" "T2P5" "T2P8" "T2P9" "T3P10"
## [19] "T3P1" "T3P2" "T3P3" "T3P4" "T3P5" "T3P6" "T3P7" "T3P8" "T3P9"
## [28] "T4P10" "T4P1" "T4P2" "T4P3" "T4P4" "T4P5" "T4P6" "T4P7" "T4P8"
## [37] "T4P9" "T5P10" "T5P1" "T5P2" "T5P3" "T5P4" "T5P5" "T5P6" "T5P7"
## [46] "T5P8" "T5P9"
```

```
# check for and remove singletons if needed
ps_pine <- phyloseq::prune_taxa(phyloseq::taxa_sums(ps_pine) > 1, ps_pine)
phyloseq::ntaxa(ps_pine) # no losses
```

```
## [1] 140
```

```
# check for and remove samples with zero row sums if needed
nsamples(ps_pine)
```

```
## [1] 47
```

```
ps_pine <- phyloseq::prune_samples(phyloseq::sample_sums(ps_pine)>0, ps_pine)
nsamples(ps_pine) # no change 47 samples
```

```
## [1] 47
```

2. Examine core microbiome for *Vaccinium* only

- for one detection and prevalence level, compare clr and rel abund data transforms
- vary detection and prevalence for clr data
 - adjust only detection up and down (at least 3 levels)
 - adjust only prevalence up and down (at least 3 levels)
- describe the effects on the size and characteristics of the core community
- optional: if you want more practice, repeat for *Pinus*

```
# Repeat the following code set with different detection and prevalence settings
ps_vacc_ra_core <- microbiome::core(ps_vacc_ra, detection = 0.001, prevalence = 50/100)
ps_vacc_ra_core

## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 8 taxa and 38 samples ]
## sample_data() Sample Data: [ 38 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 8 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 8 reference sequences ]

microbiome::taxa(ps_vacc_ra_core)

## [1] "OTU8" "OTU3130" "OTU35" "OTU42" "OTU110" "OTU4625" "OTU366"
## [8] "OTU4549"

ps_vacc_clr_core <- microbiome::core(ps_vacc_clr, detection = 0.001, prevalence = 50/100)
ps_vacc_clr_core

## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 18 taxa and 38 samples ]
## sample_data() Sample Data: [ 38 samples by 31 sample variables ]
## tax_table() Taxonomy Table: [ 18 taxa by 5 taxonomic ranks ]
## refseq() DNASTringSet: [ 18 reference sequences ]

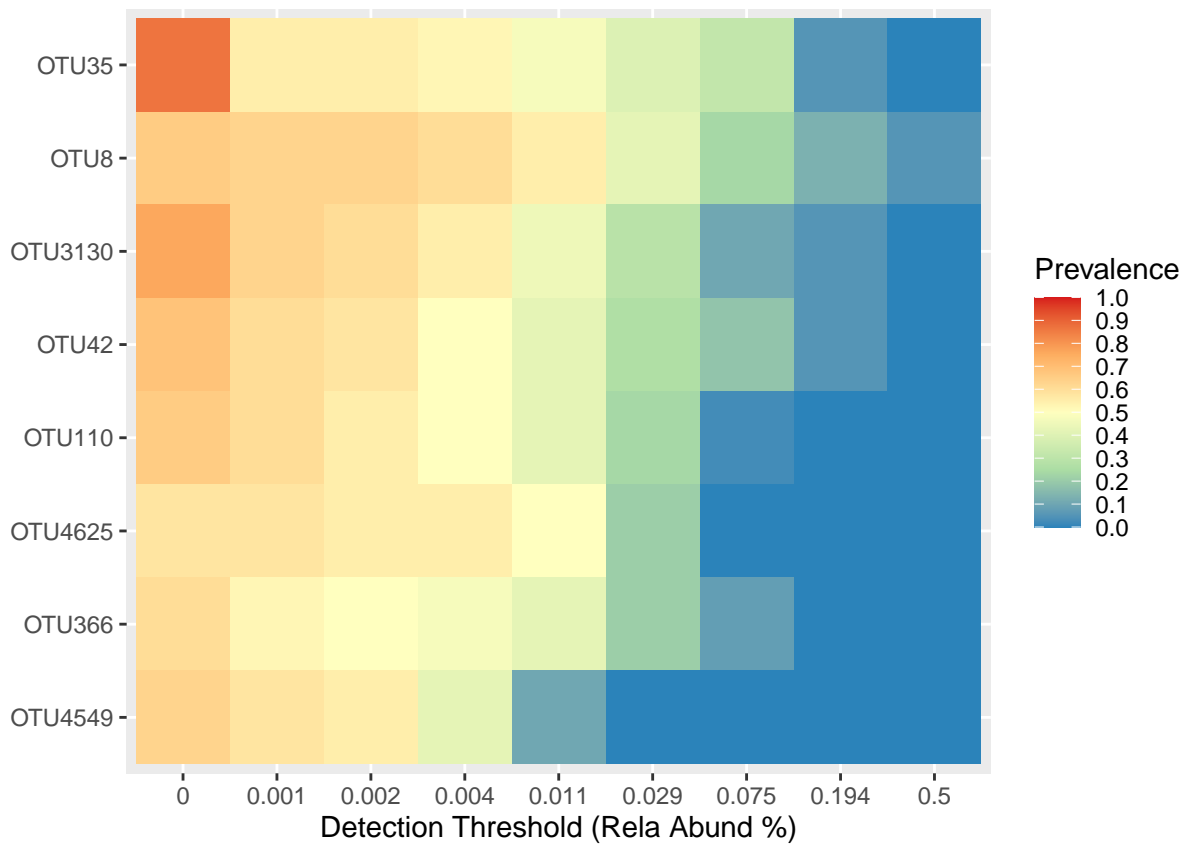
microbiome::taxa(ps_vacc_clr_core)

## [1] "OTU8" "OTU3130" "OTU35" "OTU26" "OTU42" "OTU107" "OTU110"
## [8] "OTU4032" "OTU4625" "OTU89" "OTU366" "OTU4518" "OTU4549" "OTU356"
## [15] "OTU1638" "OTU2502" "OTU4573" "OTU2728"

# Heatmap of core - note that clr shifts everything left to lower rel abund
prevalences <- seq(0.05, 1, 0.05)
detections <- round(10^seq(log10(1e-4), log10(0.5), length = 10), 3)

microbiome::plot_core(ps_vacc_ra_core, plot.type = "heatmap",
  colours = rev(RColorBrewer::brewer.pal(5, "Spectral")),
  prevalences = prevalences,
  detections = detections) +
  ggplot2::labs(x = "Detection Threshold (Rela Abund %)")
```

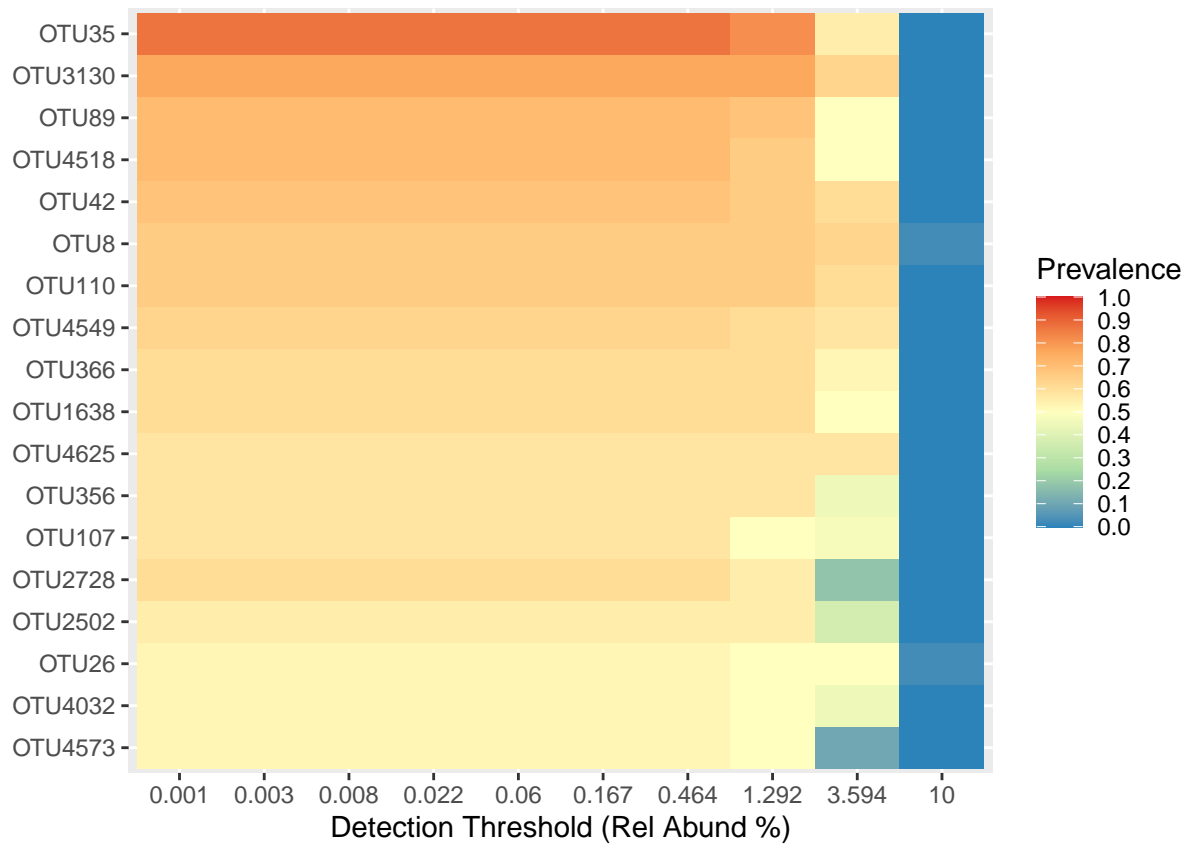
```
## Warning in microbiome::plot_core(ps_vacc_ra_core, plot.type = "heatmap", : The plot_core function is
##      data. The data is not compositional. Make sure that you
##      intend to operate on non-compositional data.
```



```
prevalences <- seq(0.05, 1, 0.05)
detections <- round(10^seq(log10(1e-3), log10(10), length = 10), 3)

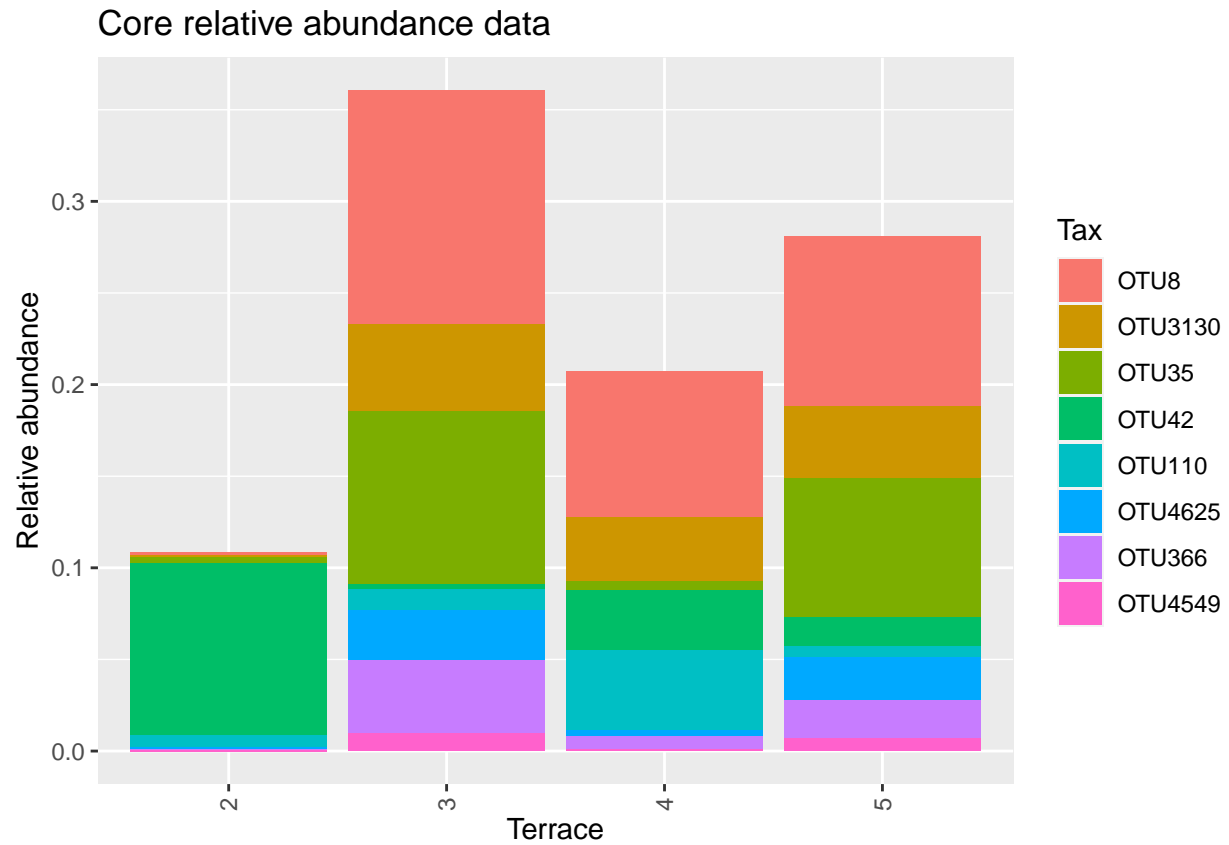
microbiome::plot_core(ps_vacc_clr_core, plot.type = "heatmap",
  colours = rev(RColorBrewer::brewer.pal(5, "Spectral")),
  prevalences = prevalences,
  detections = detections) +
  ggplot2::labs(x = "Detection Threshold (Rel Abund %)")
```

```
## Warning in microbiome::plot_core(ps_vacc_clr_core, plot.type = "heatmap", : The plot_core function is
##      data. The data is not compositional. Make sure that you
##      intend to operate on non-compositional data.
```

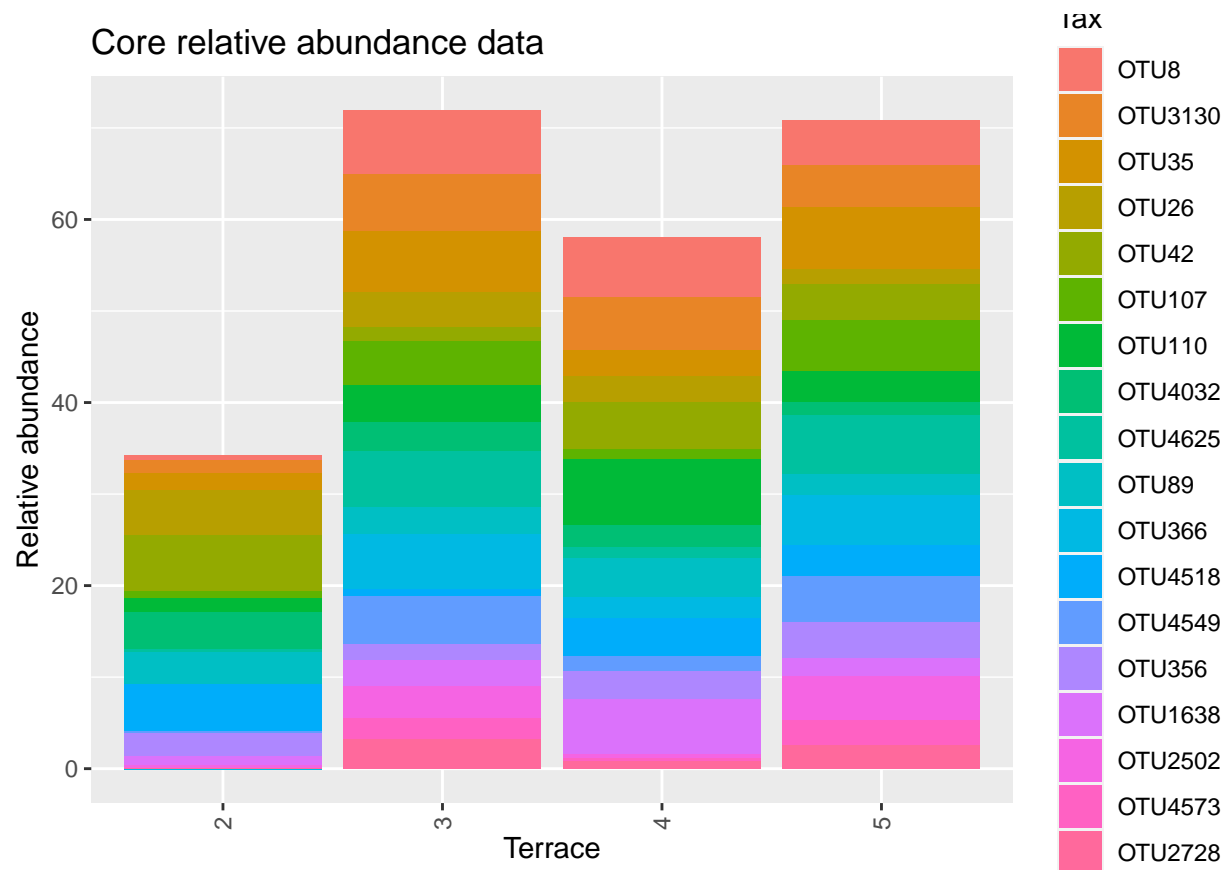


Barplot of core taxa by terrace

```
microbiome::plot_composition(ps_vacc_ra_core,
                             average_by="Terrace",
                             plot.type = "barplot",
                             sample.sort="Terrace") +
  guides(fill = guide_legend(ncol = 1)) +
  labs(x = "Terrace",
       y = "Relative abundance",
       title = "Core relative abundance data")
```

```
microbiome::plot_composition(ps_vacc_clr_core,  
                             average_by="Terrace",  
                             plot.type = "barplot",  
                             sample.sort="Terrace") +  
  guides(fill = guide_legend(ncol = 1)) +  
  labs(x = "Terrace",  
       y = "Relative abundance",  
       title = "Core relative abundance data")
```



3. Identify the core microbiota of built-in soilrep data

- load built-in soilrep data and examine
- remove clipped samples with `phyloseq::subset_samples`
- remove singletons with `phyloseq::prune_taxa`
- identify core with `microbiome::core`
 - indicate why you selected your specific prevalence and detection settings
- produce a table of core ASVs using `kable` and specify column name
- plot results as heatmap, barplot, or other plot of your choice

```
library(phyloseq)
data(soilrep)
ps_sr <- soilrep
sample_data(ps_sr)
```

```
##      Treatment warmed clipped Sample
## a_C026      UC      no      yes    6CC
## a_C066      UU      no      no     3UC
## a_C070      WU     yes      no     5UW
## a_C074      UU      no      no     2UC
## a_C075      WC     yes     yes     5CW
## a_C077      WU     yes      no     4UW
## a_C079      UU      no      no     6UC
## a_C081      UC      no      yes     3CC
## a_C082      UC      no      yes     1CC
## a_C083      UU      no      no     6UC
## a_C084      WC     yes     yes     4CW
## a_C085      UC      no      yes     1CC
## a_C086      UC      no      yes     3CC
## a_C088      WC     yes     yes     5CW
## a_C089      WC     yes     yes     1CW
## a_C090      WC     yes     yes     3CW
## a_C091      UU      no      no     2UC
## a_C093      UU      no      no     2UC
## a_C095      WU     yes      no     6UW
## a_C096      WC     yes     yes     1CW
## a_C098      WU     yes      no     5UW
## a_C099      UU      no      no     4UC
## a_C100      WU     yes      no     2UW
## a_C101      WU     yes      no     3UW
## a_C102      WC     yes     yes     3CW
## a_C116      UC      no      yes     3CC
## a_C125      UC      no      yes     4CC
## a_C126      WU     yes      no     2UW
## a_C127      WU     yes      no     6UW
## a_C128      WU     yes      no     4UW
## a_C130      WC     yes     yes     4CW
## a_C131      UU      no      no     6UC
```

```
## a_C132      WC      yes      yes      5CW
## a_C134      UC      no       yes      6CC
## a_C136      WC      yes      yes      1CW
## a_C137      WC      yes      yes      6CW
## a_C139      UC      no       yes      2CC
## a_C140      UU      no       no       3UC
## a_C141      UU      no       no       1UC
## a_C143      WU      yes      no       3UW
## a_C144      WU      yes      no       4UW
## a_C145      WC      yes      yes      6CW
## a_C146      UU      no       no       5UC
## a_C147      WU      yes      no       2UW
## a_C149      UU      no       no       1UC
## a_C150      UC      no       yes      4CC
## a_C151      WC      yes      yes      3CW
## a_C153      WU      yes      no       1UW
## a_C154      UC      no       yes      2CC
## a_C156      UC      no       yes      1CC
## a_C157      UC      no       yes      5CC
## a_C158      WU      yes      no       6UW
## a_C159      UU      no       no       4UC
## a_C160      WC      yes      yes      2CW
## a_C161      UC      no       yes      5CC
## a_C162      UU      no       no       5UC
```

```
# remove clipped samples (goes from 56 to 28 samples)
ps_sr <- phyloseq::subset_samples(ps_sr, clipped == "no")
ps_sr
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table:      [ 16825 taxa and 28 samples ]
## sample_data() Sample Data:  [ 28 samples by 4 sample variables ]
```

```
# remove singletons (reduces from 16825 to 7250 taxa)
ps_sr <- phyloseq::prune_taxa(taxa_sums(ps_sr) > 1, ps_sr)
phyloseq::ntaxa(ps_sr)
```

```
## [1] 7250
```

```
# check for any samples with zero counts
phyloseq::sample_sums(ps_sr) #none
```

```
## a_C066 a_C070 a_C074 a_C077 a_C079 a_C083 a_C091 a_C093 a_C095 a_C098 a_C099
## 1587 1858 3868 1986 1368 1991 1128 3852 3321 1506 1572
## a_C100 a_C101 a_C126 a_C127 a_C128 a_C131 a_C140 a_C141 a_C143 a_C144 a_C146
## 1543 1833 832 939 1445 1139 1241 862 1217 1166 1038
## a_C147 a_C149 a_C153 a_C158 a_C159 a_C162
## 1065 1218 2104 1104 1192 1393
```

```
# identify core - here using stringent prevalence criteria and moderate detection
ps_sr_core <- microbiome::core(ps_sr, detection = 50/100, prevalence = 90/100)
ps_sr_core
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table:      [ 3 taxa and 28 samples ]
## sample_data() Sample Data:  [ 28 samples by 4 sample variables ]
```

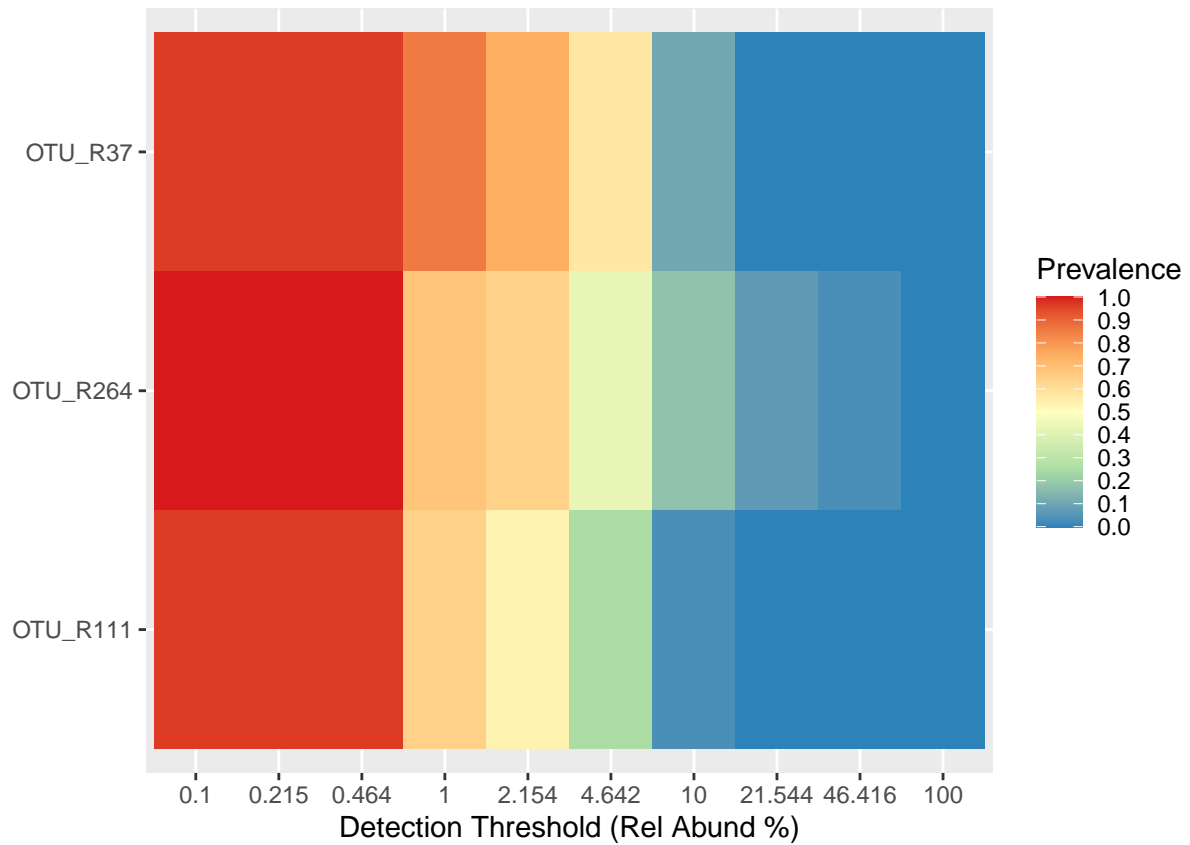
```
# core table
knitr::kable(taxa(ps_sr_core), col.names=c("core ASVs"))
```

core ASVs
OTU_R111
OTU_R264
OTU_R37

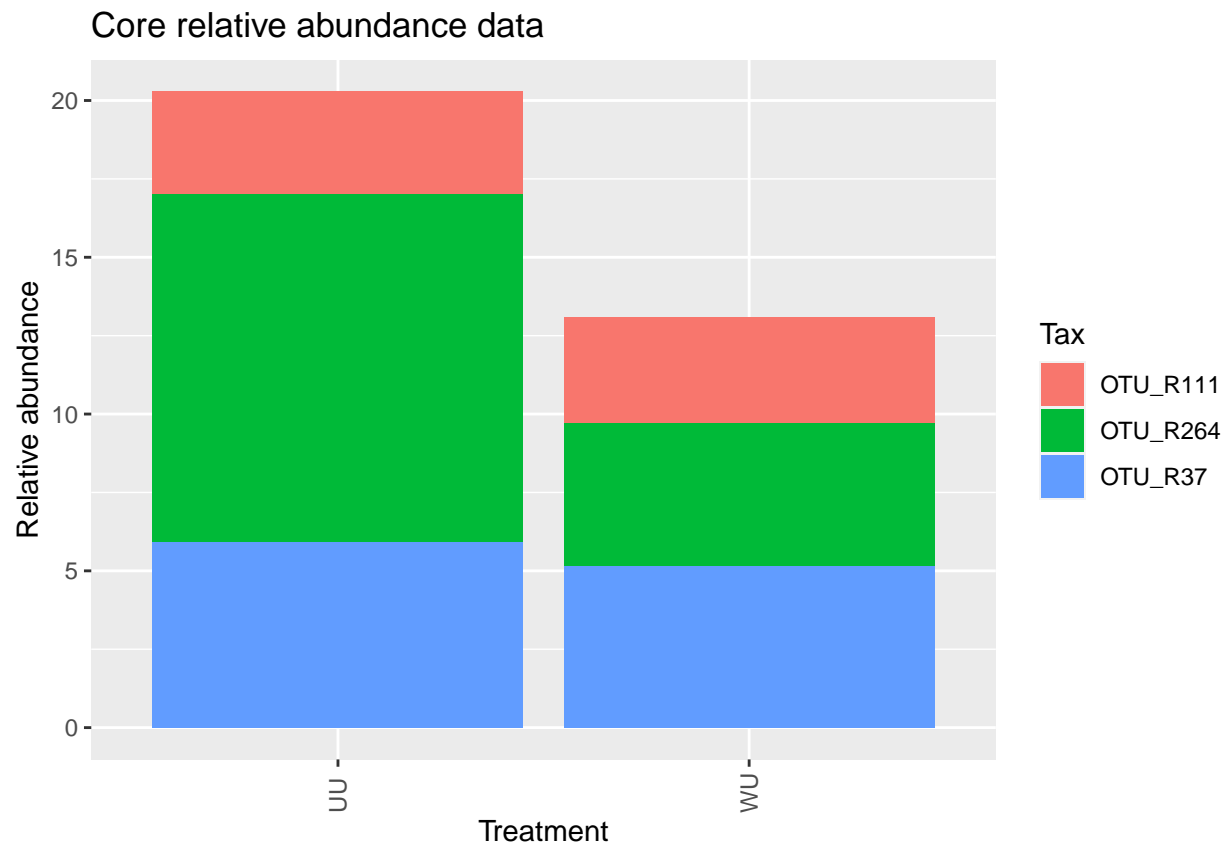
```
# plot options
prevalences <- seq(0.05, 1, 0.05)
detections <- round(10^seq(log10(1e-1), log10(100), length = 10), 3)

microbiome::plot_core(ps_sr_core, plot.type = "heatmap",
  colours = rev(RColorBrewer::brewer.pal(5, "Spectral")),
  prevalences = prevalences,
  detections = detections) +
  ggplot2::labs(x = "Detection Threshold (Rel Abund %)")
```

```
## Warning in microbiome::plot_core(ps_sr_core, plot.type = "heatmap", colours = rev(RColorBrewer::brew
##                               data. The data is not compositional. Make sure that you
##                               intend to operate on non-compositional data.
```



```
microbiome::plot_composition(ps_sr_core,
                             average_by="Treatment",
                             plot.type = "barplot",
                             sample.sort="Treatment") +
  guides(fill = guide_legend(ncol = 1)) +
  labs(x = "Treatment",
       y = "Relative abundance",
       title = "Core relative abundance data")
```



Session Info

```
sessionInfo()
```

```
## R version 4.1.1 (2021-08-10)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 19042)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_United States.1252
## [2] LC_CTYPE=English_United States.1252
## [3] LC_MONETARY=English_United States.1252
## [4] LC_NUMERIC=C
## [5] LC_TIME=English_United States.1252
##
## attached base packages:
## [1] stats4      parallel  stats      graphics  grDevices  utils      datasets
## [8] methods     base
##
## other attached packages:
## [1] phylosmith_1.0.6      MicEco_0.9.17      reshape2_1.4.4
## [4] RColorBrewer_1.1-2    vegan_2.5-7        lattice_0.20-44
## [7] permute_0.9-7         Biostrings_2.60.2  GenomeInfoDb_1.28.4
## [10] XVector_0.32.0        IRanges_2.26.0     S4Vectors_0.30.2
## [13] BiocGenerics_0.38.0   knitr_1.37         rmarkdown_2.12
## [16] compositions_2.0-4    forcats_0.5.1      stringr_1.4.0
## [19] dplyr_1.0.8           purrr_0.3.4        readr_2.1.2
## [22] tidyr_1.2.0           tibble_3.1.6       tidyverse_1.3.1
## [25] microbiome_1.14.0     ggplot2_3.3.5      phyloseq_1.36.0
##
## loaded via a namespace (and not attached):
## [1] readxl_1.3.1          snow_0.4-4          backports_1.4.1
## [4] Hmisc_4.6-0           plyr_1.8.6          igraph_1.2.11
## [7] polylablr_0.2.0       splines_4.1.1       digest_0.6.29
## [10] foreach_1.5.2         htmltools_0.5.2     viridis_0.6.2
## [13] fansi_1.0.2           magrittr_2.0.2      checkmate_2.0.0
## [16] cluster_2.1.2         tzdb_0.2.0          graphlayouts_0.8.0
## [19] modelr_0.1.8          RcppParallel_5.1.5  bayesm_3.1-4
## [22] bdsmatrix_1.3-4       jpeg_0.1-9          colorspace_2.0-3
## [25] ggrepel_0.9.1         rvest_1.0.2         haven_2.4.3
## [28] xfun_0.29             crayon_1.5.0        RCurl_1.98-1.6
## [31] jsonlite_1.8.0        survival_3.2-11     iterators_1.0.14
## [34] ape_5.6-2            glue_1.6.2          polyclip_1.10-0
## [37] gtable_0.3.0          zlibbioc_1.38.0     Rhdf5lib_1.14.2
## [40] DEoptimR_1.0-10       abind_1.4-5         scales_1.1.1
## [43] pheatmap_1.0.12      mvtnorm_1.1-3       DBI_1.1.2
## [46] Rcpp_1.0.8           viridisLite_0.4.0   htmlTable_2.4.0
```


## [49] units_0.8-0	proxy_0.4-26	foreign_0.8-81
## [52] Formula_1.2-4	htmlwidgets_1.5.4	httr_1.4.2
## [55] ellipsis_0.3.2	pkgconfig_2.0.3	farver_2.1.0
## [58] nnet_7.3-16	dbplyr_2.1.1	utf8_1.2.2
## [61] tidyselect_1.1.2	labeling_0.4.2	rlang_1.0.1
## [64] munsell_0.5.0	cellranger_1.1.0	tools_4.1.1
## [67] cli_3.2.0	generics_0.1.2	ade4_1.7-18
## [70] broom_0.7.12	evaluate_0.15	biomformat_1.20.0
## [73] fastmap_1.1.0	yaml_2.3.5	fs_1.5.2
## [76] tidygraph_1.2.0	robustbase_0.93-9	ggraph_2.0.5
## [79] nlme_3.1-152	xml2_1.3.3	compiler_4.1.1
## [82] rstudioapi_0.13	png_0.1-7	e1071_1.7-9
## [85] reprex_2.0.1	tweenr_1.0.2	stringi_1.7.6
## [88] highr_0.9	Matrix_1.3-4	classInt_0.4-3
## [91] tensorA_0.36.2	multtest_2.48.0	vctrs_0.3.8
## [94] pillar_1.7.0	lifecycle_1.0.1	rhdf5filters_1.4.0
## [97] eulerr_6.1.1	data.table_1.14.2	bitops_1.0-7
## [100] R6_2.5.1	latticeExtra_0.6-29	KernSmooth_2.23-20
## [103] gridExtra_2.3	codetools_0.2-18	MASS_7.3-54
## [106] assertthat_0.2.1	picante_1.8.2	rhdf5_2.36.0
## [109] withr_2.5.0	GenomeInfoDbData_1.2.6	mgcv_1.8-36
## [112] hms_1.1.1	doSNOW_1.0.20	grid_4.1.1
## [115] rpart_4.1-15	class_7.3-19	Rtsne_0.15
## [118] sf_1.0-7	ggforce_0.3.3	bbmle_1.0.24
## [121] numDeriv_2016.8-1.1	Biobase_2.52.0	lubridate_1.8.0
## [124] base64enc_0.1-3		