

deseq2_analysis.R

arisely

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```
## R script for publication: "Active migration associated with repeatable and specific changes to gut microbiota in migratory shorebirds"
# Alice Risely, David Waite, Beata Ujvari, Bethany Hoye, & Marcel Klaassen

##R version 3.4.1

##workflows:
#https://f1000research.com/articles/5-1492/v1
#http://deneflab.github.io/MicrobeMiseq/demos/mothur_2_phyloseq.html
#DESeq2
#https://www.bioconductor.org/packages/devel/bioc/vignettes/phyloseq/inst/doc/phyloseq-mixture-models.html

##good threads on data transformation
#https://github.com/joey711/phyloseq/issues/492
#https://github.com/joey711/phyloseq/issues/283

#To download phyloseq and DESeq2 from bioconductor run following two lines:

#source('http://bioconductor.org/biocLite.R')
#biocLite('phyloseq')
#biocLite("DESeq2")

#It might mess up some other packages, so packages may need reinstalling (either from CRAN or Bioconductor) as errors come up when loading phyloseq and DESeq2 into library
#I had to reinstall ~10 packages

##setwd() somewhere and download all github files to this location

library(phyloseq)
library(DESeq2)
```

```
## Loading required package: S4Vectors
```

```
## Loading required package: stats4
```

```
## Loading required package: BiocGenerics
```

```
## Loading required package: parallel
```

```
##
## Attaching package: 'BiocGenerics'
```

```
## The following objects are masked from 'package:parallel':
##
##   clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
##   clusterExport, clusterMap, parApply, parCapply, parLapply,
##   parLapplyLB, parRapply, parSapply, parSapplyLB
```

```
## The following objects are masked from 'package:stats':
##
##   IQR, mad, sd, var, xtabs
```

```
## The following objects are masked from 'package:base':
##
##   anyDuplicated, append, as.data.frame, cbind, colMeans,
##   colnames, colSums, do.call, duplicated, eval, evalq, Filter,
##   Find, get, grep, grepl, intersect, is.unsorted, lapply,
##   lengths, Map, mapply, match, mget, order, paste, pmax,
##   pmax.int, pmin, pmin.int, Position, rank, rbind, Reduce,
##   rowMeans, rownames, rowSums, sapply, setdiff, sort, table,
##   tapply, union, unique, unsplit, which, which.max, which.min
```

```
##
## Attaching package: 'S4Vectors'
```

```
## The following object is masked from 'package:base':
##
##   expand.grid
```

```
## Loading required package: IRanges
```

```
##
## Attaching package: 'IRanges'
```

```
## The following object is masked from 'package:phyloseq':
##
##   distance
```

```
## Loading required package: GenomicRanges
```

```
## Loading required package: GenomeInfoDb
```

```
## Loading required package: SummarizedExperiment
```

```
## Loading required package: Biobase
```

```
## Welcome to Bioconductor
##
##   Vignettes contain introductory material; view with
##   'browseVignettes()'. To cite Bioconductor, see
##   'citation("Biobase")', and for packages 'citation("pkgname")'.
```

```
##
## Attaching package: 'Biobase'
```

```
## The following object is masked from 'package:phyloseq':
##
##   sampleNames
```

```
## Loading required package: DelayedArray
```

```
## Loading required package: matrixStats
```

```
## Warning: package 'matrixStats' was built under R version 3.4.2
```

```
##
## Attaching package: 'matrixStats'
```

```
## The following objects are masked from 'package:Biobase':
##
##   anyMissing, rowMedians
```

```
##
## Attaching package: 'DelayedArray'
```

```
## The following objects are masked from 'package:matrixStats':
##
##      colMaxs, colMins, colRanges, rowMaxs, rowMins, rowRanges
```

```
## The following object is masked from 'package:base':
##
##      apply
```

##other packages you may need, althoguh I think a few are included with the phyloseq package so you may not need to load them

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.4.2
```

```
library(vegan)
```

```
## Warning: package 'vegan' was built under R version 3.4.2
```

```
## Loading required package: permute
```

```
## Warning: package 'permute' was built under R version 3.4.2
```

```
## Loading required package: lattice
```

```
## This is vegan 2.4-4
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 3.4.2
```

```
##
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:matrixStats':
##
##      count
```

```
## The following object is masked from 'package:Biobase':
##
##      combine
```

```
## The following objects are masked from 'package:GenomicRanges':
##
##      intersect, setdiff, union
```

```
## The following object is masked from 'package:GenomeInfoDb':
##
##      intersect
```

```
## The following objects are masked from 'package:IRanges':
##
##      collapse, desc, intersect, setdiff, slice, union
```

```
## The following objects are masked from 'package:S4Vectors':
##
##      first, intersect, rename, setdiff, setequal, union
```

```
## The following objects are masked from 'package:BiocGenerics':
##
##   combine, intersect, setdiff, union
```

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

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

```
library(scales)
```

```
## Warning: package 'scales' was built under R version 3.4.2
```

```
library(grid)
library(reshape2)
```

```
## Warning: package 'reshape2' was built under R version 3.4.2
```

```
library(ape)
library(gridExtra)
```

```
## Warning: package 'gridExtra' was built under R version 3.4.2
```

```
##
## Attaching package: 'gridExtra'
```

```
## The following object is masked from 'package:dplyr':
##
##   combine
```

```
## The following object is masked from 'package:Biobase':
##
##   combine
```

```
## The following object is masked from 'package:BiocGenerics':
##
##   combine
```

```
library(ade4)
```

```
## Warning: package 'ade4' was built under R version 3.4.2
```

```
##
## Attaching package: 'ade4'
```

```
## The following object is masked from 'package:GenomicRanges':
##
##   score
```

```
## The following object is masked from 'package:IRanges':
##
##   score
```

```
## The following object is masked from 'package:BiocGenerics':
##
##   score
```

```
library(plyr)
```

```
## Warning: package 'plyr' was built under R version 3.4.2
```

```
## -----
```

```
## 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
```

```
## The following object is masked from 'package:matrixStats':  
##  
##   count
```

```
## The following object is masked from 'package:IRanges':  
##  
##   desc
```

```
## The following object is masked from 'package:S4Vectors':  
##  
##   rename
```

```
library(tidyr)
```

```
## Warning: package 'tidyr' was built under R version 3.4.2
```

```
##  
## Attaching package: 'tidyr'
```

```
## The following object is masked from 'package:reshape2':  
##  
##   smiths
```

```
## The following object is masked from 'package:S4Vectors':  
##  
##   expand
```

```
library(data.table)
```

```
##  
## Attaching package: 'data.table'
```

```
## The following objects are masked from 'package:reshape2':  
##  
##   dcast, melt
```

```
## The following objects are masked from 'package:dplyr':  
##  
##   between, first, last
```

```
## The following object is masked from 'package:SummarizedExperiment':  
##  
##   shift
```

```
## The following object is masked from 'package:GenomicRanges':
##
##      shift
```

```
## The following object is masked from 'package:IRanges':
##
##      shift
```

```
## The following objects are masked from 'package:S4Vectors':
##
##      first, second
```

```
library(stringr)
```

```
## Warning: package 'stringr' was built under R version 3.4.2
```

```
source("miseqR.R")

theme_set(theme_bw())

##### ONLY OTUS WITH OVER 10 SEQUENCES (AND WITH PHYLO TREE) #####
#####

setwd("C:\\Users\\arisely\\Dropbox\\PhD\\Microbiome\\CHAPTER 3 - PHYSIOLOGY\\ANALYSIS\\R\\Phyloseq analysis")

# Assign variables for imported data

sharedfile = "shorebird.microbiome.shared"
taxfile = "shorebird.microbiome.taxonomy"

##import mothur shared and tax files

mothur_data <- import_mothur(mothur_shared_file = sharedfile,
                             mothur_constaxonomy_file = taxfile)

##import metadata
map <- read.csv("shorebird.metadata.csv", header=T, row.names=1)

##make meta data into phyloseq format

map <- sample_data(map)
str(map)
```

```
## 'data.frame':   188 obs. of  21 variables:
## Formal class 'sample_data' [package "phyloseq"] with 4 slots
##   ..@ .Data      :List of 21
##   .. ..$ : Factor w/ 180 levels "8546","8548",...: 1 2 3 4 5 6 7 8 9 10 ...
##   .. ..$ : Factor w/ 6 levels "Curlew sandpiper",...: 4 4 4 4 4 4 4 4 4 4 ...
##   .. ..$ : Factor w/ 5 levels "", "Broome", "Flinders",...: 2 2 2 2 2 2 2 2 2 2 ...
##   .. ..$ : Factor w/ 7 levels "", "August", "December",...: 2 2 2 2 2 2 2 2 2 2 ...
##   .. ..$ : Factor w/ 6 levels "", "Adult", "Juvenile",...: 5 5 2 5 5 5 5 5 5 2 ...
##   .. ..$ : Factor w/ 4 levels "", "one", "three",...: 4 4 4 4 4 4 4 4 4 4 ...
##   .. ..$ : Factor w/ 161 levels "", "3611876", "3648481",...: 20 21 10 22 23 11 12 13 14 15 ...
##   .. ..$ : Factor w/ 3 levels "bird", "env", "neg": 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : Factor w/ 3 levels "na", "No", "Yes": 2 2 2 2 2 2 2 2 2 2 ...
##   .. ..$ : Factor w/ 2 levels "No", "Yes": 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : int   NA NA NA NA NA NA NA NA NA NA NA ...
##   .. ..$ : Factor w/ 2 levels "No", "Yes": 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : Factor w/ 2 levels "No", "Yes": 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : int   NA NA NA NA NA NA NA NA NA NA NA ...
##   .. ..$ : num   NA NA NA NA NA NA NA NA NA NA NA ...
##   .. ..$ : num   NA NA NA NA NA NA NA NA NA NA NA ...
##   .. ..$ : Factor w/ 3 levels "", "sink", "source": 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : Factor w/ 4 levels "", "env", "gut",...: 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ : Factor w/ 10 levels "1/04/2017", "1/10/2016",...: 6 6 6 6 6 6 6 6 6 6 ...
##   .. ..$ : int    2 2 2 2 2 2 2 2 2 2 ...
##   .. ..$ : Factor w/ 9 levels "", "Broome-curlewsand-adult",...: 5 5 4 5 5 5 5 5 4 ...
##   ..@ names      : chr  "Group" "species" "site" "month" ...
##   ..@ row.names: chr  "8546" "8548" "8549" "8551" ...
##   ..@ .S3Class   : chr  "data.frame"
```

```
head(map)
```

```
##      Group      species  site  month      age group bird_id type
## 8546  8546 Red-necked stint Broome August Second year    two 3674227 bird
## 8548  8548 Red-necked stint Broome August Second year    two 3674229 bird
## 8549  8549 Red-necked stint Broome August      Adult    two 3671906 bird
## 8551  8551 Red-necked stint Broome August Second year    two 3674231 bird
## 8552  8552 Red-necked stint Broome August Second year    two 3674232 bird
## 8556  8556 Red-necked stint Broome August Second year    two 3674214 bird
##      Recap recap_delete Recap_num replicate remove_replicate weight wing
## 8546    No              No        NA          No              No      NA   NA
## 8548    No              No        NA          No              No      NA   NA
## 8549    No              No        NA          No              No      NA   NA
## 8551    No              No        NA          No              No      NA   NA
## 8552    No              No        NA          No              No      NA   NA
## 8556    No              No        NA          No              No      NA   NA
##      hb SourceSink Env      Date Plate      migration
## 8546 NA              22/08/2015    2    Broome-stint-SY
## 8548 NA              22/08/2015    2    Broome-stint-SY
## 8549 NA              22/08/2015    2    Broome-stint-adult
## 8551 NA              22/08/2015    2    Broome-stint-SY
## 8552 NA              22/08/2015    2    Broome-stint-SY
## 8556 NA              22/08/2015    2    Broome-stint-SY
```

```
tail(map)
```

```
##          Group species      site      month age group bird_id type Recap
## F1      Flind_env      env Flinders      March  na              na  env   No
## F2      Flind_env      env Flinders      March  na              na  env   No
## Neg      Neg1       neg      na          na  na              na  neg   na
## Neg2      Neg2       neg              na          na              na  neg   No
## WTP1      WTP_env      env      WTP December  na              na  env   No
## WTP2      WTP_env      env      WTP December  na              na  env   No
##          recap_delete Recap_num replicate remove_replicate weight wing hb
## F1              No          NA          Yes              No      NA   NA NA
## F2              No          NA          Yes              No      NA   NA NA
## Neg              No          NA          No              No      NA   NA NA
## Neg2              No          NA          No              No      NA   NA NA
## WTP1              No          NA          No              No      NA   NA NA
## WTP2              No          NA          No              No      NA   NA NA
##          SourceSink Env      Date Plate migration
## F1              sink env 11/03/2016      1
## F2              sink env 11/03/2016      1
## Neg              sink neg  1/10/2016      1
## Neg2              1/04/2017      2
## WTP1              sink env 29/12/2015      1
## WTP2              sink env 29/12/2015      1
```

```
##merge the metadata into the phyloseq object

moth_merge <- merge_phyloseq(mothur_data, map)

##import phylogenetic tree

tree<-read.tree("shorebird.microbiome.tree")

##make tree into phyloseq format

tree<-phy_tree(tree)

##merge tree into current phyloseq object so all object now contained in moth_merge

moth_merge <-merge_phyloseq(moth_merge, tree)

moth_merge
```

```
## phyloseq-class experiment-level object
## otu_table()   OTU Table:         [ 8791 taxa and 188 samples ]
## sample_data() Sample Data:      [ 188 samples by 21 sample variables ]
## tax_table()   Taxonomy Table:    [ 8791 taxa by 6 taxonomic ranks ]
## phy_tree()    Phylogenetic Tree: [ 8791 tips and 8789 internal nodes ]
```

```
##everything looks fie and no errors!

##rename columns to relevant taxonomic group

colnames(tax_table(moth_merge))
```

```
## [1] "Rank1" "Rank2" "Rank3" "Rank4" "Rank5" "Rank6"
```

```
colnames(tax_table(moth_merge)) <- c("Kingdom", "Phylum", "Class",
                                     "Order", "Family", "Genus")

data_all <- moth_merge %>%
  subset_taxa(
    Kingdom == "Bacteria" &
    Family  != "mitochondria" &
    Class   != "Chloroplast"
  )

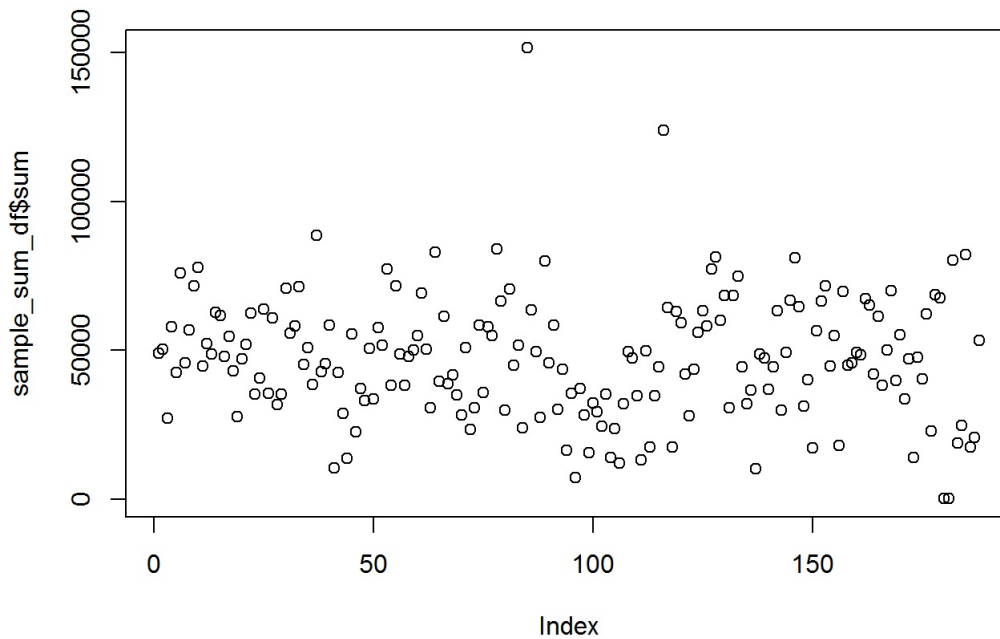
data_all
```



```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 8791 taxa and 188 samples ]
## sample_data() Sample Data: [ 188 samples by 21 sample variables ]
## tax_table() Taxonomy Table: [ 8791 taxa by 6 taxonomic ranks ]
## phy_tree() Phylogenetic Tree: [ 8791 tips and 8789 internal nodes ]
```

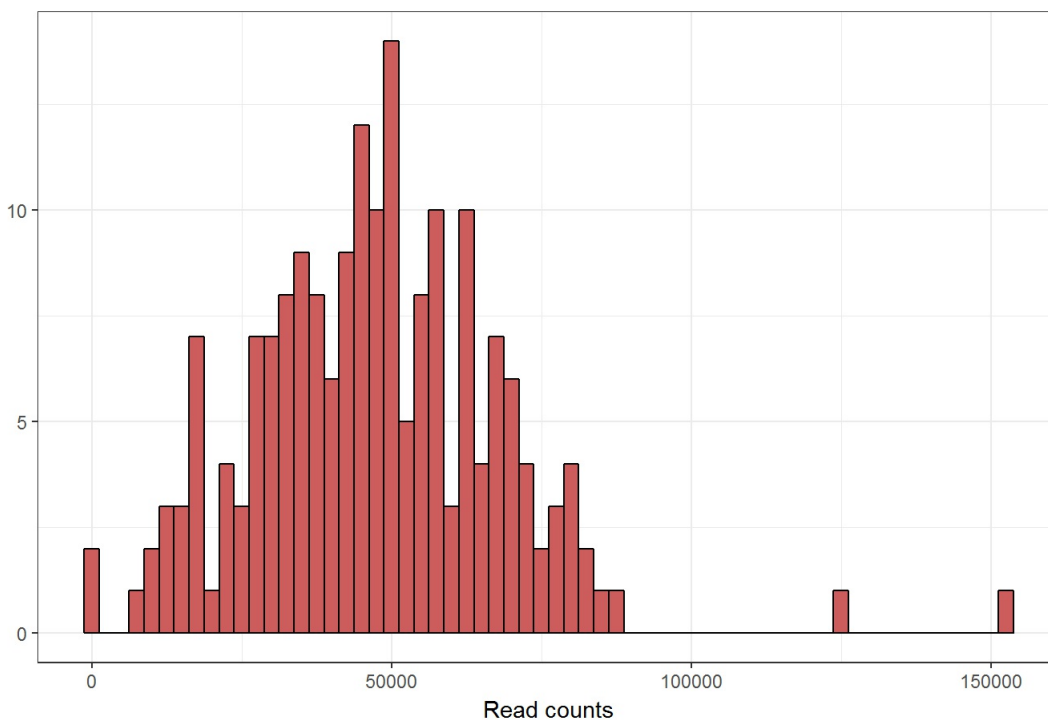
```
#####look at read library
```

```
sample_sum_df <- data.frame(sum = sample_sums(data_all))
plot(sample_sum_df$sum)
```



```
ggplot(sample_sum_df, aes(x = sum)) +
  geom_histogram(color = "black", fill = "indianred", binwidth = 2500) +
  ggtitle("Distribution of sample sequencing depth") +
  xlab("Read counts") +
  ylab("Frequency")+
  theme(axis.title.y = element_blank())
```

Distribution of sample sequencing depth



```
#####  
#####
```

```
##subset samples to just birds (there are environmental samples in here too)
```

```
data_all<-data_all %>%  
  subset_samples(Group != "Env_Broome_A")
```

```
data_all<-data_all %>%  
  subset_samples(Group != "Env_Broome_E")
```

```
data_all<-prune_taxa(taxa_sums(data_all)>0, data_all)
```

```
## remove OTUs that are common in the neg controls (Neg = just lab control, Neg2 = field and lab control)  
neg_control<-data_all %>% subset_samples(type == "neg")
```

```
neg_control <- prune_taxa(taxa_sums(neg_control) > 5, neg_control)  
neg_control
```

```
## phyloseq-class experiment-level object  
## otu_table() OTU Table: [ 97 taxa and 2 samples ]  
## sample_data() Sample Data: [ 2 samples by 21 sample variables ]  
## tax_table() Taxonomy Table: [ 97 taxa by 6 taxonomic ranks ]  
## phy_tree() Phylogenetic Tree: [ 97 tips and 96 internal nodes ]
```

```
## 97 OTUs which have over 1 reads
```

```
## we should get rid of these
```

```
badtaxa<-taxa_names(neg_control)  
alltaxa<-taxa_names(data_all)  
alltaxa1 <- alltaxa[!(alltaxa %in% badtaxa)]
```

```
data_all = prune_taxa(alltaxa1, data_all)
```

```
##get rid of negative control altogether
```

```
data_all<-data_all%>%subset_samples(species!="Neg")
```

```
#delete replicates
```

```
data_all<-data_all%>%subset_samples(remove_replicate=="No")
```

```
## just birds
```

```
birds<-data_all%>%subset_samples(type=="bird")  
birds <- prune_taxa(taxa_sums(birds) > 0, birds)
```

```
#####
```

```
## data has more samples than we need. We are going to prune the dataset to just the relevant samples for this  
analysis
```

```
## for migration analysis (migrant/resident comparison)
```

```
migration<-birds%>%subset_samples(species=="Red-necked stint"|species=="Curlew sandpiper")  
migration<-migration%>%subset_samples(month=="August"|month=="September")  
migration <- prune_taxa(taxa_sums(migration) > 0, migration)  
migration
```

```
## phyloseq-class experiment-level object  
## otu_table() OTU Table: [ 5262 taxa and 77 samples ]  
## sample_data() Sample Data: [ 77 samples by 21 sample variables ]  
## tax_table() Taxonomy Table: [ 5262 taxa by 6 taxonomic ranks ]  
## phy_tree() Phylogenetic Tree: [ 5262 tips and 5260 internal nodes ]
```

```
## 77 samples, 5262 OTUs

##### DESeq 2 #####

## Will do this analysis first (for Fig2 and Table S2) before rarefying data

## Analyse differential abundances between migrants and residents for each group

## subset RNS (Red-necked stint) in Broome

rns_br<-migration%>%subset_samples(species=="Red-necked stint")
rns_br<-rns_br%>%subset_samples(site=="Broome")
rns_br <- prune_taxa(taxa_sums(rns_br) > 0, rns_br)

#####

##the following analysis is from the Phyloseq-DESeq2 vignette, linked at top of script.

diagdds = phyloseq_to_deseq2(rns_br, ~ age)
```

```
## converting counts to integer mode
```

```
# calculate geometric means prior to estimate size factors
gm_mean = function(x, na.rm=TRUE){
  exp(sum(log(x[x > 0])), na.rm=na.rm) / length(x))
}
geoMeans = apply(counts(diagdds), 1, gm_mean)
diagdds = estimateSizeFactors(diagdds, geoMeans = geoMeans)
diagdds = DESeq(diagdds, fitType="local")
```

```
## using pre-existing size factors
```

```
## estimating dispersions
```

```
## gene-wise dispersion estimates
```

```
## mean-dispersion relationship
```

```
## final dispersion estimates
```

```
## fitting model and testing
```

```
## -- replacing outliers and refitting for 985 genes
## -- DESeq argument 'minReplicatesForReplace' = 7
## -- original counts are preserved in counts(dds)
```

```
## estimating dispersions
```

```
## fitting model and testing
```

```
#diagdds = DESeq(diagdds, fitType="parametric")

res = results(diagdds, cooksCutoff = FALSE)
alpha = 0.01
sigtab = res[which(res$padj < alpha), ]
sigtab_rns.br = cbind(as(sigtab, "data.frame"), as(tax_table(rns_br)[rownames(sigtab), ], "matrix"))
head(sigtab_rns.br)
```

```
##          baseMean log2FoldChange      lfcSE      stat      pvalue
## Otu000765      8.943807          6.451209 1.473355  4.378585 1.194523e-05
## Otu000279     12.561703          6.487397 1.733635  3.742078 1.825051e-04
## Otu000053     448.890074          4.714085 1.207552  3.903836 9.468006e-05
## Otu000013    1452.245588        -10.263594 1.174882 -8.735852 2.418274e-18
## Otu000117      43.826023         -8.519601 1.381634 -6.166322 6.989664e-10
## Otu000009     451.397495        -11.616548 1.627590 -7.137267 9.520456e-13
##          padj      Kingdom      Phylum      Class
## Otu000765 5.492657e-04 Bacteria Proteobacteria Gammaproteobacteria
## Otu000279 4.994002e-03 Bacteria Proteobacteria Gammaproteobacteria
## Otu000053 3.562337e-03 Bacteria Bacteroidetes Sphingobacteriia
## Otu000013 3.639502e-16 Bacteria Actinobacteria Actinobacteria
## Otu000117 5.259722e-08 Bacteria Actinobacteria Actinobacteria
## Otu000009 9.552191e-11 Bacteria Actinobacteria Actinobacteria
##          Order      Family
## Otu000765 Pseudomonadales Moraxellaceae
## Otu000279 Pseudomonadales Moraxellaceae
## Otu000053 Sphingobacteriales Chitinophagaceae
## Otu000013 Corynebacteriales Corynebacteriaceae
## Otu000117 Corynebacteriales Corynebacteriaceae
## Otu000009 Corynebacteriales Corynebacteriales_unclassified
##          Genus
## Otu000765 Acinetobacter
## Otu000279 Acinetobacter
## Otu000053 Vibrionimonas
## Otu000013 Corynebacteriaceae_unclassified
## Otu000117 Corynebacteriaceae_unclassified
## Otu000009 Corynebacteriales_unclassified
```

```
sigtab_rns.br<- sigtab_rns.br[order(sigtab_rns.br$log2FoldChange),]

#add variable to end
sigtab_rns.br$Group<-"Red-necked stint Broome"
#####

#####

##RNS in Vic

rns_vic<-migration%>%subset_samples(species=="Red-necked stint")
rns_vic<-rns_vic%>%subset_samples(month=="September")
rns_vic <- prune_taxa(taxa_sums(rns_vic) > 0, rns_vic)

#####

diagdds = phyloseq_to_deseq2(rns_vic, ~ age)
```

```
## converting counts to integer mode
```

```
# calculate geometric means prior to estimate size factors
gm_mean = function(x, na.rm=TRUE){
  exp(sum(log(x[x > 0])), na.rm=na.rm) / length(x))
}
geoMeans = apply(counts(diagdds), 1, gm_mean)
diagdds = estimateSizeFactors(diagdds, geoMeans = geoMeans)
diagdds = DESeq(diagdds, fitType="local")
```

```
## using pre-existing size factors
```

```
## estimating dispersions
```

```
## gene-wise dispersion estimates
```

```
## mean-dispersion relationship
```

```
## final dispersion estimates
```

```
## fitting model and testing
```

```
## -- replacing outliers and refitting for 1034 genes
## -- DESeq argument 'minReplicatesForReplace' = 7
## -- original counts are preserved in counts(dds)
```

```
## estimating dispersions
```

```
## fitting model and testing
```

```
#diagdds = DESeq(diagdds, fitType="parametric")

res = results(diagdds, cooksCutoff = FALSE)
alpha = 0.01
sigtab = res[which(res$padj < alpha), ]
sigtab_rns.vic = cbind(as(sigtab, "data.frame"), as(tax_table(rns_vic)[rownames(sigtab), ], "matrix"))
head(sigtab_rns.vic)
```

```
##          baseMean log2FoldChange      lfcSE      stat      pvalue
## Otu000013   310.74839      -7.603356  0.9299704  -8.175912  2.936371e-16
## Otu000117    11.77106      -6.999681  1.2864956  -5.440890  5.301508e-08
## Otu000009  6020.41474      -9.169087  0.9653096  -9.498597  2.127372e-21
## Otu000575    17.49598      -7.299684  1.2980066  -5.623765  1.868398e-08
## Otu000265    12.17622      -7.131328  1.6612481  -4.292753  1.764711e-05
## Otu000757    13.11934      -6.617123  1.3422332  -4.929936  8.225670e-07
##          padj      Kingdom      Phylum      Class
## Otu000013  1.642900e-13  Bacteria  Actinobacteria  Actinobacteria
## Otu000117  1.483097e-05  Bacteria  Actinobacteria  Actinobacteria
## Otu000009  2.380529e-18  Bacteria  Actinobacteria  Actinobacteria
## Otu000575  6.969125e-06  Bacteria  Actinobacteria  Actinobacteria
## Otu000265  2.194124e-03  Bacteria  Actinobacteria  Actinobacteria
## Otu000757  1.534087e-04  Bacteria  Actinobacteria  Actinobacteria
##          Order      Family
## Otu000013  Corynebacteriales  Corynebacteriaceae
## Otu000117  Corynebacteriales  Corynebacteriaceae
## Otu000009  Corynebacteriales  Corynebacteriales_unclassified
## Otu000575  Corynebacteriales  Corynebacteriaceae
## Otu000265  Corynebacteriales  Corynebacteriaceae
## Otu000757  Corynebacteriales  Corynebacteriaceae
##          Genus
## Otu000013  Corynebacteriaceae_unclassified
## Otu000117  Corynebacteriaceae_unclassified
## Otu000009  Corynebacteriales_unclassified
## Otu000575  Corynebacteriaceae_unclassified
## Otu000265  Corynebacteriaceae_unclassified
## Otu000757  Corynebacterium
```

```
sigtab_rns.vic<- sigtab_rns.vic[order(sigtab_rns.vic$log2FoldChange),]
```

```
sigtab_rns.vic$Group<-"Red-necked stint Flinders"
```

```
#####
```

```
#curlew sandpiper
```

```
cs<-migration%>%subset_samples(species=="Curlew sandpiper")
cs <- prune_taxa(taxa_sums(cs) > 0, cs)
```

```
#####
```

```
diagdds = phyloseq_to_deseq2(cs, ~ age)
```

```
## converting counts to integer mode
```

```
# calculate geometric means prior to estimate size factors
gm_mean = function(x, na.rm=TRUE){
  exp(sum(log(x[x > 0])), na.rm=na.rm) / length(x))
}
geoMeans = apply(counts(diagdds), 1, gm_mean)
diagdds = estimateSizeFactors(diagdds, geoMeans = geoMeans)
diagdds = DESeq(diagdds, fitType="local")
```

```
## using pre-existing size factors
```

```
## estimating dispersions
```

```
## gene-wise dispersion estimates
```

```
## mean-dispersion relationship
```

```
## final dispersion estimates
```

```
## fitting model and testing
```

```
## -- replacing outliers and refitting for 509 genes
## -- DESeq argument 'minReplicatesForReplace' = 7
## -- original counts are preserved in counts(dds)
```

```
## estimating dispersions
```

```
## fitting model and testing
```

```
#diagdds = DESeq(diagdds, fitType="parametric")

res = results(diagdds, cooksCutoff = FALSE)
alpha = 0.01
sigtab = res[which(res$padj < alpha), ]
sigtab_cs = cbind(as(sigtab, "data.frame"), as(tax_table(cs)[rownames(sigtab), ], "matrix"))
head(sigtab_cs)
```

```
##           baseMean log2FoldChange    lfcSE      stat      pvalue
## Otu000319  36.83220      -23.122692 3.172221 -7.289117 3.119939e-13
## Otu000017  211.66346       6.298750 1.498019  4.204720 2.614058e-05
## Otu000023  648.40735      -8.741014 1.963501 -4.451749 8.517358e-06
## Otu000134   17.35786     -22.100447 3.161498 -6.990498 2.739122e-12
## Otu000157   17.97271     -22.155554 3.172759 -6.983056 2.888269e-12
## Otu000028  177.39850      -7.319617 2.008101 -3.645044 2.673464e-04
##           padj    Kingdom      Phylum      Class
## Otu000319 1.164777e-11 Bacteria Proteobacteria Betaproteobacteria
## Otu000017 4.182493e-04 Bacteria Proteobacteria Gammaproteobacteria
## Otu000023 1.467606e-04 Bacteria Actinobacteria Actinobacteria
## Otu000134 7.188580e-11 Bacteria Actinobacteria Actinobacteria
## Otu000157 7.188580e-11 Bacteria Actinobacteria Actinobacteria
## Otu000028 3.522682e-03 Bacteria Actinobacteria Actinobacteria
##           Order      Family
## Otu000319   Rhodocyclales Rhodocyclaceae
## Otu000017   Aeromonadales Succinivibrionaceae
## Otu000023 Corynebacteriales Corynebacteriaceae
## Otu000134 Corynebacteriales Corynebacteriaceae
## Otu000157 Corynebacteriales Corynebacteriaceae
## Otu000028 Corynebacteriales Corynebacteriaceae
##           Genus
## Otu000319      Methyloversatilis
## Otu000017 Succinivibrionaceae_unclassified
## Otu000023 Corynebacteriaceae_unclassified
## Otu000134 Corynebacteriaceae_unclassified
## Otu000157      Corynebacterium
## Otu000028      <NA>
```

```
sigtab_cs<- sigtab_cs[order(sigtab_cs$log2FoldChange),]

sigtab_cs$Group<-"Curlew sandpiper Broome"

#####

##plot all together
##bind objects

deseq2<-rbind(sigtab_rns.br, sigtab_rns.vic, sigtab_cs)

#edit small details for figure

deseq2[3,11]<-"Corynebacteriaceae"
deseq2[12,11]<-"Corynebacteriaceae"
revalue(deseq2$Family, c("Gammaproteobacteria_unclassified"="Gammaproteobacteria uncl."))
```

```
##          Otu000118          Otu000028
## Corynebacteriaceae Corynebacteriaceae
##          Otu000009          Otu000013
## Corynebacteriaceae Corynebacteriaceae
##          Otu000117          Otu000011
## Corynebacteriaceae Helicobacteraceae
##          Otu000053          Otu000106
## Chitinophagaceae Lachnospiraceae
##          Otu000765          Otu000279
## Moraxellaceae Moraxellaceae
##          Otu000358          Otu000009
## Gammaproteobacteria uncl. Corynebacteriaceae
##          Otu000013          Otu000575
## Corynebacteriaceae Corynebacteriaceae
##          Otu000265          Otu000117
## Corynebacteriaceae Corynebacteriaceae
##          Otu000757          Otu000769
## Corynebacteriaceae Corynebacteriaceae
##          Otu000136          Otu000028
## Corynebacteriaceae Corynebacteriaceae
##          Otu000020          Otu000006
## Enterococcaceae Helicobacteraceae
##          Otu000319          Otu000157
## Rhodocyclaceae Corynebacteriaceae
##          Otu000134          Otu000023
## Corynebacteriaceae Corynebacteriaceae
##          Otu000028          Otu000044
## Corynebacteriaceae Ruminococcaceae
##          Otu000017          Otu000070
## Succinivibrionaceae Lachnospiraceae
##          Otu000124          Otu000126
## Lachnospiraceae Ruminococcaceae
##          Otu000227          Otu000208
## Peptostreptococcaceae Clostridiaceae_1
##          Otu000078          Otu000628
## Campylobacteraceae Peptostreptococcaceae
##          Otu000358          Otu000343
## Gammaproteobacteria uncl. Lachnospiraceae
##          Otu000094          Otu000316
## Mycoplasmataceae Lachnospiraceae
##          Otu000183
## Clostridiaceae_1
## 15 Levels: Chitinophagaceae ... Succinivibrionaceae
```

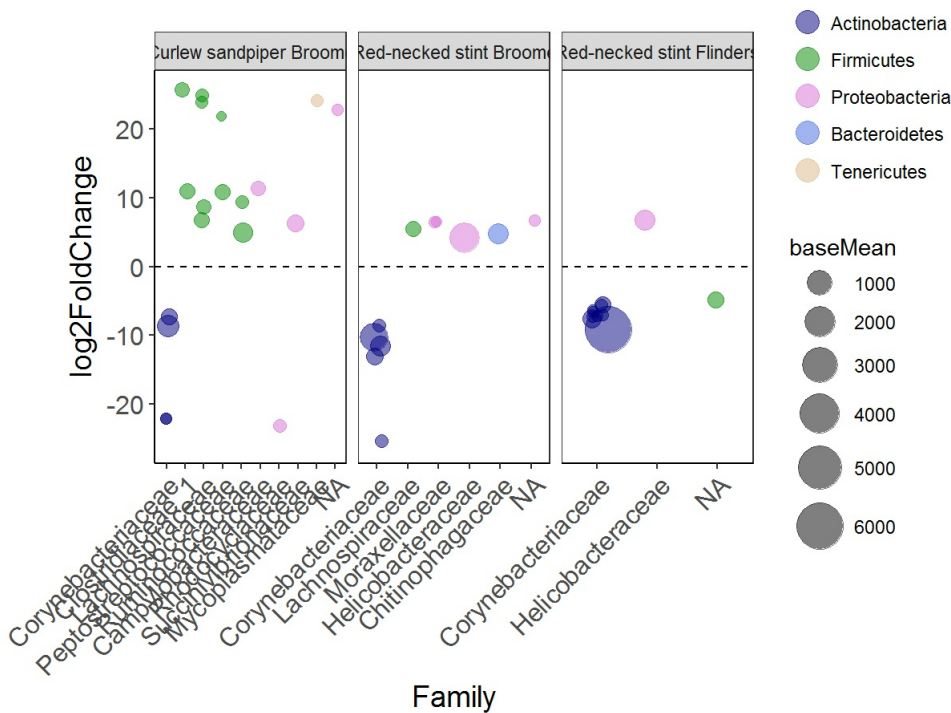
```
deseq2$Family<-factor(deseq2$Family, levels = c("Corynebacteriaceae",
                                                "Clostridiaceae_1", "Lachnospiraceae" , "Peptostreptococcaceae",
                                                "Ruminococcaceae",

                                                "Moraxellaceae",
                                                "Campylobacteraceae" ,
                                                "Helicobacteraceae" ,
                                                "Rhodocyclaceae" ,
                                                "Gammaproteobacteria uncl.",
                                                "Succinivibrionaceae",
                                                "Chitinophagaceae" ,
                                                "Mycoplasmataceae"))

deseq2$Phylum<-factor(deseq2$Phylum, levels = c("Actinobacteria", "Firmicutes", "Proteobacteria", "Bacteroidetes"
, "Tenericutes"))
```

#Figure 2

```
ggplot(deseq2, aes(x=Family, y=log2FoldChange, color=Phylum)) +
  geom_jitter(aes(size = baseMean), width = 0.2, alpha = 0.5)+ scale_size(range = c(2,10))+
  theme(axis.text.x = element_text(angle = 45, hjust = 1))+
  geom_hline(yintercept = 0, linetype = 2)+facet_grid(.~Group, scales = "free")+ylim(-26,26)+
  scale_color_manual(values = c("navyblue", "green4", "orchid", "royalblue", "burlywood"))+
  theme( axis.text = element_text(size=12), axis.title=element_text(size=14))+
  theme(plot.margin=unit(c(0.5,1,1,2), "cm"))+guides(color = guide_legend(override.aes = list(size=5)))+
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
        panel.background = element_blank(), axis.line = element_line(colour = "black"))
```



```
#####

##Now rest of analyses

migration1 <-migration

##rarefy
migration <-rarefy_even_depth(migration1, sample.size = min(sample_sums(migration1)),
                              rngseed = 3, replace = TRUE, trimOTUs = TRUE, verbose = TRUE)
```

```
## `set.seed(3)` was used to initialize repeatable random subsampling.
```

```
## Please record this for your records so others can reproduce.
```



```
## Try `set.seed(3); .Random.seed` for the full vector
```

```
## ...
```

```
## 856OTUs were removed because they are no longer  
## present in any sample after random subsampling
```

```
## ...
```

```
migration
```

```
## phyloseq-class experiment-level object  
## otu_table() OTU Table: [ 4406 taxa and 77 samples ]  
## sample_data() Sample Data: [ 77 samples by 21 sample variables ]  
## tax_table() Taxonomy Table: [ 4406 taxa by 6 taxonomic ranks ]  
## phy_tree() Phylogenetic Tree: [ 4406 tips and 4404 internal nodes ]
```

```
migration2 <- migration1 %>%  
  scale_reads(n=5815) ##scale to smallest library size  
  
#migration1 = unrarefied  
#migration = rarefied  
#migration2 = scaled  
  
#using rarefied data, but can test with others - makes no difference  
  
##### Prevalence #####  
##prevalence whole group  
  
prev0 = apply(X = otu_table(migration),  
  MARGIN = ifelse(taxa_are_rows(migration), yes = 1, no = 2),  
  FUN = function(x){sum(x > 0)})  
prevdf = data.frame(Prevalence = prev0,  
  TotalAbundance = taxa_sums(migration),  
  tax_table(migration))  
keepPhyla = table(prevdf$Phylum)[(table(prevdf$Phylum) > 5)]  
prevdf1 = subset(prevdf, Phylum %in% names(keepPhyla))  
  
prevdf$Prevalence<-(prevdf$Prevalence/77)*100  
  
head(prevdf)
```

```
##          Prevalence TotalAbundance Kingdom      Phylum  
## Otu001177   1.298701              3 Bacteria Proteobacteria  
## Otu000824   3.896104             15 Bacteria Proteobacteria  
## Otu004711   2.597403              7 Bacteria Proteobacteria  
## Otu005756   1.298701              4 Bacteria Proteobacteria  
## Otu007032   1.298701              2 Bacteria Proteobacteria  
## Otu001047   3.896104             17 Bacteria Proteobacteria  
##              Class          Order      Family  
## Otu001177 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
## Otu000824 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
## Otu004711 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
## Otu005756 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
## Otu007032 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
## Otu001047 Alphaproteobacteria Rhodobacterales Rhodobacteraceae  
##              Genus  
## Otu001177      Rubellimicrobium  
## Otu000824      Wenxinia  
## Otu004711 Rhodobacteraceae_unclassified  
## Otu005756      <NA>  
## Otu007032      <NA>  
## Otu001047      Pseudoruegeria
```

```
str(prevdf)
```

```
## 'data.frame': 4406 obs. of 8 variables:
## $ Prevalence : num 1.3 3.9 2.6 1.3 1.3 ...
## $ TotalAbundance: num 3 15 7 4 2 17 1 1 3 3 ...
## $ Kingdom : Factor w/ 1 level "Bacteria": 1 1 1 1 1 1 1 1 1 1 ...
## $ Phylum : Factor w/ 27 levels "Acidobacteria",...: 21 21 21 21 21 21 21 21 21 21 ...
## $ Class : Factor w/ 75 levels "Acidimicrobiia",...: 4 4 4 4 4 4 4 4 4 4 ...
## $ Order : Factor w/ 150 levels "Acidimicrobiales",...: 110 110 110 110 110 110 110 110 110 110 ...
## $ Family : Factor w/ 270 levels "0319-6G20", "Acetobacteraceae",...: 203 203 203 203 203 203 203 203 203 203 ...
## $ Genus : Factor w/ 554 levels "0319-6G20_ge",...: 443 548 422 NA NA 407 407 422 422 NA ...
```

```
summary(prevdf$Prevalence)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 1.299 1.299 2.597 4.638 3.896 93.506
```

```
###only keep those phyla which occur in at least 5 samples
```

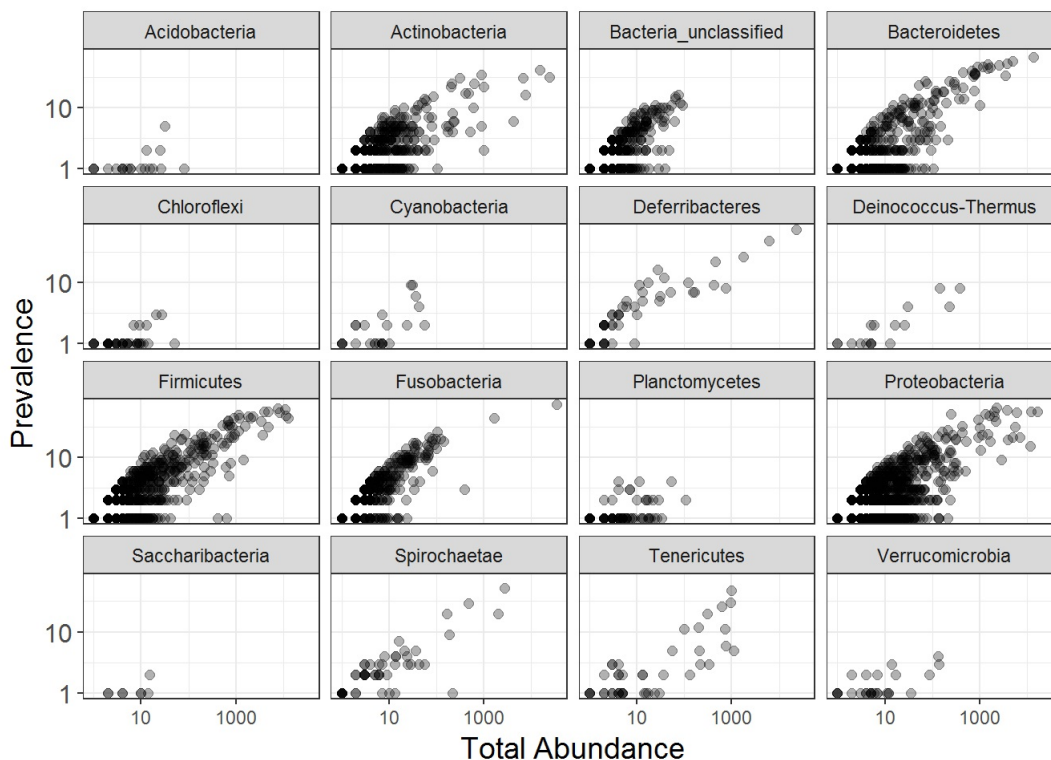
```
# Define prevalence threshold as 5% of total samples
prevalenceThreshold = 0.05 * nsamples(migration)
prevalenceThreshold
```

```
## [1] 3.85
```

```
migration1 = prune_taxa((prev0 > prevalenceThreshold), migration)
migration1
```

```
## phyloseq-class experiment-level object
## otu_table() OTU Table: [ 1016 taxa and 77 samples ]
## sample_data() Sample Data: [ 77 samples by 21 sample variables ]
## tax_table() Taxonomy Table: [ 1016 taxa by 6 taxonomic ranks ]
## phy_tree() Phylogenetic Tree: [ 1016 tips and 1015 internal nodes ]
```

```
ggplot(prevdf1, aes(TotalAbundance, Prevalence)) +
  geom_point(size = 2, alpha = 0.3) +
  scale_y_log10() + scale_x_log10() +
  xlab("Total Abundance") +
  facet_wrap(~Phylum)+theme_bw() +
  theme( axis.text.y = element_text(size=12), axis.title=element_text(size=14))
```



```
#write.csv(prevdf, "prevalence_all_samples.csv")

#####

##table S1

#Subset each group seperately to get top 10 OTUs
# b = broome, f = flinders, a = adult/migrants, sy = second year/resident

curlew.b<-migration%>%subset_samples(species=="Curlew sandpiper"& site=="Broome")
curlew.b<- prune_taxa(taxa_sums(curlew.b) > 0, curlew.b)
curlew.b.a<-curlew.b%>%subset_samples(age=="Adult")
curlew.b.a<- prune_taxa(taxa_sums(curlew.b.a) > 0, curlew.b.a)
curlew.b.sy<-curlew.b%>%subset_samples(age=="Second year")
curlew.b.sy<- prune_taxa(taxa_sums(curlew.b.sy) > 0, curlew.b.sy)

stint.b<-migration%>%subset_samples(site=="Broome" & species=="Red-necked stint")
stint.b<- prune_taxa(taxa_sums(stint.b) > 0, stint.b)
stint.b.a<-stint.b%>%subset_samples(age=="Adult")
stint.b.a<- prune_taxa(taxa_sums(stint.b.a) > 0, stint.b.a)
stint.b.sy<-stint.b%>%subset_samples(age=="Second year")
stint.b.sy<- prune_taxa(taxa_sums(stint.b.sy) > 0, stint.b.sy)

stint.f<-migration%>%subset_samples(month=="September")
stint.f<-stint.f%>%subset_samples(species=="Red-necked stint")
stint.f<- prune_taxa(taxa_sums(stint.f) > 0, stint.f)
stint.f.sy<-stint.f%>%subset_samples(age=="Second year")

stint.f.sy<- prune_taxa(taxa_sums(stint.f.sy) > 0, stint.f.sy)
stint.f.a<-stint.f%>%subset_samples(age=="Adult")
stint.f.a<- prune_taxa(taxa_sums(stint.f.a) > 0, stint.f.a)

##### TABLE S1A)

# broome curlew sand migrants

prev0 = apply(X = otu_table(curlew.b.a),
             MARGIN = ifelse(taxa_are_rows(curlew.b.a), yes = 1, no = 2),
             FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                   TotalAbundance = taxa_sums(curlew.b.a),
                   tax_table(curlew.b.a))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/12
prevdf<-prevdf[order(-prevdf$RelAbund),]
head(prevdf)
```

```
##              Prevalence TotalAbundance Kingdom      Phylum
## Otu000023          9           7688 Bacteria Actinobacteria
## Otu000027          8           5928 Bacteria Proteobacteria
## Otu000002         11           4683 Bacteria Fusobacteria
## Otu000058          4           4295 Bacteria Actinobacteria
## Otu000010         10           3326 Bacteria Deferribacteres
## Otu000028          7           2829 Bacteria Actinobacteria
##              Class      Order      Family
## Otu000023 Actinobacteria Corynebacteriales Corynebacteriaceae
## Otu000027 Gammaproteobacteria Aeromonadales Succinivibrionaceae
## Otu000002 Fusobacteriia Fusobacteriales Fusobacteriaceae
## Otu000058 Actinobacteria Corynebacteriales Corynebacteriaceae
## Otu000010 Deferribacteres Deferribacterales Deferribacteraceae
## Otu000028 Actinobacteria Corynebacteriales Corynebacteriaceae
##              Genus RelAbund RelPrev
## Otu000023 Corynebacteriaceae_unclassified 0.11017484 0.7500000
## Otu000027 Anaerobiospirillum 0.08495271 0.6666667
## Otu000002 Cetobacterium 0.06711092 0.9166667
## Otu000058 Corynebacteriaceae_unclassified 0.06155059 0.3333333
## Otu000010 Mucispirillum 0.04766409 0.8333333
## Otu000028 <NA> 0.04054170 0.5833333
```

```
#write.csv(prevdf, "curlew.b.a.csv")

#####

# broome curlew sand residents

prev0 = apply(X = otu_table(curlew.b.sy),
              MARGIN = ifelse(taxa_are_rows(curlew.b.sy), yes = 1, no = 2),
              FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                    TotalAbundance = taxa_sums(curlew.b.sy),
                    tax_table(curlew.b.sy))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/6
prevdf<-prevdf[order(-prevdf$RelAbund),]
head(prevdf)
```

```
##              Prevalence TotalAbundance Kingdom      Phylum
## Otu000002          5          4223 Bacteria  Fusobacteria
## Otu000010          5          3457 Bacteria  Deferribacteres
## Otu000018          6          2711 Bacteria  Firmicutes
## Otu000016          6          2427 Bacteria  Bacteroidetes
## Otu000034          6          2238 Bacteria  Firmicutes
## Otu000008          5          1973 Bacteria  Firmicutes
##              Class      Order      Family
## Otu000002  Fusobacteriia  Fusobacteriales  Fusobacteriaceae
## Otu000010  Deferribacteres  Deferribacterales  Deferribacteraceae
## Otu000018  Clostridia  Clostridiales  Lachnospiraceae
## Otu000016  Bacteroidia  Bacteroidales  Bacteroidaceae
## Otu000034  Erysipelotrichia  Erysipelotrichales  Erysipelotrichaceae
## Otu000008  Bacilli  Lactobacillales  Enterococcaceae
##              Genus RelAbund RelPrev
## Otu000002  Cetobacterium 0.12103755 0.8333333
## Otu000010  Mucispirillum 0.09908283 0.8333333
## Otu000018  Tyzzerella_3 0.07770135 1.0000000
## Otu000016  Bacteroides 0.06956148 1.0000000
## Otu000034  Erysipelotrichaceae_unclassified 0.06414445 1.0000000
## Otu000008  Catellicoccus 0.05654915 0.8333333
```

```
#write.csv(prevdf, "curlew.b.sy.csv")

#####

##table S1b)

# broome stint migrants

prev0 = apply(X = otu_table(stint.b.a),
              MARGIN = ifelse(taxa_are_rows(stint.b.a), yes = 1, no = 2),
              FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                    TotalAbundance = taxa_sums(stint.b.a),
                    tax_table(stint.b.a))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/13
prevdf<-prevdf[order(-prevdf$RelAbund),]
head(prevdf)
```

```
##          Prevalence TotalAbundance  Kingdom          Phylum
## Otu000013          11              8616 Bacteria  Actinobacteria
## Otu000002          13              7614 Bacteria  Fusobacteria
## Otu000017          12              5778 Bacteria  Proteobacteria
## Otu000009          6              5195 Bacteria  Actinobacteria
## Otu000010          13              3510 Bacteria  Deferribacteres
## Otu000016          12              3106 Bacteria  Bacteroidetes
##
##          Class          Order
## Otu000013  Actinobacteria  Corynebacteriales
## Otu000002  Fusobacteriia  Fusobacteriales
## Otu000017  Gammaproteobacteria  Aeromonadales
## Otu000009  Actinobacteria  Corynebacteriales
## Otu000010  Deferribacteres  Deferribacterales
## Otu000016  Bacteroidia    Bacteroidales
##
##          Family          Genus
## Otu000013  Corynebacteriaceae  Corynebacteriaceae_unclassified
## Otu000002  Fusobacteriaceae    Cetobacterium
## Otu000017  Succinivibrionaceae  Succinivibrionaceae_unclassified
## Otu000009  Corynebacteriales_unclassified  Corynebacteriales_unclassified
## Otu000010  Deferribacteraceae    Mucispirillum
## Otu000016  Bacteroidaceae        Bacteroides
##
## RelAbund  RelPrev
## Otu000013 0.11397579 0.8461538
## Otu000002 0.10072095 1.0000000
## Otu000017 0.07643363 0.9230769
## Otu000009 0.06872148 0.4615385
## Otu000010 0.04643164 1.0000000
## Otu000016 0.04108737 0.9230769
```

```
#write.csv(prevdf, "stint.b.a.csv")
#####

## broome stint resident

prev0 = apply(X = otu_table(stint.b.sy),
              MARGIN = ifelse(taxa_are_rows(stint.b.sy), yes = 1, no = 2),
              FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                    TotalAbundance = taxa_sums(stint.b.sy),
                    tax_table(stint.b.sy))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/16
prevdf<-prevdf[order(-prevdf$RelAbund),]
head(prevdf)
```

```
##          Prevalence TotalAbundance  Kingdom          Phylum
## Otu000010          16              9725 Bacteria  Deferribacteres
## Otu000011          15              7912 Bacteria  Proteobacteria
## Otu000020          10              5754 Bacteria  Firmicutes
## Otu000002          15              4614 Bacteria  Fusobacteria
## Otu000018          9              4421 Bacteria  Firmicutes
## Otu000036          15              4284 Bacteria  Proteobacteria
##
##          Class          Order          Family
## Otu000010  Deferribacteres  Deferribacterales  Deferribacteraceae
## Otu000011  Epsilonproteobacteria  Campylobacterales  Helicobacteraceae
## Otu000020  Bacilli          Lactobacillales  Enterococcaceae
## Otu000002  Fusobacteriia  Fusobacteriales  Fusobacteriaceae
## Otu000018  Clostridia      Clostridiales  Lachnospiraceae
## Otu000036  Alphaproteobacteria  Rhizobiales  Methylobacteriaceae
##
##          Genus  RelAbund  RelPrev
## Otu000010  Mucispirillum 0.10452494 1.0000
## Otu000011  Helicobacter 0.08503869 0.9375
## Otu000020  Catellicoccus 0.06184437 0.6250
## Otu000002  Cetobacterium 0.04959157 0.9375
## Otu000018  Tyzzerella_3 0.04751720 0.5625
## Otu000036  Methylobacterium 0.04604471 0.9375
```

```
#write.csv(prevdf, "stint.b.sy.csv")

#####

#Table Slc)

## vic stint migrants

prev0 = apply(X = otu_table(stint.f.a),
              MARGIN = ifelse(taxa_are_rows(stint.f.a), yes = 1, no = 2),
              FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                    TotalAbundance = taxa_sums(stint.f.a),
                    tax_table(stint.f.a))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/15
prevdf<-prevdf[order(-prevdf$RelAbund),]

head(prevdf)
```

```
##          Prevalence TotalAbundance Kingdom      Phylum
## Otu000009          14          19124 Bacteria Actinobacteria
## Otu000002          14           6743 Bacteria  Fusobacteria
## Otu000011          13           5365 Bacteria Proteobacteria
## Otu000020          10           4261 Bacteria  Firmicutes
## Otu000013          14           3865 Bacteria Actinobacteria
## Otu000014          15           2560 Bacteria  Firmicutes
##                  Class              Order
## Otu000009 Actinobacteria Corynebacteriales
## Otu000002 Fusobacteriia  Fusobacteriales
## Otu000011 Epsilonproteobacteria Campylobacteriales
## Otu000020 Bacilli       Lactobacillales
## Otu000013 Actinobacteria Corynebacteriales
## Otu000014 Bacilli       Lactobacillales
##                  Family
## Otu000009 Corynebacteriales_unclassified Corynebacteriales_unclassified
## Otu000002 Fusobacteriaceae                Cetobacterium
## Otu000011 Helicobacteraceae                Helicobacter
## Otu000020 Enterococcaceae                Catellicoccus
## Otu000013 Corynebacteriaceae Corynebacteriaceae_unclassified
## Otu000014 Enterococcaceae                Catellicoccus
##          RelAbund RelPrev
## Otu000009 0.21924907 0.9333333
## Otu000002 0.07730582 0.9333333
## Otu000011 0.06150760 0.8666667
## Otu000020 0.04885067 0.6666667
## Otu000013 0.04431069 0.9333333
## Otu000014 0.02934938 1.0000000
```

```
#write.csv(prevdf, "stint.f.a.csv")

#####

##vic stint resident

prev0 = apply(X = otu_table(stint.f.sy),
              MARGIN = ifelse(taxa_are_rows(stint.f.sy), yes = 1, no = 2),
              FUN = function(x){sum(x > 0)})
prevdf = data.frame(Prevalence = prev0,
                    TotalAbundance = taxa_sums(stint.f.sy),
                    tax_table(stint.f.sy))

prevdf$RelAbund<-prevdf$TotalAbundance/sum(prevdf$TotalAbundance)
prevdf$RelPrev<-prevdf$Prevalence/15
prevdf<-prevdf[order(-prevdf$RelAbund),]
head(prevdf)
```

```
##          Prevalence TotalAbundance  Kingdom          Phylum
## Otu000006          7          11596 Bacteria Proteobacteria
## Otu000002         14          6271 Bacteria  Fusobacteria
## Otu000008         15          4866 Bacteria  Firmicutes
## Otu000031          8          4102 Bacteria  Firmicutes
## Otu000039          3          3774 Bacteria Proteobacteria
## Otu000014         13          2963 Bacteria  Firmicutes
##
##          Class          Order          Family
## Otu000006 Epsilonproteobacteria Campylobacteriales Helicobacteraceae
## Otu000002 Fusobacteriia Fusobacteriales Fusobacteriaceae
## Otu000008 Bacilli Lactobacillales Enterococcaceae
## Otu000031 Erysipelotrichia Erysipelotrichales Erysipelotrichaceae
## Otu000039 Gammaproteobacteria Enterobacteriales Enterobacteriaceae
## Otu000014 Bacilli Lactobacillales Enterococcaceae
##
##          Genus RelAbund RelPrev
## Otu000006 Helicobacteraceae_unclassified 0.13294354 0.4666667
## Otu000002 Cetobacterium 0.07189453 0.9333333
## Otu000008 Catellicoccus 0.05578676 1.0000000
## Otu000031 Erysipelotrichaceae_unclassified 0.04702780 0.5333333
## Otu000039 Escherichia-Shigella 0.04326741 0.2000000
## Otu000014 Catellicoccus 0.03396962 0.8666667
```

```
#write.csv(prevdf, "stint.f.sy.csv")
```

```
#####stacked barplots #####
#####
```

```
#Figure 1)
```

```
##### migration
```

```
migration_phylum <- migration %>%
  tax_glom(taxrank = "Phylum") %>% # agglomerate at phylum level
  transform_sample_counts(function(x) {x/sum(x)} ) %>% # Transform to rel. abundance
  psmelt() %>% # Melt to long format
  arrange(Phylum) # Sort data frame alphabetically by phylum
```

```
#filter(Abundance > 0.05) %>% # Filter out low abundance taxa
head(migration_phylum)
```

```
##          OTU Sample  Abundance Group          species      site      month
## 1 Otu000944      8680 0.016852966 8680 Red-necked stint Flinders September
## 2 Otu000944      8607 0.004815133 8607 Red-necked stint Broome      August
## 3 Otu000944      8667 0.004815133 8667 Red-necked stint Flinders September
## 4 Otu000944      8639 0.003955288 8639 Red-necked stint Flinders September
## 5 Otu000944      8560 0.003439381 8560 Red-necked stint Broome      August
## 6 Otu000944      8566 0.002063629 8566 Red-necked stint Broome      August
##          age group bird_id type Recap recap_delete Recap_num replicate
## 1      Adult three 3685834 bird    No              No          NA          No
## 2      Adult two 3678451 bird    No              No          NA          No
## 3 Second year three 3686272 bird    No              No          NA          No
## 4      Adult three 3686361 bird   Yes              Yes           1          Yes
## 5      Adult two 3674218 bird    No              No          NA          No
## 6 Second year two 3674234 bird    No              No          NA          No
##          remove_replicate weight wing hb SourceSink Env      Date Plate
## 1              No      29  NA NA      source gut 20/09/2015      1
## 2              No      NA  NA NA      source gut 29/08/2015      2
## 3              No      28  NA NA      source gut 20/09/2015      1
## 4              No      31  NA NA      source gut 20/09/2015      1
## 5              No      NA  NA NA      source gut 22/08/2015      2
## 6              No      NA  NA NA      source gut 22/08/2015      2
##          migration Kingdom          Phylum
## 1 Flinders-stint-adult Bacteria Acidobacteria
## 2 Broome-stint-adult Bacteria Acidobacteria
## 3 Flinders-stint-SY Bacteria Acidobacteria
## 4 Flinders-stint-adult Bacteria Acidobacteria
## 5 Broome-stint-adult Bacteria Acidobacteria
## 6 Broome-stint-SY Bacteria Acidobacteria
```

```
#####
```

```
migration_phylum$Phylum<-factor(migration_phylum$Phylum)
#broome_phylum$Order<-factor(broome_phylum$Order)
unique(migration_phylum$Phylum )
```

```
## [1] Acidobacteria          Actinobacteria
## [3] Armatimonadetes         Bacteria_unclassified
## [5] Bacteroidetes           Chloroflexi
## [7] Cyanobacteria           Deferribacteres
## [9] Deinococcus-Thermus     FBP
## [11] Fibrobacteres           Firmicutes
## [13] Fusobacteria            Gemmatimonadetes
## [15] Gracilibacteria         Lentisphaerae
## [17] Microgenomates          Parcubacteria
## [19] Peregrinibacteria       Planctomycetes
## [21] Proteobacteria          Saccharibacteria
## [23] Spirochaetae            SR1_(Absconditabacteria)
## [25] Tenericutes             TM6_(Dependentiae)
## [27] Verrucomicrobia
## 27 Levels: Acidobacteria Actinobacteria ... Verrucomicrobia
```

```
migration_phylum$Phylum<-factor(migration_phylum$Phylum)
```

```
unique(migration_phylum$Phylum)
```

```
## [1] Acidobacteria          Actinobacteria
## [3] Armatimonadetes         Bacteria_unclassified
## [5] Bacteroidetes           Chloroflexi
## [7] Cyanobacteria           Deferribacteres
## [9] Deinococcus-Thermus     FBP
## [11] Fibrobacteres           Firmicutes
## [13] Fusobacteria            Gemmatimonadetes
## [15] Gracilibacteria         Lentisphaerae
## [17] Microgenomates          Parcubacteria
## [19] Peregrinibacteria       Planctomycetes
## [21] Proteobacteria          Saccharibacteria
## [23] Spirochaetae            SR1_(Absconditabacteria)
## [25] Tenericutes             TM6_(Dependentiae)
## [27] Verrucomicrobia
## 27 Levels: Acidobacteria Actinobacteria ... Verrucomicrobia
```



```
migration_phylum$Phylum<-factor(migration_phylum$Phylum, level = c("Actinobacteria",
                                                                           "Firmicutes",
                                                                           "Proteobacteria",
                                                                           "Bacteroidetes",
                                                                           "Deferribacteres",
                                                                           "Fusobacteria",
                                                                           "Tenericutes",
                                                                           "Spirochaetae",
                                                                           "Deinococcus-Thermus",
                                                                           "Acidobacteria",
                                                                           "Armatimonadetes",
                                                                           "Bacteria_unclassified",
                                                                           "Chloroflexi",
                                                                           "Cyanobacteria",
                                                                           "FBP",
                                                                           "Fibrobacteres",
                                                                           "Gemmatimonadetes",
                                                                           "Gracilibacteria",
                                                                           "Ignavibacteriae",
                                                                           "Lentisphaerae",
                                                                           "Microgenomates",
                                                                           "Parcubacteria",
                                                                           "Peregrinibacteria",
                                                                           "Planctomycetes",
                                                                           "Saccharibacteria",
                                                                           "SR1_(Absconditabacteria)",
                                                                           "TM6_(Dependentiae)",
                                                                           "Verrucomicrobia"))
```

```
unique(migration_phylum$Sample)
```

```
## [1] "8680" "8607" "8667" "8639" "8560" "8566" "8645" "8662" "8558" "8559"
## [11] "8610" "8632" "8628" "8563" "8581" "8671" "8549" "8552" "8548" "8660"
## [21] "8601" "8604" "8546" "8605" "8612" "8658" "8556" "8570" "8551" "8659"
## [31] "8657" "8646" "8665" "8629" "8655" "8627" "8633" "8609" "8616" "8649"
## [41] "8589" "8606" "8586" "8575" "8644" "8557" "8571" "8677" "8672" "8663"
## [51] "8630" "8689" "8565" "8569" "8615" "8564" "8590" "8561" "8682" "8583"
## [61] "8634" "8683" "8685" "8588" "8675" "8674" "8613" "8626" "8688" "8686"
## [71] "8578" "8625" "8573" "8618" "8684" "8679" "8638"
```

```
colors <- c("navyblue","green4","orchid","royalblue","black","red","brown","grey","bisque",
            "white","white","white","white","white","white","white","white","white","white","white",
            "white","white","white","white","white","white","white","white")
```

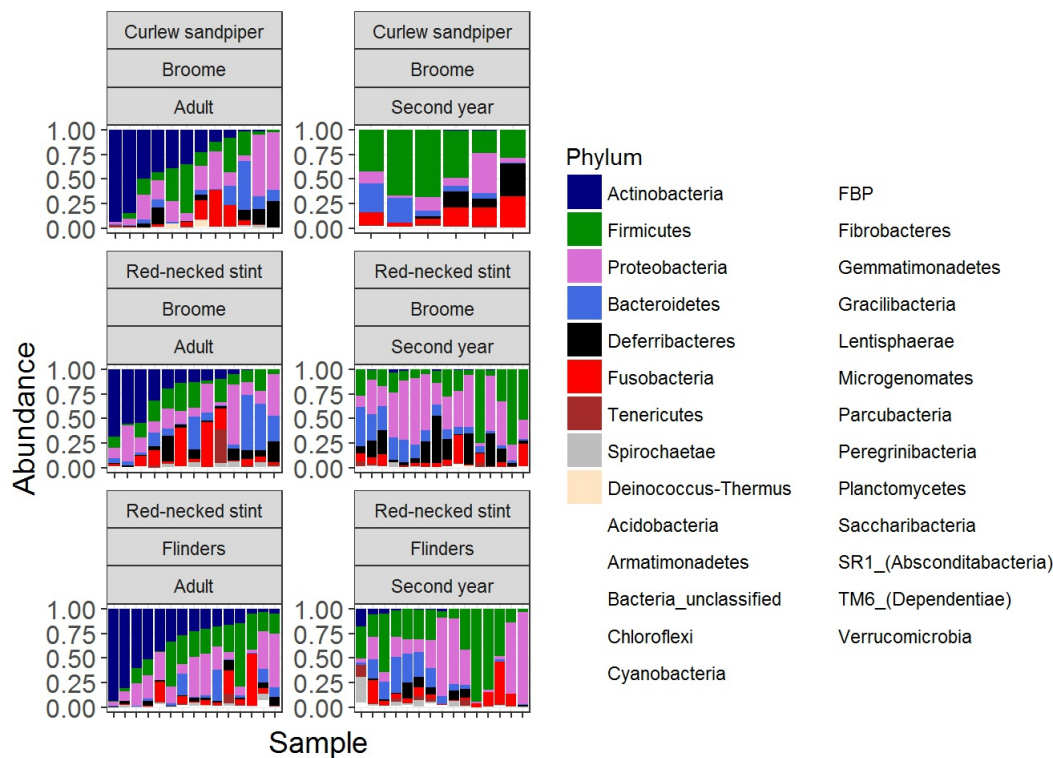
```
names(colors) <- levels(migration_phylum$Phylum)
colScale <- scale_fill_manual(name = "Phylum", values = colors)
```

```
migration_phylum$Sample<-factor(migration_phylum$Sample, level = c("8682", "8629", "8615", "8685", "8610", "866
2", "8560", "8607", "8671", "8589", "8657", "8573", "8625", "8627", "8639", "8549", "8645", "8632", "8660", "86
49", "8601",
                                                                           "8658", "8638", "8646", "8634", "8606", "861
2", "8581", "8618", "8609", "8628", "8672", "8605", "8689", "8659", "8616", "8626", "8604", "8569", "8677", "85
88", "8675",
                                                                           "8551", "8613", "8546", "8683", "8590", "855
7", "8558", "8674", "8559", "8680", "8630", "8563", "8586", "8556", "8561", "8644", "8667", "8566", "8665", "86
33", "8564",
                                                                           "8583", "8575", "8571", "8565", "8578", "857
0", "8684", "8688", "8655", "8552", "8548", "8679", "8663", "8686"))
```

```
##Figure 1a
```

```
ggplot(migration_phylum, aes(x = Sample, y = Abundance, fill = Phylum)) +

  geom_bar(stat = "identity") +
  theme(axis.text.x = element_blank(), axis.text.y = element_text(size=12), axis.title=element_text(size=14))+
  facet_wrap(species~site~age, scales="free", nrow=3, ncol=2)+colScale
```



#####repeat but for Family Fig 1b #####

```
migration_family <- migration %>%
  tax_glom(taxrank = "Family") %>% # agglomerate at phylum level
  transform_sample_counts(function(x) {x/sum(x)} ) %>% # Transform to rel. abundance
  psmelt() %>% # Melt to long format
  low_abundance_taxa # Filter out
  arrange(Family) # Sort data frame alphabetically by family
#
head(migration_family)
```

```
##      OTU Sample  Abundance Group      species  site  month
## 1 Otu000437  8558 0.003439381 8558 Red-necked stint Broome August
## 2 Otu000437  8563 0.002407567 8563 Red-necked stint Broome August
## 3 Otu000437  8552 0.001031814 8552 Red-necked stint Broome August
## 4 Otu000437  8564 0.001031814 8564 Red-necked stint Broome August
## 5 Otu000437  8556 0.000000000 8556 Red-necked stint Broome August
## 6 Otu000437  8583 0.000000000 8583 Curlew sandpiper Broome August
##      age group bird_id type Recap recap_delete Recap_num replicate
## 1 Second year  two 3674216 bird  No          No          NA          No
## 2 Second year  two 3674221 bird  No          No          NA          No
## 3 Second year  two 3674232 bird  No          No          NA          No
## 4 Second year  two 3674222 bird  No          No          NA          No
## 5 Second year  two 3674214 bird  No          No          NA          No
## 6 Second year  one 4267072 bird  No          No          NA          No
##      remove_replicate weight wing hb SourceSink Env      Date Plate
## 1                    No    NA   NA NA          22/08/2015      2
## 2                    No    NA   NA NA          22/08/2015      2
## 3                    No    NA   NA NA          22/08/2015      2
## 4                    No    NA   NA NA          22/08/2015      2
## 5                    No    NA   NA NA          22/08/2015      2
## 6                    No    NA   NA NA          22/08/2015      2
##      migration Kingdom      Phylum      Class
## 1 Broome-stint-SY Bacteria Proteobacteria Deltaproteobacteria
## 2 Broome-stint-SY Bacteria Proteobacteria Deltaproteobacteria
## 3 Broome-stint-SY Bacteria Proteobacteria Deltaproteobacteria
## 4 Broome-stint-SY Bacteria Proteobacteria Deltaproteobacteria
## 5 Broome-stint-SY Bacteria Proteobacteria Deltaproteobacteria
## 6 Broome-curlewsand-SY Bacteria Proteobacteria Deltaproteobacteria
##      Order      Family
## 1 Oligoflexales 0319-6G20
## 2 Oligoflexales 0319-6G20
## 3 Oligoflexales 0319-6G20
## 4 Oligoflexales 0319-6G20
## 5 Oligoflexales 0319-6G20
## 6 Oligoflexales 0319-6G20
```

```

migration_family$Family<-factor(migration_family$Family)
#broome_phylum$Order<-factor(broome_phylum$Order)
unique(migration_family$Family)

```

```

## [1] 0319-6G20 Acetobacteraceae
## [3] Acidaminococcaceae Acidimicrobiaceae
## [5] Acidimicrobiales_unclassified Acidobacteriaceae_(Subgroup_1)
## [7] Actinobacteria_unclassified Actinomycetaceae
## [9] Aeromonadaceae Aeromonadales_unclassified
## [11] Alcaligenaceae Alphaproteobacteria_unclassified
## [13] Alteromonadaceae Anaerolineaceae
## [15] Anaeroplasmataceae Anaplasmataceae
## [17] Archangiaceae Ardentcatenales_fa
## [19] Arenicellaceae Armatimonadales_unclassified
## [21] Aurantimonadaceae Bacillaceae
## [23] Bacilli_unclassified Bacteria_unclassified
## [25] Bacteriovoracaceae Bacteroidaceae
## [27] Bacteroidales_S24-7_group Bacteroidales_unclassified
## [29] Bacteroidetes_unclassified BD2-11_terrestrial_group_fa
## [31] BD7-8_marine_group_fa Bdellovibrionaceae
## [33] Beijerinckiaceae Bifidobacteriaceae
## [35] Blastocatellaceae_(Subgroup_4) Brachyspiraceae
## [37] Bradymonadaceae Bradyrhizobiaceae
## [39] Brevibacteriaceae Brucellaceae
## [41] Burkholderiaceae Burkholderiales_unclassified
## [43] Caldilineaceae Campylobacteraceae
## [45] Campylobacteriales_unclassified Cardiobacteriaceae
## [47] Carnobacteriaceae Caulobacteraceae
## [49] Cellulomonadaceae Cellvibrionaceae
## [51] Cellvibrionales_unclassified Chitinophagaceae
## [53] Chloroflexaceae Chloroflexi_unclassified
## [55] Chromatiaceae Clostridia_unclassified
## [57] Clostridiaceae_1 Clostridiaceae_4
## [59] Clostridiales_unclassified Clostridiales_vadinBB60_group
## [61] Comamonadaceae Coriobacteriaceae
## [63] Corynebacteriaceae Corynebacteriales_unclassified
## [65] Coxiellaceae Cryomorphaceae
## [67] CS-B046_fa Cyanobacteria_unclassified
## [69] Cyclobacteriaceae Cytophagaceae
## [71] Cytophagales_unclassified dl42_fa
## [73] DA111 DBS1
## [75] Deferribacteraceae Deinococcaceae
## [77] Deltaproteobacteria_unclassified Dermabacteraceae
## [79] Dermacoccaceae Dermatophilaceae
## [81] Desulfobacteraceae Desulfobulbaceae
## [83] Desulfovibrionaceae Desulfovibrionales_unclassified
## [85] Desulfuromonadaceae DEV007
## [87] Dietziaceae Ectothiorhodospiraceae
## [89] Eel-36e1D6 Elev-16S-1332
## [91] Enterobacteriaceae Enterococcaceae
## [93] Entomoplasmatales_Incertae_Sedis Epsilonproteobacteria_unclassified
## [95] Erysipelotrichaceae Erythrobacteraceae
## [97] Eubacteriaceae Euzebyaceae
## [99] EV818SWSAP88 Family_XI
## [101] Family_XII Family_XIII
## [103] FamilyI FamilyII
## [105] FBP_fa FD035
## [107] Firmicutes_unclassified Flammeovirgaceae
## [109] Flavobacteriaceae Flavobacteriales_unclassified
## [111] Fusobacteriaceae Fusobacteriales_unclassified
## [113] Gammaproteobacteria_unclassified Gastranaerophilales_fa
## [115] Gemmatimonadaceae Geodermatophilaceae
## [117] Gitt-GS-136_fa Gracilibacteria_fa
## [119] Granulosicoccaceae Hahellaceae
## [121] Halanaerobiales_unclassified Haliangiaceae
## [123] Halieaceae Halomonadaceae
## [125] Helicobacteraceae Herpetosiphonaceae
## [127] HOC36_fa Holosporaceae
## [129] Hydrogenophilaceae Hyphomicrobiaceae
## [131] Hyphomonadaceae Intrsporangiaceae
## [133] JG30-KF-CM45_fa JG34-KF-361
## [135] JTB255_marine_benthic_group KD4-96_fa
## [137] KF-JG30-B3 Kineosporiaceae
## [139] Lachnospiraceae Lactobacillaceae

```

## [141]	Lactobacillales_unclassified	LD29
## [143]	Legionellaceae	Lentisphaeraceae
## [145]	Leptospiraceae	Leptotrichiaceae
## [147]	Leuconostocaceae	Listeriaceae
## [149]	Longimicrobiaceae	Methylobacteriaceae
## [151]	Methylophilaceae	MgMjR-022
## [153]	Microbacteriaceae	Micrococcaceae
## [155]	Micrococcales_unclassified	Microgenomates_unclassified
## [157]	Milano-WF1B-44_fa	Mollicutes_unclassified
## [159]	Moraxellaceae	MSB-1E8
## [161]	Mycobacteriaceae	Mycoplasmataceae
## [163]	NB1-n_fa	Neisseriaceae
## [165]	Nitriliruptoraceae	Nitrosomonadaceae
## [167]	Nocardiaceae	Nocardioidaceae
## [169]	NS9_marine_group	Obscuribacterales_fa
## [171]	Oceanospirillaceae	Oceanospirillales_unclassified
## [173]	Oligoflexaceae	OM1_clade
## [175]	OM182_clade	OM190_fa
## [177]	Oxalobacteraceae	Paenibacillaceae
## [179]	Parcubacteria_fa	Parvularculaceae
## [181]	Pasteurellaceae	PeM15_fa
## [183]	Peptostreptococcaceae	Peregrinibacteria_fa
## [185]	Phycisphaeraceae	Phyllobacteriaceae
## [187]	Pla3_lineage_fa	Planctomycetaceae
## [189]	Planococcaceae	Porphyromonadaceae
## [191]	Prevotellaceae	Propionibacteriaceae
## [193]	Proteobacteria_unclassified	Pseudoalteromonadaceae
## [195]	Pseudomonadaceae	Pseudonocardiaceae
## [197]	Psychromonadaceae	Puniceicoccaceae
## [199]	PYR10d3_fa	Rhizobiaceae
## [201]	Rhizobiales_Incertae_Sedis	Rhizobiales_unclassified
## [203]	Rhodobacteraceae	Rhodobiaceae
## [205]	Rhodocyclaceae	Rhodospirillaceae
## [207]	Rhodospirillales_Incertae_Sedis	Rhodospirillales_unclassified
## [209]	Rhodothermaceae	Rickettsiaceae
## [211]	Rickettsiales_Incertae_Sedis	Rickettsiales_unclassified
## [213]	Rikenellaceae	Rs-D42
## [215]	Rubritaleaceae	Ruminococcaceae
## [217]	S0134_terrestrial_group_fa	S085_fa
## [219]	Saccharibacteria_fa	Sandaracinaceae
## [221]	Saprospiraceae	SAR116_clade
## [223]	SAR324_clade (Marine_group_B)_fa	Shewanellaceae
## [225]	Solibacteraceae_(Subgroup_3)	Solirubrobacteraceae
## [227]	Sphingobacteriaceae	Sphingobacteriales_unclassified
## [229]	Sphingomonadaceae	Sphingomonadales_unclassified
## [231]	Spirochaetaceae	Spirochaetales_unclassified
## [233]	Spongiibacteraceae	SR1_(Absconditabacteria)_fa
## [235]	Staphylococcaceae	Streptococcaceae
## [237]	Streptomycetaceae	Subgroup_21_fa
## [239]	Subgroup_22_fa	Subgroup_23_fa
## [241]	Subgroup_6_fa	Subgroup_7_fa
## [243]	Succinivibrionaceae	Surface_1
## [245]	Sva0071_fa	Sva0725
## [247]	Sva0996_marine_group	Sva1033
## [249]	Syntrophaceae	T9d
## [251]	Tepidisphaeraceae	Thermoactinomycetaceae
## [253]	Thiotrichaceae	TM6_(Dependentiae)_fa
## [255]	TRA3-20_fa	Trueperaceae
## [257]	uncultured	uncultured_fa
## [259]	Unknown_Family	Veillonellaceae
## [261]	Verrucomicrobia_fa	Verrucomicrobia_unclassified
## [263]	Verrucomicrobiaceae	Verrucomicrobiales_unclassified
## [265]	Vibrionaceae	X35_fa
## [267]	Xanthobacteraceae	Xanthomonadaceae
## [269]	Xanthomonadales_Incertae_Sedis	Xanthomonadales_unclassified
## 270	Levels: 0319-6G20 Acetobacteraceae ...	Xanthomonadales_unclassified

```
#write.csv(migration_family,"family.csv")
```

```
##all common Corynebacteriales_unclassified are genus Corynebacterium (we ran sequences through ARB):
```

```
migration_family$Family[migration_family$Family=="Corynebacteriales_unclassified"]<-"Corynebacteriaceae"
```

```
#####plot
```

```

migration_family$Family<-factor(migration_family$Family, level = c("Corynebacteriaceae",
"Acidaminococcaceae", "Clostridiaceae_1", "Clostridiales_unclassified" , "Enterococcaceae" , "Erysipelotrichaceae" , "Firmicutes_unclassified" , "Lachnospiraceae" , "Peptostreptococcaceae" , "Ruminococcaceae" , "Staphylococcaceae" ,
"Comamonadaceae", "Desulfovibrionaceae", "Enterobacteriaceae", "Gammaproteobacteria_unclassified" , "Helicobacteraceae" , "Methylobacteriaceae" , "Oxalobacteraceae" , "Rhizobiales_unclassified" , "Rhodobacteraceae", "Sphingomonadaceae" , "Succinivibrionaceae", "Bacteroidaceae" , "Bacteroidales_unclassified" , "Chitinophagaceae" , "Flavobacteriaceae", "Porphyromonadaceae", "Rikenellaceae" ,
"Deferribacteraceae" ,
"Fusobacteriaceae",
"Anaeroplasmataceae", "Mycoplasmataceae",
"Brachyspiraceae" ,
"Deinococcaceae",

"0319-6G20",
"Acetobacteraceae",

"Acidimicrobiaceae",
"Acidimicrobiales_unclassified",
"Acidobacteriaceae_(Subgroup_1)",
"Actinobacteria_unclassified",
"Actinomycetaceae",
"Aeromonadaceae",
"Aeromonadales_unclassified",
"Alcaligenaceae",
"Alphaproteobacteria_unclassified",
"Alteromonadaceae",
"Anaerolineaceae",

"Anaplasmataceae",
"Ardenticatenales_fa",
"Arenicellaceae",
"Armatimonadales_unclassified",
"Aurantimonadaceae",
"Bacillaceae",
"Bacilli_unclassified",
"Bacteria_unclassified",
"Bacteriovoracaceae",

"Bacteroidales_S24-7_group",

"Bacteroidetes_unclassified",
"Bartonellaceae",
"BD2-11_terrestrial_group_fa",
"BD7-8_marine_group_fa",
"Bdellovibrionaceae",
"Beijerinckiaceae",
"Bifidobacteriaceae",
"Blastocatellaceae_(Subgroup_4)",

"Bradymonadaceae",
"Bradymonadales_fa",
"Bradyrhizobiaceae",
"Brevibacteriaceae",
"Brucellaceae",
"Burkholderiaceae",
"Burkholderiales_unclassified",
"Caldilineaceae",
"Campylobacteraceae",
"Campylobacterales_unclassified",
"Cardiobacteriaceae",
"Carnobacteriaceae",
"Caulobacteraceae",
"Cellulomonadaceae",
"Cellvibrionaceae",
"Cellvibrionales_unclassified",

"Chloroflexaceae",
"Chloroflexi_unclassified",
"Chromatiaceae",
"Clostridia_unclassified",

"Clostridiaceae_4",

"Clostridiales_vadinBB60_group",

```

"Coriobacteriaceae",

"Coxiellaceae",
"Cryomorphaceae",
"CS-B046_fa",
"Cyanobacteria_unclassified",
"Cyclobacteriaceae",
"Cytophagaceae",
"Cytophagales_unclassified",
"d142_fa",
"DA111",
"DBS1",

"Deltaproteobacteria_unclassified",
"Dermabacteraceae",
"Dermacoccaceae",
"Dermatophilaceae",
"Desulfobacteraceae",
"Desulfobulbaceae",

"Desulfovibrionales_unclassified",
"Desulfuromonadaceae",
"Desulfuromonadales_unclassified",
"DEV007",
"Dietziaceae",
"Eel-36elD6",
"Elev-16S-1332",

"Entomoplasmatales_Incertae_Sedis",
"Epsilonproteobacteria_unclassified",

"Erythrobacteraceae",
"Eubacteriaceae",
"Euzebyaceae",
"Family_XI",
"Family_XII",
"Family_XIII",
"FamilyI",
"FamilyII",
"FBP_fa",
"FD035",

"Flammeovirgaceae",

"Flavobacteriales_unclassified",

"Fusobacteriales_unclassified",

"Gastranaerophilales_fa",
"Gemmatimonadaceae",
"Geodermatophilaceae",
"Gitt-GS-136_fa",
"Gracilibacteria_fa",
"Granulosicoccaceae",
"Hahellaceae",
"Halanaerobiales_unclassified",
"Haliangiaceae",
"Halieaceae",
"Halomonadaceae",

"Herpetosiphonaceae",
"HOC36_fa",
"Holosporaceae",
"Hydrogenophilaceae",
"Hyphomicrobiaceae",
"Hyphomonadaceae",
"Iamiaceae",
"Intrasporangiaceae",
"JG30-KF-CM45_fa",
"JG34-KF-361",
"JTB255_marine_benthic_group",
"KD4-96_fa",
"KF-JG30-B3",

"Kineosporiaceae",

"Lactobacillaceae",
"Lactobacillales_unclassified",
"LD29",
"Legionellaceae",
"Lentisphaeraceae",
"Leptospiraceae",
"Leptotrichiaceae",
"Leuconostocaceae",
"Listeriaceae",
"Longimicrobiaceae",

"Methylophilaceae",
"MgMjR-022",
"Microbacteriaceae",
"Micrococcaceae",
"Micrococcales_unclassified",
"Microgenomates_unclassified",
"Milano-WF1B-44_fa",
"Mollicutes_unclassified",
"Moraxellaceae",
"MSB-1E8",
"Mycobacteriaceae",

"NB1-n_fa",
"Neisseriaceae",
"Nitriliruptoraceae",
"Nitrosomonadaceae",
"Nocardiaceae",
"Nocardioidaceae",
"NS9_marine_group",
"Obscuribacteriales_fa",
"Oceanospirillaceae",
"Oceanospirillales_unclassified",
"Oligoflexaceae",
"OM1_clade",
"OM182_clade",
"OM190_fa",

"Paenibacillaceae",
"Parcubacteria_fa",
"Parvularculaceae",
"Pasteurellaceae",
"PeM15_fa",

"Peregrinibacteria_fa",
"Phycisphaeraceae",
"Phyllobacteriaceae",
"Pla3_lineage_fa",
"Planctomycetaceae",
"Planococcaceae",

"Prevotellaceae",
"Propionibacteriaceae",
"Proteobacteria_unclassified",
"Pseudoalteromonadaceae",
"Pseudomonadaceae",
"Pseudonocardiaceae",
"Psychromonadaceae",
"Puniceicoccaceae",
"PYR10d3_fa",
"Rhizobiaceae",
"Rhizobiales_Incertae_Sedis",

"Rhodobiaceae",
"Rhodocyclaceae",
"Rhodospirillaceae",
"Rhodospirillales_Incertae_Sedis",
"Rhodospirillales_unclassified",
"Rhodothermaceae",
"Rickettsiaceae",
"Rickettsiales_Incertae_Sedis",
"Rickettsiales_unclassified",

```

"Rs-D42",
"Rubritaleaceae",

"S0134_terrestrial_group_fa",
"S085_fa",
"Saccharibacteria_fa",
"Sandaracinaceae",
"Saprospiraceae",
"SAR116_clade",
"SAR324_clade (Marine_group_B)_fa",
"Shewanellaceae",
"Solibacteraceae (Subgroup_3)",
"Solirubrobacteraceae",
"Sphingobacteriaceae",
"Sphingobacteriales_unclassified",

"Sphingomonadales_unclassified",
"Spirochaetaceae",
"Spirochaetales_unclassified",
"Spongiibacteraceae",
"SR-FBR-L83",
"SR1_(Absconditabacteria)_fa",

"Streptococcaceae",
"Streptomycetaceae",
"Subgroup_21_fa",
"Subgroup_22_fa",
"Subgroup_23_fa",
"Subgroup_6_fa",
"Subgroup_7_fa",

"Surface_1",
"Sva0071_fa",
"Sva0725",
"Sva0996_marine_group",
"Sva1033",
"Syntrophaceae",
"T9d",
"Tepidisphaeraceae",
"Thermoactinomycetaceae",
"Thiotrichaceae",
"TM146",
"TM6_(Dependentiae)_fa",
"TRA3-20_fa",
"Trueperaceae",
"uncultured",
"uncultured_fa",
"Unknown_Family",
"Veillonellaceae",
"Verrucomicrobia_fa",
"Verrucomicrobia_unclassified",
"Verrucomicrobiaceae",
"Verrucomicrobiales_unclassified",
"Vibrionaceae",
"X35_fa",
"Xanthobacteraceae",
"Xanthomonadaceae",
"Xanthomonadales_Incertae_Sedis",
"Xanthomonadales_unclassified"))

```

##only families that make up more than 5% total abundance are coloured. The rest are coloured white

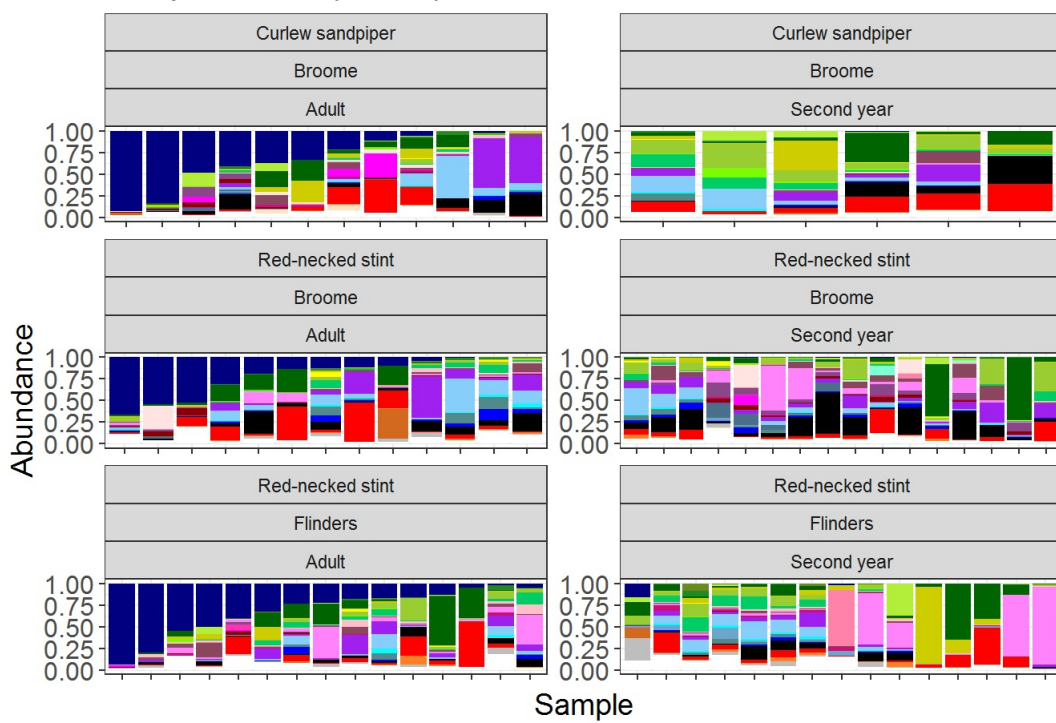
```

colors <- c("navy",
            "olivedrab4","olivedrab2","forestgreen","darkgreen","yellow3","yellow1","yellowgreen","lawngreen",
springgreen3","aquamarine",
            "mistyrose","pink","palevioletred1","palevioletred4", "orchid1","orchid4", "maroon1","magenta","dee
ppink3","red4","purple",

            "lightskyblue", "cyan","skyblue4", "skyblue3","darkslategray4","blue",
            "black",
            "red",

```


Family abundance per sample



```
##### ORDINATION ANALYSIS#####
#
```

```
##Figure S3)
```

```
set.seed(1)
```

```
migration_unifrac <- ordinate(
  physeq = migration,
  method = "NMDS",
  distance = "unifrac"
)
```

```
## Warning in UniFrac(physeq, ...): Randomly assigning root as -- Otu000566 --
## in the phylogenetic tree in the data you provided.
```

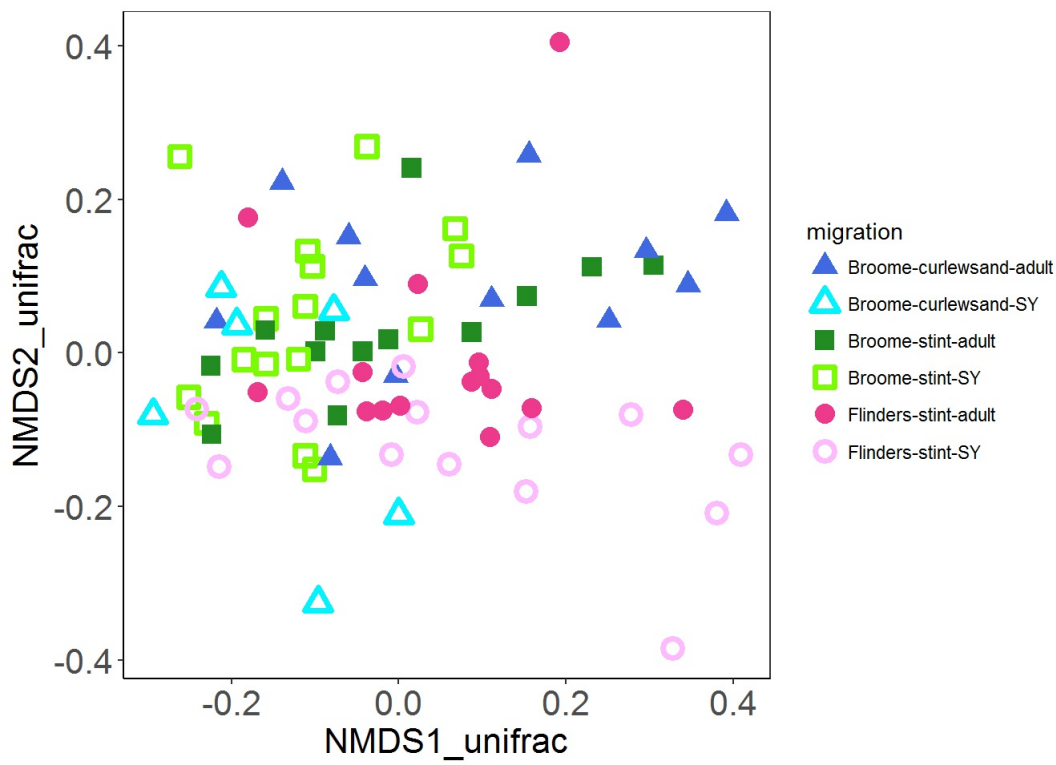
```
## Run 0 stress 0.2178487
## Run 1 stress 0.2322504
## Run 2 stress 0.2155708
## ... New best solution
## ... Procrustes: rmse 0.06628759  max resid 0.2890222
## Run 3 stress 0.2271085
## Run 4 stress 0.2286939
## Run 5 stress 0.2196326
## Run 6 stress 0.2349434
## Run 7 stress 0.2226841
## Run 8 stress 0.2133707
## ... New best solution
## ... Procrustes: rmse 0.02225117  max resid 0.1186602
## Run 9 stress 0.2190638
## Run 10 stress 0.2108544
## ... New best solution
## ... Procrustes: rmse 0.06605822  max resid 0.2885335
## Run 11 stress 0.2209799
## Run 12 stress 0.2172837
## Run 13 stress 0.22129
## Run 14 stress 0.2105399
## ... New best solution
## ... Procrustes: rmse 0.02824563  max resid 0.1835601
## Run 15 stress 0.2231375
## Run 16 stress 0.2185725
## Run 17 stress 0.2123678
## Run 18 stress 0.2131853
## Run 19 stress 0.2172729
## Run 20 stress 0.2181311
## *** No convergence -- monoMDS stopping criteria:
##      20: stress ratio > sratmax
```

```
## add unifrac ordination axes to dataframe and plot
```

```
unifrac<-data.frame(migration_unifrac$points)
sample_data(migration)$NMDS1_unifrac<-unifrac$MDS1
sample_data(migration)$NMDS2_unifrac<-unifrac$MDS2
```

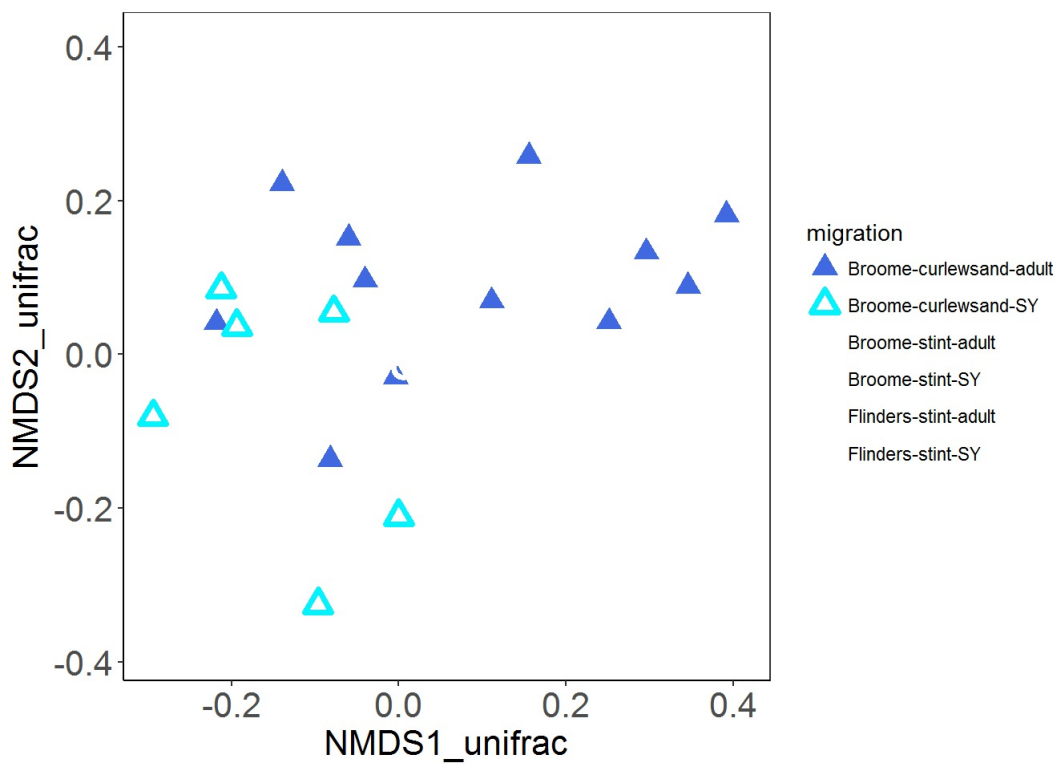
```
migration.df<-data.frame(sample_data(migration))
```

```
ggplot(data=migration.df, aes(x = NMDS1_unifrac, y = NMDS2_unifrac))+ geom_point(aes(colour = migration, shape
= migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("royalblue","turquoise1",
                                           "forestgreen","lawngreen","violetred2","plum1"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(), panel.grid.minor = ele
ment_blank(),
                                                    panel.background = element_blank(), axis.line = element_li
ne(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



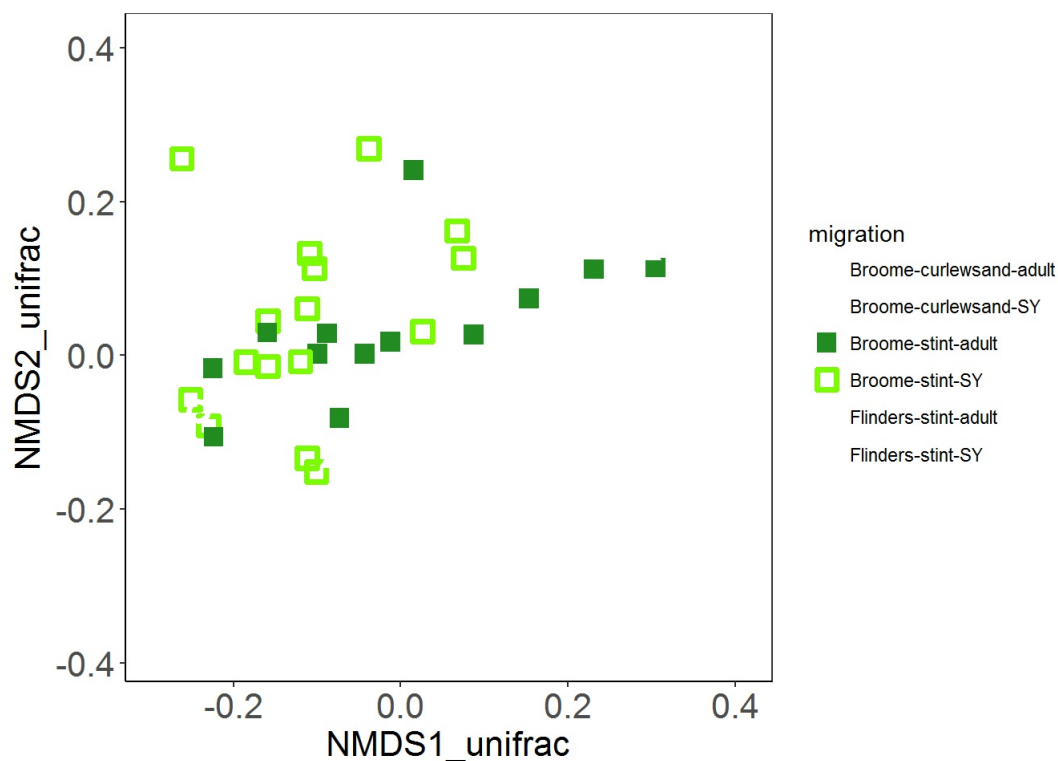
```
#####
```

```
ggplot(data=migration.df, aes(x = NMDS1_unifrac, y = NMDS2_unifrac))+ geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("royalblue","turquoise1",
                                         "white","white","white","white"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
                                                    panel.background = element_blank(), axis.line = element_line(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



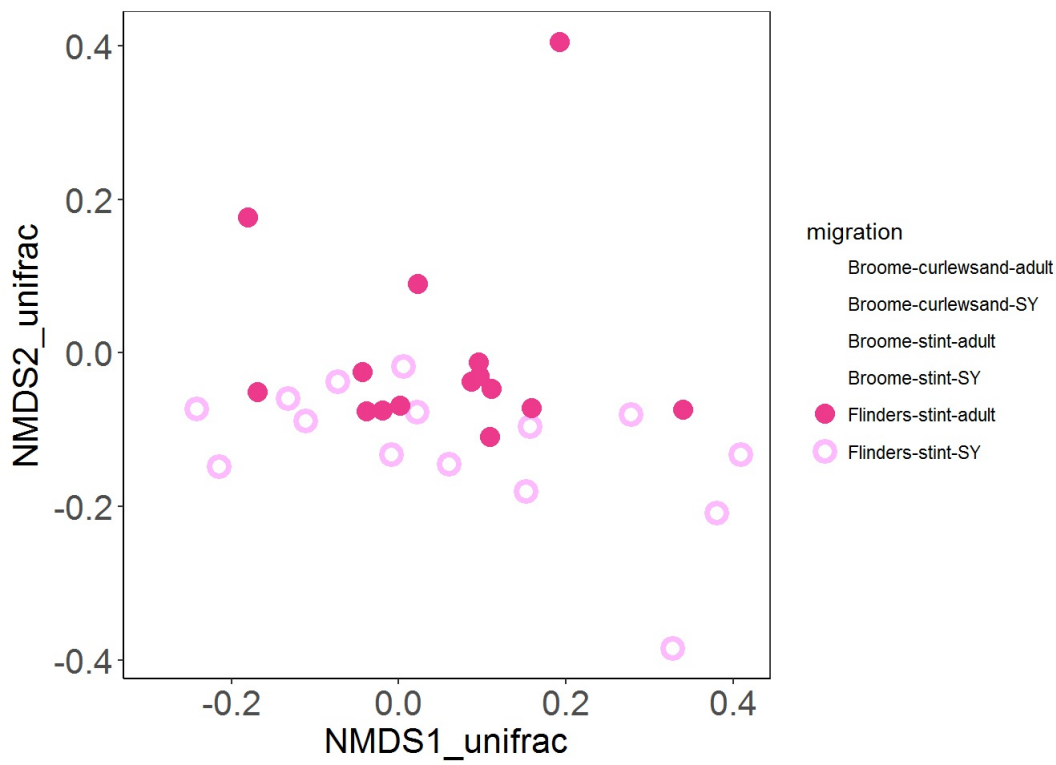
```
#####
```

```
ggplot(data=migration.df, aes(x = NMDS1_unifrac, y = NMDS2_unifrac))+ geom_point(aes(colour = migration, shape
= migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("white","white",
                                         "forestgreen","lawngreen","white","white"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(), panel.grid.minor = ele
ment_blank(),
                                         panel.background = element_blank(), axis.line = element_li
ne(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



```
#####
```

```
ggplot(data=migration.df, aes(x = NMDS1_unifrac, y = NMDS2_unifrac))+ geom_point(aes(colour = migration, shape
= migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("white","white",
                                         "white","white","violetred2","plum1"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(), panel.grid.minor = ele
ment_blank(),
                                         panel.background = element_blank(), axis.line = element_li
ne(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



```
#####
```

```
##stats
```

```
migration_unifrac <- phyloseq::distance(migration, method = "unifrac")
```

```
## Warning in UniFrac(physeq, ...): Randomly assigning root as -- Otu004229 --
## in the phylogenetic tree in the data you provided.
```

```
# make a data frame from the sample_data
sampledf <- data.frame(sample_data(migration))
```

```
# Adonis test
```

```
adonis(migration_unifrac ~ site+species+age, data = sampledf)
```

```
##
## Call:
## adonis(formula = migration_unifrac ~ site + species + age, data = sampledf)
##
## Permutation: free
## Number of permutations: 999
##
## Terms added sequentially (first to last)
##
##              Df SumsOfSqs MeanSqs F.Model    R2 Pr(>F)
## site          1    0.8129 0.81295   3.3509 0.04143 0.001 ***
## species        1    0.5368 0.53678   2.2126 0.02736 0.001 ***
## age            1    0.5606 0.56056   2.3106 0.02857 0.001 ***
## Residuals     73   17.7102 0.24261         0.90264
## Total         76   19.6205         1.00000
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#####
```

```
#####bray curtis
```

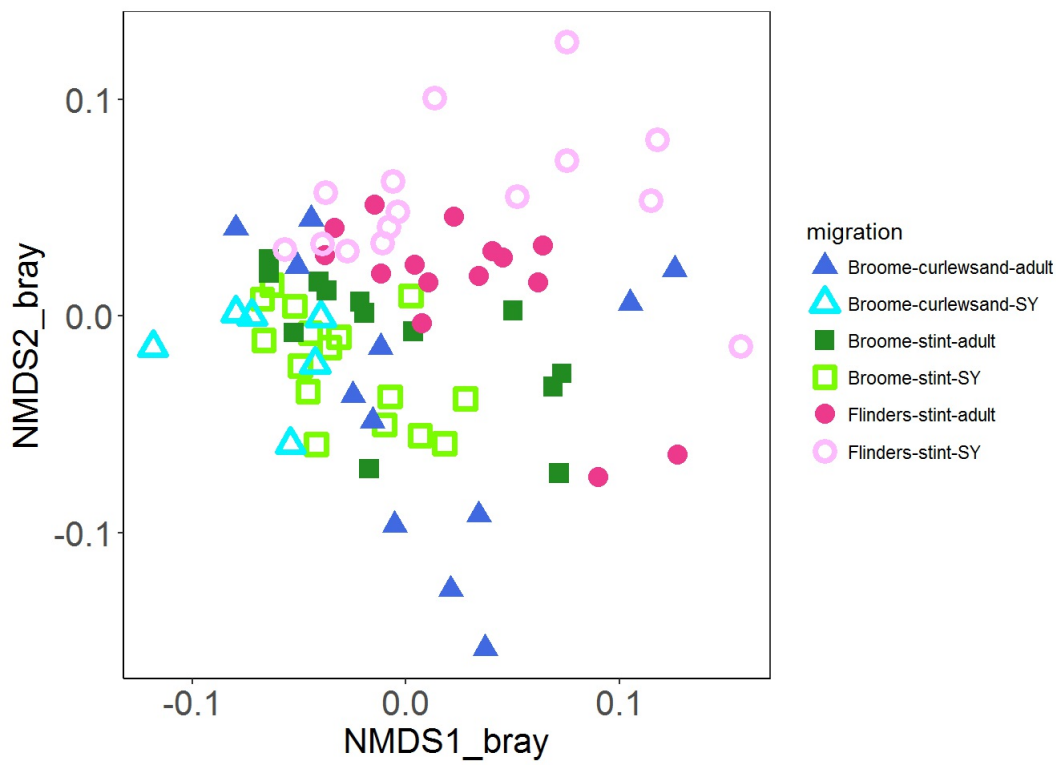
```
##Figure 3)
```

```
set.seed(1)
```

```
migration_bray <- ordinate(  
  physeq = migration,  
  method = "NMDS",  
  distance = "bray"  
)
```

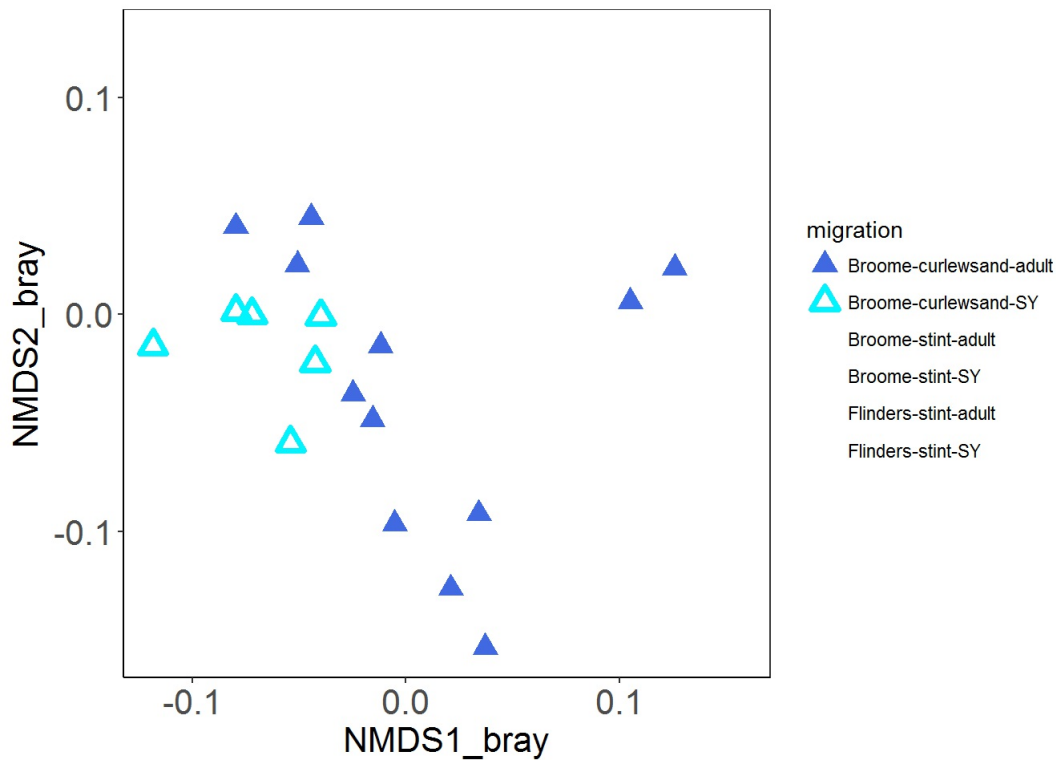
```
## Square root transformation  
## Wisconsin double standardization  
## Run 0 stress 0.2453602  
## Run 1 stress 0.2607537  
## Run 2 stress 0.255154  
## Run 3 stress 0.2561769  
## Run 4 stress 0.2518526  
## Run 5 stress 0.2718302  
## Run 6 stress 0.2533695  
## Run 7 stress 0.2584876  
## Run 8 stress 0.2453605  
## ... Procrustes: rmse 0.0003253219 max resid 0.002527825  
## ... Similar to previous best  
## Run 9 stress 0.261244  
## Run 10 stress 0.2517643  
## Run 11 stress 0.2439716  
## ... New best solution  
## ... Procrustes: rmse 0.04242624 max resid 0.3241008  
## Run 12 stress 0.2649426  
## Run 13 stress 0.2657925  
## Run 14 stress 0.2448462  
## Run 15 stress 0.243968  
## ... New best solution  
## ... Procrustes: rmse 0.0320396 max resid 0.2619141  
## Run 16 stress 0.26698  
## Run 17 stress 0.2443094  
## ... Procrustes: rmse 0.04255965 max resid 0.2075804  
## Run 18 stress 0.2705507  
## Run 19 stress 0.2493743  
## Run 20 stress 0.2500642  
## *** No convergence -- monoMDS stopping criteria:  
##      20: stress ratio > sratmax
```

```
bray<-data.frame(migration_bray$points)  
sample_data(migration)$NMDS1_bray<-bray$MDS1  
sample_data(migration)$NMDS2_bray<-bray$MDS2  
  
migration.df<-data.frame(sample_data(migration))  
  
##plot  
  
ggplot(data=migration.df, aes(x = NMDS1_bray, y = NMDS2_bray))+  
  geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +  
  theme_bw()+scale_color_manual(values=c("royalblue","turquoise1",  
                                           "forestgreen","lawngreen","violetred2","plum1"))+  
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(),  
                                                    panel.grid.minor = element_blank(),  
                                                    panel.background = element_blank(),  
                                                    axis.line = element_line(colour = "black"))+  
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



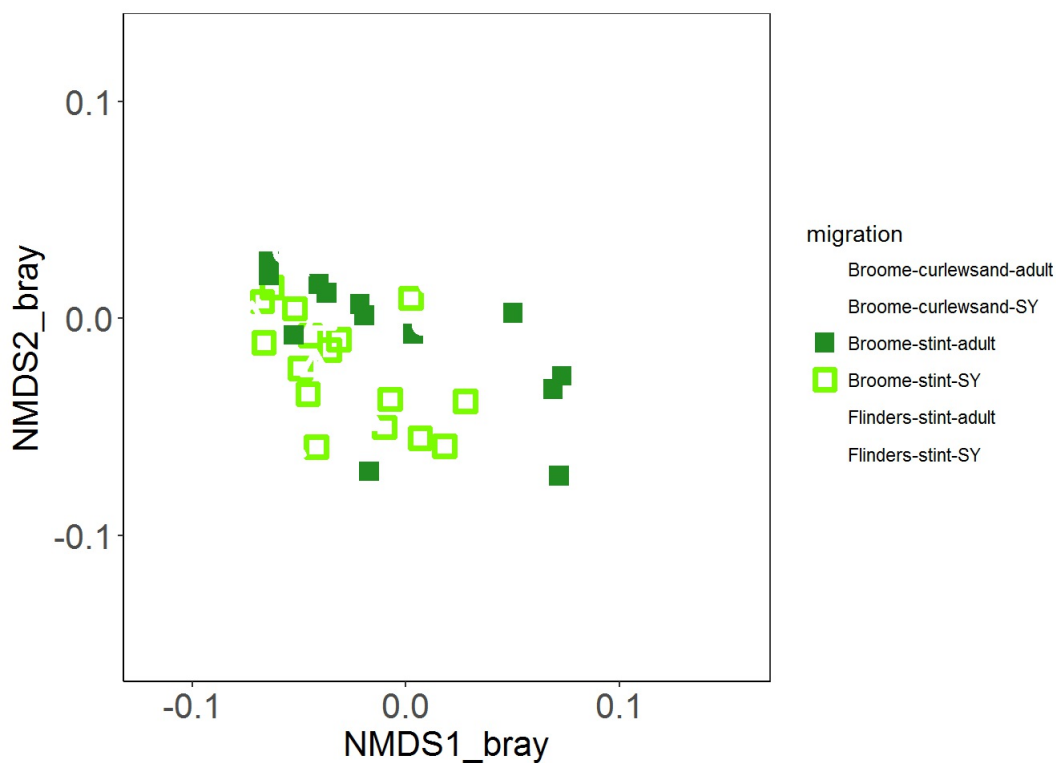
```
#####
#####

ggplot(data=migration.df, aes(x = NMDS1_bray, y = NMDS2_bray))+
  geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("royalblue","turquoise1",
                                         "white","white","white","white"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(),
                                                    panel.grid.minor = element_blank(),
                                                    panel.background = element_blank(),
                                                    axis.line = element_line(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



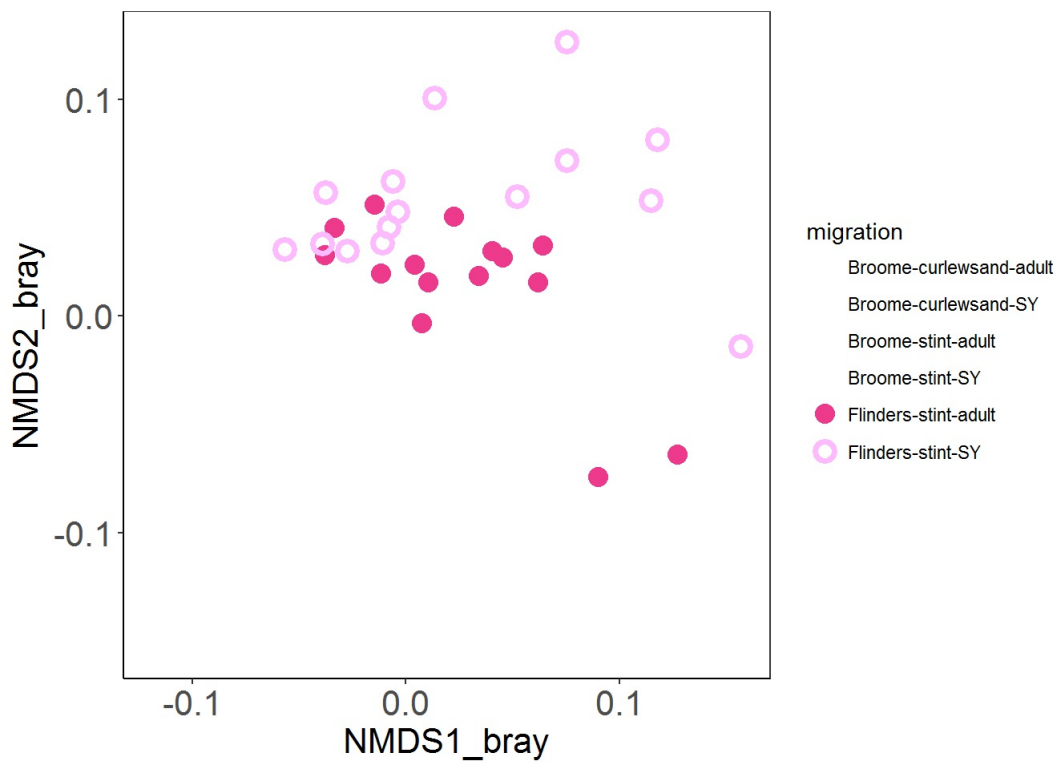

```
#####
```

```
ggplot(data=migration.df, aes(x = NMDS1_bray, y = NMDS2_bray))+
  geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("white","white",
                                          "forestgreen","lawngreen","white","white"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(),
                                                    panel.grid.minor = element_blank(),
                                                    panel.background = element_blank(),
                                                    axis.line = element_line(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



```
#####
```

```
ggplot(data=migration.df, aes(x = NMDS1_bray, y = NMDS2_bray))+
  geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("white","white",
                                          "white","white","violetred2","plum1"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(),
                                                    panel.grid.minor = element_blank(),
                                                    panel.background = element_blank(),
                                                    axis.line = element_line(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



```
#####

##stats

migration_bray <- phyloseq::distance(migration, method = "bray")

# make a data frame from the sample_data
sampledf <- data.frame(sample_data(migration))

# Adonis test
adonis(migration_bray ~ site+species+age, data = sampledf)
```

```
##
## Call:
## adonis(formula = migration_bray ~ site + species + age, data = sampledf)
##
## Permutation: free
## Number of permutations: 999
##
## Terms added sequentially (first to last)
##
##           Df SumsOfSqs MeanSqs F.Model      R2 Pr(>F)
## site      1   1.1939 1.19392   3.4024 0.04114 0.001 ***
## species   1   0.9877 0.98765   2.8146 0.03403 0.001 ***
## age       1   1.2225 1.22247   3.4837 0.04212 0.001 ***
## Residuals 73   25.6163 0.35091         0.88270
## Total     76   29.0203         1.00000
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#####

##remove corynebacteria to see if any other differences

migration1<-subset_taxa(migration, Order != "Corynebacteriales")
migration1<- prune_taxa(taxa_sums(migration1)>0, migration1)

set.seed(1)

migration_bray <- ordinate(
  physeq = migration1,
  method = "NMDS",
  distance = "bray"
)
```

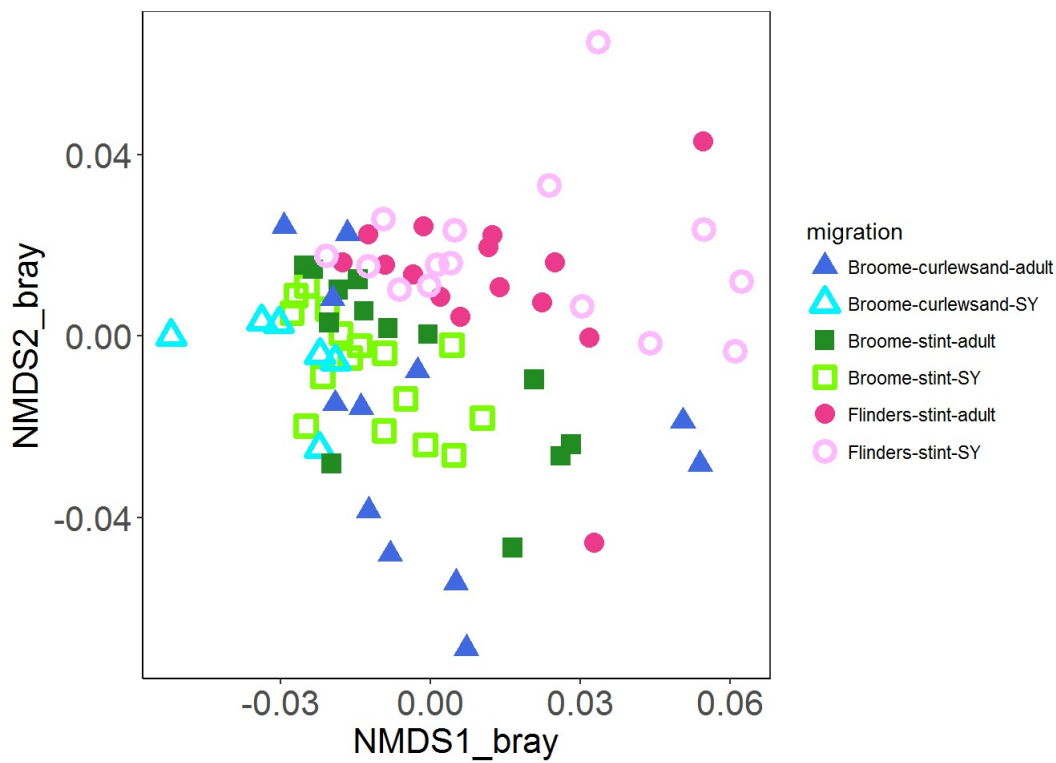
```
## Square root transformation
## Wisconsin double standardization
## Run 0 stress 0.2434372
## Run 1 stress 0.2520435
## Run 2 stress 0.2505532
## Run 3 stress 0.2414003
## ... New best solution
## ... Procrustes: rmse 0.05723422 max resid 0.3325888
## Run 4 stress 0.2407164
## ... New best solution
## ... Procrustes: rmse 0.04558418 max resid 0.3847929
## Run 5 stress 0.2636338
## Run 6 stress 0.2525901
## Run 7 stress 0.2554168
## Run 8 stress 0.2445606
## Run 9 stress 0.2592545
## Run 10 stress 0.2590405
## Run 11 stress 0.2401147
## ... New best solution
## ... Procrustes: rmse 0.04837525 max resid 0.3577633
## Run 12 stress 0.2539493
## Run 13 stress 0.248226
## Run 14 stress 0.2423354
## Run 15 stress 0.2504476
## Run 16 stress 0.2600155
## Run 17 stress 0.2488418
## Run 18 stress 0.2526451
## Run 19 stress 0.240373
## ... Procrustes: rmse 0.03543252 max resid 0.2868491
## Run 20 stress 0.2567804
## *** No convergence -- monoMDS stopping criteria:
##      20: stress ratio > sratmax
```

```
bray<-data.frame(migration_bray$points)
sample_data(migration1)$NMDS1_bray<-bray$MDS1
sample_data(migration1)$NMDS2_bray<-bray$MDS2

migration1.df<-data.frame(sample_data(migration1))

##plot

ggplot(data=migration1.df, aes(x = NMDS1_bray, y = NMDS2_bray))+
  geom_point(aes(colour = migration, shape = migration), alpha = 1, size = 3, stroke = 2) +
  theme_bw()+scale_color_manual(values=c("royalblue","turquoise1",
                                         "forestgreen","lawngreen","violetred2","plum1"))+
  scale_shape_manual(values=c(17,2,15,0,16,1))+theme(panel.grid.major = element_blank(),
                                                    panel.grid.minor = element_blank(),
                                                    panel.background = element_blank(),
                                                    axis.line = element_line(colour = "black"))+
  theme(axis.text=element_text(size=16), axis.title=element_text(size=16))
```



```
##stats

migration_bray <- phyloseq::distance(migration1, method = "bray")
#migration_bray <- phyloseq::distance(migration1, method = "unifrac")

# make a data frame from the sample_data
sampledf <- data.frame(sample_data(migration1))

# Adonis test
adonis(migration_bray ~ site+species+age, data = sampledf)

##
## Call:
## adonis(formula = migration_bray ~ site + species + age, data = sampledf)
##
## Permutation: free
## Number of permutations: 999
##
## Terms added sequentially (first to last)
##
##           Df SumsOfSqs MeanSqs F.Model      R2 Pr(>F)
## site       1    1.0202 1.02019   2.8284 0.03539 0.001 ***
## species    1    0.9369 0.93687   2.5974 0.03250 0.001 ***
## age        1    0.5401 0.54012   1.4974 0.01874 0.042 *
## Residuals 73   26.3305 0.36069         0.91338
## Total     76   28.8277         1.00000
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#####

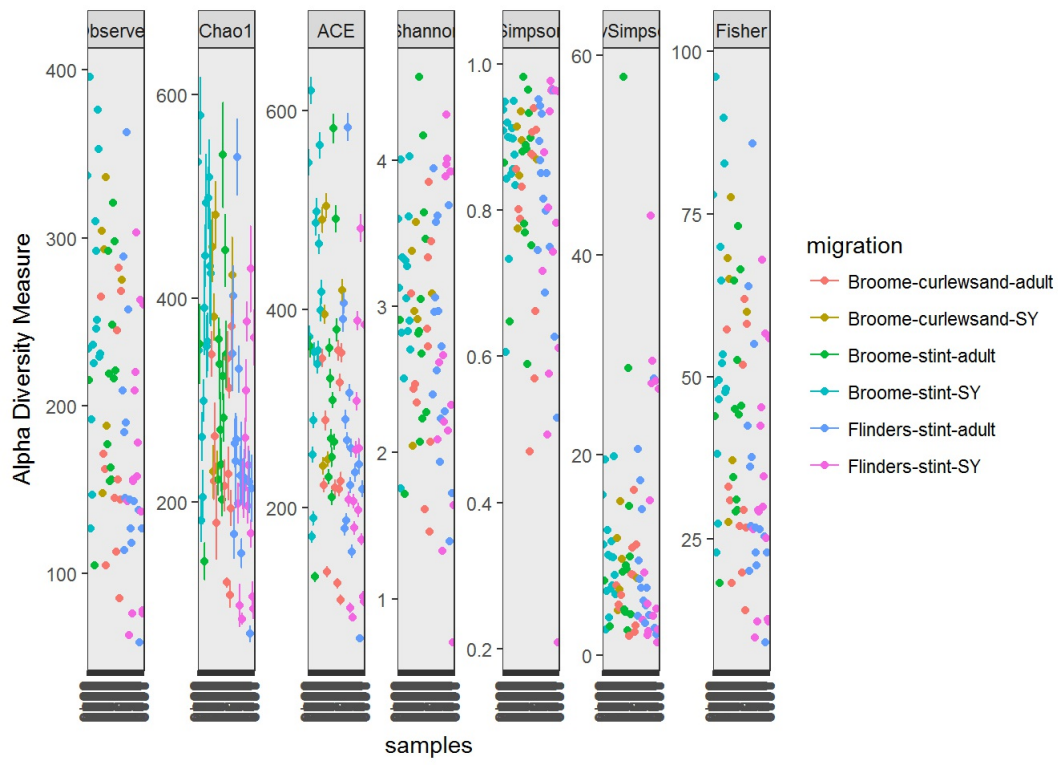
#####

#####

####alpha diversity
### Figure 3c)

plot_richness(migration, color = "migration")
```

```
## Warning: Removed 385 rows containing missing values (geom_errorbar).
```



```

min_lib <- min(sample_sums(migration))

# Initialize matrices to store richness and evenness estimates
nsamp = nsamples(migration)
trials = 100

richness <- matrix(nrow = nsamp, ncol = trials)
row.names(richness) <- sample_names(migration)

evenness <- matrix(nrow = nsamp, ncol = trials)
row.names(evenness) <- sample_names(migration)

###

set.seed(3)

for (i in 1:100) {
  # Subsample
  r <- rarefy_even_depth(migration, sample.size = min_lib, verbose = FALSE, replace = TRUE)

  # Calculate richness
  rich <- as.numeric(as.matrix(estimate_richness(r, measures = "Observed")))
  richness[,i] <- rich

  # Calculate evenness
  even <- as.numeric(as.matrix(estimate_richness(r, measures = "InvSimpson")))
  evenness[,i] <- even
}

SampleID <- row.names(richness)
mean <- apply(richness, 1, mean)
sd <- apply(richness, 1, sd)
measure <- rep("Richness", nsamp)
rich_stats <- data.frame(SampleID, mean, sd, measure)

SampleID <- row.names(evenness)
mean <- apply(evenness, 1, mean)
sd <- apply(evenness, 1, sd)
measure <- rep("Inverse Simpson", nsamp)
even_stats <- data.frame(SampleID, mean, sd, measure)

s <- data.frame(sample_data(migration))

alphadiv_rich_observed <- merge(rich_stats, s, by = "row.names")
alphadiv_even_inverse <- merge(even_stats, s, by = "row.names")

#summary stats

tapply(alphadiv_rich_observed$mean, alphadiv_rich_observed$migration, summary)

```

```

## $`Broome-curlewsand-adult`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  74.05  116.02  132.22  152.53  210.12  251.35
##
## $`Broome-curlewsand-SY`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  118.6   176.5   236.0   212.4   248.5   274.6
##
## $`Broome-stint-adult`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   91.52  139.22  176.77  179.97  215.02  262.04
##
## $`Broome-stint-SY`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   107.2   185.0   206.9   216.7   257.8   324.4
##
## $`Flinders-stint-adult`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   52.94  103.52  118.65  145.55  162.16  294.01
##
## $`Flinders-stint-SY`
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   54.81   92.65  141.46  141.66  189.76  256.55

```

```
unique(alphadiv_rich_observed$migration)
```

```
## [1] Broome-stint-SY      Broome-stint-adult      Broome-curlewsand-adult
## [4] Broome-curlewsand-SY    Flinders-stint-adult    Flinders-stint-SY
## 6 Levels: Broome-curlewsand-adult ... Flinders-stint-SY
```

```
group1a<-subset(alphadiv_rich_observed, migration=="Broome-curlewsand-adult")
sd(group1a$mean)
```

```
## [1] 59.45963
```

```
group1b<-subset(alphadiv_rich_observed, migration=="Broome-curlewsand-SY")
sd(group1b$mean)
```

```
## [1] 60.1231
```

```
group2a<-subset(alphadiv_rich_observed, migration=="Broome-stint-adult")
sd(group2a$mean)
```

```
## [1] 53.34191
```

```
group2b<-subset(alphadiv_rich_observed, migration=="Broome-stint-SY")
sd(group2b$mean)
```

```
## [1] 62.84926
```

```
group3a<-subset(alphadiv_rich_observed, migration=="Flinders-stint-adult")
sd(group3a$mean)
```

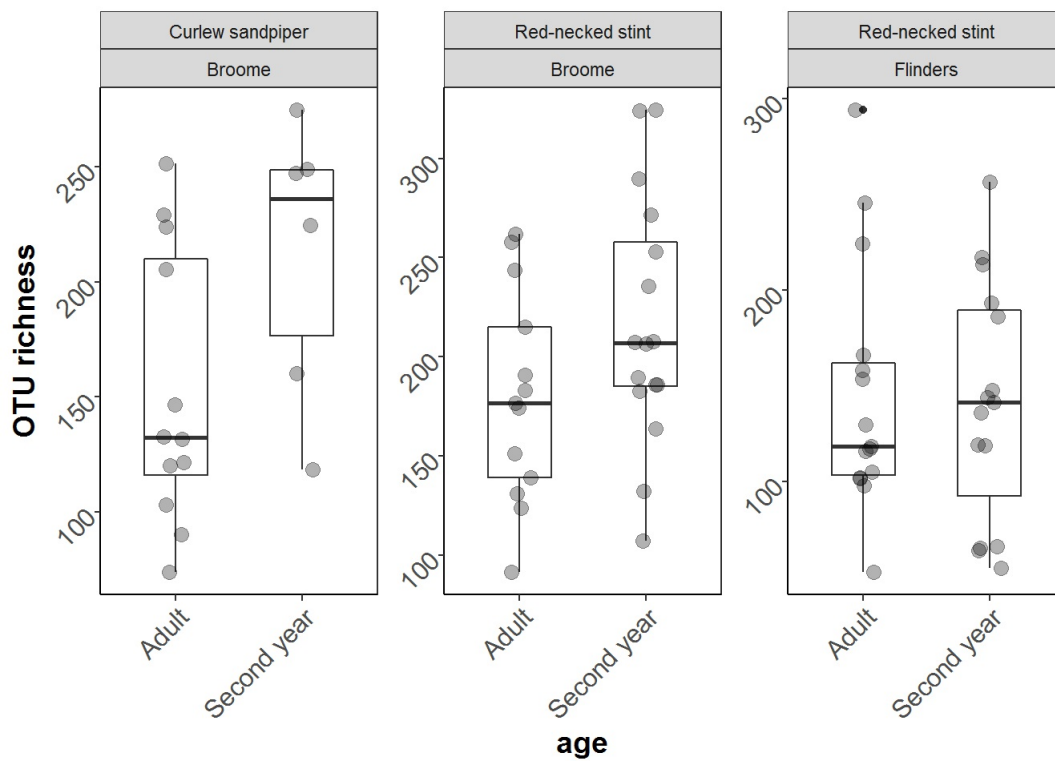
```
## [1] 64.27512
```

```
group3b<-subset(alphadiv_rich_observed, migration=="Flinders-stint-SY")
sd(group3b$mean)
```

```
## [1] 62.57409
```

```
##observed richness
```

```
ggplot(alphadiv_rich_observed, aes(x = age, y = mean))+geom_boxplot(width=0.5)+
  geom_jitter(width=0.1, alpha = 0.3, size=3)+ylab("OTU richness")+
  theme(axis.text=element_text(size=12, angle=45, hjust=1), axis.title=element_text(size=14, face="bold"))+
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
        panel.background = element_blank(), axis.line = element_line(colour = "black"))+
  facet_wrap(species~site, scales="free", nrow=1, ncol=3)
```



```
#####SHANNON

set.seed(3)

for (i in 1:100) {
  # Subsample
  r <- rarefy_even_depth(migration, sample.size = min_lib, verbose = FALSE, replace = TRUE)

  # Calculate richness
  rich <- as.numeric(as.matrix(estimate_richness(r, measures = "Shannon")))
  richness[,i] <- rich

  # Calculate evenness
  even <- as.numeric(as.matrix(estimate_richness(r, measures = "InvSimpson")))
  evenness[,i] <- even
}

SampleID <- row.names(richness)
mean <- apply(richness, 1, mean)
sd <- apply(richness, 1, sd)
measure <- rep("Richness", nsamp)
rich_stats <- data.frame(SampleID, mean, sd, measure)

SampleID <- row.names(evenness)
mean <- apply(evenness, 1, mean)
sd <- apply(evenness, 1, sd)
measure <- rep("Inverse Simpson", nsamp)
even_stats <- data.frame(SampleID, mean, sd, measure)

s <- data.frame(sample_data(migration))

alphadiv_rich_observed <- merge(rich_stats, s, by = "row.names")
alphadiv_even_inverse <- merge(even_stats, s, by = "row.names")

#summary stats

tapply(alphadiv_rich_observed$mean, alphadiv_rich_observed$migration, summary)
```



```
## $`Broome-curlewsand-adult`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    1.449   2.263   2.585   2.624   3.128   3.824
##
## $`Broome-curlewsand-SY`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    2.032   2.908   3.000   2.971   3.279   3.545
##
## $`Broome-stint-adult`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    1.708   2.257   2.814   2.936   3.433   4.546
##
## $`Broome-stint-SY`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    1.744   2.803   3.176   3.112   3.381   3.990
##
## $`Flinders-stint-adult`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    1.392   2.244   2.717   2.723   3.307   3.918
##
## $`Flinders-stint-SY`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    0.6943  2.1060  2.6075  2.7683  3.8974  4.2952
```

```
unique(alphadiv_rich_observed$migration)
```

```
## [1] Broome-stint-SY      Broome-stint-adult      Broome-curlewsand-adult
## [4] Broome-curlewsand-SY    Flinders-stint-adult    Flinders-stint-SY
## 6 Levels: Broome-curlewsand-adult ... Flinders-stint-SY
```

```
group1a<-subset(alphadiv_rich_observed, migration=="Broome-curlewsand-adult")
sd(group1a$mean)
```

```
## [1] 0.720232
```

```
group1b<-subset(alphadiv_rich_observed, migration=="Broome-curlewsand-SY")
sd(group1b$mean)
```

```
## [1] 0.5237586
```

```
group2a<-subset(alphadiv_rich_observed, migration=="Broome-stint-adult")
sd(group2a$mean)
```

```
## [1] 0.8243023
```

```
group2b<-subset(alphadiv_rich_observed, migration=="Broome-stint-SY")
sd(group2b$mean)
```

```
## [1] 0.5628283
```

```
group3a<-subset(alphadiv_rich_observed, migration=="Flinders-stint-adult")
sd(group3a$mean)
```

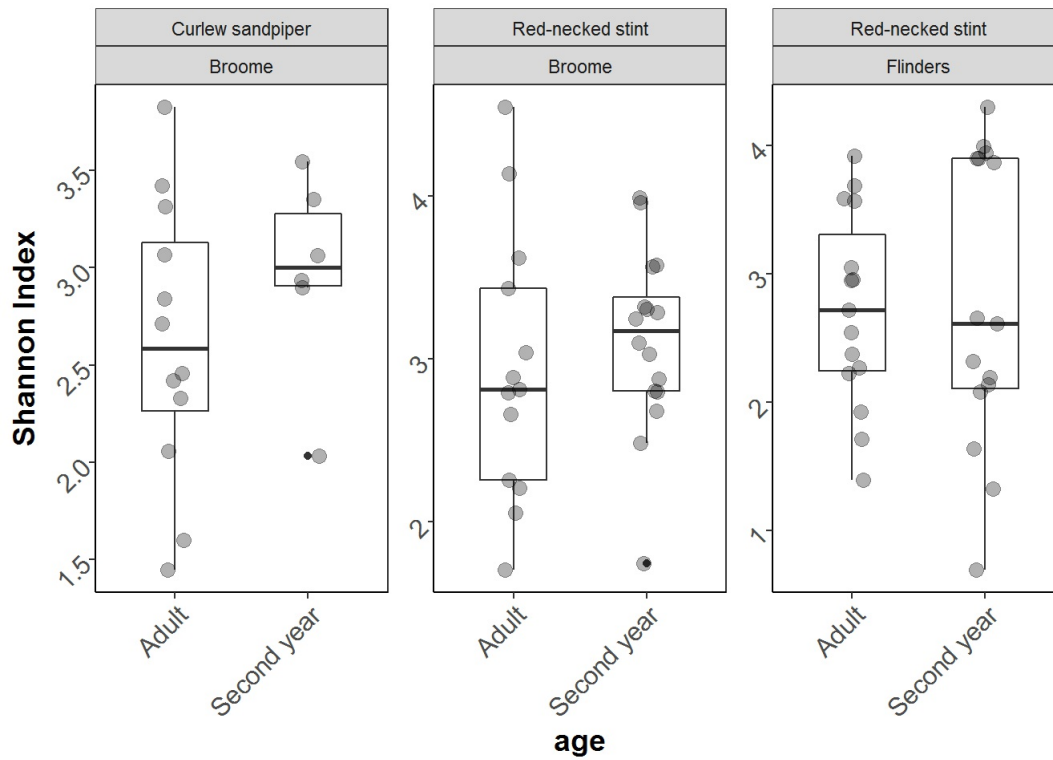
```
## [1] 0.7589452
```

```
group3b<-subset(alphadiv_rich_observed, migration=="Flinders-stint-SY")
sd(group3b$mean)
```

```
## [1] 1.137082
```

```
##observed richness
```

```
ggplot(alphadiv_rich_observed, aes(x = age, y = mean))+geom_boxplot(width=0.5)+  
  geom_jitter(width=0.1, alpha = 0.3, size=3)+ylab("Shannon Index")+  
  theme(axis.text=element_text(size=12, angle=45, hjust=1), axis.title=element_text(size=14,face="bold"))+  
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),  
        panel.background = element_blank(), axis.line = element_line(colour = "black"))+  
  facet_wrap(species~site, scales="free", nrow=1, ncol=3)
```



```
##### END #####
```