





Microbial community analysis using Chipster: data tidying, visualization and statistics

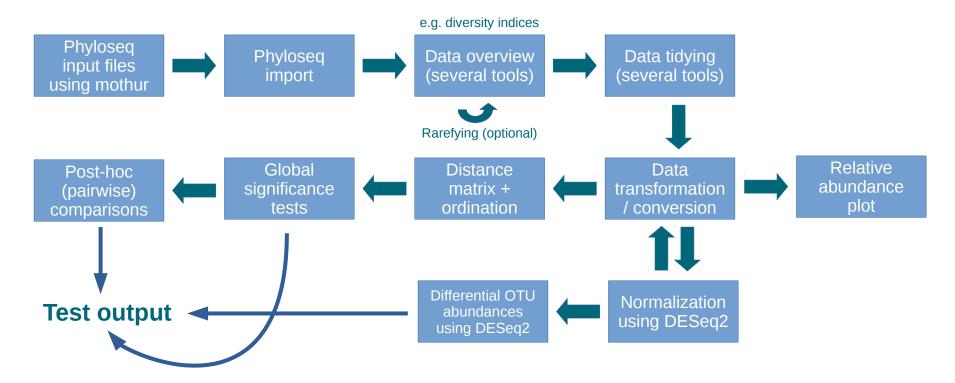
Part 1: Tool overview and data importing

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Workflow for data tidying, analysis and statistics

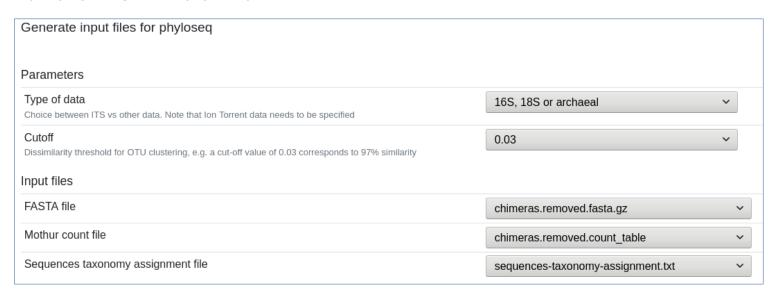




Generating phyloseq input files

Phyloseq is a multi-use R package for microbial community data processing and analysis

https://joey711.github.io/phyloseq/

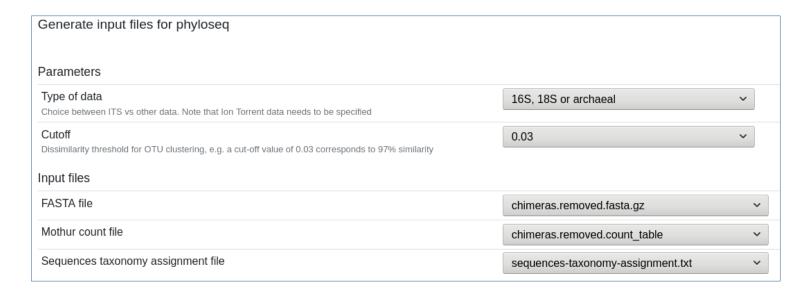




Generating phyloseq input files

Specifications for creating phyloseq input files:

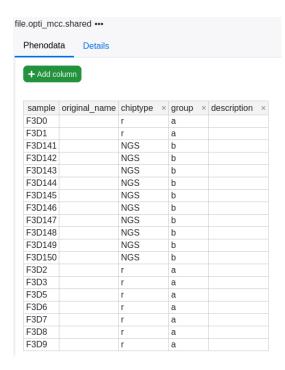
Type of data, % cutoff and files produced by mothur (FASTA, count file, taxonomy file)





The phenodata file

Generated input files: .shared + phenodata file, consensus taxonomy file



The phenodata file is an editable table with unique IDs for each sample and sample groupings



Converting input files into a phyloseq object





Converting input files into a phyloseq object

```
### Imported phyloseg object ###
phyloseq-class experiment-level object
otu table() OTU Table:
                                 [ 1114 taxa and 19 samples ]
sample data() Sample Data:
                                 [ 19 samples by 5 sample variables ]
tax table() Taxonomy Table:
                                 [ 1114 taxa by 6 taxonomic ranks ]
### Sample names ###
              "F3D1"
                       "F3D141" "F3D142" "F3D143"
                                                  "F3D144" "F3D145"
                                                                    "F3D146'
                                                                     "F3D6"
 [9] "F3D147" "F3D148" "F3D149" "F3D150" "F3D2"
[17] "F3D7"
              "F3D8"
                       "F3D9"
### Sample variables ###
[1] "sample"
                    "original name" "chiptype"
                                                     "aroup"
[5] "description"
```

Produces a phyloseq object (.Rda) and a text summary

The Rda file is used as the input for downstream analyses





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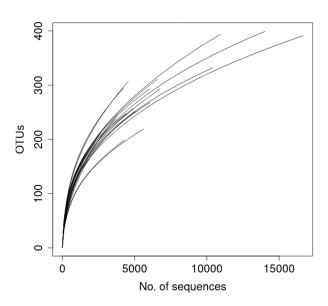
Part 2: Data inspection and tidying

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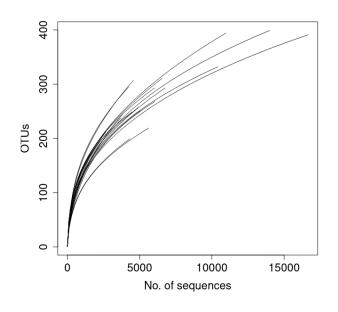
Sequence no.s, rarefaction curves and alpha diversity



```
### Per-sample sequence no.s ###
  F3D0
         F3D1 F3D141 F3D142 F3D143 F3D144 F3D145 F3D146 F3D147 F3D148 F3D149
         4904
                5046
                       2629
                              2635
                                      3843
  6568
                                             6063
                                                                 10413 10964
F3D150
         F3D2
                F3D3
                       F3D5
                              F3D6
                                      F3D7
                                             F3D8
                                                    F3D9
        16662
                5626
                       3682
                              6768
                                      4318
                                                    6036
 4563
                                             4445
### Alpha diversity estimates (observed OTUs, Chao1, Shannon's index, Pielou's evenness) ###
                                               pielou group
       0bserved
                   Chao1 se.chao1 Shannon
F3D0
            311 634.0357 83.85414 4.079427 0.7107272
F3D1
            257 519.6500 79.02103 4.209831 0.7586544
F3D141
            251 376.8378 34.66995 3.767491 0.6818430
F3D142
            211 358.0968 41.63567 3.624253 0.6771952
F3D143
            211 321.7576 32.37996 3.779248 0.7061562
```



Sequence no.s, rarefaction curves and alpha diversity



Previously often used for library size normalization, but increasing evidence for drawbacks

Alternatives: data transformation / corrections for unequal library size

PLOS COMPUTATIONAL BIOLOGY



Waste Not, Want Not: Why Rarefying Microbiome Data Is Inadmissible

Paul J. McMurdie, Susan Holmes 🖪

Published: April 3, 2014 • https://doi.org/10.1371/journal.pcbi.1003531

CSC

Taxon-level clean-up tools

- Taxon composition overview (user-specified level)
- Removing non-specific sequences (keep e.g. Bacteria or Archaea only)
- Remove chloroplast and/or mitochondrial sequences
- Manually remove specific taxa

Split into a total of three different tools



Visualizing and filtering low-abundance features

Prevalence

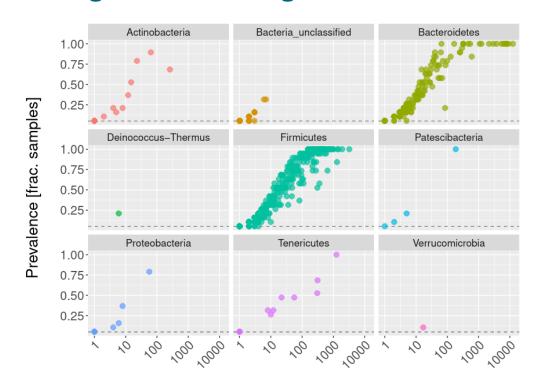
- Definition: no. of samples in which a taxon appears at least once
- Visualization and filtering tool

Singletons and doubletons

Text summary and filtering tool



Visualizing and filtering low-abundance features



Total abundance







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Part 3: Transformations and ordinations

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Data transformation

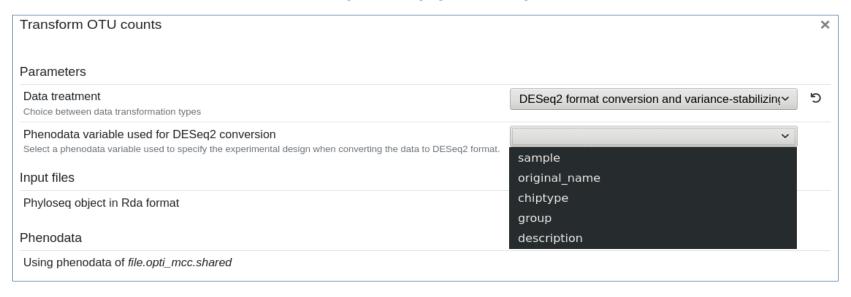
Four options (April 2021)





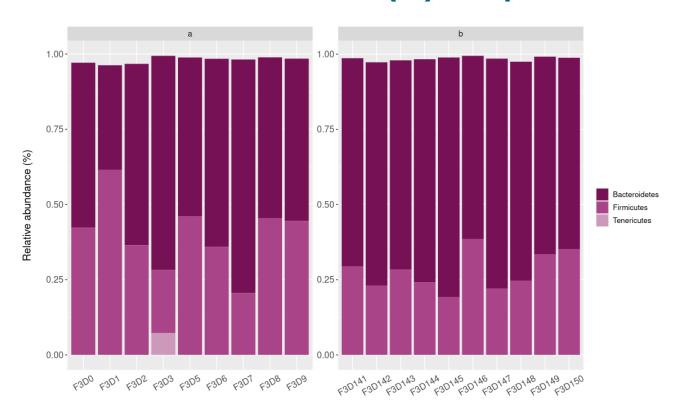
Data transformation

Four options (April 2021)





Relative abundance (%) bar plots





Distance matrices and ordinations

Distance measures: Euclidean or Bray-Curtis
Types of ordination: nMDS or db-RDA

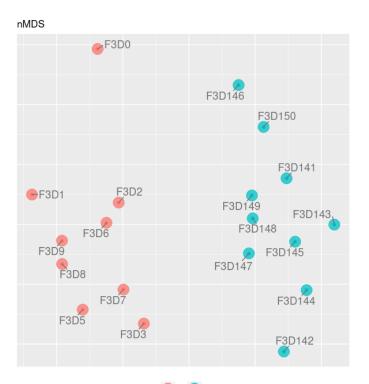
(Aitchison distance = CLR + Euclidean)

Parameters		
Type of distance measure Choice between Euclidean and Bray-Curtis distances	Euclidean	~
Type of ordination Choice between using non-metric multidimensional scaling (nMDS) or distance-based redundancy analysis (db-RDA)	nMDS	v
Phenodata variable with sequencing sample IDs Phenodata variable with unique IDs for each community profile.	sample	v

Guide to Statistical Analysis in Microbial Ecology: https://sites.google.com/site/mb3gustame/



Non-metric multidimensional scaling (nMDS)





Distance-based redundancy analysis (db-RDA)

Requires specification of a phenodata variable

Parameters

Type of distance measure Choice between Euclidean and Bray-Curtis distances	Euclidean	~
Type of ordination Choice between using non-metric multidimensional scaling (nMDS) or distance-based redundancy analysis (db-RDA)	db-RDA	~
Phenodata variable with sequencing sample IDs Phenodata variable with unique IDs for each community profile.	sample	~
Show sample IDs in ordination? Should sample labels be plotted next to data points in the ordination?	Yes	~
Phenodata variable for grouping ordination points by colour Phenodata variable used for grouping ordination points by colour.	group	~
Phenodata variable for grouping ordination points by shape Phenodata variable used for grouping ordination points by shape.		~
Phenodata variable 1 for db-RDA formula specification 1st phenodata variable used in the "formula" argument when performing db-RDA (minimum requirement is 1	group	~



Distance-based redundancy analysis (db-RDA)













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Part 4: Statistics

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PERMANOVA + PERMDISP

Require a distance matrix as the input

PERMANOVA (permutational multivariate analysis of variance)

- Global test: "Does community structure differ between sample groups?"
- Pairwise test: "Which particular groups differ from one another?"
- Influenced by both *location* and *dispersion* (more on these terms on the next slide)

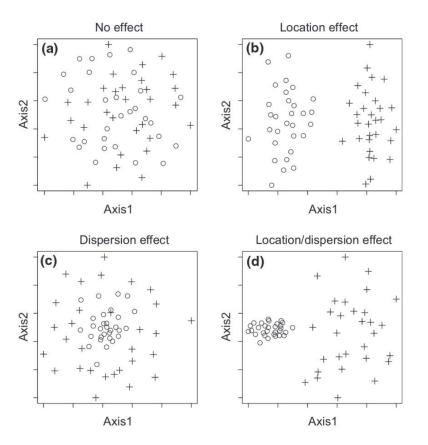
PERMDISP (test for the homogeneity of multivariate dispersions)

- Run only if get a significant (p < 0.05) PERMANOVA result
- Can help tell why the PERMANOVA result is significant



A significant PERMANOVA result can be due to:

- A location effect
- A dispersion effect
- A combination of both



Source: https://doi.org/10.1111/j.2041-210X.2011.00127.x



PERMANOVA output

```
### Global PERMANOVA summary ###
Call:
adonis(formula = ps_dist ~ get(pheno1), data = ps_df)
Permutation: free
Number of permutations: 999
Terms added sequentially (first to last)
           Df SumsOfSqs MeanSqs F.Model
                                             R2 Pr(>F)
                                                 0.001 ***
get(pheno1)
                 3118.6 3118.58 6.9211 0.28933
Residuals
                 7660.1 450.59
                                        0.71067
Total
                10778.6
                                        1.00000
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Df: Degrees of freedom

F.Model: Test statistic (pseudo-*F*)

Pr(>F):
Statistical significance
(p value)



Post-hoc comparisons

Following significant global PERMANOVA:

Pairwise PERMANOVA (similar as global test but for sample pairs)

Following significant PERMDISP:

Tukey's Honestly Significant Difference (HSD) test

Both methods use a correction for multiple testing (Benjamini-Hochberg correction)

DESeq2



- Originates from the RNAseq field
- Used to address the question: "Which taxa are differentially abundant between sample groups?"
- Enables inferences such as: "Illness x is associated with a reduction in the abundance of beneficial gut microbes y and z"
- Untransformed data used as input (internally corrects for differences in library size)
- Results given as log fold changes
- Link to more info online:
 - https://joey711.github.io/phyloseq-extensions/DESeq2.html

DESeq2



Current tool configuration (April 2021)

- Focuses on comparisons of two groups at a time
 - If selected phenodata column has >2 groups, can specify a pair (Group 1 and Group 2)
- Phenodata column with two groups:
 - Reference level selected alphabetically (e.g. 'b vs a' or 'sick vs healthy')
- Phenodata column with >2 groups:
 - Reference level corresponds to 'Group 2'



DESeq2

