Basic storage, access, and manipulation of phylogenetic sequencing data with *phyloseq*

Paul J. McMurdie

Holmes Group

Statistics Department, Stanford University

Stanford, CA

October 28, 2011

Contents

1 Introduction

There are already several ecology and phylogenetic packages available in R, including the adephylo, vegan, ade4, picante, ape, phangorn, phylobase, and OTUbase packages. These can already take advantage of many of the powerful statistical and graphics tools available in R. However, at present a user must devise their own methods for parsing the output of their favorite OTU clustering application, and, as a consequence, there is also no standard within Bioconductor (or R generally) for storing or sharing the suite of related data objects that describe a phylogenetic sequencing project. The phyloseq package[1] seeks to address these issues by providing a related set of S4 classes that internally manage the handling tasks associated with organizing, linking, storing, and analyzing phylogenetic sequencing data. *phyloseq* additionally provides some convenience wrappers for input from common clustering applications, common analysis pipelines, and native implementation of methods that are not available in other R packages.

---

[1]Throughout this vignette we use regular or *italics* font for packages/applications with names that are capitalized or uncapitalized, respectively. We further use a `courier` style font for R code, including function and class names.

2 Load *phyloseq* and import data

2.1 Load *phyloseq*

To use *phyloseq* in a new R session, it will have to be loaded. This can be done in your package manager, or at the command line using the `library()` command:

```
> library("phyloseq")
```

2.2 Import data

An important feature of *phyloseq* are methods for importing phylogenetic sequencing data from common taxonomic clustering pipelines. These methods take file pathnames as input, read and parse those files, and return a single object that contains all of the data.

2.3 Import from QIIME

QIIME is a free, open-source OTU clustering and analysis pipeline written for Unix (mostly Linux) [Caporaso et al.(2010)Capora It is distributed in a number of different forms (including a pre-installed virtual machine), and relevant links for obtaining and using QIIME should be found at:

    http://qiime.org/

2.3.1 Input

One QIIME input file (sample map), and two QIIME output files (`"otutable.txt"`, `".tre"`) are recognized by the `import_qiime()` function. Only one of the three input files is required to run, although an `"otutable.txt"` file is required if `import_qiime()` is to return a complete experiment object.

In practice, you will have to find the relevant QIIME files among a number of other files created by the QIIME pipeline. A screenshot of the directory structure created during a typical QIIME run is shown in Figure 1.

2.3.2 Output

The class of the object returned by `import_qiime()` depends upon which filenames are provided. The most comprehensive class is chosen automatically, based on the input files listed as arguments. At least one argument needs to be provided.

2.3.3 Example

The following lines of code would create a `phyloseqTaxTree` object (see Appendix A for class definitions) from files on your computer, had they been created by the QIIME pipeline.

```
> otufilename <- "../data/ex1_otutable.txt"
> mapfilename <- "../data/ex1_samplemap.txt"
> trefilename <- "../data/ex1_tree.tre"
> MyExpmt1 <- import_qiime(otufilename, mapfilename, trefilename)
```

A separate object containing taxonomic data is not necessary, because this information is included in the "otu_table.txt" file, and parsed into a proper `taxonomyTable` component object by `import_qiime()`.

⊟ 📁 wf_da	2 items
⊟ 📁 uclust_picked_otus	4 items
⊟ 📁 rep_set	4 items
⊟ 📁 pynast_aligned_seqs	5 items
⊟ 📁 fasttree_phylogeny	2 items
📄 seqs_rep_set.tre	54.7 KB
📄 seqs_rep_set_phylogeny.log	173 bytes
📄 seqs_rep_set_aligned.fasta	15.1 MB
📄 seqs_rep_set_aligned_pfiltered.fasta	1.5 MB
📄 seqs_rep_set_failures.fasta	10.3 KB
📄 seqs_rep_set_log.txt	81.5 KB
⊟ 📁 rdp_assigned_taxonomy	3 items
⊟ 📁 otu_table	1 item
📄 seqs_otu_table.txt	301.5 KB
📄 seqs_rep_set_tax_assignments.log	401 bytes
📄 seqs_rep_set_tax_assignments.txt	187.0 KB
📄 seqs_rep_set.fasta	545.2 KB
📄 seqs_rep_set.log	257 bytes
📄 seqs_clusters.uc	55.0 MB
📄 seqs_otus.log	399 bytes
📄 seqs_otus.txt	4.1 MB
📄 log_20110823081355.txt	4.0 KB

Figure 1: **A** typical QIIME output directory. The two output files suitable for import by *phyloseq* are highlighted. A third file describing the samples, their barcodes and covariates, is created by the user and required as *input* to QIIME. It is a good idea to import this file, as it can be converted directly to a `sampleMap` object and can be extremely useful for certain analyses.

2.4 Import from mothur

The open-source, platform-independent, locally-installed software package, "mothur", can also process barcoded amplicon sequences and perform OTU-clustering [Schloss et al.(2009)Schloss, Westcott, Ryabin, Hall, Hartmann, Holliste
It is extensively documented on a wiki at the following URL:

    http://www.mothur.org/wiki/

2.4.1 Input

Currently, there are three different files produced by the *mothur* package (Ver 1.22.0) that can be imported by *phyloseq*. At minimum, a user must supply a `".list"` file, and at least one of the following two files: `".groups"` or `".tree"`

The group file is produced by *mothur*'s `"make.group()"` function. Details on its use can be found at:

    http://www.mothur.org/wiki/Make.group

The tree file is a phylogenetic tree calculated by *mothur*.

2.4.2 Output

The output from `import_mothur()` depends on which file types are provided. If all three file types are provided, an `"otuTree"` object is returned (which contains both an OTU abundance table and its associated phylogenetic tree).

2.4.3 Example

The path on your machine may (probably will) vary, but here is an example import with all three files:

```
> mothur_list_file <- "~/Downloads/mothur/Esophagus/esophagus.an.list"
> mothur_group_file <- "~/Downloads/mothur/Esophagus/esophagus.good.groups"
> mothur_tree_file <- "~/Downloads/mothur/Esophagus/esophagus.tree"
> show_mothur_list_cutoffs(mothur_list_file)
> MyExpmt1 <- import_mothur(mothur_list_file, mothur_group_file,
+     mothur_tree_file)
```

2.5 Import from PyroTagger

PyroTagger is an OTU-clustering pipeline for barcoded 16S rRNA amplicon sequences, served and maintained by the Department of Energy's (DOE's) Joint Genome Institute (JGI). It can be used through a straightforward web interface at:

http://pyrotagger.jgi-psf.org/

PyroTagger takes as input the untrimmed sequence (`.fasta`) and sequence-quality (`.qual`) files, as well as a sample mapping file that contains the bar code sequence for each sample and its name. It uses a 97% identity threshold for defining OTU clusters, and provides no options for specifying otherwise. It does allow users to modify the threshold setting for low-quality bases.

2.5.1 Input

PyroTagger returns a single excel spreadsheet file (`.xls`) containing both abundance and taxonomy data, as well as some associated confidence information related to each taxonomic assignment. This spreadsheet also reports on potential chimeric sequences.

This single output file is sufficient for `import_RDP_tab()`, provided the file has been converted to a tab-delimited plain-text format. Any spreadsheet application should suffice. No other changes should be made to the `.xls` file.

2.5.2 Output

`import_RDP_tab()` returns an object from the `"otuTax"` class, containing the OTU abundance table and taxonomy table. To my knowledge, PyroTagger does not calculate a tree of the representative sequences from each OTU cluster, nor a distance object, so analyses like `tipglom()`.

2.5.3 Example

Here is an example, importing a PyroTagger file:

```
> pyrotagger_tab_file <- "path/to/my/filename.txt"
> myData1 <- import_pyrotagger_tab(pyrotagger_tab_file, strict_taxonomy = FALSE,
+     keep_potential_chimeras = FALSE)
```

For completeness, the optional arguments were shown with their default values selected. If you do not need to modify the optional arguments, the import command simplifies to:

```
> myData1 <- import_pyrotagger_tab(pyrotagger_tab_file)
```

2.6 Import from RDP pipeline

The Ribosomal Database Project (RDP [Cole et al.(2009)Cole, Wang, Cardenas, Fish, Chai, Farris, Kulam-Syed-Mohideen, Mc
`http://rdp.cme.msu.edu/`) provides a web-based barcoded 16S rRNA amplicon sequence processing pipeline
called the "RDP Pyrosequencing Pipeline" (`http://pyro.cme.msu.edu/`). A user must run all three of the
"Data Processing" steps sequentially through the web interface in order to acquire the output from Complete
Linkage Clustering, the approach to OTU clustering used by the RDP Pipeline. Note that this import
function assumes that the sequence names in the resulting cluster file follow a particular naming convention
with underscore delimiter. (See Section 2.6.3, below.)

2.6.1 Input

The output from the Complete Linkage Clustering, `".clust"`, is the only input to the RDP pipeline importer:

```
> myOTU1 <- import_RDP_cluster("path/to/my/filename.clust")
```

2.6.2 Output

This importer returns an `otuTable` object.

2.6.3 Expected Naming Convention

The RDP cluster pipeline (specifically, the output of the complete linkage clustering step) has no formal
documentation for the ".clust" file structure or its apparent sequence naming convention.

The cluster file itself contains the names of all sequences contained in the input alignment. If the upstream
barcode and aligment processing steps are also done with the RDP pipeline, then the sequence names follow
a predictable naming convention wherein each sequence is named by its sample and sequence ID, separated
by a "_" as delimiter:

"sampleName_sequenceIDnumber"

This import function assumes that the sequence names in the cluster file follow this convention, and that
the sample name does not contain any "_". It is unlikely to work if this is not the case. It is likely to work
if you used the upstream steps in the RDP pipeline to process your raw (barcoded, untrimmed) fasta/fastq
data.

2.7 Example Data (included)

An example data set is included in *phyloseq*, called "ex1". It is derived from a preliminary investigation of human intestinal microbiome. The user need only invoke the `data()` command to bring this object into the environment:

```
> data(ex1)
```

2.8 phyloseq object summaries

In small font, the following is the summary of object `ex1` that prints to the terminal. These summaries are consistent among all object classes defined by *phyloseq*. Although the components of `ex1` have many thousands of elements, the command-line returns only a short summary of each by default. This encourages you to check that an object is still what you expect, without needing to let thousands of elements scroll across the terminal. In the cases in which you do want to see all elements of an object, use `print()`.

```
> ex1

otuSamTaxTree Object

Sample Map [21 by 2]:
Samples: sa1, sa2 ... sa20, sa21
Variables: Diet Gender
    Diet Gender
sa1   0       B
sa2   0       A
sa3   0       B
...

OTU Table [7077 by 21]:
Species: otuID_3, otuID_4 ... otuID_9997, otuID_10000
Samples: sa1, sa2 ... sa20, sa21
        sa1 sa2 sa3 sa4
otuID_3   0  16   0   1
otuID_4   0   0   0   0
otuID_5   2   0   0   1
...

Taxonomy Table [7077 by 9]:
Species: otuID_3, otuID_4 ... otuID_9997, otuID_10000
Taxonomic Level: Root, Domain ... Species, Strain
        Root    Domain     Phylum           Class
otuID_3 "Root" "Bacteria" "Actinobacteria" "Actinobacteria"
otuID_4 "Root" "Bacteria" "Firmicutes"     "Clostridia"
otuID_5 "Root" "Bacteria" "Firmicutes"     "Clostridia"
...

<<< tree >>>
"phylo4"-class phylogenetic tree with
7077 tips, and 7070 internal nodes.
Tips: otuID_4769 otuID_8891 otuID_4100 ...
Unrooted.
<<< tree >>>
```

2.9　Convert raw data to phyloseq components

Suppose you have already imported raw data from an experiment into R, and their indices are labeled correctly. How do you get *phyloseq* to recognize these tables as the appropriate class of data? And further combine them together? Table 1 lists key functions for converting these core data formats into specific component data objects recognized by *phyloseq*. These will also

Functions for building component data objects

Function	Input Class	Output Description
`otuTable`	numeric matrix	`otuTable` object storing taxa abundance
`otuTable`	data.frame	`otuTable` object storing taxa abundance
`sampleMap`	data.frame	`sampleMap` object storing sample variables
`taxTab`	character string	`taxonomyTable` object storing taxonomic identities
`tre`	file path char	phylo4-class tree, read from file
`tre`	phylo-class tree	phylo4-class tree, converted from argument
`read.table`	table file path	A matrix or data.frame (Std Rcore function)
`read.tree`	Newick file path	phylo-class tree object (ape)
`read.nexus`	Nexus file path	phylo-class tree object (ape)
`readNexus`	Nexus file path	phylo4-class tree object (phylobase)

Functions for building complex data objects

Function	Input Class	Output Description
`phyloseq`	2 or more component objects	Complete experiment object
`merge_phyloseq`	2 or more component/complex objects	Complete experiment object

Table 1: Constructors: functions for building *phyloseq* objects.

The following example illustrates using the constructor methods for component data tables.

```
> otu1 <- otuTable(raw_abundance_matrix, speciesAreRows = FALSE)
> sam1 <- sampleMap(raw_sample_data.frame)
> tax1 <- taxTab(raw_taxonomy_matrix)
> tre1 <- read.nexus(my_nexus_file)
```

2.10　`phyloseq()` function: building complex phyloseq objects

Once you've converted the data tables to their appropriate class, combining them into one object requires only one additional function call, `phyloseq()`:

```
> ex1b <- phyloseq(my_otuTable, my_sampleMap, my_taxonomyTable,
+     my_tree)
```

You do not need to have all four data types in the example above in order to combine them. The `phyloseq()` method will create an object of whichever class is appropriate, according to the data you provide. Downstream analysis methods will know which class and which data types are required, and throw a warning if something is missing. For most downstream methods you will only need to supply the combined object returned by `phyloseq()` (usually as the first argument) and appropriate options.

```
> ex1c <- phyloseq(my_otuTable, my_sampleMap)
```

We refer to these classes that contain more than one component data type as *higher-order* classes. Whenever an instance of these classes is created by *phyloseq* — for example, when we use the `import_qiime()`

function to import data, or combine manually imported tables using `phyloseq()` — the row and column indices representing taxa or samples are internally checked/trimmed for compatibility, such that all component data describe exactly the same species and samples.

2.11 `merge_phyloseq()` function: merge multiple phyloseq objects

What if you have multiple objects describing parts of the same experimental project (say, because they came from different files)? What if you had already built a combined object for the earlier trials with the `phyloseq()` function, but now want to add additional data tables to that new object?

For all of these merging situations, the suggested function is `merge_phyloseq()`.

3 Accessor functions

Once you have a phyloseq object available, many accessor functions are available to query aspects of the data set. The function name and its purpose are summarized in Table 2.

Function	Description
`[`	Standard extraction operator. works on otuTable, sampleMap, and taxonomyTable
`access`	General slot accessor function for phyloseq-package
`getslots.phyloseq`	Return the slot names of phyloseq objects
`getSpecies`	Returns the abundance values of sample 'i' for all species in 'x'
`getSamples`	Returns the abundance values of species 'i' for all samples in 'x'
`nsamples`	Get the number of samples described by an object
`nspecies`	Get the number of species (taxa) described by an object
`otuSam`	Subset just the otuSam portion of a H.O. object
`otuTree`	Subset just the otuTree portion of a H.O. object
`otuTax`	Subset just the otuTax portion of a H.O. object
`otuSamTax`	Subset just the otuSamTax portion of a H.O. object
`otuSamTree`	Subset just the otuSamTree portion of a H.O. object
`otuSamTaxTree`	Subset just the otuSamTaxTree portion of a H.O. object
`otuTable`	Build or access otuTable objects
`sampleMap`	Build or access sampleMap objects
`taxTab`	Build or access taxTab objects
`tre`	Access the tree contained in a phyloseq object
`sample.names`	Return the names of the samples described by an object
`species.names`	Return the names of the species described by an object
`sampleSums`	Returns the total number of individuals observed from each sample
`speciesSums`	Returns the total number of individuals observed from each species
`speciesAreRows`	returns the orientation of the abundance table

Table 2: Accessor functions for *phyloseq* objects.

4 Trimming, subsetting, filtering phyloseq data

4.1 Trimming: `prune_species()` and `prune_samples()`

Trimming high-throughput phylogenetic sequencing data can be useful, or even necessary, for certain types of analyses. However, it is important that the original data always be available for reference and reproducibility; and that the methods used for trimming be transparent to others, so they can perform the same trimming or filtering steps on the same or related data.

To facilitate this, *phyloseq* contains many ways to trim/filter the data from a phylogenetic sequencing project. Because matching indices for taxa and samples is strictly enforced, subsetting one of the data components automatically subsets the corresponding indices from the others. Variables holding trimmed versions of your original data can be declared, and further trimmed, without losing track of the original data.

In general, most trimming should be accomplished using the S4 methods `prune_species()` or `prune_samples()`.

4.2 Simple filtering example

For example, lets make a new object that only holds the most abundant 20 taxa in the experiment. To accomplish this, we will use the `prune_species()` function.

```
> data(ex1)
> most_abundant_taxa <- sort(speciesSums(ex1), TRUE)[1:topN]
> ex2 <- prune_species(names(most_abundant_taxa), ex1)
```

An alternative, replacement-style approach is also supported by *phyloseq*:

```
> species.names(ex2) <- names(most_abundant_taxa)
```

Now we can ask the question, "what taxonomic Family are these OTUs?" (Subsetting still returns a `taxonomyTable` object, which is summarized. We will need to convert to a vector)

```
> topFamilies <- taxTab(ex2)[, "Family"]
> as(topFamilies, "vector")

 [1] "Bacteroidaceae"  "Bacteroidaceae"  "Bacteroidaceae"  "Lachnospiraceae"
 [5] "Ruminococcaceae" NA                "Bacteroidaceae"  "Ruminococcaceae"
 [9] "Bacteroidaceae"  "Prevotellaceae"  "Ruminococcaceae" "Prevotellaceae"
[13] "Bacteroidaceae"  "Lachnospiraceae" "Lachnospiraceae" "Ruminococcaceae"
[17] "Lachnospiraceae" "Rikenellaceae"   "Bacteroidaceae"  "Prevotellaceae"
```

4.3 Arbitrarily complex abundance filtering

The previous example was a relatively simple filtering in which we kept only the most abundant 20 in the whole experiment. But what if we wanted to keep the most abundant 20 taxa of each sample? And of those, keep only the taxa that are also found in at least one-third of our samples?

For this more complicated filtering *phyloseq* contains a function, `genefilterSample()`, that takes as an argument a *phyloseq* object, as well as a list of one or more filtering functions that will be applied to each sample in the abundance matrix (`otuTable`), as well as an integer argument, `A`, that specifies for how many samples the filtering function must return `TRUE` for a particular taxa to avoid removal from the object. A supporting function `filterfunSample` is also included in *phyloseq* to facilitate creating a properly formatted function (enclosure) if more than one function is going to be applied simultaneously. `genefilterSample` returns a logical vector suitable for sending directly to prune_species() for the actual trimming.

Here is an example on a completely fabricated `otuTable` called `testOTU`.

```
> testOTU <- otuTable(matrix(sample(1:50, 25, replace = TRUE),
+     5, 5), speciesAreRows = FALSE)
```

```
> f1 <- filterfunSample(topk(2))
> wh1 <- genefilterSample(testOTU, f1, A = 2)
> wh2 <- c(T, T, T, F, F)
> prune_species(wh1, testOTU)
> prune_species(wh2, testOTU)
```

Here is a second example using the included dataset, ex1. The most abundant taxa are kept only if they are in the most abundant 10% of taxa in at least half of the samples in dataset ex1. Note that it is not necessary to subset ex1 in order to do this filtering. The S4 method prune_species() subsets each of the relavent component objects, and returns the complex object back.

```
> data(ex1)
> f1 <- filterfunSample(topp(0.1))
> wh1 <- genefilterSample(ex1, f1, A = (1/2 * nsamples(ex1)))
> sum(wh1)

[1] 734

> ex2 <- prune_species(wh1, ex1)

> print(ex2)

otuSamTaxTree Object

Sample Map [21 by 2]:
Samples: sa1, sa2 ... sa20, sa21
Variables: Diet Gender
    Diet Gender
sa1   0      B
sa2   0      A
sa3   0      B
...

OTU Table [734 by 21]:
Species: otuID_3, otuID_22 ... otuID_9972, otuID_9985
Samples: sa1, sa2 ... sa20, sa21
        sa1 sa2 sa3 sa4
otuID_3   0  16   0   1
otuID_22  7   1   0   0
otuID_39  3   8   0   1
...

Taxonomy Table [734 by 9]:
Species: otuID_3, otuID_22 ... otuID_9972, otuID_9985
Taxonomic Level: Root, Domain ... Species, Strain
         Root    Domain     Phylum           Class
otuID_3  "Root" "Bacteria" "Actinobacteria" "Actinobacteria"
otuID_22 "Root" "Bacteria" "Bacteroidetes"  "Bacteroidetes"
otuID_39 "Root" "Bacteria" "Firmicutes"     "Clostridia"
...

<<< tree >>>
"phylo4"-class phylogenetic tree with
734 tips, and 732 internal nodes.
Tips: otuID_1118 otuID_8946 otuID_9507 ...
Unrooted.
<<< tree >>>
```

If instead of the most abundant fraction of taxa, you are interested in the most abundant fraction of individuals (aka sequences, observations), then the topf function is appropriate. For steep rank-abundance curves, topf will seem to be much more conservative (trim more taxa) because it is based on the cumulative sum of relative abundance. It does not guarantee that a certain number or fraction of total taxa (richness) will be retained.

```
> data(ex1)
> f1 <- filterfunSample(topf(0.9))
```

```
> wh1 <- genefilterSample(ex1, f1, A = (1/3 * nsamples(ex1)))
> sum(wh1)
> prune_species(wh1, ex1)
```

4.4 subset_samples(): subset by sample variates

It is possible to subset the samples in a *phyloseq* object based on the sample variables using the **sub-set_samples()** function. For example to subset **ex1** such that only *Gender A* is present, the following line is needed (the related tables are subsetted automatically as well):

```
> ex3 <- subset_samples(ex1, Gender == "A")
```

```
> ex3

otuSamTaxTree Object

Sample Map [11 by 2]:
Samples: sa2, sa4 ... sa17, sa21
Variables: Diet Gender
    Diet Gender
sa2    0      A
sa4    1      A
sa5    1      A
...

OTU Table [7077 by 11]:
Species: otuID_3, otuID_4 ... otuID_9997, otuID_10000
Samples: sa2, sa4 ... sa17, sa21
         sa2 sa4 sa5 sa6
otuID_3  16   1   0   0
otuID_4   0   0   0   0
otuID_5   0   1   0   0
...

Taxonomy Table [7077 by 9]:
Species: otuID_3, otuID_4 ... otuID_9997, otuID_10000
Taxonomic Level: Root, Domain ... Species, Strain
        Root    Domain     Phylum           Class
otuID_3 "Root" "Bacteria" "Actinobacteria" "Actinobacteria"
otuID_4 "Root" "Bacteria" "Firmicutes"     "Clostridia"
otuID_5 "Root" "Bacteria" "Firmicutes"     "Clostridia"
...

<<< tree >>>
"phylo4"-class phylogenetic tree with
7077 tips, and 7070 internal nodes.
Tips: otuID_4769 otuID_8891 otuID_4100 ...
Unrooted.
<<< tree >>>
```

For this example only a categorical variable is shown, but in principle a continuous variable could be specified and a logical expression provided just as for the **subset** function. In fact, because **sampleMap** component objects are an extension of the data.frame class, they can also be subsetted with the **subset** function:

```
> subset(sampleMap(ex1), Gender == "A")

     Diet Gender
sa2    0      A
sa4    1      A
sa5    1      A
sa6    1      A
sa7    0      A
sa10   1      A
sa11   0      A
sa12   0      A
```

```
sa14    1      A
sa17    1      A
sa21    0      A
```

4.5 subset_species(): subset by taxonomic categories

It is possible to subset by specific taxonomic category using the subset_species() function. For example, if we wanted to subset ex1 so that it only contains data regarding the phylum *Firmicutes*:

```
> ex4 <- subset_species(ex1, Phylum == "Firmicutes")

> ex4

otuSamTaxTree Object

Sample Map [21 by 2]:
Samples: sa1, sa2 ... sa20, sa21
Variables: Diet Gender
    Diet Gender
sa1    0      B
sa2    0      A
sa3    0      B
...

OTU Table [4960 by 21]:
Species: otuID_4, otuID_5 ... otuID_9997, otuID_10000
Samples: sa1, sa2 ... sa20, sa21
        sa1 sa2 sa3 sa4
otuID_4   0   0   0   0
otuID_5   2   0   0   1
otuID_6   0   0   0   0
...

Taxonomy Table [4960 by 9]:
Species: otuID_4, otuID_5 ... otuID_9997, otuID_10000
Taxonomic Level: Root, Domain ... Species, Strain
        Root    Domain    Phylum       Class
otuID_4 "Root" "Bacteria" "Firmicutes" "Clostridia"
otuID_5 "Root" "Bacteria" "Firmicutes" "Clostridia"
otuID_6 "Root" "Bacteria" "Firmicutes" "Clostridia"
...

<<< tree >>>
"phylo4"-class phylogenetic tree with
4960 tips, and 4954 internal nodes.
Tips: otuID_4769 otuID_8891 otuID_4100 ...
Unrooted.
<<< tree >>>
```

4.6 random subsample abundance data

Can also randomly subset, for example a random subset of 100 taxa from the full dataset.

```
> randomSpecies100 <- sample(species.names(ex1), 100, replace = FALSE)
> ex5 <- prune_species(randomSpecies100, ex1)
```

5 Transform abundance data

Sample-wise transformation can be achieved with the `transformsamplecounts()` function. It requires two arguments, (1) the *phyloseq* object that you want to transform, and the function that you want to use to perform the transformation. Any arbitrary function can be provided as the second argument, as long as it returns a numeric vector with the same length as its input. In the following trivial example, we create a second object, `ex2`, that has been "transformed" by the identity function such that it is actually identical to `ex1`.

```
> data(ex1)
> ex2 <- transformsamplecounts(ex1, I)
```

For certain kinds of analyis we may want to transform the abundance data. For example, for RDA we want to transform abundance counts to within-sample ranks, and to further include a threshold beyond which all taxa receive the same rank value. The ranking for each sample is performed independently, so that the rank of a particular taxa within a particular sample is not influenced by that sample's total quantity of sequencing relative to the other samples in the project.

The following example shows how to perform such a thresholded-rank transformation of the abundance table in the complex *phyloseq* object `ex1` with an arbitrary threshold of 500.

```
> ex4 <- transformsamplecounts(ex1, threshrankfun(500))
```

6   Phylogenetic smoothing

6.1   `taxglom()` method

Suppose we are skeptical about the importance of species-level distinctions in our dataset. For this scenario, *phyloseq* includes a taxonomic-agglommeration method, `taxglom()`, which merges taxa of the same taxonomic category for a user-specified taxonomic level. In the following code, we merge all taxa of the same Genus, and store that new object as `ex6`.

```
> ex6 <- taxglom(ex1, taxlevel = "Genus")
```

6.2   `tipglom()` method

Similarly, our original example object (`ex1`) also contains a phlyogenetic tree corresponding to each OTU, which we could also use as a means to merge taxa in our dataset that are closely related. In this case, we specify a threshold patristic distance. Taxa more closely related than this threshold are merged. This is especially useful when a dataset has many taxa that lack a taxonomic assignment at the level you want to investigate, a problem when using `taxglom()`. Note that for datasets with a large number of taxa, `taxglom` will be noticeably faster than `tipglom`. Also, keep in mind that `tipglom` requires that its first argument be an object that contains a tree, while `taxglom` instead requires a `taxonomyTable` (See Appendix A).

```
> ex7 <- tipglom(ex1, speciationMinLength = 0.05)
```

Command output not provided here to save time during compilation of the vignette. The user is encouraged to try this out on your dataset, or even this example, if interested. It may take a while to run on the full, untrimmed data.

A *phyloseq* classes

The class structure in the *phyloseq* package follows the inheritance diagram shown in Fig. 2. The *phyloseq* package contains multiple inherited classes with incremental complexity so that methods can be extended to handle exactly the data types that are present in a particular object. Currently, *phyloseq* uses 4 core data classes. They are the taxonomic abundance table (`otuTable`), a table of sample data (`sampleMap`), a table of taxonomic descriptors (`taxonomyTable`), and a phylogenetic tree (`phylo4`, *phylobase* package). The `otuTable` class can be considered the central data type, as it directly represents the number and type of sequences observed in each sample. `otuTable` extends the numeric matrix class in the R base, and has a few additonal feature slots. The most important of these feature slots is the `speciesAreRows` slot, which holds a single logical that indicates whether the table is oriented with taxa as rows (as in the *genefilter* package in Bioconductor [Gentleman et al.(2004)Gentleman, Carey, Bates, and others]) or with taxa as columns (as in *vegan* and *picante* packages). In *phyloseq* methods, as well as its extensions of methods in other packages, the `speciesAreRows` value is checked to ensure proper orientation of the otuTable. A *phyloseq* user is only required to specify the `otuTable` orientation during initialization, following which all handling is internal.

The `sampleMap` class directly inherits R's `data.frame` class, and thus effectively stores both categorical and numerical data about each sample. The orientation of a `data.frame` in this context requires that samples/trials are rows, and variables are columns (consistent with *vegan* and other packages). The `taxonomyTable` class directly inherits the `matrix` class, and is oriented such that rows are taxa (e.g. *species*) and columns are taxonomic levels (e.g. *Phylum*).

We use the term "higher-order classes" for those that contain two or more of the previously-described core data classes. We assume that *phyloseq* users will be interested in analyses that utilize their abundance counts derived from the phylogenetic sequencing data, and so all higher-order classes contain an `otuTable` slot. There are a number of common methods that require either an `otuTable` and `sampleMap` combination, or an `otuTable` and phylogenetic tree combination. These methods can operate on instances of the `otuSam` or `otuTree` classes, respectively, or their children. In addition, a virtual class has been defined, `phyloseqFather`, that is inherited by all other higher-order classes. In many cases a method will only need to be defined for one of these classes in order to work properly for other relevant classes as well.
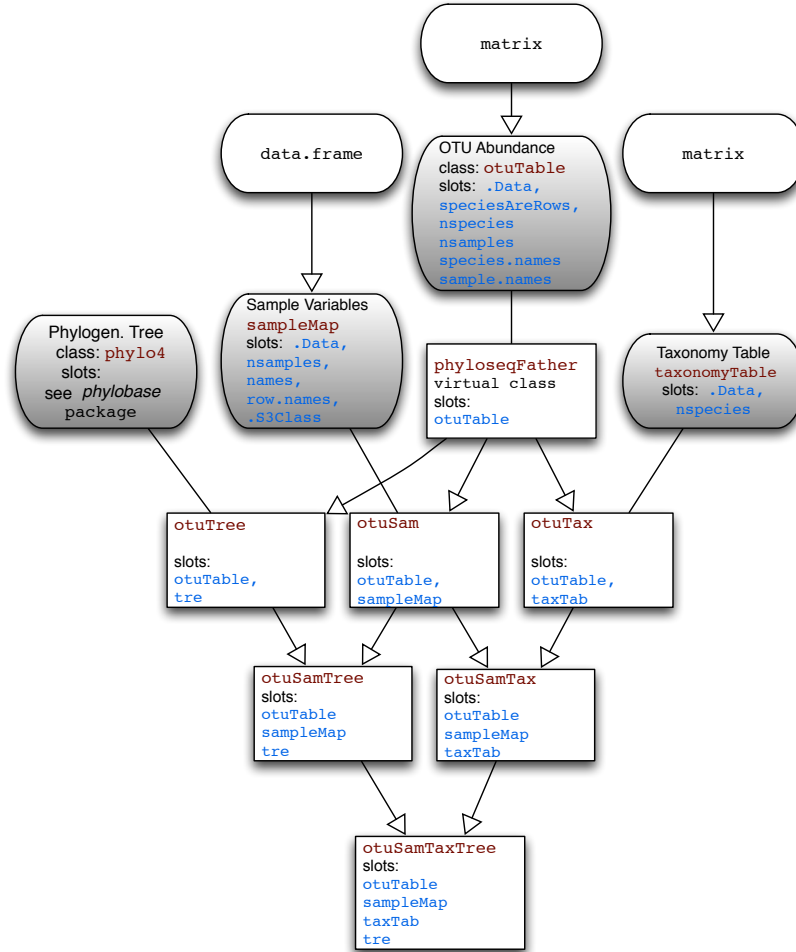
Figure 2: **C**lasses and inheritance in the *phyloseq* package. Core data classes are shown with grey fill and rounded corners. The class name and its slots are shown with red- or blue-shaded text, respectively. Inheritance is indicated graphically by arrows. Lines without arrows indicate that a higher-order object contains a slot with the associated class as one of its components.

B Install Development Version

For development version of the *phyloseq* package when it is not yet available from Bioconductor, see the development homepage on GitHub (`https://github.com/joey711/phyloseq`). This is also the best place to post issues, bug reports, feature requests, etc. The most convenient way to install this and other R packages available on GitHub is to first make sure you have the latest R development tools installed (*devtools* package), and then using the special github variant of the `install()` command, `install_github()`. Step-by-step:

```
> install.packages("devtools")
> library("devtools")
> install_github("devtools")
> library("devtools")
> install("ape")
> install("ggplot2")
> install("igraph")
> source("http://bioconductor.org/biocLite.R")
> biocLite("multtest")
> install("phylobase")
> install("picante")
> install("vegan")
> install("foreach")
> install_github("phyloseq", "joey711")
```

Note that the above instructions include installation of a number of additional packages upon which *phyloseq* is dependent. This may take some time to install. If there is a portion of these initial instructions not working as listed, please notify me by e-mail or at the github page. For running parallel implementation of functions/methods in *phyloseq* (e.g. `UniFrac(ex1, parallel=TRUE)`), you will need also to install a function for registering a parallel "backend". Only one working parallel backend is needed, but there are several options, depending on the details of your particular system:

```
> install("doMC")
> install("doSNOW")
> install("doMPI")
```

C Bibliography

References

[Caporaso et al.(2010)Caporaso, Kuczynski, Stombaugh, Bittinger, Bushman, Costello, Fierer, Peña, Goodrich, Gordon, Hutt
J Gregory Caporaso, Justin Kuczynski, Jesse Stombaugh, Kyle Bittinger, Frederic D Bushman, Eliz-
abeth K Costello, Noah Fierer, Antonio Gonzalez Peña, Julia K Goodrich, Jeffrey I Gordon, Gavin A
Huttley, Scott T Kelley, Dan Knights, Jeremy E Koenig, Ruth E Ley, Catherine A Lozupone, Daniel
McDonald, Brian D Muegge, Meg Pirrung, Jens Reeder, Joel R Sevinsky, Peter J Turnbaugh, William A
Walters, Jeremy Widmann, Tanya Yatsunenko, Jesse Zaneveld, and Rob Knight. QIIME allows analysis
of high-throughput community sequencing data. *Nature Methods*, 7(5):335–336, 2010.

[Cole et al.(2009)Cole, Wang, Cardenas, Fish, Chai, Farris, Kulam-Syed-Mohideen, McGarrell, Marsh, Garrity, and Tiedje]
J R Cole, Q Wang, E Cardenas, J Fish, B Chai, R J Farris, A S Kulam-Syed-Mohideen, D M McGarrell,
T Marsh, G M Garrity, and J M Tiedje. The Ribosomal Database Project: improved alignments and
new tools for rRNA analysis. *Nucleic Acids Research*, 37(Database issue):D141–5, 2009.

[Gentleman et al.(2004)Gentleman, Carey, Bates, and others] Robert C Gentleman, Vincent J. Carey, Dou-
glas M. Bates, and others. Bioconductor: Open software development for computational biology and
bioinformatics. *Genome Biology*, 5:R80, 2004.

[Schloss et al.(2009)Schloss, Westcott, Ryabin, Hall, Hartmann, Hollister, Lesniewski, Oakley, Parks, Robinson, Sahl, Stres, T
P D Schloss, S L Westcott, T Ryabin, J R Hall, M Hartmann, E B Hollister, R A Lesniewski, B B
Oakley, D H Parks, C J Robinson, J W Sahl, B Stres, G G Thallinger, D J Van Horn, and C F
Weber. Introducing mothur: Open-Source, Platform-Independent, Community-Supported Software for
Describing and Comparing Microbial Communities. *Applied and Environmental Microbiology*, 75(23):
7537–7541, 2009.