





Bioinformatics for microbiome research Day 2: microbial community analysis

Jyväskylä Summer School 2023

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Outline of Day 1: Preprocessing reads



Quality check



Remove primers



Filter reads for quality and length



Remove identical sequences



Alignment



Filter and trim alignment



Precluster



Remove chimeras



Classify sequences (=assign taxonomy)

Output so far:

- 1. FASTA file of processed reads
- 2. count file (which read in which sample)
- 3. taxonomy file (taxonomy of each read)

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Day 2: Community analysis

- Clustering into OTUs
- Import into Phyloseq
- Data tidying & transformations
- Taxonomy plots
- Alpha diversity
- Beta diversity: ordinations & statistics





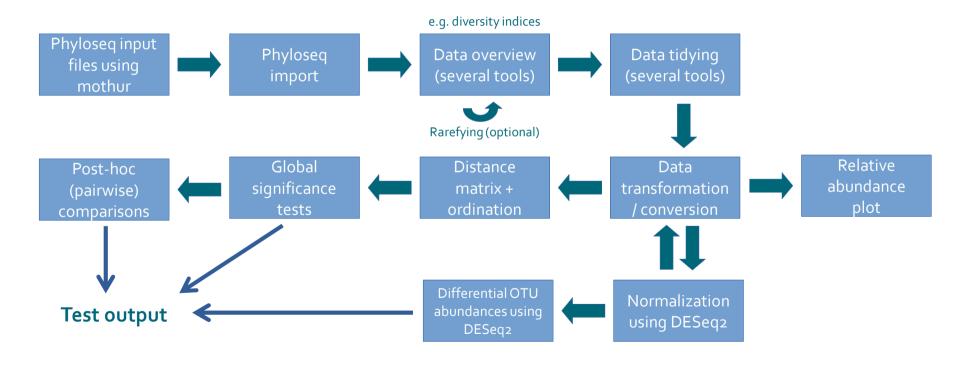


Part 1: Tool overview and data importing



Workflow for microbial community analysis in Chipster



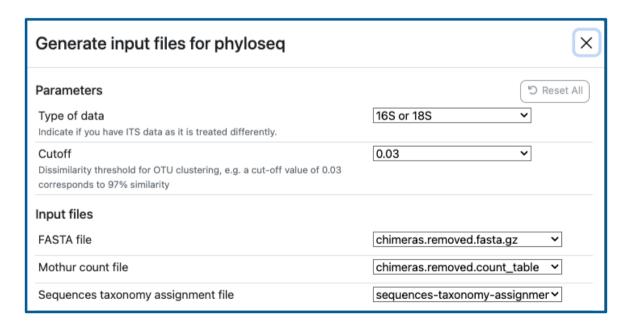




Generating input files for phyloseq

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

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

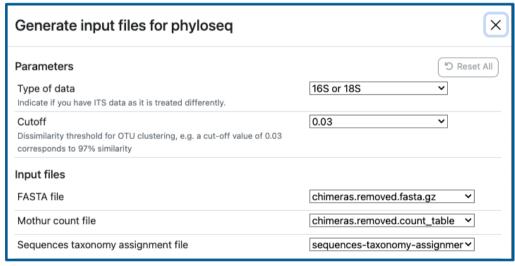


Generating input files for phyloseq



Specifications for creating phyloseq input files:

- type of data (16S/18S or ITS)
- % cut off for <u>OTU clustering</u>
- files produced by mothur
 - o final FASTA
 - o count file
 - taxonomy file = taxonomy assignment of each <u>read</u>







Generated input files:

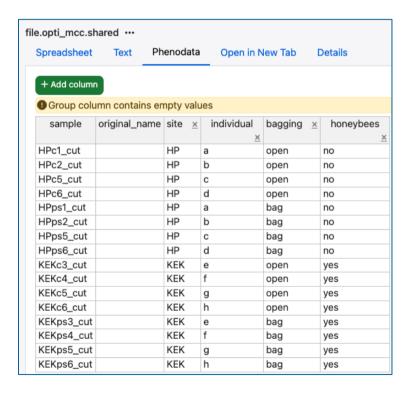
- .shared file (mothur file format)
 - o samples in rows, OTUs in columns
 - how many reads of each OTU in each sample (OTU table)
- consensus taxonomy file
 - o taxonomy assignments of <u>OTUs</u>
- phenodata file

Spread	dsheet Te	xt Pher	nodata	Open in N	lew Tab	Details				
Showing all 16 rows and the first 500 columns.										
label	Group	numOtus	Otu0001	Otu0002	Otu0003	Otu0004	Otu0005	Otu0006		
0.03	HPc1_cut	1437	10	2	152	126	10	314		
0.03	HPc2_cut	1437	2	0	1872	951	521	382		
0.03	HPc5_cut	1437	6	1	770	631	1	708		
0.03	HPc6_cut	1437	7	1	288	717	1	595		
0.03	HPps1_cut	1437	4	0	115	270	695	213		
0.03	HPps2_cut	1437	0	0	119	128	2712	104		
0.03	HPps5 cut	1437	3	0	234	657	1	442		

Spreadsheet		Text Open in New Tab Details					
Showing the	first 10	0 of 1437 rows and the first 500 columns. View in full screen to see all rows.					
OTU	Size						
Otu0001	17741	Bacteria(100);Proteobacteria(100);Gammaproteobacteria(100);Pseudomonadales(100);					
Otu0002	13712	Bacteria(100);Proteobacteria(100);Gammaproteobacteria(100);Xanthomonadales(100);					
Otu0003	8762	Bacteria (100); Proteobacteria (100); Gamma proteobacteria (100); Burkholderiales (100); Me					
Otu0004	8259	Bacteria (100); Proteobacteria (100); Gamma proteobacteria (100); Burkholderiales (100); Control of the control					
Otu0005	4458	Bacteria(100);Cyanobacteria(100);Cyanobacteriia(100);Chloroplast(100);Chloroplast_fa					
Otu0006	4162	Bacteria(100); Proteobacteria(100); Alphaproteobacteria(100); Rhodobacterales(100); Rhod					
Otu0007	3448	Bacteria(100);Proteobacteria(100);Gammaproteobacteria(100);Burkholderiales(100);Co					
Otu0008	2695	Bacteria (100); Proteobacteria (100); Alphaproteobacteria (100); Sphingomonadales (100); S					
Otu0009	2298	Bacteria (100); Proteobacteria (100); Gamma proteobacteria (100); Burkholderiales (100); Co					
Otu0010	1347	Bacteria(100); Proteobacteria(100); Gammaproteobacteria(100); Gammaproteobacteria_I					

Phenodata file: fill in sample information



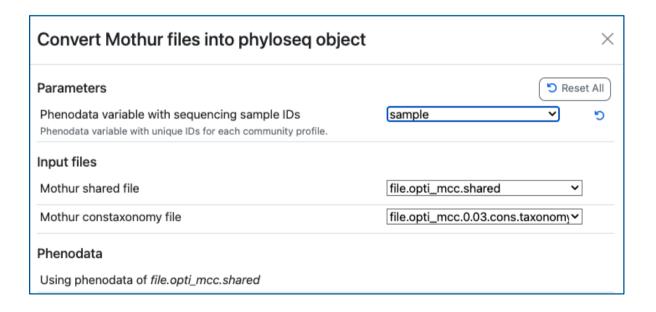


The phenodata file is an editable table with

- 1) unique IDs for each sample
- 2) information on sample groupings



Converting input files into a phyloseq object





Converting input files into a phyloseq object

```
### Imported phyloseg object ###
phyloseq-class experiment-level object
                                 [ 1437 taxa and 16 samples ]
otu table()
             OTU Table:
                                 [ 16 samples by 6 sample variables ]
sample data() Sample Data:
tax table() Taxonomy Table:
                                 [ 1437 taxa by 6 taxonomic ranks ]
### Sample names ###
 [1] "HPc1 cut"
                  "HPc2 cut"
                               "HPc5 cut"
                                            "HPc6 cut"
                                                         "HPps1 cut"
 [6] "HPps2 cut"
                  "HPps5_cut"
                               "HPps6 cut" "KEKc3 cut"
                                                         "KEKc4 cut"
[11] "KEKc5 cut"
                  "KEKc6_cut"
                               "KEKps3 cut" "KEKps4 cut" "KEKps5 cut"
[16] "KEKps6 cut"
### Sample variables ###
[1] "sample"
                    "original name" "site"
                                                    "individual"
[5] "badding"
                    "honevbees"
```

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

The Rda file is used as the input for downstream analyses

OTU table, taxonomy table and sample data can be exported for use outside Chipster (Microbial amplicon data analyses / Extract information from the Phyloseq object)







Part 2: Data tidying and alpha diversity

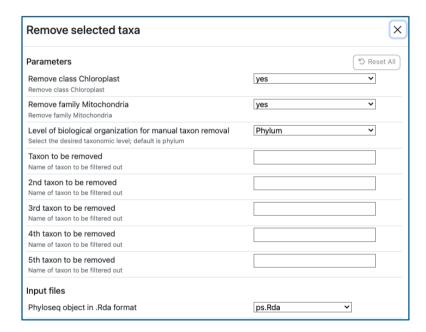




Taxon-level clean-up tools

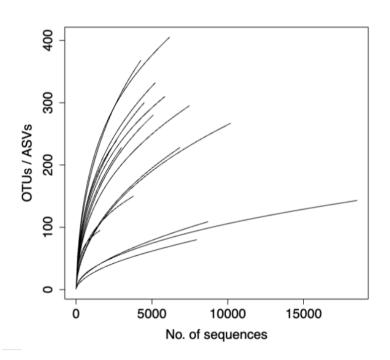
Under Microbial amplicon data analyses:

- Filter by taxonomic group
 - Remove non-specific sequences (keep e.g. Bacteria or Archaea only)
- Remove selected taxa
 - Remove chloroplast and/or mitochondrial sequences
 - (Manually remove specific taxa)
- Overview of taxon composition
 - user-specified level



Data inspection: Sequence numbers, rarefaction curve and alpha diversity estimates





```
### Per-sample sequence no.s ###
  HPc1 cut
             HPc2_cut
                        HPc5_cut
                                   HPc6_cut HPps1_cut HPps2_cut
      5084
                 7467
                            5859
                                       5218
                                                  2664
                                                              1585
                                                                         4495
                                  KEKc5_cut KEKc6_cut KEKps3_cut KEKps4_cut
 HPps6 cut KEKc3 cut KEKc4 cut
      6161
                10198
                            3000
                                       4277
                                                   8703
                                                              6851
                                                                         3789
KEKps5 cut KEKps6 cut
      7943
                18524
### Alpha diversity estimates (observed OTUs, Chao1, Shannon's index, Pielou's evenness) ###
           Observed
                       Chao1 se.chao1 Shannon
                                                  pielou
                                                              sample
HPc1 cut
                280 438.8095 40.27640 3.424902 0.6078137
                                                           HPc1 cut
                                                           HPc2 cut
HPc2 cut
                295 475.0244 44.97236 3.318920 0.5836002
HPc5_cut
                310 488.5789 45.77474 3.715068 0.6476111
                                                           HPc5_cut
HPc6 cut
                332 536.1395 49.12739 4.019842 0.6924632
                                                           HPc6 cut
HPps1 cut
                239 320,2222 23,13010 4,044690 0,7385587
                                                          HPps1 cut
HPps2 cut
                 95 118.0769 12.23199 3.474902 0.7630644
                                                          HPps2 cut
HPps5_cut
                300 473.3182 42.63040 3.798141 0.6658987
                                                          HPps5_cut
HPps6_cut
                405 669.0000 62.88645 4.397870 0.7325038
                                                          HPps6 cut
                267 462.5385 48.92997 2.625653 0.4699367 KEKc3 cut
KEKc3 cut
```



Filtering low-abundance OTUs

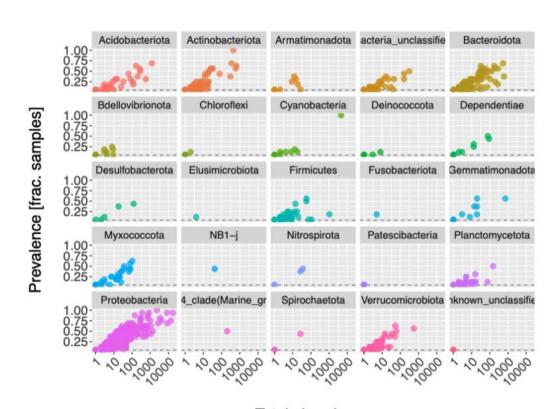
- Filter out OTUs that occur in less than x % of samples
 - Proportional prevalence filtering
- Remove singletons and doubletons
 - Remove OTUs with o-2 occurrences

Visualizing low-abundance OTUs



Additional prevalence summaries

- Visualization of OTU prevalence at phylum level
- Text summary on low prevalence OTUs



Total abundance



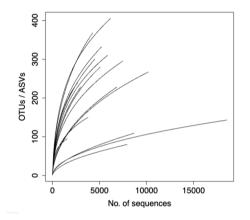
Alpha diversity

Diversity within a habitat: how many species are present, at which relative amounts?

- Species richness how many species?
 - o observed number of OTUs
 - richness estimators (Chao1)
- Evenness abundance distribution of species?
 - o all OTUs equally abundant vs. a few dominant and a lot of rare OTUs
 - o Pielou's evenness
- Many diversity indices **combine richness and evenness** (e.g. Shannon index)

CSC

Rarefaction?



Uneven sequence numbers among samples can bias comparisons, especially with alpha diversity. Solutions:

- rarefying: equal number of reads picked from all samples (Rarefy OTU data to even depth)
- data transformation (Transform OTU counts)

PLOS COMPUTATIONAL BIOLOGY OPEN ACCESS PERREVIEWED RESEARCH ARTICLE 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

Waste not, want not: Revisiting the analysis that called into question the practice of rarefaction

Patrick D. Schloss

doi: https://doi.org/10.1101/2023.06.23.546312

Rarefaction is currently the best approach to control for uneven sequencing effort in amplicon sequence analyses

Patrick D. Schloss

doi: https://doi.org/10.1101/2023.06.23.546313







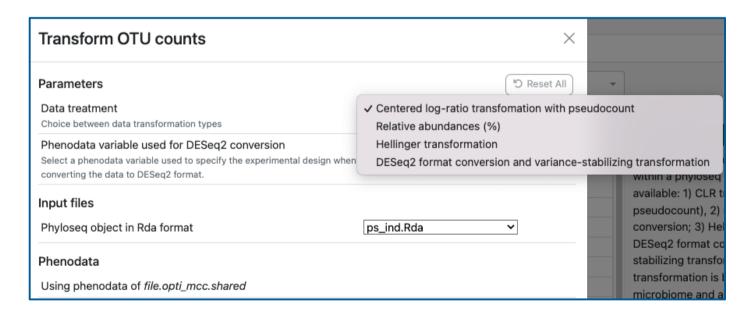
Part 3: Transformations and ordinations





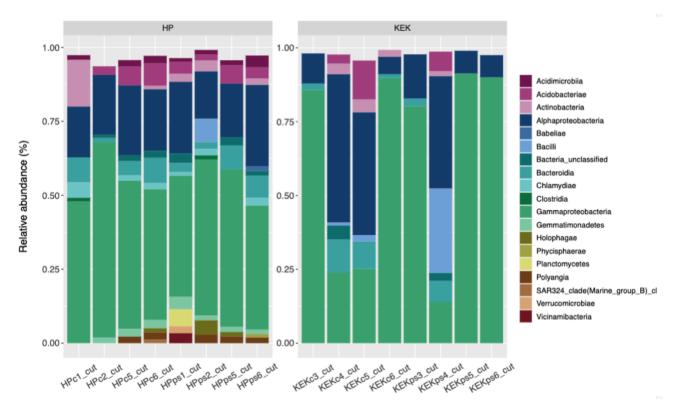
Transformation of OTU data

Four options (August 2023)





Relative abundance (%) bar plots





Beta diversity

Change in community composition among habitats

 Does the microbial community composition in treatment A differ from treatment B?

- Unlike in alpha diversity, OTU identity matters
- Both identity and relative abundance of OTUs usually included
- Quantified with distance or dissimilarity measures



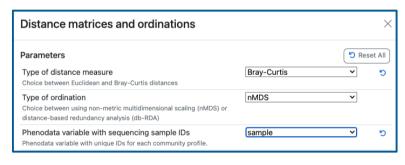
Distance matrices and ordinations

Distance measures available: Bray-Curtis or Euclidean

centered log ratio (CLR) transformation + Euclidean distance =
 Aitchinson distance

Ordinations: visualizing beta diversity

- o nMDS (non-metric multidimensional scaling)
 - o overall variation among samples displayed
- odb-RDA (distance-based redundancy analysis)
 - focus on the variation explained by phenodata variable(s)

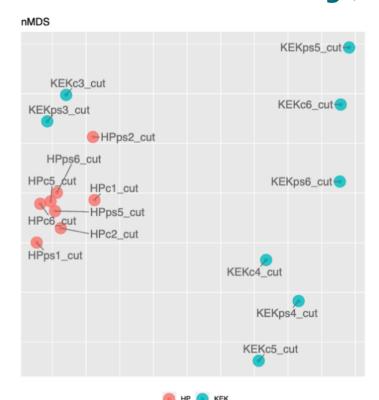


Recommended for more information:

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



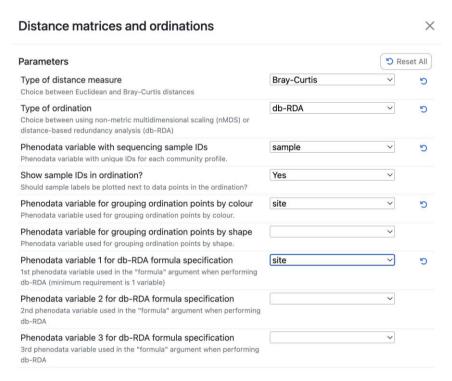
Non-metric multidimensional scaling (nMDS)



Distance-based redundancy analysis (db-RDA)



Requires specifying one or more phenodata variables

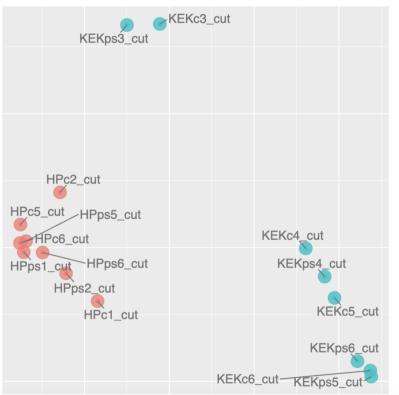




Distance-based redundancy analysis (db-RDA)







Constrained ordination = focus on the community variation explained by the phenodata variable(s)

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







Part 4: Statistics

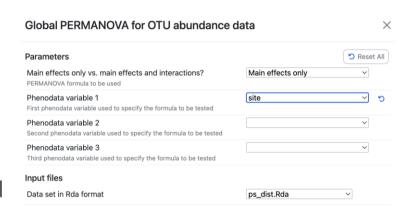


PERMANOVA (permutational multivariate analysis of variance)



Input: distance matrix (ps_dist.Rda)

- Global test: 'Does community structure differ between sample groups?'
 - Pairwise test: 'Which groups differ from one another?'
- Currently: several phenodata variables -> added sequentially -> order matters!
- Influenced by both location and dispersion

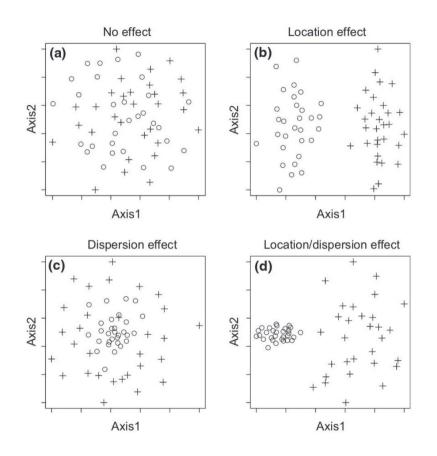




Location vs. dispersion

A significant PERMANOVA result can be due to:

- Location effect
- Dispersion effect
- Combination of both



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

PERMANOVA output



```
### Global PERMANOVA summary ###
$aov.tab
Permutation: free
Number of permutations: 999
Terms added sequentially (first to last)
           Df SumsOfSqs MeanSqs F.Model
                                             R2 Pr(>F)
get(pheno1)
                 1.5086 1.50863 6.0983 0.30342
                                                 0.001 ***
Residuals
                 3.4634 0.24738
                                        0.69658
Total
                 4.9720
                                        1.00000
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
$call
adonis(formula = ps_dist ~ get(pheno1), data = ps_df)
```

Df: Degrees of freedom

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

R2: Variation explained by the model

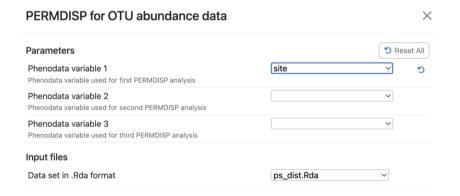
Pr(>F): Statistical significance (*p* value)

PERMDISP: test for the homogeneity of multivariate dispersion



Input: distance matrix

- Test if a significant PERMANOVA result is due to dispersion, not (only) location
- Significant result -> be careful with PERMANOVA interpretation





Post-hoc comparisons

Sample group comparisons following significant global PERMANOVA:

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

Dispersion comparisons following significant PERMDISP:

• Tukey's Honestly Significant Difference (HSD) test

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



DESeq2

- Originates from the RNAseq field
- Addresses 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'
- Input: untransformed data -> convert to DESeq2 format with
 Transform OTU counts (corrects for differences in sequencing depth)
- Results given as log fold changes

More information: https://joey711.github.io/phyloseq-extensions/DESeq2.html



DESeq2

Current tool configuration (August 2023):

Focused on comparison of two groups at a time

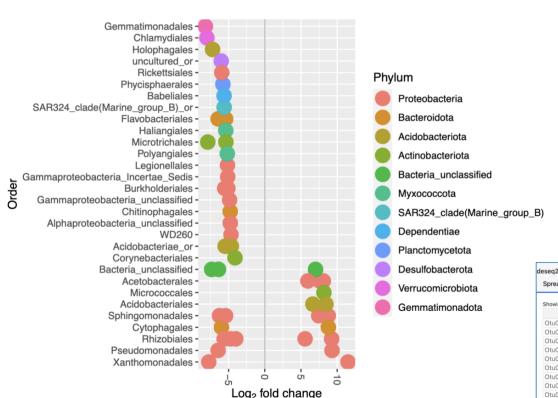
 If selected phenodata column has >2 groups, specify a pair (Group 1 and Group 2)

Reference level selection:

- Phenodata column with two groups -> first in alphabet is the reference level (e.g. 'b vs. a' or 'sick vs. healthy'=
- Phenodata column with >2 groups -> 'Group 2' is the reference level



DESeq2



8-fold increase compared to reference level = log_2 fold change 3 (because 2^3 =8)

each dot = OTU with adjusted p value < 0.01

Spreadsh	eet Text Volca	ano Plot Open in Ne	w Tab Details		
Showing all	50 rows.				[] Full Scree
	baseMean	log2FoldChange	IfcSE	padj	Phylum
Otu0001	1409.91917064308	9.28902619591325	1.18286988891193	1.2962293903996e-12	Proteobacteria
Otu0002	1105.19832748892	11.4960029127497	1.62958657068022	2.7618686719035e-10	Proteobacteria
Otu0018	38.895406507353	-8.21549247207248	1.20763958116883	1.09001829250313e-09	Gemmatimonadota
Otu0026	30.9633612180588	-7.88598154713512	1.19109114653925	2.84906353648758e-09	Actinobacteriota
Otu0008	203.100644341426	8.77278020719472	1.35913709266973	6.9194678347515e-09	Proteobacteria
Otu0032	33.7041539395581	-8.00673929826559	1.2531798742985	8.8694406626066e-09	Verrucomicrobiota
Otu0013	50.021527652905	9.23355358020544	1.6450425877987	9.06405910265283e-07	Proteobacteria
Otu0051	15.1741860288095	-7.22190000316191	1.30754637637176	1.32701482507549e-06	Acidobacteriota
Otu0043	16.5181434544254	-7.34421691851304	1.35671188338474	2.19372482320632e-06	Bacteria_unclassified
Otu0021	37.6825791339695	-6.26794763204544	1.266459089238	2.37727339785904e-05	Proteobacteria
04-0004	20.0040505507445	0.70040007000400	4.70000400004400	0.004000270402705- 05	Destauridate