

Report for the analysis of 16S microbiome data

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Experimental design and goals:

The aim of the project is to identify the presence and composition of the microorganisms isolated from compost and compost tee samples using a 16S rRNA sequencing strategy in order to investigate the reproducibility of EDAPRO's brewing system as well as the diversity of the microbiome across different biological conditions (i.e. different nutrients).

- We have 3 brewing systems (Treatment: nutrient A, nutrient B, no nutrient C)
- Two different composts as starting material: Compost charge 1 = K2, charge 2 = K3
- We have 5 replicates for each brewing system processed over several weeks producing 30 tee compost samples (K=2 x T=3 x R=5)
- Some control samples (starting Compost K2 and K3 in 5 replicates, MCStd, NTC)

The different questions we would like to answer:

- determine the reproducibility of the brewing system
- diversity and identification / relative abundance of the microbiome of the tee compost samples upon addition of nutrients A and B.
- diversity and identification / relative abundance of the microbiome of tee compost samples versus the microbiome of the compost samples (charge 1 and 2)
- diversity and identification / relative abundance of the microbiome between the compost samples (charge 1 and 2)

Sequencing

The 42 samples were amplified with primers for 16S V3-V4 region (Klindworth et al, 2013), the library prepared following Illumina recommendations and sequenced on one lane of a MiSeq (2x 300bp PE).

>Forward_Primer

```
TCGTCGGCAGCGTCAGATGTGTATAAGAGACAGCCTACGGGNGGCWGCAG
```

>Reverse_Primer

```
GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAGGACTACHVGGGTATCTAATCC
```

Analysis

1) QC of reads (FastQC)

The reads were quality controlled with FastQC, revealing no particular issues.

Library Size Overview



2) DADA2 processing -> ASVs (script dada2.R)

Reads were corrected for errors, merged to form ASV, chimera removed and annotated with Silva database 132 to the genus level following DADA2 protocol.

Export files: rep-seqs.fna, seqtab-nochim.txt, seqtab-taxa.txt

3) Convert to biom format with Qiime2

source activate qiime2-2019.10

#for conversion of dada2 to biom

```
qiime tools import --input-path rep-seqs.fna --type 'FeatureData[Sequence]' --output-path rep-seqs.qza
```

```
echo -n "#OTU Table" | cat - seqtab-nochim.txt > biom-table.txt
```

```
biom convert -i biom-table.txt -o table_v1.biom --table-type="OTU table" --to-json
```

#invert seqtab-taxa.txt with in-house perl (much faster than R):

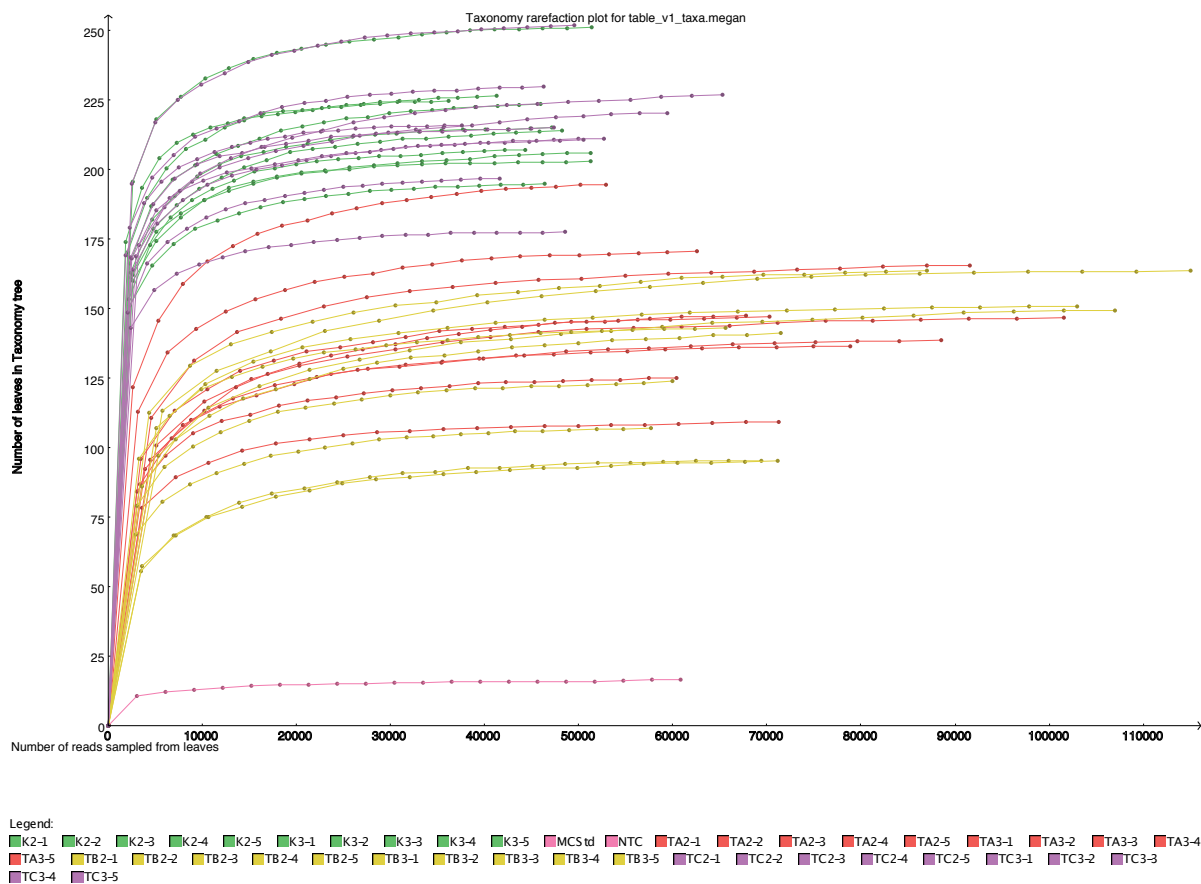
```
./inverter.pl seqtab-taxa.txt > seqtab-taxa_inv.txt
```

#add "ASV" to first line manually into seqtab-taxa_inv.txt

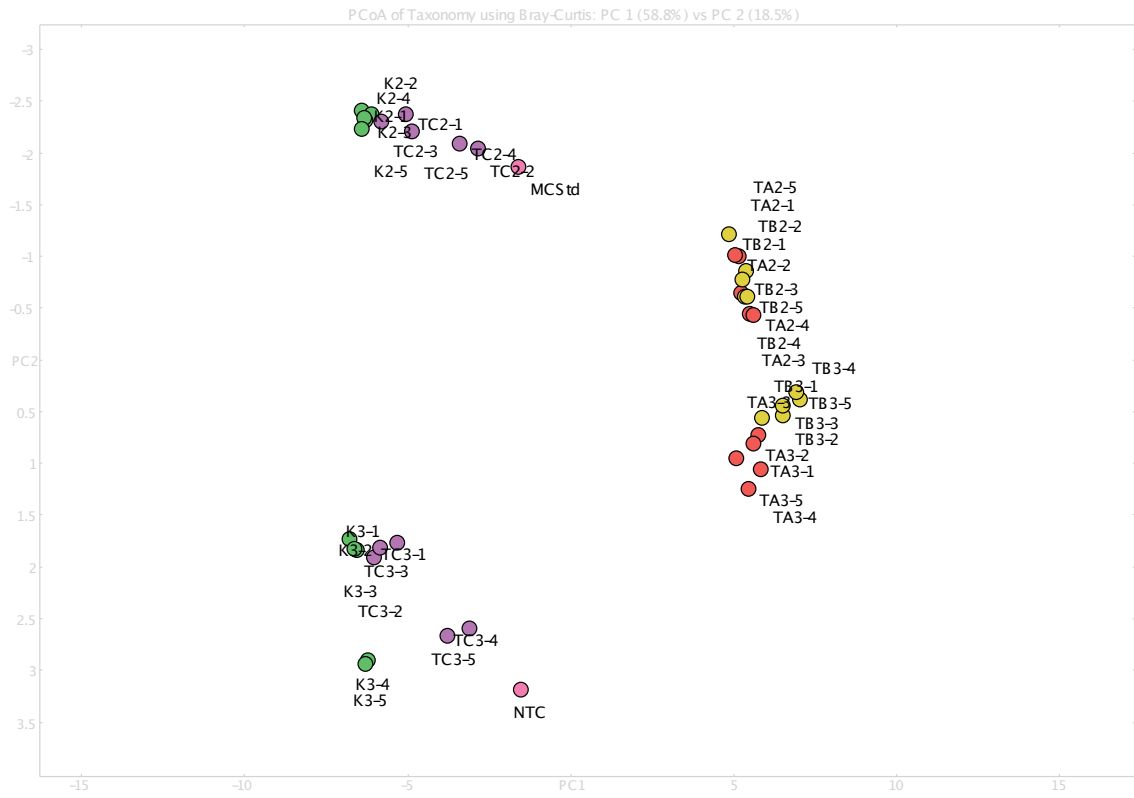
```
biom add-metadata -i table_v1.biom -o table_v1_taxa.biom --observation-metadata-fp seqtab-taxa_inv.txt --observation-header OTUID,taxonomy --sc-separated taxonomy
```

4) Analysis with MEGAN

Import table_v1_taxa.biom and metadata.txt, then calculate rarefaction curves

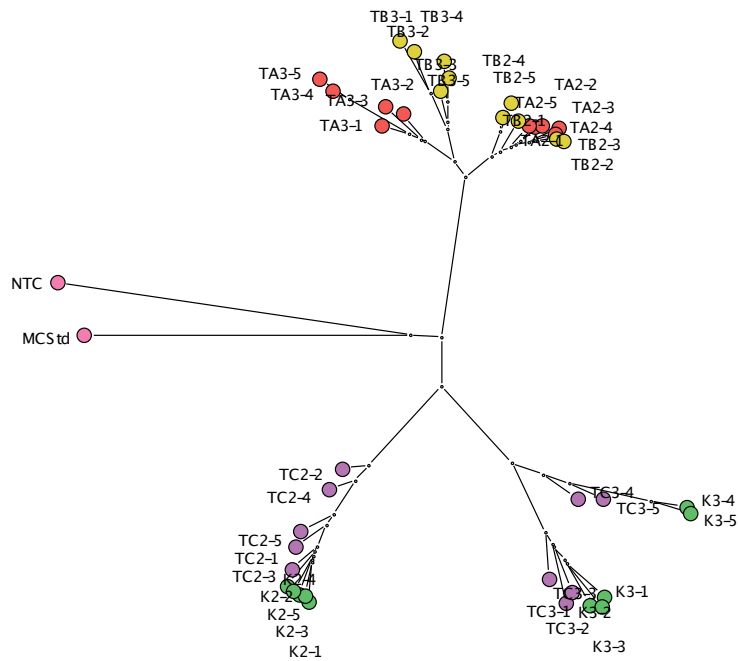


Calculate the PCoA, with Bray-Curtis distance

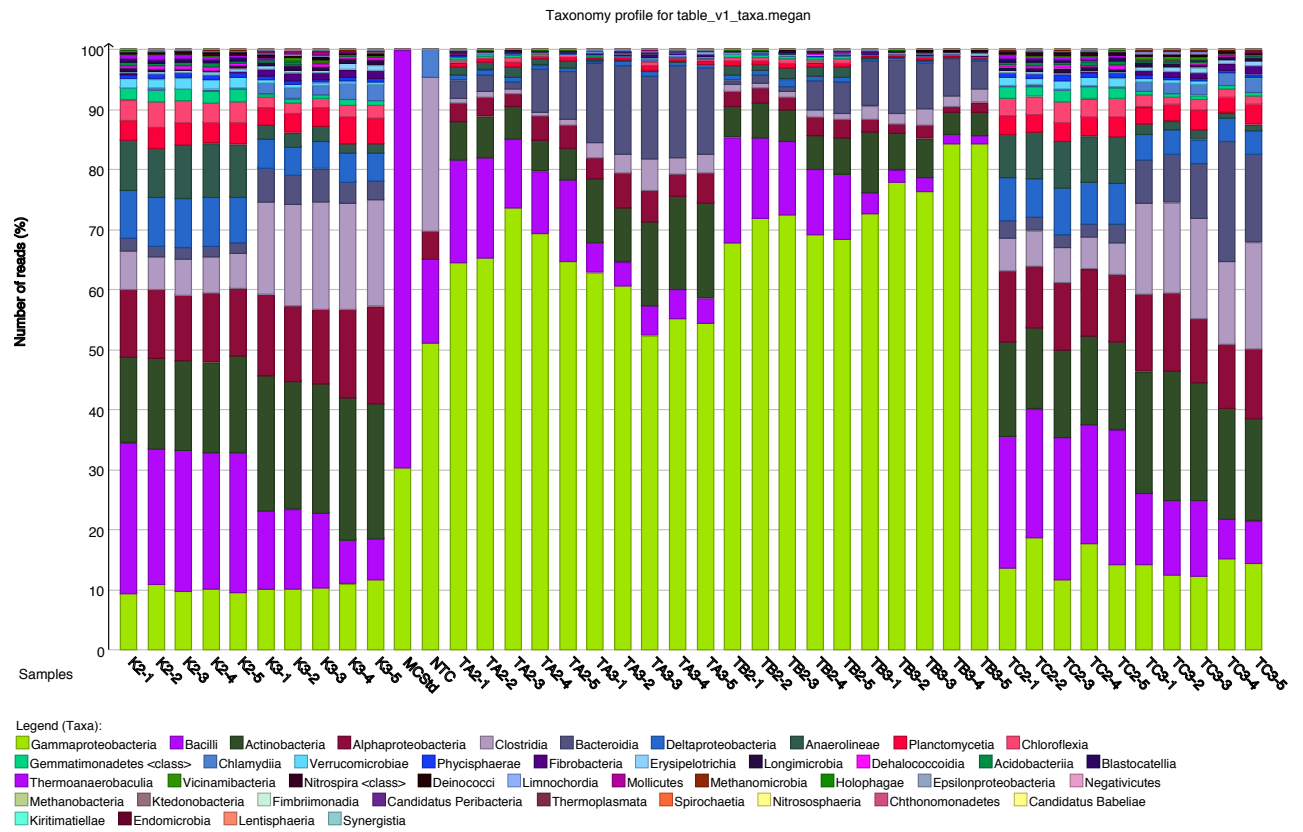


The Neighbour-Joining tree.

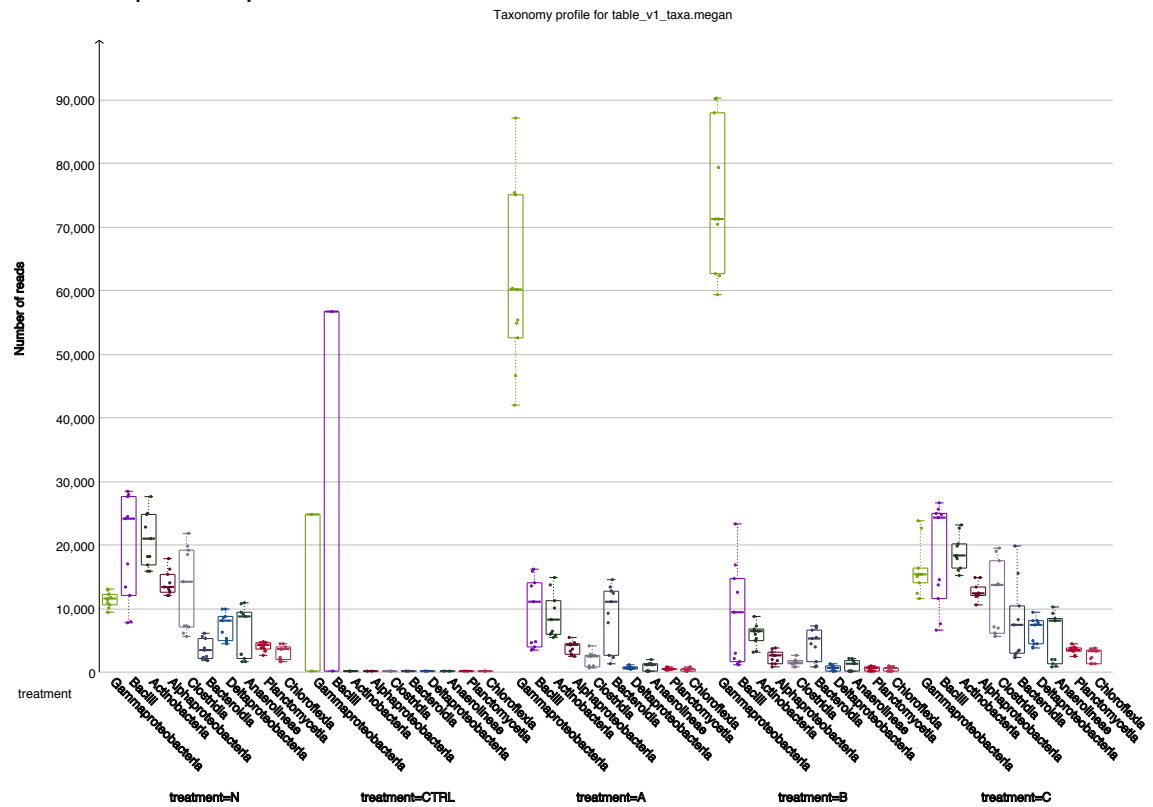
0.1



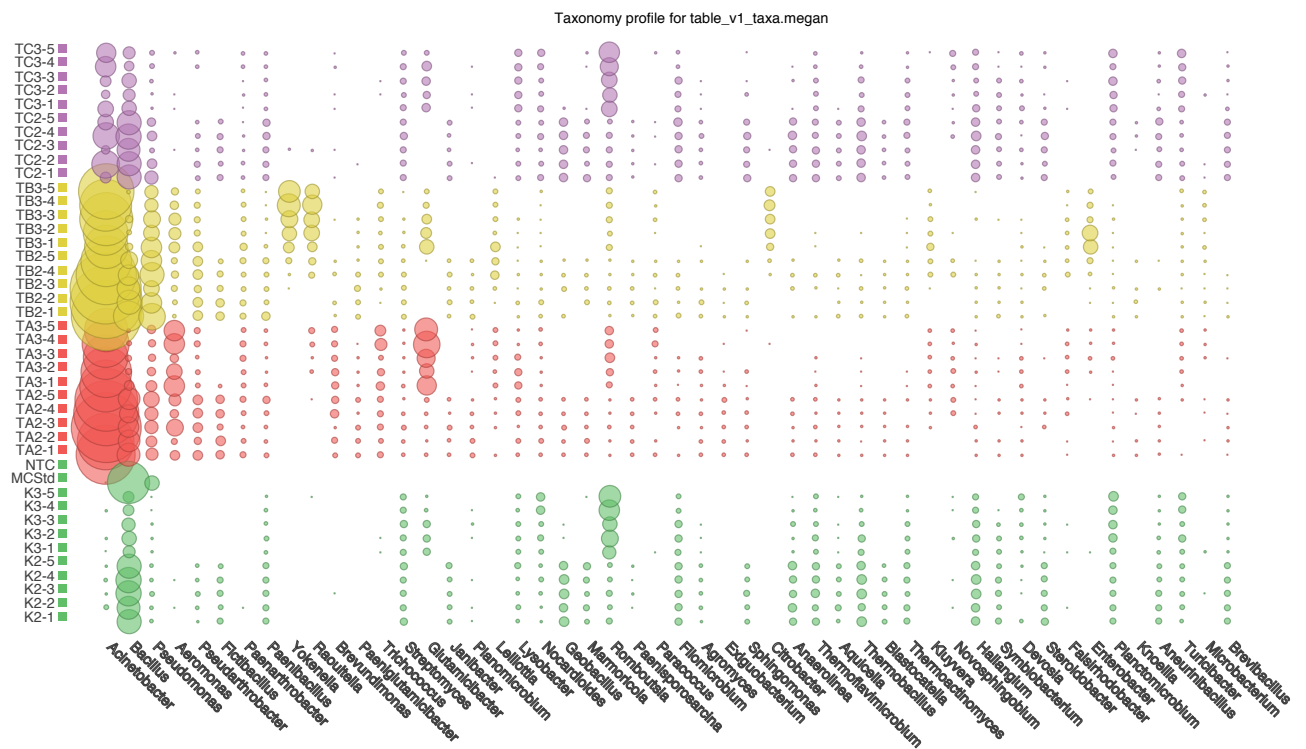
The stacked percentage by taxonomy class



The class profile by treatment

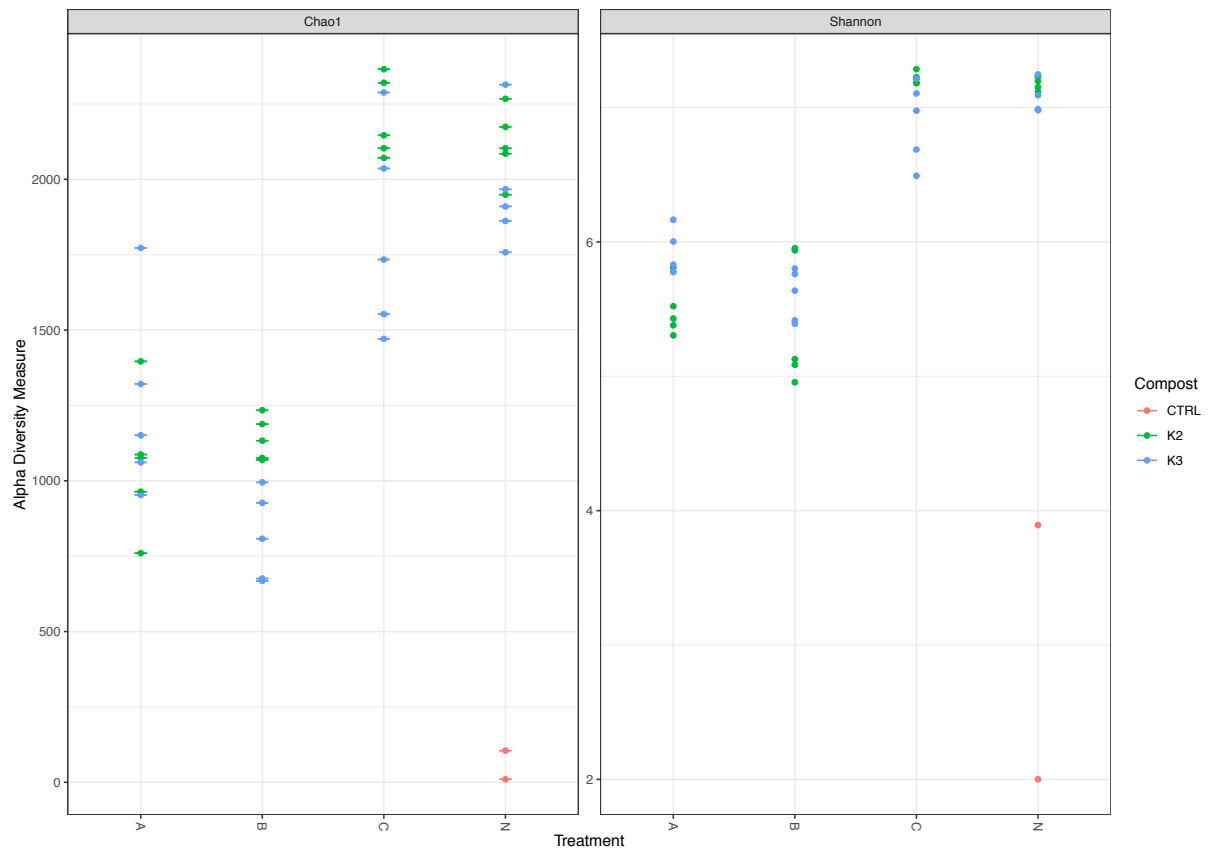


The bubble plot of the top50 most abundant genus overall samples (from left to right):

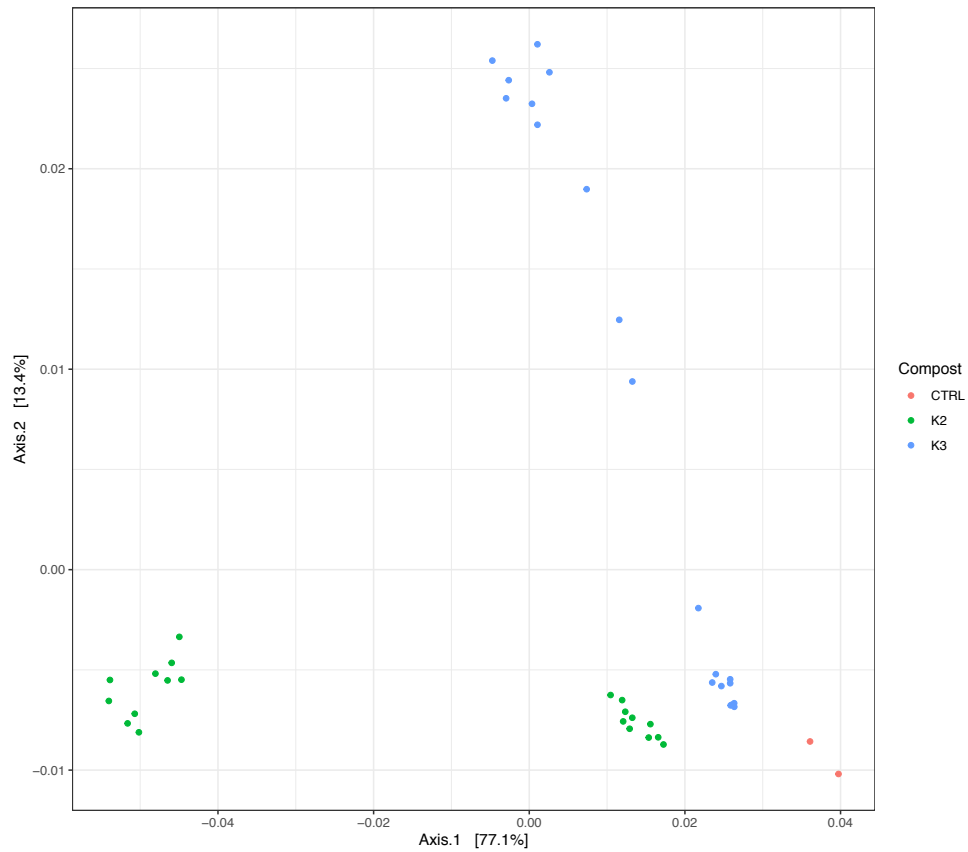


5) Qiime2 was used to build tree of the representative sequences (ASV) source activate qiime2-2019.10
 qiime alignment mafft --i-sequences rep-seqs.qza --o-alignment mafft-rep-seqs.gza
 #Then, eliminate the highly variable positions, to avoid overestimate distances:
 qiime alignment mask --i-alignment mafft-rep-seqs.qza --o-masked-alignment masked-msa-rep-seqs.qza
 #Finally, build the ML tree with FastTree:
 qiime phylogeny fasttree --i-alignment masked-msa-rep-seqs.qza --o-tree unroot-ml-tree-masked.qza
 #Additionally, root your unrooted tree based on midpoint rooting method:
 qiime phylogeny midpoint-root --i-tree unroot-ml-tree-masked.qza --o-rooted-tree root-ml-tree.qza
 qiime tools export --input-path root-ml-tree.qza --output-path root-ml-tree.nwk

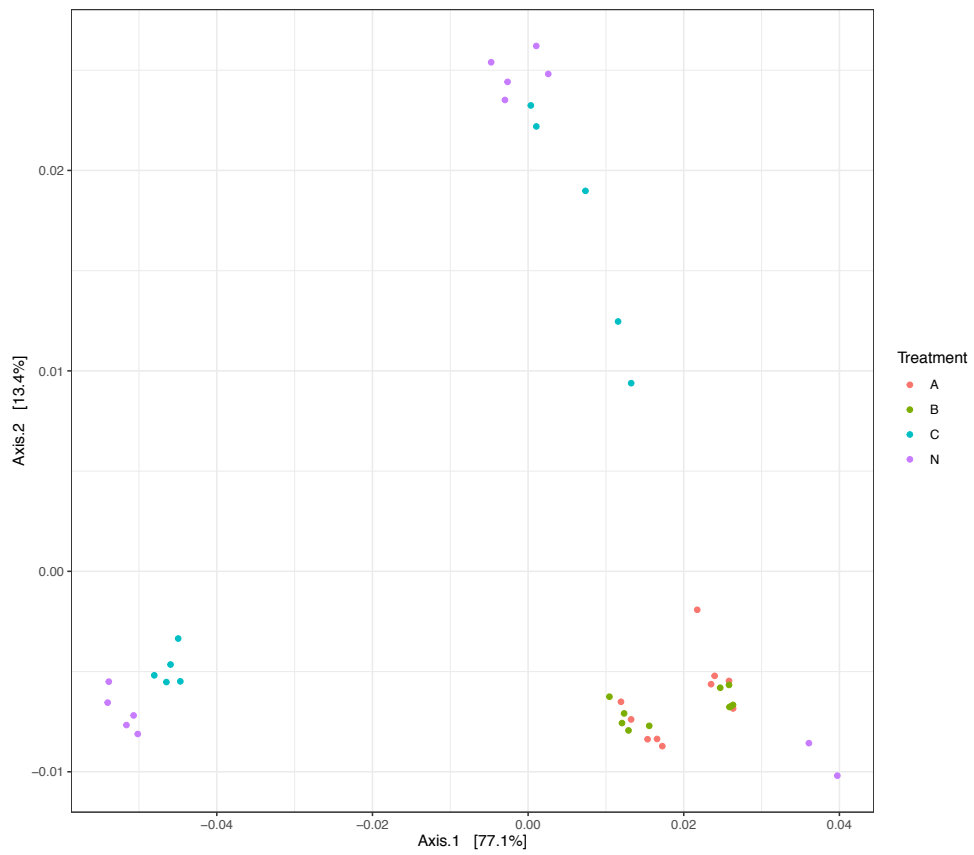
- 6) Import into phyloseq object (script phyloseq.R)
- 7) Calculate plot richness alpha diversity



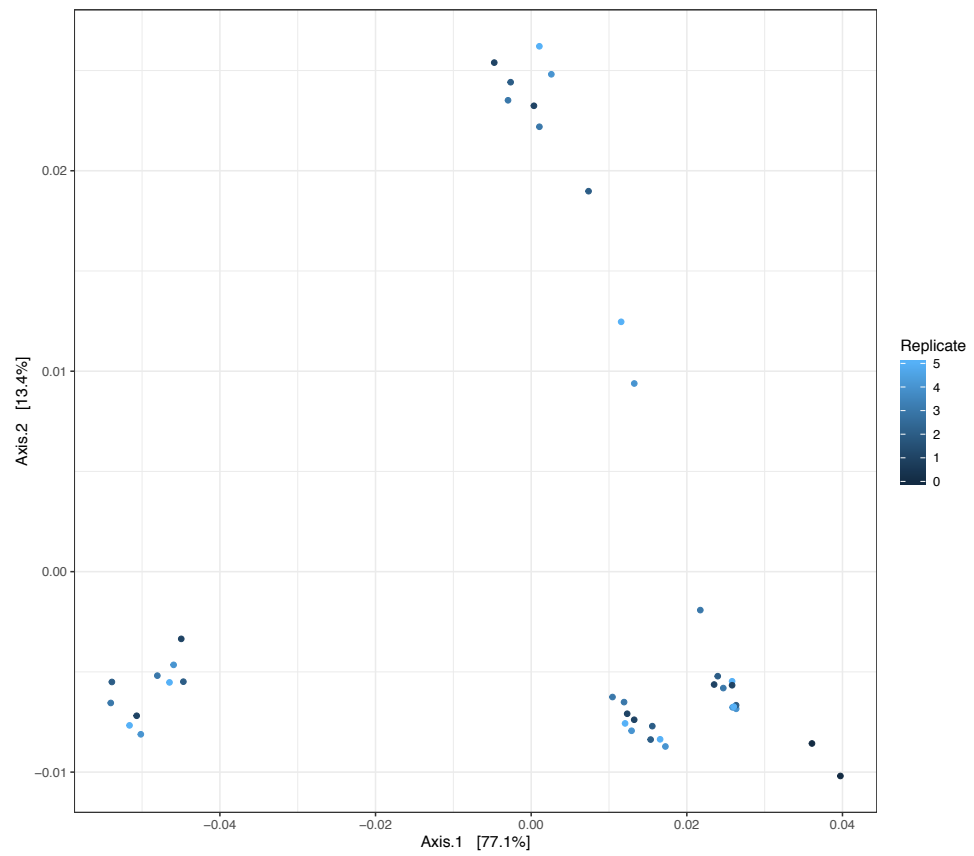
8) Calculate beta diversity (wunifrac, PCoA) and comparisons with PERMANOVA Composts



Treatments



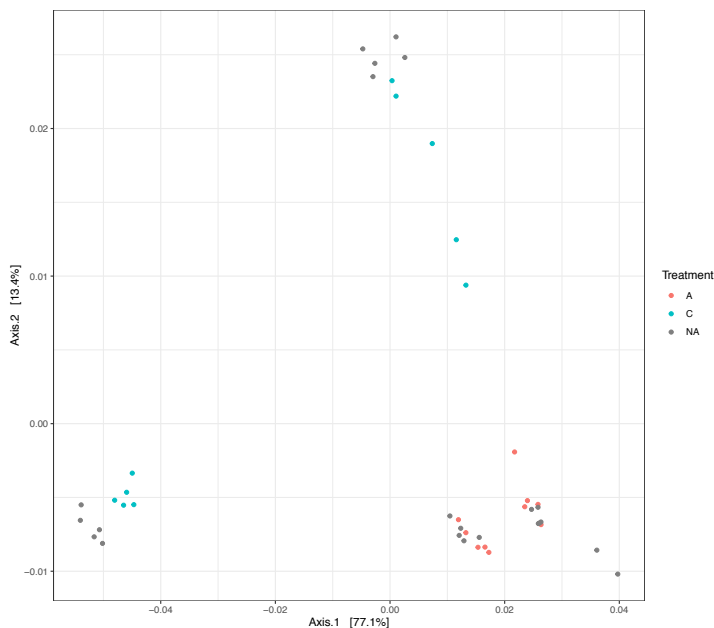
Replicates



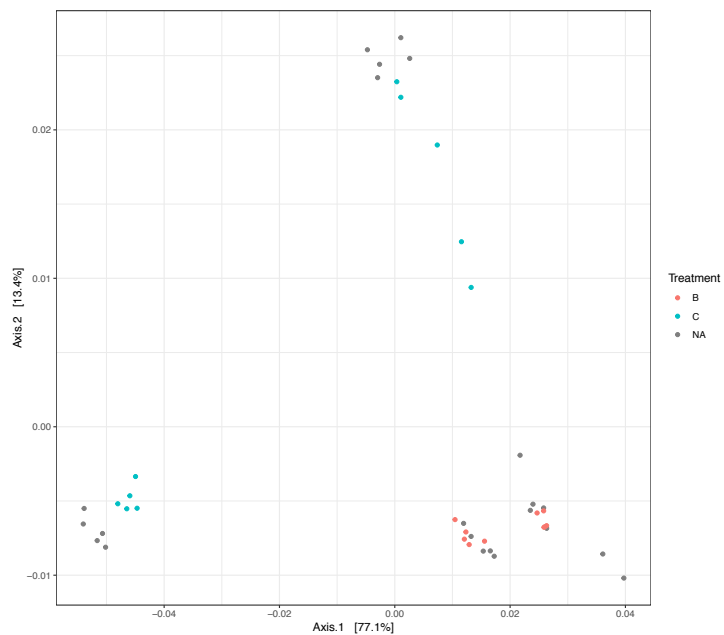
PERMANOVA comparison

| Sample | Df | SumOfSqs | MeanSqs | F.Model | R2 | Pr(>F) | Signif |
|---|----|------------|------------|---------|---------|--------|--------|
| sample_data(sichh)\$Replicate | 1 | 0.000504 | 0.00050394 | 0.44034 | 0.01089 | 0.612 | |
| Residuals | 40 | 0.045777 | 0.00114442 | | 0.98911 | | |
| Total | 41 | 0.046281 | | | 1.00000 | | |
| sample_data(sichh)\$Treatment | 3 | 0.016564 | 0.0055213 | 7.0603 | 0.3579 | 0.001 | *** |
| Residuals | 38 | 0.029717 | 0.0007820 | | 0.6421 | | |
| Total | 41 | 0.046281 | | | 1.0000 | | |
| sample_data(sichh)\$Compost | 2 | 0.015341 | 0.0076704 | 9.6685 | 0.33147 | 0.001 | *** |
| Residuals | 39 | 0.030940 | 0.0007933 | | 0.66853 | | |
| Total | 41 | 0.046281 | | | 1.00000 | | |
| Compare only A, B, C | | | | | | | |
| sample_data(AvsB)\$Treatment | 1 | 0.00014621 | 1.4621e-04 | 1.5106 | 0.07742 | 0.17 | |
| Residuals | 18 | 0.00174229 | 9.6794e-05 | | 0.92258 | | |
| Total | 19 | 0.00188850 | | | 1.00000 | | |
| sample_data(AvsC)\$Treatment | 1 | 0.0086708 | 0.0086708 | 14.67 | 0.44904 | 0.001 | *** |
| Residuals | 18 | 0.0106387 | 0.0005910 | | 0.55096 | | |
| Total | 19 | 0.0193095 | | | 1.00000 | | |
| sample_data(BvsC)\$Treatment | 1 | 0.0085147 | 0.0085147 | 14.329 | 0.44323 | 0.001 | *** |
| Residuals | 18 | 0.0106959 | 0.0005942 | | 0.55677 | | |
| Total | 19 | 0.0192106 | | | 1.00000 | | |
| Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 | | | | | | | |

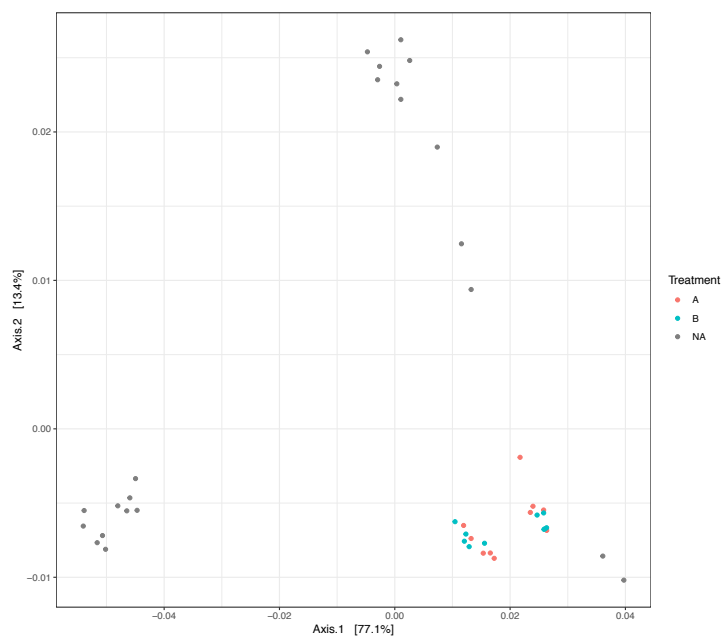
A vs C



B vs C



A vs B



Conclusion

The results explain a clear separation of the microbiomes of the 2 starting composts K2 and K3. Both treatments with nutrient A or nutrient B create a shift of the microbiome leading to a reduction of the diversity with a convergence of the treated compost microbiomes. The genus *Acinetobacter* (class gammaproteobacteria) is highly dominant in the treated composts. The 5 replicates show a good reproducibility in all cases.

References

FastQC

URL: <https://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

Dada2

Callahan B, McMurdie P, Rosen M et al. DADA2: High-resolution sample inference from Illumina amplicon data. *Nat Methods* 13, 581–583 (2016).
<https://doi.org/10.1038/nmeth.3869>

Phyloseq

McMurdie PJ, Holmes S (2013) phyloseq: An R Package for Reproducible Interactive Analysis and Graphics of Microbiome Census Data. *PLoS ONE* 8(4): e61217.
<https://doi.org/10.1371/journal.pone.0061217>

Qiime2

Bolyen E, Rideout JR, et al. 2019. Reproducible, interactive, scalable and extensible microbiome data science using QIIME 2. *Nature Biotechnology* 37: 852–857.
<https://doi.org/10.1038/s41587-019-0209-9>

MEGAN

Huson, D, Beier, S, Flade, I, Gorska, A, El-Hadidi, M, Mitra, S, Ruscheweyh, H, and Rewati Tappu, D (2016). MEGAN Community Edition - Interactive exploration and analysis of large-scale microbiome sequencing data. *PLoS Computational Biology*, 12(6):e1004957

Silva database

Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, Peplies J, Glöckner FO (2013) The SILVA ribosomal RNA gene database project: improved data processing and web-based tools. *Nucl. Acids Res.* 41 (D1): D590-D596.

Primers V3-V4 (kindworth)

Kindworth A, Pruesse E, Schweer T, Peplies J, Quast C, Horn M, Glöckner FO. Evaluation of general 16S ribosomal RNA gene PCR primers for classical and next-generation sequencing-based diversity studies, *Nucleic Acids Research*, Volume 41, Issue 1, 1 January 2013, Page e1,
<https://doi.org/10.1093/nar/gks808>