Bacterial communities on classroom surfaces

2 Manuscript demo

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7 Introduction

- ⁸ The data used here are a small subset (first 20,000 quality-filtered sequences) of those previously published
- 9 [1]. This demo illustrates a few basic multivariate analysis methods with a sample dataset. In the original
- manuscript, we investigated the sources of microbes on classroom surfaces, and whether those microbial
- communities reflect common human contact with indoor surfaces.

12 Methods

- 13 This sequence dataset was processed using QIIME 1.8 [2] with a default MacQIIME installation (http:
- 14 //www.wernerlab.org/software/macqiime). Scripts for processing raw data are in the ../QIIME/ folder. To
- pick OTUs in that folder, you will execute the pickTheseOTUs.sh script sitting in that folder. This script
- wants to run MacQIIME, so if you are not using MacQIIME, you'll need to alter the top line to reflect your
- 17 system.
- 18 For statistical analyses, we primiarily used the phyloseq package to handle QIME output files, and vegan
- and labdsv for multivariate ecology stats [3–5]. All sequences were rarefied to an equal sampling depth (100
- 20 sequences per sample) prior to analysis. Beta-diversity was calculated using the Canberra taxonomic metric.
- 21 The Canberra metric is defined as:

$$d_{jk} = \frac{1}{NZ} \sum \frac{x_{ij} - x_{ik}}{x_{ij} + x_{ik}}$$

- where NZ is the number of non-zero entries. Reproducible documents were created with the knitr package
- 23 in R [6].

24 Results

- Out of a total 1.5923×104 sequences that passed quality filtering, we analyzed 5800 sequences in 58 samples
- distributed among 966 OTUs (97% sequence similarity). The most abundant OTU in the dataset was a
- 27 Cyanobacterium (2.67% of all sequences). The most abundant taxa are shown in Table 1.

	Phylum	Family	Genus	Species	RelAbu
505954	Cyanobacteria	Xenococcaceae	=	-	2.67
1039477	Firmicutes	Staphylococcaceae	Staphylococcus	epidermidis	2.52
4449609	Proteobacteria	Sphingomonadaceae	Sphingomonas	-	2.40
359689	Actinobacteria	Corynebacteriaceae	Corynebacterium		2.16
4482309	Proteobacteria	Acetobacteraceae	-	-	2.14

Table 1: Most abundant taxa across all surfaces.

	Df	SumsOfSqs	MeanSqs	F.Model	R2	Pr(>F)
map\$SurfaceType	3	2.14	0.71	1.80	0.09	0.001
Residuals	54	21.39	0.40		0.91	
Total	57	23.52			1.00	

Table 2: Surface type explains a significant amount of variation among communities.

- We found that surface type explained a significant amount of community variation (p = 0.001; from PERMANOVA on Canberra distances).
- Next, we tested for a quasi-distance-decay relationship. This is the sort of pattern we see in just about every
- ecosystem with most forms of life. We even found this to be a stong predictor in the dust sampled from
- the entire building [7]. So we can use the x and y coordinates as a map of samples, and then calculate the
- Euclidean pairwise distance between all samples. Then that goes through a mantel test to determine if these
- 34 distance are correlated with the community distances.
- We did not find any significant coorelation between community similarity and spatial distance (p = 0.456;
- ₃₆ from Mantel test) when considering all samples together. Likewise, individual sample types tested alone
- showed no relationship with spatial distance (p > 0.1 for all four sample types).

38 Discussion

- 39 So it looks like the type of surface, potentially as a proxy for human contact, explains a significant amount of
- 40 variation, in the microbial communities on those surfaces, but their proximity to each other around the room
- doesn't seem to matter at all.

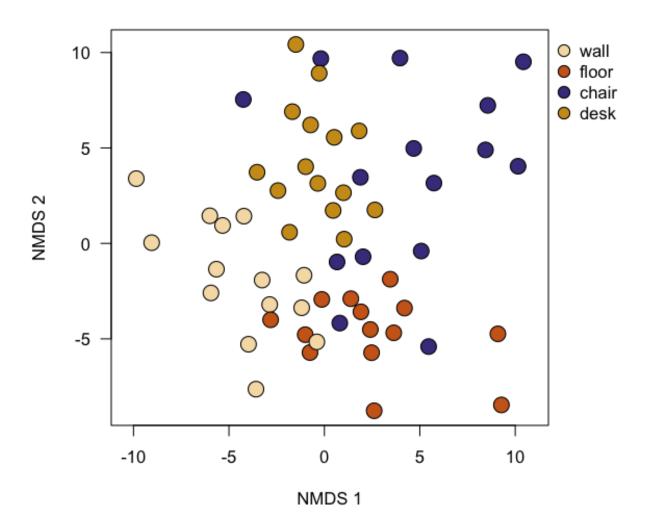


Figure 1: Samples cluster by the type of surface.

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3 References

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