

This is a self-archived version of an original article. This version may differ from the original in pagination and typographic details.

Author(s): Davies, Lucy R.; Barbero-López, Aitor; Lähteenmäki, Veli-Matti; Salonen, Antti; Fedorik, Filip; Haapala, Antti; Watts, Phillip C.

Title: Microbes within the building envelope : a case study on the patterns of colonization and potential sampling bias

Year: 2023

Version: Published version

Copyright: © 2023 Davies et al.

Rights: CC BY 4.0

Rights url: https://creativecommons.org/licenses/by/4.0/

Please cite the original version:

Davies, L. R., Barbero-López, A., Lähteenmäki, V.-M., Salonen, A., Fedorik, F., Haapala, A., & Watts, P. C. (2023). Microbes within the building envelope: a case study on the patterns of colonization and potential sampling bias. PeerJ, 11, Article e16355.

https://doi.org/10.7717/peerj.16355



Microbes within the building envelope—a case study on the patterns of colonization and potential sampling bias

Lucy R. Davies¹, Aitor Barbero-López², Veli-Matti Lähteenmäki², Antti Salonen³, Filip Fedorik³, Antti Haapala² and Phillip C. Watts¹

- ¹ Department of Biological and Environmental Science, University of Jyväskylä, Jyväskylä, Finland
- ² Department of Chemistry, University of Eastern Finland, Joensuu, Finland
- ³ Civil Engineering, Faculty of Technology, University of Oulu, Oulu, Finland

ABSTRACT

Humans are exposed to diverse communities of microbes every day. With more time spent indoors by humans, investigations into the communities of microbes inhabiting occupied spaces have become important to deduce the impacts of these microbes on human health and building health. Studies so far have given considerable insight into the communities of the indoor microbiota humans interact with, but mainly focus on sampling surfaces or indoor dust from filters. Beneath the surfaces though, building envelopes have the potential to contain environments that would support the growth of microbial communities. But due to design choices and distance from ground moisture, for example, the temperature and humidity across a building will vary and cause environmental gradients. These microenvironments could then influence the composition of the microbial communities within the walls. Here we present a case study designed to quantify any patterns in the compositions of fungal and bacterial communities existing in a building envelope and determine some of the key variables, such as cardinal direction, distance from floor or distance from wall joinings, that may influence any microbial community composition variation. By drilling small holes across walls of a house, we extracted microbes onto air filters and conducted amplicon sequencing. We found sampling height (distance from the floor) and cardinal direction the wall was facing caused differences in the diversity of the microbial communities, showing that patterns in the microbial composition will be dependent on sampling location within the building. By sampling beneath the surfaces, our approach provides a more complete picture of the microbial condition of a building environment, with the significant variation in community composition demonstrating a potential sampling bias if multiple sampling locations across a building are not considered. By identifying features of the built environment that promote/retard microbial growth, improvements to building designs can be made to achieve overall healthier occupied spaces.

Subjects Microbiology, Molecular Biology, Environmental Health **Keywords** Indoor microbiota, Built environment, Occupant health, Building mould

INTRODUCTION

Investigations into microbial communities of occupied spaces of the built environment (the indoor microbiota) typically use methods that collect filtered air particles or dust that

Submitted 21 April 2023 Accepted 4 October 2023 Published 17 November 2023

Corresponding author Lucy R. Davies, lucy.r.davies@jyu.fi

Academic editor Jonathan Thomas

Additional Information and Declarations can be found on page 12

DOI 10.7717/peerj.16355

© Copyright 2023 Davies et al.

Distributed under Creative Commons CC-BY 4.0

OPEN ACCESS

accumulate on indoor surfaces (e.g., Maestre et al., 2018; Pekkanen et al., 2018; Fu et al., 2020). These methods facilitate large-scale projects (especially citizen science (Barberán et al., 2015a; Barberán et al., 2015b) and are both straightforward and non-destructive, but these samples cannot separate the microbial contributions of the structural aspects of the building (e.g., the building envelope) from those derived from the building occupants (e.g., hair or skin) (Cao et al., 2021) or the surrounding environment (e.g., soil, pollen grains) (Barberán et al., 2015a), with limited information on how any structural aspects of the building could be contributing to the microbiota of the built environment.

Buildings are complex, three-dimensional structures that contain significant spaces beneath the occupant accessible surfaces. By design, these areas are intended to be dry environments, but humidity can be found in these spaces due to air flow (Fedorik et al., 2021). Evidence of microbial communities inhabiting extreme environments such as hyper-arid areas within the Atacama desert (Schulze-Makuch et al. 2018; Hwang et al., 2021) or the International Space Station (Checinska Sielaff et al., 2019), raises the potential for these building areas to host viable microbes. Some fungi genera and actinomycetes, for example, have been cultured from insulation material samples (Pessi et al., 2002) and can penetrate building structures (Pessi et al., 2002; Airaksinen et al., 2004). The presence and potential contamination of indoor air of these microbes is of significant concern due to their detrimental effects on health (Järvi et al., 2018). Investigating factors influencing microbiota formation, such as building design, environmental conditions and materials, that have an impact on these microbial communities would allow for better predictions and control over building health problems, including material degradation, indoor air quality and human health concerns due to microbial interactions. Establishing a clearer picture and better understanding of the microbes occupying the whole built environment could influence building policies and demonstrate the need to adapt for a healthier home.

Building design and structural condition are therefore likely to be a key determinant of the microbiota composition within building structures due to microenvironmental variation. For example, gradients in temperature, and/or moisture in a building will be an expected consequence of material choice and envelope depth (*Fedorik et al.*, 2021), or cardinal direction in which a structure is facing (*Kabátováa & Ďuricaa*, 2019). Examining the microbes that live beneath the surface within an occupied space is essential for establishing a fuller picture of the microbiota of the built environment. In-wall sampling though must take these potential gradients into consideration as sampling in just few locations may result in sampling bias.

The goal of this study was to investigate possible differences in microbial (bacteria and fungi) communities, across different locations within a building, that have accumulated over time; thereby highlighting the potential bias that may arise from insufficient in-wall sampling. Additionally, we aimed to identify potential factors driving any variation in microbial communities. To test our approach, we selected a residential dwelling situated within a forested area in Finland. As a stand-alone building with walls facing all cardinal directions, it served as an ideal case-study. Samples were taken by drilling small holes (\emptyset = 12 mm) into different areas of a home to make a contrast between cardinal directions in which the wall faces, the distance from floor level and the distance from wall joints. Air

samples were then extracted from spaces between internal and external walls to obtain a sample of microbes which was then quantified using amplicon sequencing. This novel approach identified significant spatial variation in the bacterial and fungal communities to demonstrate the diversity of microbes within building materials.

METHODS

Sampling plan and microbial extraction

On September 20th, 2020, we sampled a traditional Finnish wood-framed house with a natural gravity-driven ventilation located in municipality of Vaala, Finland, that had traditional sawdust and wood shive insulation from 1962 on cast concrete foundation by inserting a sterile tygon tube into 12 mm holes that had been drilled into the walls, with the hole then sealed using modelling clay to prevent collection of 'indoor air' from living space. Air from within the building element was pumped (20 min at 3 litres/min) using a SKC universal air pump (model 224-PCMTX4K) over a SureSeal Blank Styrene cassette containing a 25 mm PTFE membrane filter. Holes were drilled in different sampling areas: (1) direction (*i.e.*, the cardinal direction in which the wall is facing), (2) height from floor level (lower, middle and upper) and (3) position (near left wall joining, centre of the wall and near right wall joining) (Fig. 1). Additionally, air samples were taken throughout the day around the outside of the building and from various rooms within the building.

DNA extraction and sequencing

Filters were soaked in molecular grade water for 12 h. This was to ensure microbes were washed 'out' of the filters and easily accessible for DNA extraction. DNA was extracted from the filter and its water using Qiagen DNeasy[®] PowerSoil[®] Pro Kit according to the manufacturer's protocol, but with the following adjustments: (1) PowerBead Pro Tubes were vortexed for 20 min at maximum rpm speed. DNA was also extracted from control samples: only buffer, molecular grade water and buffer, and unused filters and buffer.

Bacterial and fungal taxa was then identified by amplicon sequencing performed on an Illumina NovoSeq by NovoGene Ltd (https://www.novogene.com/us-en/). Bacterial 16S V4 region (universal primers GTGCCAGCMGCCGCGGTAA and GGACTACHVGGGTWTCTAAT) and the fungal ITS2 region (universal primers CATCGATGAAGAACGCAGC, TCCTCCGCTTATTGATATGC) were targeted.

Processing and analysing fungal and bacterial sequences

Primers were removed using *cutadapt* v1.10 (*Marcel*, *2011*) on ITS reads, while *DADA2* v1.18 (*Callahan et al.*, *2016*) was used to remove primers and truncate (forward reads at base 220, reverse reads at base 200) the 16S reads. *DADA2* was used for both ITS and 16S data to merge paired-end reads, remove chimaeras and identify amplicon sequence variants (ASVs) (*Callahan et al.*, *2016*). Taxonomy was assigned to ITS and 16S ASVs using UNITE v. 10.05.2021 (*Nilsson et al.*, *2015*) and the SILVA v.138 database (*Quast et al.*, *2013*) respectively. *decontam* v.1.14.0 (*Davis et al.*, *2018*) was used to eliminate potential contaminant ASVs identified in the control samples, using the prevalence method and the probability that a read is a contaminant threshold of 0.5 (*Davis et al.*, *2018*). To remove low

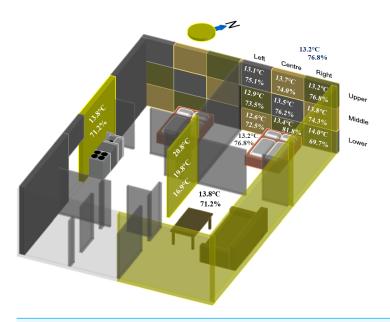


Figure 1 Schematic showing the layout of the sampling site. Holes were drilled into the walls of a traditional Finnish wood-framed house which contained two bedrooms, a living room and a kitchen. The toilet was located in an out-house. Samples were taken from walls facing different cardinal direction, at different heights (lower, middle and upper), and at different positions (left, centre and right). The grid squares show the different areas sampled. Although only shown on a couple of walls, this scheme was followed on all sampled walls. Sampled walls are highlighted in yellow. Wall temperatures and humidity measurements, where taken, are labelled in white font, outside temperature and humidity in blue, inside in darkgrey, and floor in black. Measurements were taken during the day of sampling.

Full-size DOI: 10.7717/peerj.16355/fig-1

frequency ASVs which were most likely a result of sequencing errors, while also avoiding the removal of rare ASVs, ASVs that had a total count of 20 across the entire dataset were removed.

Statistical analysis

Statistical analysis was performed using R v.4.1.3 (*R Core Team, 2021*). The package *FEAST* v.1.0 (Fast Expectation-mAximization microbial Source Tracking) (*Shenhav et al., 2019*) was used to estimate the contributions of the microbial communities from indoor air and outdoor air to the microbial communities of the in-wall samples. To do this, the air samples were labelled as 'sources', and the wall samples were as labelled 'sinks'. Total Sum Scaling (TSS) normalization was used to remove any bias related to differences in sequencing depth in different libraries by dividing each ASV count with the total library size per sample. Analyses of the microbial communities was then performed on the relative abundance of each ASV using *phyloseq* v.1.38.0 (*McMurdie & Holmes, 2013*) and the package *vegan* v.2.5-7 (*Oksanen et al., 2019*). The package vegan was used to calculate the alpha diversity measures (the number of different taxa groups and their abundance, and the number of distinct taxa) and beta diversity measures (the diversity differences between two samples). Predictors of variation in alpha diversity was assessed using a generalized linear model that contained alpha diversity as the response variable and direction, height and position as predictor variables. We detected no over-dispersion in the model. We compared the full

model with reduced models from which a predictor variable was omitted using a F-test to obtain *p*-values. The package vegan was used to conduct PERMANOVA tests (using the adonis2() command, using the Bray Curtis distance method, and setting permutations to 999) to assess predictors of variation in beta diversity, and was also used to conduct a distance-based redundancy analysis (dbRDA) ordination method which we used to visualise differences between the microbial communities across different groups (*Legendre & Anderson, 1999*). To examine the ASVs driving any variation in the beta diversity, the SIMPER() command from vegan was used to calculate similarity percentages (*Clarke, 1993*). From the identified genera, a Kruskal–Wallis test was conducted, followed by a Dunn test to assess pair-wise significant differences (*Dinno, 2017*). Statistical significance was based on adjusted *p*-values using the Bonferroni method (*Dunn, 1961*). Plots were produced using *ggplot2* v.3.3.6 (*Wickham, 2016*) and *patchwork* v.1.1.2 (*Pedersen, 2022*).

RESULTS AND DISCUSSION

To quantify patterns of microbial colonization, we first measured alpha diversity using Shannon diversity (the number of different taxa groups and their abundance), and amplicon sequence variant (ASV) richness (the number of distinct taxa). For fungal communities, the height at which the samples were taken had a significant effect on Shannon diversity (Table 1), with the lowest diversity found in the middle of the wall (although a Tukey posthoc test did not show pairwise significant differences: Lower vs Middle Padjusted = 0.24; Middle vs Upper $P_{\text{adjusted}} = 0.12$; Lower vs Upper $P_{\text{adjusted}} = 0.96$) (Fig. 2B). There were no significant differences between the cardinal direction of the wall and positions (Figs. 2A, 2C; Table 1). Although not significant (Table 1), height from ground showed a qualitative pattern where the alpha diversity measurements of the bacterial communities decreased from the lower to the upper part of the wall (Fig. 3B). We were therefore interested in examining whether the Shannon diversity and richness correlated with a quantitative measurement of the height of the wall. When taking the distance from the floor whereby the holes were drilled into consideration as linear variables (20 cm, 120 cm, 220 cm), we found the Shannon diversity of bacterial communities significantly decreased as the distance from the floor increases ($F_{1,21} = 4.12$, P < 0.04). As with the alpha diversity of fungal communities, there were no significant differences between the directions and positions (Figs. 3A, 3C; Table 1).

Fungi and bacteria can inhabit diverse environments (*Storze & Hengge*, 2011; *Haruta & Kanno*, 2015), with the community composition dependent on the outcome of selection and competition among taxa. In natural environments, environmental variation in, for example, humidity (*Wang et al.*, 2021), moisture (*Borowik & Wyszkowska*, 2016), temperature (*Nottingham et al.*, 2022), and pH (*Scholier et al.*, 2022) are important drivers of fungal and bacterial composition and activity. Just as there are environmental gradients in nature (*Wang et al.*, 2021), typically, increasing higher up you go in a wall is associated with a dryer and warmer environment (*Fedorik et al.*, 2021). It is these variations that would elicit changes in microbiota composition.

For fungal and bacterial richness, the variable that had the greatest variance was the cardinal direction in which the sampled wall was facing (Table 1). Samples taken from west

Table 1 Statistical results from the F-tests for the alpha diversity of fungal and bacteria communities. Results were obtained by comparing a generalized linear model containing either Shannon diversity or number of individual ASVs as the response variable and direction, height and position as the predictor variables with a reduced model where a predictor variable was omitted.

		Shannon diversity index			Number of individual ASVs				
Kingdom	Variable	Estimate ± s.e.	R-squared	$F_{ m d.f}$	P-value	Estimate ± s.e.	R-squared	$F_{ m d.f}$	P-value
Fungi	Direction	2.06 ± 0.39	0.13	1.84 _{1,21}	0.16	21.5 ± 10.34	0.28	2.09 _{1,21}	0.12
	Height	2.15 ± 0.27	0.13	$4.11_{1,21}$	0.03*	29.64 ± 8.59	0.01	$0.38_{1,21}$	0.69
	Position	2.04 ± 0.25	0.04	$0.37_{1,21}$	0.68	35.71 ± 7.23	0.11	1.67 _{1,21}	0.21
Bacteria	Direction	3.05 ± 0.26	0.13	$1.76_{1,21}$	0.18	202.67 ± 38.55	0.18	$2.50_{1,21}$	0.08
	Height	3.31 ± 0.18	0.15	$2.88_{1,21}$	0.08	236.73 ± 28.54	0.11	$2.84_{1,21}$	0.08
	Position	2.93 ± 0.18	0.02	$0.99_{1,21}$	0.38	182.93 ± 26.28	0.03	1.57 _{1,21}	0.23

Notes.

facing walls had the highest number of individual ASVs (Figs. 2B, 3A). Previous studies on indoor microbiota have also shown cardinal direction to have an important impact on bacterial communities (*Fahimipour et al.*, 2018; *Horve et al.*, 2020). Though focusing on viable bacteria on surfaces, rather than in-wall sampling, *Horve et al.* (2020) found west facing rooms with windows to have a higher abundance of viable bacteria compared to rooms facing other cardinal directions due to direct sunlight (*Horve et al.*, 2020). The intensity of direct sunlight will vary across different seasons and as such, studies have shown exposure to indoor microbes can vary across seasons (*Garrett et al.*, 1997; *Rintala et al.*, 2008). The heat generated by direct sunlight can increase indoor humidity - a factor that contributes to these seasonal variations (*Frankel et al.*, 2012; *Knudsen, Gunnarsen & Madsen, 2017*). Therefore, when investigating microbial compositions, humidity within the building envelope would likely be an important environmental factor causing variation.

To examine the most relatively abundant taxa, we grouped the ASVs at genus level and identified the top fifteen (see Figs. S1 and S2 for fungi and bacteria, respectively). Several of the top relatively abundant genera identified in this study are associated with building and occupant health. For example, fungal species of Antrodia, and Heterobasidion are key house-rot fungi, as they decay wooden building materials (Huckfeldt & Schmidt, 2006; Schmidt, 2007; Schmidt & Huckfeldt, 2011; Gabriel & Švec, 2017; Haas et al., 2019). Sarocladium species are associated with problems in biodegradation of mineral-based materials (Ponizovskaya et al., 2019). Species of Aspergillus are reported from multiple studies of the indoor environments (Tanaka-Kagawa et al., 2005; Mousavi et al., 2016; Chen et al., 2017) and some species from this genus, along with species of *Phialocephala*, affect the severity of asthmatic symptoms (Hedayati, Mayahi & Denning, 2010; Dannemiller et al., 2016; Mousavi et al., 2016). Many of the top relatively abundant bacterial genera have also been identified in studies on the indoor environment. Acinetobacter (Hui et al., 2019; Wu et al., 2022), Cutibacterium (Sun et al., 2022), Staphylococcus (Moon, Huh & Jeong, 2014; Madsen et al., 2018) and Blaudia (Fu et al., 2021), for example, occur in samples of indoor dust and swabs. Previous studies have also made links between these genera and negative impacts on human health (Kozajda, Jeżak & Kapsa, 2019; Fu et al., 2021; Sun et al., 2022; Wu et al., 2022). Further investigations are necessary to determine if similar risks

^{*}Significant P-value.

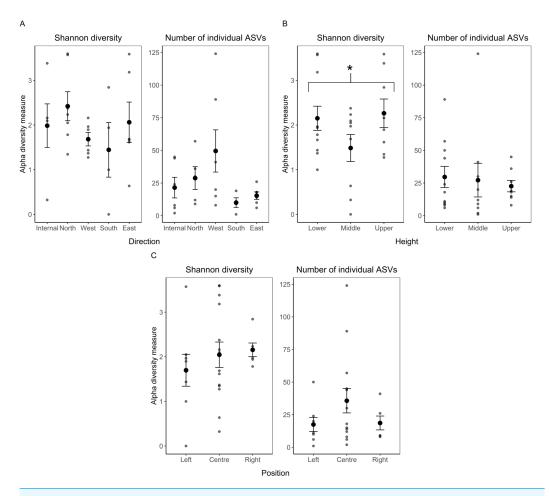


Figure 2 ITS alpha diversity analysis: The average Shannon diversity index and number of individual ASVs (±standard error of mean). Comparing (A) direction, (B) height, (C) position. Smaller points represent raw data from each sample. Sample sizes: internal = 5, north = 7, west = 6, south = 4, east = 6; lower = 11, middle = 9, upper = 8; left = 8, centre = 14, right = 6. An * indicates a significant F-test result.

Full-size DOI: 10.7717/peerj.16355/fig-2

exist when these genera are inhabiting areas that are not frequently in direct contact with humans. Regarding fungal genera, it is possible that they may remain dormant and only become problematic when certain environmental changes occur, such as excess moisture. But identifying the presence of these microbial genera is important because it provides valuable insights into the potential risks and could allow for mitigation of problems.

As the microbial communities observed in household surfaces can be sourced from building occupants (*Cao et al.*, 2021), or the surrounding environment (*Barberán et al.*, 2015a) and geographic location (*Chen et al.*, 2017), we wanted to examine the potential impact of the indoor and outdoor microbial communities on the communities found within the walls. Source tracking revealed that the fungal communities found within the wall are likely to be independent of the communities found in the indoor and outdoor air (Fig. S3). Within the bacterial communities, half of the wall samples had more than 50% of ASVs that could not be sourced to the air samples (Fig. S4), while the remaining had more

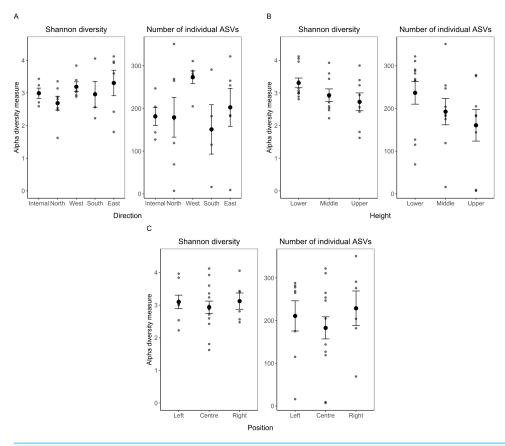


Figure 3 16S alpha diversity analysis: The average Shannon diversity index and number of individual ASVs (±standard error of mean). Comparing (A) direction, (B) position, (C) height. Smaller points represent raw data from each sample. Sample sizes: internal = 5, north = 7, west = 6, south = 4, east = 6; lower = 11, middle = 9, upper = 8; left = 8, centre = 14, right = 6.

Full-size DOI: 10.7717/peerj.16355/fig-3

of a mix of air and unknown sources (Fig. S4). Of the top fifteen bacterial genera (Fig. S4), a number of these could be traced to human as a source, such as *Faecalbacterium* (*Bai et al.*, 2023), *Blaudia* (*Dobay et al.*, 2019) and *Cutibacterium* (*Sun et al.*, 2022), or could be traced to soil, for example *Sphingomonas* (*White, Sutton & Ringelberg, 1996*) and *Acinetobacter* (*Hui et al.*, 2019). Understanding when microbes could colonize the building, such as during material processing or when being stored on a building site, and investigating if their relative abundance increases over time, would provide information on their potential sources and dynamics within the built environment. This knowledge could contribute to a better understanding of the factors influencing the microbiota of a building. Additionally, as many microbes have dormant stages, it would be interesting to determine which of these taxa are viable.

We next quantified patterns in beta diversity (the diversity between two microbial communities). To determine the sampling variable most influencing differences in communities, we analysed Brays-Curtis dissimilarity (index based on the abundance of individual ASV groups) and Jaccard distance (index based on the presence and absence

Table 2 Statistical results from the ADONIS tests for the beta diversity of fungal and bacteria communities. Distance matrices Brays—Curtis dissimilarity and Jaccard were both tested with each variable used as the predictor variable.

		Brays-	Curtis	Jaccard		
Kingdom	Variable	R-squared	P-value	R-squared	P-value	
	Direction	0.16	0.31	0.15	0.35	
Fungi	Height	0.08	0.71	0.07	0.68	
	Position	0.07	0.67	0.07	0.74	
	Direction	0.21	0.08	0.19	0.06	
Bacteria	Height	0.03	0.90	0.04	0.97	
	Position	0.08	0.25	0.07	0.41	

of individual ASV groups). For both measures, direction particularly caused variation across the fungal and bacterial communities (Table 2, Figs. 4A, 5B). As with height, differences in macroenvironment will influence the survival and selection of different microbial taxa (*Storze & Hengge*, 2011; *Haruta & Kanno*, 2015). Variation among walls in their exposure to wind and/or sun can generate differences in envelope temperature that could explain variation in the community composition.

To investigate how the most abundant ASVs contributed to the variability, we examined the differential abundance patterns of the top fifteen genera across the cardinal directions (see Figs. S1 and S2 for fungi and bacteria, respectively). To do this, we used similarity percentages (Clarke, 1993) and identified the genera that most contributed to the betadiversity measures based on Bray-Curtis dissimilarity indices. Among the top fifteen relatively abundant genera, eleven fungal and ten bacterial genera were present across the pairwise comparisons and were therefore considered to be influencing the observed variations (Table S1). A Kruskal-Wallis and subsequent Dunn test on the eleven fungal genera found a significant difference in the relative abundance of Ceriporiopsis; the relative abundance was significantly higher in west-facing walls compared to east (Table S2). The relative abundance of the bacterial genus Bradyrhizobium was significantly higher in internal walls compared to north-facing walls (Table S2). A limitation with our data, though, is sample size, and a larger dataset would have greater statistical power. We would expect buildings with different uses and built using different materials to exhibit different patterns across the structure. Further research involving a diverse range of building types and materials would provide an interesting and comprehensive understanding of the factors driving compositional changes in building microbiota. Additionally, a further direction could be to also quantify the microbial load at different locations, although this would be challenging to do so (Galazzo et al., 2020). Nonetheless, despite being limited to one building, we show the occurrence of compositional changes in microbiota, even across a relatively small building. As such, our data shows how in-wall sampling needs to encompass multiple locations within and across different regions of any building element to avoid sampling bias when studying the building microbiota.

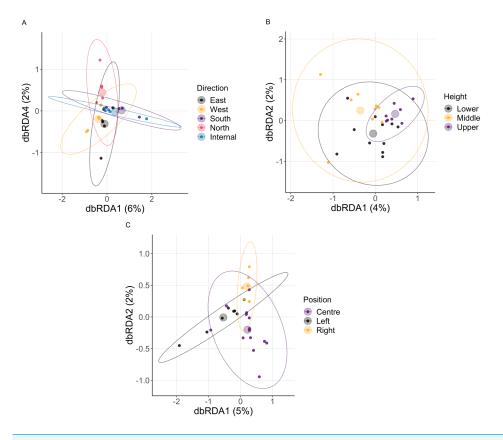


Figure 4 Fungi beta diversity analysis. The ordination was obtained by conducting a distance-based redundancy analysis based on Brays–Curtis dissimilarity matrices for (A) direction, (B) height, (C) position. Each smaller point represents the fungal community in a sample. Ellipses represent a 95% CI centred around a centroid shown by a transparent, larger point. Sample sizes: internal = 5, north = 7, west = 6, south = 4, east = 6; lower = 11, middle = 9, upper = 8; left = 8, centre = 14, right = 6.

Full-size DOI: 10.7717/peerj.16355/fig-4

CONCLUSION

Here we present a case study using an overlooked approach in sampling and understanding the processes that determine the composition of the building microbiota. Our aim was to characterize the microbial communities within the building envelope and determine which building elements, if any, have the largest impact on compositional variation found within structures. Our results show significant differences in the alpha diversity across different heights of the wall and that factors such as cardinal direction can elicit variation in the community composition. While our study focused on one building, we show the potential for diverse microbiota across the building envelope and that to get a fuller picture of the microbiota of the built environment, variation across height or the cardinal direction of the building must be taken into account during sampling. The identification of independent communities within the walls shows that future investigations should therefore think of a building as its own ecosystem amongst the indoor biome (*Adams et al.*, 2015). Studies should take into consideration these 'hidden' microbial communities that

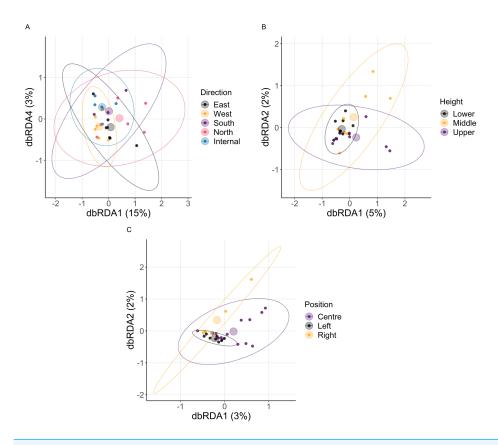


Figure 5 Bacteria beta diversity analysis. The ordination was obtained by conducting a distance-based redundancy analysis based on Brays–Curtis dissimilarity matrices for (A) direction, (B) height, (C) position. Each smaller point represents the bacterial community in a sample. Ellipses represent a 95% CI centred around a centroid shown by a transparent, larger point. Sample sizes: internal = 5, north = 7, west = 6, south = 4, east = 6; lower = 11, middle = 9, upper = 8; left = 8, centre = 14, right = 6.

Full-size DOI: 10.7717/peerj.16355/fig-5

have the potential to cause damage to buildings and cause problems to occupants' health while also accounting for microenvironmental changes across building structures. These preliminary findings serve as the foundation for expanding our approach and delving deeper into investigating the microbiotas beneath the surface.

ACKNOWLEDGEMENTS

We would like to thank Pekka Ylimäki for his assistance in sampling the cottage and Ilze Brila for her valuable guidance on the analysis. The authors would also like to thank the reviewers for their time. We sincerely appreciate their valuable comments and expertise, which helped improve the quality of the manuscript.

ADDITIONAL INFORMATION AND DECLARATIONS

Funding

The work was funded by The Academy of Finland; award no. 329883 to Phillip Watts as part of the CLIHE programme. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Grant Disclosures

The following grant information was disclosed by the authors: The Academy of Finland: 329883.

Competing Interests

The authors declare there are no competing interests.

Author Contributions

- Lucy R. Davies conceived and designed the experiments, performed the experiments, analyzed the data, prepared figures and/or tables, authored or reviewed drafts of the article, and approved the final draft.
- Aitor Barbero-López conceived and designed the experiments, performed the
 experiments, analyzed the data, authored or reviewed drafts of the article, and approved
 the final draft.
- Veli-Matti Lähteenmäki conceived and designed the experiments, performed the
 experiments, analyzed the data, authored or reviewed drafts of the article, and approved
 the final draft.
- Antti Salonen analyzed the data, authored or reviewed drafts of the article, and approved the final draft.
- Filip Fedorik conceived and designed the experiments, performed the experiments, analyzed the data, authored or reviewed drafts of the article, and approved the final draft.
- Antti Haapala conceived and designed the experiments, performed the experiments, analyzed the data, authored or reviewed drafts of the article, and approved the final draft.
- Phillip C. Watts conceived and designed the experiments, performed the experiments, analyzed the data, authored or reviewed drafts of the article, and approved the final draft.

Data Availability

The following information was supplied regarding data availability:

The 16S and ITS raw sequences are available at GenBank: PRJNA958164.

The metadata is available at figshare: Davies, Lucy (2023). Metadata. figshare. Dataset. https://doi.org/10.6084/m9.figshare.22059653.v2.

The code for analysis is available at figshare: Davies, Lucy (2023). R script for analysis. figshare. Online resource. https://doi.org/10.6084/m9.figshare.22059686.v2.

Supplemental Information

Supplemental information for this article can be found online at http://dx.doi.org/10.7717/peerj.16355#supplemental-information.

REFERENCES

- Adams RI, Bateman A, Bik HM, Hawks J, Hird SM, Hughes D, Kembel SW, Kinney K, Kolokotronis SO, Levy G, McClain C, Meadow JF, Medina RF, Mhuireach G, Moreau CS, Munshi-South J, Nichols LM, Palmer C, Popova L, Schal C, Täubel M, Trautwein M, Ugalde JA, Dunn RR. 2015. Evolution of the indoor biome. *Trends in Ecology & Evolution* 30(4):223–232 DOI 10.1016/j.tree.2015.02.001.
- **Airaksinen M, Kurnitski J, Pasanen P, Seppänen O. 2004.** Microbial growth in building material samples and occupants' health in severely moisture-damaged homes. *Indoor Air* **14(2)**:92–104 DOI 10.1046/j.1600-0668.2003.00215.x.
- Bai Z, Zhang N, Jin Y, Chen L, Mao Y, Sun L, Fang F, Liu Y, Han M, Li G. 2023. Comprehensive analysis of 84 *Faecalibacterium prausnitzii* strains uncovers their genetic diversity, functional characteristics, and potential risks. *Frontiers in Cellular and Infection Microbiology* 12:919701 DOI 10.3389/fcimb.2022.919701.
- Barberán A, Dunn RR, Reich BJ, Pacifici K, Laber EB, Menninger HL, Morton JM, Henley JB, Leff JW, Miller SL, Fierer N. 2015b. The ecology of microscopic life in household dust. *Proceedings of the Royal Society B: Biological Sciences* 282:20151139 DOI 10.1098/rspb.2015.1139.
- Barberán A, Ladau J, Leff JW, Pollard KS, Menninger HL, Dunn Fierer RRN. 2015a.

 Continental-scale distributions of dust-associated bacteria and fungi. *Proceedings of the National Academy of Sciences of the United States of America* 112(18):5756–5761

 DOI 10.1073/pnas.1420815112.
- Borowik A, Wyszkowska J. 2016. Soil moisture as a factor affecting the microbiological and biochemical activity of soil. *Plant, Soil and Environment* 26(6):250–255 DOI 10.17221/158/2016-PSE.
- Callahan BJ, McMurdie PJ, Rosen MJ, Han AW, Johnson AJA, Holmes SP. 2016.

 DADA2: high-resolution sample inference from Illumina amplicon data. *Nature Methods* 13:581–583 DOI 10.1038/nmeth.3869.
- Cao L, Yang L, Swanson CS, He Q. 2021. Comparative analysis of impact of human occupancy on indoor microbiomes. *Frontiers of Environmental Science & Engineering* 15:89 DOI 10.1007/s11783-020-1383-1.
- Chase J, Fouquier J, Zare M, Sonderegger DL, Knight R, Kelly ST, Siegel J, Caporaso JG. 2016. Geography and location are the primary drivers of office microbiome composition. *mStystems* 1(2):e00022-16 DOI 10.1128/mSystems.00022-16.
- Checinska Sielaff A, Urbaniak C, Mohan GBM, Stepanov VG, Tran Q, Wood JM, Minich J, McDonald D, Mayer T, Knight R, Karouia F, Fox GE, Venkateswaran K. 2019. Characterization of the total and viable bacterial and fungal communities associated with the International Space Station surfaces. *Microbiome* 7:50 DOI 10.1186/s40168-019-0666-x.

- Chen AJ, Hubka V, Visgagie CM, Houbraken J, Meijer M, Varga J, Demirel R, Jurjević Ž, Kubátová A, Sklenář F, Zhou YG, Samson RA. 2017. Polyphasic taxonomy of Aspergillus section Aspergillus (formerly Eurotium), and its occurrence in indoor environments and food. *Studies in Mycology* 88:37–135 DOI 10.1016/j.simyco.2017.07.001.
- Clarke KR. 1993. Non-parametric multivariate analyses of changes in community structure. *Austral Ecology* 18:117–143 DOI 10.1111/j.1442-9993.1993.tb00438.x.
- Dannemiller KC, Gent JF, Leaderer BP, Peccia J. 2016. Indoor microbial communities: influence on asthma severity in atopic and nonatopic children. *Journal of Allergy and Clinical Immunology* 138:76–83 DOI 10.1016/j.jaci.2015.11.027.
- Davis NM, Proctor DM, Holmes SP, Relman DA, Callahan BJ. 2018. Simple statistical identification and removal of contaminant sequences in marker-gene and metagenomics data. *Microbiome* 6:226 DOI 10.1186/s40168-018-0605-2.
- **Dinno A. 2017.** dunn.test: Dunn's test of multiple comparisons using rank sums. R package version 1.3.5. *Available at https://CRAN.R-project.org/package=dunn.test* (accessed on May 2023).
- Dobay A, Haas C, Fucile G, Downey N, Morrison HG, Kratzer A, Arora N. 2019.

 Microbiome-based body fluid identification of samples exposed to indoor conditions. *Forensic Science International: Genetics* **40**:105–113

 DOI 10.1016/j.fsigen.2019.02.010.
- **Dunn OJ. 1961.** Multiple comparisons among means. *JASA* **56**:52–64 DOI 10.1080/01621459.1961.10482090.
- Fahimipour A, Hartmann E, Siemens A, Kline J, Levin DA, Wilson H, Betancourt-Roman CM, Brown GZ, Fretz M, Northcutt D, Siemens KN, Huttenhower C, Green JL, Ven Den Wymelenberg K. 2018. Daylight exposure modulates bacterial communities associated with household dust. *Microbiome* 6:175 DOI 10.1186/s40168-018-0559-4.
- **Fedorik F, Alitalo S, Savolainen J-P, Räina I, Illikainen K. 2021.** Analysis of hygrothermal performance of low-energy house in Nordic climate. *Journal of Building Physics* **45**:344–367 DOI 10.1177/1744259120984187.
- Frankel M, Bekö G, Timm M, Gustavsen S, Hansen EW, Madsen AM. 2012. Seasonal variations of indoor microbial exposures and their relation to temperature, relative humidity, and air exchange rate. *Applied and Environmental Microbiology* 78(23):8289–8297 DOI 10.1128/AEM.02069-12.
- Fu X, Li Y, Meng Y, Yuan Q, Zhang Z, Wen H, Deng Y, Norbäck D, Hu Q, Zhang X, Sun Y. 2021. Derived habitats of indoor microbes are associated with asthma symptoms in Chinese university dormitories. *Environmental Research* 194:110501 DOI 10.1016/j.envres.2020.110501.
- Fu X, Norbäack D, Yuan Q, Li Y, Zhu X, Hashim JH, Hashim Z, Ali F, Zheng Y-W, Lai X-X, Spangfort M, Deng Y, Sun Y. 2020. Indoor microbiome, environmental characteristics and asthma among junior high school students in Johor Bahru, Malaysia. *Environment International* 138:105664

 DOI 10.1016/j.envint.2020.105664.

- **Gabriel J, Švec K. 2017.** Occurrence of indoor wood decay basidiomycetes in Europe. *Fungal Biology Reviews* **31**:212–217 DOI 10.1016/j.fbr.2017.05.002.
- Galazzo G, van Best N, Benedikter BJ, Janssen K, Bervoets L, Driessen C, Oomen M, Lucchesi M, van Eijck PH, Becker HEF, Hornef MW, Savelkoul PH, Stassen FRM, Wolffs PF, Penders J. 2020. How to count our microbes? The effect of different quantitative microbiome profiling approaches. *Frontiers in Cellular and Infection Microbiology* 10:403 DOI 10.3389/fcimb.2020.00403.
- Garrett MH, Hooper BM, Cole FM, Hooper MA. 1997. Airborne fungal spores in 80 homes in the Latrobe Valley, Australia: levels, seasonality and indoor-outdoor relationship. *Aerobiologia* 13:121–126 DOI 10.1007/BF02694428.
- Haas D, Mayrhofer H, Habib J, Galler H, Reinthaler FF, Fuxjäger ML, Buzina W. 2019. Distribution of building-associated wood-destroying fungi in the federal state of Styria, Austria. *European Journal of Wood and Wood Products* 77:527–537 DOI 10.1007/s00107-019-01407-w.
- Haruta S, Kanno N. 2015. Survivability of microbes in natural environments and their ecological impacts. *Microbes and Environments* 30:123–125 DOI 10.1264/jsme2.ME3002rh.
- **Hedayati MT, Mayahi S, Denning DW. 2010.** A study on *Aspergillus* species in houses of asthmatic patients from Sari City, Iran and a brief review of the health effects of exposure to indoor *Aspergillus*. *Environmental Monitoring and Assessment* **168**:481–487 DOI 10.1007/s10661-009-1128-x.
- Horve PF, Dietz LG, Ishaq SL, Kline J, Fretz M, Van Den Wymelenberg KG. 2020. Viable bacterial communities on hospital window components in patient rooms. *PeerJ* 8:e9580 DOI 10.7717/peerj.9580.
- **Huckfeldt T, Schmidt O. 2006.** Identification key for European strand-forming houserot fungi. *Mycologist* **20(2)**:42–56 DOI 10.1016/j.mycol.2006.03.012.
- Hui N, Parajuli A, Puhakka R, Grönroos M, Roslun MI, Vari HK, Selonen VAO, Yan G, Siter N, Nurminen N, Oikarinen S, Laitinen OH, Rajaniemi J, Hyöty H, Sinkkonen A. 2019. Temporal variation in indoor transfer of dirt-associated environmental bacteria in agricultural and urban areas. *Environment International* 132:105069 DOI 10.1016/j.envint.2019.105069.
- Hwang Y, Schulze-Makuch D, Arens FL, Saenz JS, Adam PS, Sager C, Bornemann TLV, Zhao W, Zhang Y, Airo A, Schloter M, Probst AJ. 2021. Leave no stone unturned: individually adapted xerotolerant *Thaumarchaeota* sheltered below the boulders of the Atacama Desert hyperarid core. *Microbiome* 9:234 DOI 10.1186/s40168-021-01177-9.
- Järvi K, Hyvärinen A, Täubel M, Karvonen AM, Turunen M, Jalkanen K, Patovirta R, Syrjänen T, Pirinen J, Salonen H, Nevalainen A, Pekkanen J. 2018. Microbial growth in building material samples and occupants' health in severely moisture-damaged homes. *Indoor Air* 28:287–297 DOI 10.1111/ina.12440.
- **Kabátováa V, Ďuricaa P. 2019.** Measured and simulated temperature values in the chosen wall of a wooden building considering cardinal direction. *Transp* **40**:718–723 DOI 10.1016/j.trpro.2019.07.101.

- **Knudsen SM, Gunnarsen L, Madsen AM. 2017.** Airborne fungal species associated with mouldy and non-mouldy buildings—effects of air change rates, humidity, and air velocity. *Building and Environment* **122**:161–170 DOI 10.1016/j.buildenv.2017.06.017.
- **Kozajda A, Jeżak K, Kapsa A. 2019.** Airborne Staphylococcus aureus in different environments—a review. *Environmental Science and Pollution Research* **26**:34741–34753 DOI 10.1007/s11356-019-06557-1.
- **Legendre P, Anderson MJ. 1999.** Distance-based redundancy analysis: testing multispecies responses in multifactorial ecological experiments. *Ecological Monographs* **69**:1–24 DOI 10.1890/0012-9615(1999)069[0001:DBRATM]2.0.CO;2.
- Madsen AM, Moslehi-Jenabian S, Islam MDZ, Frankel M, Spilak M, Frederiksen MW. 2018. Concentrations of Staphylococcus species in indoor air as associated with other bacteria, season, relative humidity, air change rate, and S. aureus-positive occupants. *Environmental Research* 160:282–291 DOI 10.1016/j.envres.2017.10.001.
- Maestre JP, Jennings W, Wylie D, Horner SD, Siegel J, Kinney KA. 2018. Filter forensics: microbiota recovery from residential HVAC filters. *Microbiome* **6**:22 DOI 10.1186/s40168-018-0407-6.
- Marcel M. 2011. Cutadapt removes adapter sequences from high-throughput sequencing reads. *EMBnet.Journal* [S.L.] v. 17(n. 1):10–12 DOI 10.14806/ej.17.1.200.
- **McMurdie PJ, Holmes S. 2013.** Phyloseq: an R package for reproducible interactive analysis and graphics of microbiome census data. *PLOS ONE* **8**:e61217 DOI 10.1371/journal.pone.0061217.
- **Moon KW, Huh EH, Jeong HC. 2014.** Seasonal evaluation of bioaerosols from indoor air of residential apartments within the metropolitan area in South Korea. *Environmental Monitoring and Assessment* **186**:2111–2120 DOI 10.1007/s10661-013-3521-8.
- Mousavi B, Hedayati MT, Hedayati N, Ilkit M, Syedmousavi S. 2016. Aspergillus species in indoor environments and their possible occupational and public health hazards. *Current Medical Mycology* 1:36–42 DOI 10.18869/acadpub.cmm.2.1.36.
- Nilsson RH, Tedersoo L, Ryberg M, Kristiansson E, Hartmann M, Unterseher M, Porter TM, Bengtsson-Palme J, Walker DM, de Sousa F, Gamper HA, Larsson E, Larsson KH, Kõljalg U, Edgar RC, Abarenkov K. 2015. A comprehensive, automatically updated fungal ITS sequence dataset for reference-based chimera control in environmental sequencing efforts. *Microbes and Environments* 30(2):145–150 DOI 10.1264/jsme2.ME14121.
- Nottingham AT, Scott JJ, Saltonstall K, Broders K, Montero-Sanchez M, Puspök J, Bååth E, Meir P. 2022. Microbial diversity declines in warmed tropical soil and respiration rise exceed predictions as communities adapt. *Nature Microbiology* 7:1650–1660 DOI 10.1038/s41564-022-01200-1.
- Oksanen J, Guillaume Blanchet F, Friendly M, Kindt R, Legendre P, McGlinn D, Minchin P, O'Hara B, Simpson GL, Solymos P, Stevens MHH, Szöcs E, Wagener H. 2019. vegan: community ecology package. R package 2.5-7. *Available at https:*//CRAN.R-project.org/package=vegan (accessed on January 2022).
- **Pedersen TL. 2022.** patchwork: the composer of plots. R package version 1.1.2. *Available at https://CRAN.R-project.org/package=patchwork* (accessed on December 2022).

- Pekkanen J, Valkonen M, Täubel M, Tischer C, Leppänen H, Kärkkäinen PM, Rintala H, Zock J-P, Casas L, Probst-Hensch N, Forsberg B, Holm M, Janson C, Pin I, Gislason T, Jarvis D, Heinrich J, Hyvärinen A. 2018. Indoor bacteria and asthma in adults: a multicentre case—control study within ECRHS II. *European Respiratory Journal* 51(2):1701241 DOI 10.1183/13993003.01241-2017.
- Pessi AM, Suonketo J, Pentti M, Kurkilahti M, Peltola K, Rantio-Lehtimäki A. 2002. Microbial growth inside insulated external walls as an indoor air biocontamination source. *AEM* **68**:963–967 DOI 10.1128/AEM.68.2.963-967.2002.
- Ponizovskaya VB, Rebrikova NL, Kachalkin AV, Antropova AB, Bilanenko EN, Mokeeva VL. 2019. Micromycetes as colonizers of mineral building materials in historic monuments and museums. *Fungal Biology* 123:290–306 DOI 10.1016/j.funbio.2019.01.002.
- Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, Peplies J, Glöckner F. 2013. The SILVA ribosomal RNA gene database project: improved data processing and web-based tools. *Nucleic Acids Research* 41(D1):590–596 DOI 10.1093/nar/gks1219.
- **R Core Team. 2021.** R: a language and environment for statistical computing. R Foundation for Statistical Computing. *Available at https://www.R-project.org/* (accessed on January 2022).
- **Rintala H, Pitkäranta M, Toivola M, Paulin L, Nevalainen A. 2008.** Diversity and seasonal dynamics of bacterial community in indoor environment. *BMC Microbiology* **8**:56 DOI 10.1186/1471-2180-8-56.
- **Schmidt O. 2007.** Indoor wood-decay basidiomycetes: damage, causal fungi, physiology, identification and characterization, prevention and control. *Mycological Progress* **6**:261–279 DOI 10.1007/s11557-007-0534-0.
- **Schmidt O, Huckeldt . 2011.** Characteristics and identification of indoor wood-decaying basidiomycetes. In: *Fundamentals of mold growth in indoor environments and strategies for healthy living.* Wageningen: Wageningen Academic Publishers DOI 10.3920/978-90-8686-722-6_6.
- Scholier T, Lavrinienko A, Brila I, Tukalenko E, Hindström R, Vasylenko A, Cayol C, Ecke F, Sing NJ, Forsman JT, Tolvanen A, Matala J, Huitu O, Kallio ER, Koskela E, Mappes T, Watts PC. 2022. Urban forest soils harbour distinct and more diverse communities of bacteria and fungi compared to less disturbed forest soils. *Molecular Ecology* 32:504–517 DOI 10.1111/mec.16754.
- Schulze-Makuch D, Wagner D, Kounaves SP, Mangelsdorf K, Deine KG, de Vera J-P, Scmitt-Kopplin P, Grossart H-P, Parro V, Kaupenjohann M, Galy A, Schneider B, Airo A, Frösler J, Davila AF, Arens FL, Carceres L, Cornjeo FS, Carrizo D, Dartnell L, Di Ruggiero J, Flury M, Ganzert L, Gessner MO, Grathwohl P, Guan L, Heinz J, Hess M, Keppler F, Maus D, McKay CP, Meckenstock RU, Montgomery W, Oberlin EA, Probst AJ, Saenz JS, Sattler T, Shirmack J, Sephton MA, Schloter M, Uhl J, Valenzuela B, Vestergaard G, Wörmer L, Zamorano P. 2018.

 Transitory microbial habitat in the hyperarid Atacama Desert. *Proceedings of the*

- National Academy of Sciences of the United States of America 115(11):2670–2675 DOI 10.1073/pnas.1714341115.
- Shenhav L, Thompson M, Jospeh TA, Briscoe L, Furman O, Bogumil D, Mizrahi I, Pe'er I, Halperin E. 2019. FEAST: fast expectation—maximization for microbial source tracking. *Nature Methods* 16:627–632 DOI 10.1038/s41592-019-0431-x.
- **Storze G, Hengge R. 2011.** *Bacterial stress responses.* Washington, D.C.: ASM Press DOI 10.1128/9781555816841.
- Sun Y, Meng Y, Ou Z, Li Y, Zhang M, Chen Y, Zhang Z, Chen X, Mu P, Norbäck D, Zhao Z, Zhang X, Fu X. 2022. Indoor microbiome, air pollutants and asthma, rhinitis and eczema in preschool children—a repeated cross-sectional study. *Environment International* 161:107137 DOI 10.1016/j.envint.2022.107137.
- Tanaka-Kagawa T, Uchiyama S, Matsushima E, Sasaki A, Kobayashi H, Kobayashi H, Yagi M, Tsuno M, Arao M, Ikemoto K, Yamasaki M, Nakashima A, Shimizu Y, Otsubo Y, Ando M, Jinno H, Tokunaga H. 2005. Survey of volatile organic compounds found in indoor and outdoor air samples from Japan. Kokuritsu Iyakuhin Shokuhin Eisei Kenkyujo Hokoku = Bulletin of National Institute of Health Sciences 123:27–31.
- Wang S, Zuo X, Awada T, Medima-Roldan E, Freng K, Yue P, Lian J, Zhao S. 2021. Changes of soil bacterial and fungal community structure along a natural aridity gradient in desert grassland ecosystems, Inner Mongolia. *Catena* 205:105470 DOI 10.1016/j.catena.2021.105470.
- White DC, Sutton SD, Ringelberg DB. 1996. The genus Sphingomonas: physiology and ecology. *Current Opinion in Biotechnology* **7(3)**:301–306 DOI 10.1016/S0958-1669(96)80034-6.
- **Wickham H. 2016.** ggplot2: elegant graphics for data analysis. New York: Springer-Verlag. *Available at https://ggplot2.tidyverse.org* (accessed on January 2022).
- Wu Z, Lyu H, Ma X, Ren G, Song J, Jing X, Liu Y. 2022. Comparative effects of environmental factors on bacterial communities in two types of indoor dust: potential risks to university students. *Environmental Research* 203:111869 DOI 10.1016/j.envres.2021.111869.