Metagenomic Insights into the Bioaerosols in the Indoor and Outdoor Environments of Childcare Facilities

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Abstract

Airborne microorganisms have significant effects on human health, and children are more vulnerable to pathogens and allergens than adults. However, little is known about the microbial communities in the air of childcare facilities. Here, we analyzed the bacterial and fungal communities in 50 air samples collected from five daycare centers and five elementary schools located in Seoul, Korea using culture-independent high-throughput pyrosequencing. The microbial communities contained a wide variety of taxa not previously identified in child daycare centers and schools. Moreover, the dominant species differed from those reported in previous studies using culture-dependent methods. The well-known fungi detected in previous culture-based studies (Alternaria, Aspergillus, Penicillium, and Cladosporium) represented less than 12% of the total sequence reads. The composition of the fungal and bacterial communities in the indoor air differed greatly with regard to the source of the microorganisms. The bacterial community in the indoor air appeared to contain diverse bacteria associated with both humans and the outside environment. In contrast, the fungal community was largely derived from the surrounding outdoor environment and not from human activity. The profile of the microorganisms in bioaerosols identified in this study provides the fundamental knowledge needed to develop public health policies regarding the monitoring and management of indoor air quality.

Introduction

There are numerous microorganisms in the air we breathe; the number of bacterial cells and fungal spores present in air is estimated to be $10^{4}$--$10^{6}$ m$^{-3}$ [1,2] and $10^{3}$--$10^{4}$ m$^{-3}$ [3–5], respectively. Although the atmosphere is an extreme environment for microorganisms because of high levels of solar radiation, low moisture, and low nutrient content, many of these airborne microbes are metabolically active [6,7]. Both the metabolically active and inactive airborne...
microbes have clinically important effects on human health. For example, exposure to airborne fungal allergens such as hyphae, spores, and fungal fragments can cause asthma, rhinitis, atopic dermatitis, and allergic diseases [8–12]. Young children have higher susceptibility to bioaerosols than adults; hence, they are more vulnerable than adults to the diseases caused by microorganisms in bioaerosols. In addition, they breathe more air than adults (per kilogram of body weight), which may result in a higher daily intake of the pathogens or allergens present in air [13,14]. In Korea, children aged 4–9 years generally attend daycare centers or elementary schools for one quarter of each day. Therefore, determining the components of the bioaerosols in these spaces is important for improving the public health of children.

For a long time, culture-based microbiological methods have been used to investigate airborne microbial communities, and these methods have enabled direct observation of airborne bacteria and fungi. Culture-based studies have shown that most airborne bacteria are typically gram-positive bacteria belonging to the genera *Staphylococcus*, *Corynebacterium*, *Bacillus*, and *Micrococcus* [15–21]. However, some gram-negative bacteria, such as *Acinetobacter*, *Moraxella*, *Pantoea*, and *Pseudomonas* were also isolated at lower frequencies [19]. The dominant airborne fungi detected belonged to the genera *Aspergillus*, *Penicillium*, *Cladosporium*, and *Alternaria* [15,22,23]. However, culture-based studies can only provide an estimate of the true diversity of the microbial communities in air because most (>99%) microorganisms in the environment are recalcitrant to culturing [24,25]. In fact, the composition of the airborne microbial communities determined by culture-dependent and culture-independent 16S rRNA gene sequencing differ considerably [26,27].

Since the 1990s, new molecular taxonomic techniques for fungi have been introduced, and DNA sequence databases have been developed to facilitate fungal identification. These molecular identification tools have provided new insights into fungal ecology. In recent culture-independent studies of airborne fungi, the ITS region was chosen as a target molecular marker [28–30]. Yamamoto *et al.* [31] identified 558 different genera in outdoor air samples by analyzing the entire ITS region. Fröhlich-Nowoisky *et al.* [32] used DNA sequence analysis to efficiently detect and unambiguously characterize fungi in atmospheric aerosol samples. The ITS region is currently the most phylogenetically informative sequence for fungal identification, especially at lower taxonomic levels.

In recent years, metagenomic approaches have provided deeper insights into airborne microbial diversity in different environments, such as urban environments [31,33–36], subway systems [37,38], museums [39], buildings [40,41], and air particles [32,42]. However, the bioaerosols in children’s daytime spaces have not been extensively examined in metagenomic studies. Thus, we investigated the diversity of microorganisms present in the air of education facilities for children in Seoul, Korea using 454 pyrosequencing. The bacterial and fungal populations in 50 indoor and outdoor air samples from five daycare centers and five elementary schools were analyzed. The purpose of this study was (1) to characterize the microbial composition of the air at locations frequented by children, and (2) to examine the association between indoor and outdoor bioaerosols. The results of this study will improve our understanding of airborne microorganisms and provide a rationale for developing public health policies regarding the monitoring and management of indoor air quality.

**Materials and Methods**

**Sample collection**

Indoor and outdoor air samples were collected at five daycare centers (sites A to E) and five elementary schools (sites F to J) located in Seoul, Korea from August 19 to October 24, 2013. The presidents of the elementary schools and the directors of the daycare centers were provided
official documents from the Research and Business Foundation of Korea University requesting their cooperation, and they gave permission to conduct the study at each site. The results of this study were communicated to the daycare centers and elementary schools. Information about the investigated daycare centers and schools is shown in Table 1. The facilities were selected because the buildings are widely distributed in Seoul and are nearly equidistant to each other. The evaluated daycare centers and schools varied in building construction, size, surrounding environment, and the number of children served. Ventilation was provided naturally through windows. The children performed their normal activities during the measurements. Temperature and relative humidity were monitored during the sampling periods. The microbiological sampling cycle was of 10 h (from 8 am to 6 pm). To obtain duplicate or triplicate samples from the same site, the sampling sessions were conducted for 2 subsequent days (24 h periods) for the daycare centers and 3 subsequent days for the elementary schools. Samples were obtained using an air sampler equipped with Millipore cassettes (47 mm diameter) with sterile mixed cellulose ester filters (47 mm; 0.45 μm; Millipore). Although microsized (<0.45 μm) bacteria could pass through the 0.45 μm pores, we chose this filter based on the results of previous studies [36,43,44] and initial testing, in which we verified an increase in bacterial mass by agglomeration of abiotic or biotic particles on the filter membrane during sampling. The air sampler was operated at an airflow rate of 24 L/min, which was maintained by a vacuum pump (BMW-200; Total Eng) and verified with a calibrator (TSI-4045 Mass Flowmeter; TSI) during each 24-h measurement. For indoor air measurement, the air sampler cassette was fixed on a tripod 1.5 m above the floor in the rear of the classroom. The pumps for the indoor air samplers were placed inside noise-insulated enclosures to reduce the noise exposure of the occupants. The outdoor air measurements were obtained approximately 1.5 m above the ground outside the building, avoiding areas traversed by people. Before sampling, the empty sampling cassettes were sterilized in an autoclave and then dried in a drying oven. The samples were transported directly to the laboratory under chilled conditions (4°C). Then, the filters were placed in sterile plastic tubes and frozen at -20°C until DNA extraction.

DNA extraction, PCR, and pyrosequencing

DNA was extracted directly from each air sampling filter using a commercial soil DNA isolation kit (MP Biomedicals). The extracted DNA was amplified using primers targeting the V1–V3 regions of the prokaryotic 16S rRNA gene as previously described [45]. The primers used for bacteria were V1-9F (5′-CCATATGGCTGCCTGAGTCAC-TGACATCAG-3’ [the underlined sequence is the gene-specific region]) and V3-541R (5′-CCATCTCATACGCTGTCGTCGTCGAC-TCAG-barcode-ACCTGAGTTCGCTGAGT-3’ [45]). The ITS regions of the fungal rRNA operon were amplified using fusion primers ITS-3F (5′-CCATATGGCTGCCTGAGTCAC-TGACATCAG-3’ [the underlined sequence is the gene-specific region]) and ITS-5R (5′-CCATCTCATACGCTGTCGTCGTCGAC-TCAG-barcode-ACCTGAGTTCGCTGAGT-3’). The fusion primers contained 454-specific adapters, keys, linkers, barcodes, and universal fungal ITS priming sequences [46]. The barcode and primer sequences are available at http://www.ezbiocloud.net/resource/M1001. The PCR for both bacteria and fungi was performed under the following cycling conditions: an initial denaturation step at 94°C for 5 min, followed by 10 cycles of denaturation at 94°C for 30 s, annealing at 60°C to 55°C (with a touchdown program) for 45 s, and elongation at 72°C for 90 s. This was followed by an additional 20 cycles of denaturation at 94°C for 30 s, annealing at 55°C for 45 s, and elongation at 72°C for 90 s, and a final elongation step at 72°C for 5 min. The sizes of the amplicons were 500 bp–700 bp for bacteria and 600 bp–800 bp for fungi, respectively. The amplified products were purified using resin columns (Qiagen), and 1 μg of the PCR
product from each sample was mixed and purified using the AMPure bead kit (Agencourt Bioscience). The DNA was sequenced unidirectionally from universal primers (518R for bacteria and ITS-4R for fungi) at Chunlab, Inc. with a Roche/454 GS Junior system according to the manufacturer’s instructions. The sequencing data from this study were deposited in the Short Read Archive under accession number SRP043178 (Bioproject accession number PRJNA252641).

### Processing of sequencing data

The pyrosequencing data for the 16S rRNA gene sequences was processed through Java-based multi-step bioinformatics pipelines as described elsewhere [45,47–49]. The unidirectional sequencing reads from different samples were separated by their unique barcodes. To filter low quality sequences, reads <300 bp or with an average quality score <25 were omitted. Then the barcode, linker, and PCR primer sequences were removed from both sides of the reads using pairwise sequence alignment and the hmm-search program in HMMER 3.0 [50]. The trimmed sequencing reads were assembled into sets of highly similar sequences using a TBC clustering algorithm with a 97% cutoff [51]. While clustering, homopolymeric errors were ignored by allowing a mismatch error of up to 2 bp, which was based on the error rate of 454 sequencing (0.5%). Representative sequences in clusters of trimmed sequences were chosen for identification. Singletons were considered as individual OTUs. The representative sequences and singletons were assigned to taxonomic positions according to the highest pairwise similarity among the top five BLASTN hits against the EzTaxon-e database [52]. Sequences that showed no match in a BLASTN search (expectation value of >e−5) against the EzTaxon-e database were considered to be non-target sequences and were ignored. To calculate the nucleotide sequence similarity between the query and the candidate species, Myers and Miller global pairwise alignment [53] was used along with CLUSTAL [54]. Chimeric sequences were detected by UCHIME [55] and were eliminated from further processing.

To analyze the fungal sequences, the sequencing reads were processed by Fungal ITS Extractor [56], which uses a profile hidden Markov model to obtain pure ITS sequences. The extracted ITS sequences were filtered and denoised using the same pipeline used for the bacterial sequence analysis, except that the clustering was not performed. Individual reads were subjected to a BLASTN search against the UNITE (https://unite.ut.ee/) and EzFungi (http://www.ezbiocloud.net/ezfungi) databases. Sequences that showed no match in the BLASTN search (expectation value of >e−5) were considered to be non-target sequences and were ignored.

### Table 1. Characteristics of the investigated child daycare centers and elementary schools.

| Site | Sampling date       | Location          | Age of occupants | Mean temp. (°C) | Mean humidity (%) |
|------|---------------------|-------------------|------------------|-----------------|------------------|
|      | Indoor   | Outdoor | Indoor   | Outdoor | Indoor   | Outdoor | Indoor   | Outdoor | Indoor   | Outdoor |
| A    | 2013.08.19–20       | 127°1’11.8”E,37°37’39.6”N | 5               | 26.7           | 28.9           | 56.8           | 57.6           |
| B    | 2013.08.21–22       | 127°40.5”E,37°39’32.9”N | 5               | 28.8           | 29.4           | 47.7           | 53.3           |
| C    | 2013.08.26–27       | 127°0’50.9”E,37°39’15.1”N | 5               | 30.2           | 29.5           | 45.2           | 46.0           |
| D    | 2013.08.28–29       | 127°7’13.7”E,37°32’7.3”N | 5               | 29.6           | 28.3           | 56.5           | 66.4           |
| E    | 2013.09.30–01       | 127°2’36.1”E,37°40’43.0”N | 5               | 24.0           | 24.1           | 63.3           | 60.0           |
| F    | 2013.09.02–04       | 127°3’58.1”E,37°40’2.4”N | 8               | 28.4           | 24.3           | 45.9           | 49.5           |
| G    | 2013.09.09–11       | 127°2’6.5”E,37°35’41.2”N | 8               | 26.6           | 24.1           | 59.1           | 70.6           |
| H    | 2013.09.24–26       | 127°0’54.7”E,37°34’2.2”N | 8               | 26.7           | 23.7           | 45.6           | 52.4           |
| I    | 2013.10.14–16       | 126°52’55.4”E,37°34’47.3”N | 9               | 26.0           | 16.9           | 38.1           | 59.8           |
| J    | 2013.10.21–23       | 126°57’6.8”E,37°33’32.7”N | 8               | 23.1           | 20.5           | 37.9           | 41.5           |

doi:10.1371/journal.pone.0126960.t001
Statistical analyses

To avoid potential bias caused by different sequencing depths, samples with more than 3,000 reads were rarefied to a depth of 3,000 reads (2,000 reads for fungal sequences) for subsequent analysis. The species richness (rarefaction curves) and diversity indices (OTUs richness, Chao1 richness [57], and Shannon diversity [58,59]) were calculated using the rRNA Database Project’s pyrosequencing pipeline (http://pyro.cme.msu.edu/) using the CLcommunity program (http://www.chunlab.com/software_clcommunity_about). The diversity measures were calculated by using a TBC clustering algorithm, and the cutoff value for assigning a sequence to a species-level OTU was ≥97% similarity. Source tracking using CLcommunity software was performed to determine the proportion of the communities that originated from human sources. The percentage of the sequencing reads that were shared between a sample community and a reference human microbiome was calculated. The overall phylogenetic distance between each pair of communities was estimated using Fast UniFrac analysis (29) in the CLcommunity program. In brief, sequences were first identified at the species level using a similarity-based identification method with the EzTaxon-e or EzFungi database. The taxonomic assignments were applied to a reference phylogenetic tree backbone constructed from the EzTaxon-e or EzFungi taxonomic structure. Sequences that could not be assigned to known taxa were considered to belong to different species. Using the reference tree with species abundance values, a weighted Fast UniFrac distance was calculated for the samples. The resultant distance matrix was then used to generate an ordination diagram using principal coordinate analysis (PCoA) in the R program (http://www.r-project.org/). For fungal communities, Bray-Curtis dissimilarity-based PCA analysis was also performed using the genera abundance table.

To compare the microbial community structures based on categorical metadata, samples were pooled into bins (daycare/elementary school or indoor/outdoor), and statistical significance tests were performed using the R program. Differences in diversity indices, depending on categorical metadata, were evaluated by ANOSIM using the Wilcoxon t-test. The significance of differences in the microbial profiles (the PCoA vectors from UniFrac distance analysis) according to categorical metadata was determined using Wilcoxon t-test on the x- and y-coordinates. The difference in species abundance, depending on categorical metadata, was determined using Hotelling’s t-test, and multiple testing problems were adjusted with false discovery rate correction [60]. The correlation between environmental variables (temperature or humidity) and microbial community was evaluated using linear regression analysis.

Results

Overview of the microbial diversity in aerosols

Fifty samples were collected from the indoor and outdoor air of five daycare centers (sites A to E) and five elementary schools (sites F to J) (Table 1). The bacterial and fungal communities in all the samples were successfully characterized (S1 Table). Pyrosequencing of bacterial 16S rRNA gene amplicons resulted in 254,771 valid reads (average length, 458 nt) for the 50 air samples (average, 5,095 reads/sample). We observed an average of 1,440 bacterial operational taxonomic units (OTUs) for each sample (range, 144–5,027). The Chao1 estimator of species richness ranged from 173 to 6,599 (average, 1,788). The observed OTU richness of the indoor and outdoor samples did not differ significantly (Wilcoxon t-test). Bacterial community diversity, as estimated by the Shannon diversity index, between indoors and outdoors or between daycare centers and elementary schools did not differ significantly.

Pyrosequencing of fungal ITS amplicons resulted in 195,092 valid reads (average length, 339 nt) for the 50 air samples (average, 3,902 reads/sample; S1 Table). Fungal richness and diversity
were estimated based on the OTUs with 97% nucleotide sequence similarity. The number of OTUs ranged from 149 to 697 for each sample (average, 390). The Chao1 estimator of species richness ranged from 292 to 1,155 depending on the sample (average, 629), indicating that approximately 1.61-fold more fungal OTUs may have been present in the collected air samples than were actually observed in this study. The observed OTU richness and the fungal community diversity, as estimated by Shannon diversity index, did not differ between the indoors and outdoors or between the daycare centers and elementary schools.

In the rarefaction analyses of bacterial or fungal reads, although the gradients of the collector's curves decreased with increasing numbers of sequences, the number of OTUs still increased even at the highest number of sequences samples, indicating the high bacterial diversity in tested samples (data not shown).

**Community structure dynamics**

The bacterial composition of the indoor samples was generally different from that of the outdoor samples. In the Fast-UniFrac distance-based PCoA analysis, the indoor and outdoor samples formed distinct clusters (Fig 1A; Wilcoxon t-test, p < 0.0000 and p = 0.0022 for the x-coordinates and y-coordinates, respectively). A few exceptional cases were also observed, such as at sites J, G, and F. Although the J-Out samples were from outdoor air, they contained a large proportion of human-associated bacteria, including species belonging to the genera *Micrococcus*, *Paracoccus*, and *Staphylococcus*. The G-In samples (from indoor air) contained a high proportion of Actinobacteria, which are usually abundant in outdoor air. Significant differences between the F-out samples and other outdoor samples were driven by the idiosyncratic abundance of bacteria belonging to the genera *Streptomyces*, *Pseudonocardia*, and *Nocardiosis*.

In addition, differences were also observed between the indoor bacterial communities in daycare centers and elementary schools (p = 0.0001 [x-coordinates] and p = 0.8918 [y-coordinates]). The bacterial communities outside of daycare centers and elementary schools did not differ.

In contrast to the bacterial community, the fungal community composition in the indoor and outdoor air did not differ. This was evident in both the EzFungi-based UniFrac analysis and the UNITE-based Bray-Curtis dissimilarity analysis (S1 Fig). The air samples tended to cluster according to sampling site, regardless of whether they were of indoor or outdoor origin (Fig 1B). For example, the indoor and outdoor samples from the J3 site were very similar (upper left side of the ordination graph). This result indicates that the fungal communities of the indoor air reflected those of the outdoor air; therefore, human activity had little influence on the indoor fungal composition.

The effect of temperature and relative humidity on the abundance of specific microbial taxa was evaluated using regression analyses; however, no direct correlation was observed. In addition, the overall bacterial community structure (PCoA coordinates) was not linked to environmental variables. This may be because the duration of sampling was as long as 10 h, and the samplings were restricted to the same city over a short period of time.

**Bacterial community composition**

In the outdoor and indoor air, 38 and 32 bacterial phyla were detected, respectively. Most sequences (60–96.3%) were assigned to 3 dominant phyla, namely, *Actinobacteria* (37% indoor, 35% outdoor), *Proteobacteria* (34% indoor, 33% outdoor), and *Firmicutes* (18% indoor, 15% outdoor). The phyla *Cyanobacteria*, *Bacteroidetes*, *Chloroflexi*, *Acidobacteria*, and *Deinococcus-Thermus* made minor contributions (<5% on average) to the overall population. Other phyla
contributed less than 1% (on average) of the total sequences. Only a few sequences fell into candidate phyla (e.g., TM7, BRC1, and OD1; <1% of each sample).

At the generic level, 2,537 bacterial genera were detected in the indoor and outdoor samples of this study. The distributions of the most prominent bacterial genera (>1.2% of the total bacteria) in the 50 air samples are shown in Fig 2. The bacterial genera at >1.2% abundance accounted for 50.0% and 41.3% of the bacteria in the indoor and outdoor samples, respectively. Of the bacterial genera identified in the indoor air samples, the genus Micrococcus was the most abundant (13.2%), followed by Paracoccus (5.2%), Staphylococcus (4.6%), and Enhydrobacter (4.3%; S2 Table).

The composition of the dominant bacteria in the indoor air clearly differed from that in the outdoor air. The indoor air samples contained a greater number of well-documented human-associated bacteria than the outdoor air; Micrococcus (13.72% indoor, 3.3% outdoor; p < 0.0001), Staphylococcus (4.2% indoor, 1.3% outdoor; p < 0.0001), Streptococcus (4.2% indoor, 0.2% outdoor; p < 0.0001), Corynebacterium (3.3% indoor, 1.7% outdoor; p = 0.0008), and Propionibacterium (1% indoor, 0.3% outdoor; p = 0.0006; S2 Fig).

In contrast, in the outdoor samples, several genera commonly found in soil and water were generally abundant, Methylobacterium (3.9%), Streptomyces (3.5%), Pseudonocardia (2.8%), Sphingomonas (2.6%), and Bacillus (2.5%; S2 Table). The composition of the abundant bacterial genera in the outdoor air samples was highly variable. For example, the most common genus varied between the outdoor samples: Bacillus (12.1%) in D2-Out, Sphingomonas (7.2%) in E1-Out, Pseudonocardia (20.83%) in F1-Out, and Methylobacterium (12.2%) in H1-Out.
Because the indoor air samples from daycare centers and elementary schools also differentiated on the PCoA plot, the abundance of the dominant bacteria was compared. The genera *Paracoccus* (8.1% daycare centers, 3.4% elementary schools; p = 0.0001), *Streptococcus* (5.7%, 2.9%; p = 0.0313), and *Bacillus* (2.2%, 0.95%; p = 0.0049) were more abundant in daycare centers than in elementary schools.

**Fungal community composition**

Based on the EzFungi database, most ITS reads belonged to *Basidiomycota* (73.5% indoor, 59.8% outdoor) and *Ascomycota* (23.5% indoor, 35.1% outdoor; Fig 3). A very low proportion of *Chytridiomycota* (0% indoor, 0.001% outdoor) and *Glomeromycota* (0.001% indoor, 0% outdoor) was detected. The rest of the identified sequences belonged to unclassified fungi that could not be attributed to a phylum.

The airborne *Ascomycota* encompassed three major fungal classes, namely *Dothideomycetes* (10.3% indoor, 15.2% outdoor), *Eurotiomycetes* (7.2% indoor, 8.8% outdoor), and *Sordariomycetes* (4.5% indoor, 8.8% outdoor; Fig 3). Depending on the sample, 0.2% to 57.7% of the reads could not be attributed to a class and were regarded as unidentified sequences. *Dothideomycetes*, which includes genera associated with allergenic fungi such as *Alternaria*, *Epicoccum*, *Curvularia*, and *Cladosporium*, comprised almost half of the *Ascomycota* (10.0–62.5%; average, 44%). *Eurotiomycetes*, which includes *Aspergillus* and *Penicillium*, accounted for 28% and 24% of the *Ascomycota* sequences in the indoor and outdoor air, respectively. In contrast, the majority of detected *Basidiomycota* species (>95%) belonged to a single class, *Agaricomycetes*. These fungi accounted for 69.5% of the indoor samples and 56.9% of the outdoor samples.

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**Fig 2. Relative abundance of bacteria identified at the genus level in indoor and outdoor air.** The sampling sites and replicates are indicated as letters and numbers, respectively, shown in columns. The taxonomic groups that had an average abundance >1.2% are shown.

doi:10.1371/journal.pone.0126960.g002
Hyphodontia and Thanatephorus were the most common fungal genera in the indoor and outdoor air samples (S3 Table). Human skin-associated fungi, such as species belonging to the genus Malassezia, were detected in indoor air samples but were nearly absent from the outdoor air samples. Many different types of fungi, including mushrooms (e.g., Agaricomycetes), plant pathogens, wood-rotting fungi, and molds, were identified in these air samples.

The relative abundance of four representative allergic fungi, namely Aspergillus, Alternaria, Cladosporium, and Penicillium, was calculated (Fig 4). The abundance of Aspergillus (5.0% indoor; 5.2% outdoor) and Cladosporium (7.2% indoor; 7.2% outdoor) did not differ between the indoor and outdoor air. However, the abundance of Alternaria (1.8% indoor; 3.2% outdoor) showed a difference (P = 0.0033), and was 1.8 times more abundant in outdoor air than in indoor air. Penicillium (1.1% indoor; 1.9% outdoor) was also more abundant (1.8 times; P = 0.0036) in outdoor air than in indoor air.

Because fungal ITS identification varies depending on the database used, the ITS reads obtained in this study were also analyzed by BLAST searches against the UNITE database. Based on the UNITE database, 0.1–66.8% of the ITS reads were identified as uncultured fungi. The fungal genera highly represented in EzFungi were similarly observed in the UNITE database, although their abundances differed (S3 Table). Because the two databases did not yield conflicting results, EzFungi-based taxon composition was used for overall analyses.

Identification of bacterial species in bioaerosols that are shared with the human microbiome

To identify the bacterial species that originated from humans, the bacterial communities obtained in this study were compared with a previously reported human microbiome. The skin,
airways, gastrointestinal tract, oral, and urogenital tract microbiome data reported by Costello et al. [61] were used as the reference human microbiome. The proportion of the bacterial communities that came from skin and other human sources were significantly higher (p < 0.0005) in indoor aerosols than in outdoor aerosols (S3 Fig). On average, 10.6% of the bacterial sequences in indoor air came from human skin, and 6.6% or 6.5% of the bacterial sequences in indoor aerosols were from human airways or the oral microbiome, respectively. In comparison, the proportion of human originated bacteria in outdoor samples was lower.

Discussion

Previous studies of the aerosol microbiology in facilities for children primarily utilized culture-dependent techniques [15,16,19–21,23,62–65]. Despite the development of molecular techniques and the great interest in airborne microbiomes, metagenomic investigations have not been employed to identify the airborne microbes in children’s facilities. Previously, Lee et al. [66] investigated the bacterial diversity in a daycare center using culture-independent methods (16S rRNA gene sequencing of clone libraries). They used surface swabs of toys and furniture and found 190 bacterial species. According to the data, the genus *Pseudomonas* was particularly abundant in the clone libraries, followed by uncultured bacterial strains and human-associated bacteria. In the present study, a high abundance of *Pseudomonas* spp. was not observed, which may due to the difference in the sample types. Our results, which were obtained by high throughput sequencing, show that the aerosols in the daytime spaces of children harbor a diverse and dynamic microbial population. Several bacterial genera belonging to the *Proteobacteria* (e.g., *Paracoccus*, *Methylobacterium*, and *Acinetobacter*) were newly identified, as these...
genera had not been previously reported in schools or daycare centers using cultivation-dependent methods [6]. The high abundance of Firmicutes (Staphylococcus and Bacillus) and Actinobacteria (Corynebacterium and Micrococcus) that was consistently observed in previous culture-based studies was confirmed here.

Traditionally, fungi have been identified by directly observing the morphology of captured spores or after cultivation. Most culture-based fungal studies have focused on detecting several fungal genera (Alternaria, Aspergillus, Penicillium, and Cladosporium) and the reported proportion of these genera within air samples varied depending on the study and method of analysis [15,18,23,62,64]. Fungi belonging to the Ascomycota can be more easily cultivated than Basidiomycota using traditional media [32,67] and have received attention because of their allergenic effect and clinical importance. In our study, we used the ITS2 region for a metagenomic analysis of fungal communities and identified 855 genera from 50 air samples. The fungal communities were more diverse than the visible communities and contained a wide variety of previously unidentified taxa. Moreover, the dominant species were significantly different from those reported in previous studies using culture-based methods. The dominant fungi from the culture-based studies (Alternaria, Aspergillus, Penicillium, and Cladosporium) represented <12% of the total sequence reads, whereas Basidiomycota (class Agaricomycetes) represented over 60% of the total reads. Similar to the present findings, using ITS1 analysis, Adams et al. observed a diverse composition of fungal taxa and a particular abundance of Agaricomycetes and Dothideomycetes in indoor environments [68–70]. However, another study of the global indoor fungal composition using sequence analysis of the ITS2 region showed a dominant abundance of the phylum Ascomycota (particularly class Dothideomycetes) [71].

The bacteria in indoor air appeared to be a diverse combination of species associated with both humans and the outside environment. For example, in the indoor air samples, Micrococcus and Staphylococcus species [72–74], which are commonly found in the human skin microbiome, were observed together with Paracoccus [75–77] and Methylobacterium species [78,79], which have been isolated from various geographical locations and environments. Additionally, PCoA using weighted UniFrac distances detected a significant difference in the bacterial composition between the two G-In samples and other indoor samples. The G-In samples contained a large proportion of bacterial genera generally detected in outdoor environments. We postulated that the inflow of adjacent outdoor air by the naturally supplied ventilation could have a large impact on indoor air microbiomes as was shown in previous studies [38,80]. The different outdoor environments, various terrains, and land use types (mountain, stream, big market, park, etc.) may explain the differences in the outdoor and/or indoor microbial communities. However, human-associated bacteria greatly contributed to the structure and taxa of the indoor bacterial communities. The source tracking analysis results also support human skin as a primary source of bacteria in indoor air. Moreover, comparison of the bacterial communities that originated from human sources in daycare centers and schools showed that the indoor microbiome varied according to the microbial community of the occupants, similar to what was shown in other studies of indoor surfaces [81–83]. These results were in agreement with a previous report showing that the fundamental source of bacteria in indoor air was the direct shedding of microorganisms from humans during occupancy [40]. Although Meadow et al. [43] reported a close resemblance between outdoor and indoor airborne bacterial communities, the human-associated bacterial genera observed in this study were more than twice as abundant in indoor air than in outdoor air. Human occupancy in indoor environments is believed to elevate the concentration of indoor airborne bacteria due to the resuspension of settled dust and the shedding of bacteria from human skin [40,44].

Our study demonstrated that the composition of fungi and bacteria in indoor air is vastly different. In contrast to the bacteria, most fungi detected indoors appeared to originate from
diverse outdoor sources, independent of human activity. Known human-associated fungi such as *Malassezia* [84–86], which comprises lipophilic yeasts that are not detected by culturing in common media, were detected in the indoor air; however, the contribution of these fungi to the fungal communities was very low (0.08%). As demonstrated in the UniFrac analyses of this study, the indoor fungal composition did was not differ from the outdoor fungal composition. This result is congruent with the previous report of Adams et al. [69], who suggested that residential surfaces are passive collectors of airborne fungi of putative outdoor origin and that aerosolization of endogenous fungi from growth on typical household surfaces is minor compared to the fungal input from outdoors. Other studies using culture-independent techniques also showed that the composition of the indoor fungal community is predominantly of outdoor origin and is geographically patterned [70,71]. It would be not surprising that the human-associated fungal assemblages in indoor air are smaller than the bacterial ones because the total number of fungal cells on humans is orders of magnitude smaller than that of the bacterial microbiota [87].

**Conclusions**

Our 454 pyrosequencing approach provided deeper insight into the high diversity of the microbial community in the indoor and outdoor air of spaces typically occupied by children. This study also detected rare airborne microbes not found by traditional culture-based surveys and demonstrated that the airborne microbial communities were likely influenced by various environmental sources. The composition of the indoor bacterial community was influenced by human occupancy; however, the composition of the indoor fungal community originated from diverse outdoor sources rather than from humans. These findings provide a better understanding of the airborne microbes present in human environments and a clue for assessing the infections, asthma, allergy, or other respiratory diseases resulting from exposure to airborne microbes. However, additional analyses at different locations over longer time periods are needed to obtain a comprehensive understanding of the airborne microbiome and the various factors that affect the airborne microbial ecology.

**Supporting Information**

**S1 Fig.** Principal coordinate analysis (PCoA) of the fungal community based on (A) the EzFungi database and (B) UNITE database. Differences in the composition of the microbial communities were quantified using a Bray-Curtis dissimilarity matrix. (TIF)

**S2 Fig.** Relative abundance of four dominant bacterial genera in indoor and outdoor air samples. The solid black lines represent the median, and the circles are outliers. Bars denote the minimum and maximum values excluding the outliers. (TIF)

**S3 Fig.** The proportion of human originated bacteria in the aerosols. The air sample data were compared to reference human microbiome data, and the proportion of sequences observed in both the sample and reference data was calculated. The most frequently observed human bacteria and their proportion in the indoor aerosols are shown in the lower panel. (TIF)

**S1 Table.** Sequencing statistics and diversity estimates. (DOCX)
S2 Table. Bacterial genera observed in indoor and outdoor air samples.

(DOC)

S3 Table. Fungal genera observed in indoor and outdoor air samples.

(DOCX)

**Author Contributions**

Conceived and designed the experiments: SKS JK JS HY. Performed the experiments: SKS JK. Analyzed the data: SKS SMH HSO JC HY. Contributed reagents/materials/analysis tools: SKS JK HY. Wrote the paper: SKS HY.

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