ORIGINAL ARTICLE

Lung Microbiome Analysis in Steroid-Naïve Asthma Patients by Using Whole Sputum



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Background: Although recent metagenomic approaches have characterized the distinguished microbial compositions in airways of asthmatics, these results did not reach a consensus due to the small sample size, non-standardization of specimens and medication status. We conducted a metagenomics approach by using terminal restriction fragment length polymorphism (T-RFLP) analysis of the induced whole sputum representing both the cellular and fluid phases in a relative large number of steroid naïve asthmatics.

Methods: Induced whole sputum samples obtained from 36 healthy subjects and 89 steroid-naïve asthma patients were analyzed through T-RFLP analysis.

Results: In contrast to previous reports about microbiota in the asthmatic airways, the diversity of microbial composition was not significantly different between the controls and asthma patients (p=0.937). In an analysis of similarities, the global R-value showed a statistically significant difference but a very low separation (0.148, p=0.002). The dissimilarity in the bacterial communities between groups was 28.74%, and operational taxonomic units (OTUs) contributing to this difference were as follows: OTU 789 (*Lachnospiraceae*), 517 (*Comamonadaceae*, *Acetobacteraceae*, and *Chloroplast*), 633 (*Prevotella*), 645 (*Actinobacteria* and *Propionibacterium acnes*), 607 (*Lactobacillus buchneri*, *Lactobacillus otakiensis*, *Lactobacillus sunkii*, and *Rhodobacteraceae*), and 661 (*Acinetobacter*, *Pseudomonas*, and *Leptotrichiaceae*), and they were significantly more prevalent in the sputum of asthma patients than in the sputum of the controls.

Conclusion: Before starting anti-asthmatic treatment, the microbiota in the whole sputum of patients with asthma showed a marginal difference from the microbiota in the whole sputum of the controls.

Keywords: Asthma; Microbiota; Polymorphism, Restriction Fragment Length; RNA, Ribosomal, 16S; Sputum

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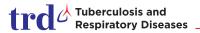
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Introduction

Asthma is a chronic inflammatory airway disease in which various cells and mediators play a role in developing episodic dyspnea with wheezing as well as coughing¹. Previously, asthma was known to be caused by external stimuli such as allergens and genetic susceptibility to allergic diseases; however, according to recent research, microbial communities of the airway may contribute to the pathogenesis of asthma.

As a representative example of bacterial involvement in asthma, *Mycoplasma pneumoniae* and *Chlamydophila pneumonia* were isolated from patients with acute asthma exacerbation². Invasive pneumococcal pneumonia was found to be 2 times more common in the asthma patients as compared to that in the control group². In addition, a risk of developing asthma and recurrence of wheezing was reported in infants having colonizers in their hypopharynx including the combination of *Streptococcus pneumoniae*, *Haemophilus influenzae*, and *Moraxella catarrhalis*³. However, these results reflect only a small culturable portion of the entire microbiota but not whole microorganisms.

The method for metagenomics study has been used recently to overcome this limitation. In metagenomic approach, the uncultured microorganisms can be now identified by targeting the 16S ribosomal RNA (16S rRNA) gene. Traditionally, gut microbiota were the main topics of human metagenomic research but interest in the role of the microbiome in asthma has been increasing recently⁴⁻⁹. Because of small sample size, diverse specimens and different asthma treatment status, results of metagenomic approaches on microbial compositions in airways of patients with asthma were not consistent amongst studies.

Previous studies on the microbiome in asthma patients were performed using the bronchoalveolar lavage (BAL) fluid. Some researchers attempted microbiome analysis by analyzing induced sputum samples ¹⁰⁻¹². However, those studies using sputum were performed with the fluid phase or the cellular phase only. This is because an appropriate sample for microbiome study in airway diseases has not been established yet.

This study was performed to reassess the characteristics of the microbiome in steroid-naïve asthma patients using induced sputum, easily obtainable airway samples, including not only fluid phase but also cellular phase in a large sample size.

Materials and Methods

1. Study population

A total of 135 subjects (36 healthy subjects and 89 asthma patients) who were non-smokers or ex-smokers (less than 10 pack-years) were selected for the study. Subjects with any re-

cent infection in either upper or lower airways prior to enrollment of the study and those who ever used antibiotics within the last 1 month were excluded from the study. Subjects who were never diagnosed with asthma and did not have any respiratory symptoms were enrolled as the healthy control group. Patients who were newly diagnosed with asthma and did not start anti-asthmatic treatment (treatment-naïve) were considered as the asthma patients group. Asthma was defined as chronic respiratory symptoms such as coughing, sputum, dyspnea, and wheezing in addition to the confirmation of airway reversibility or airway hyperresponsiveness (AHR). Airway reversibility was defined as an increase in a forced expiratory volume in 1 second (FEV₁) of 12% and 200 mL or greater when it was measured 15 minutes after inhalation of a short acting β2 agonist and AHR was defined as less than 16 mg/mL of provocative concentration of mechacholine causing a 20% fall in FEV_1 (PC₂₀)¹³. Patients with a history of using anti-inflammatory anti-asthmatic drugs including inhaled corticosteroids were excluded from the study.

The basic demographic characteristics of the patients, skin test results, the results of pulmonary function tests, serum total IgE, and peripheral blood cell counts were investigated.

2. Sputum induction and DNA extraction

Sputum induction was performed according to the standardized protocol 14 . After measurement of basal ${\rm FEV}_{\rm l}$, all participants were treated with 200 μg of salbutamol (Ventolin; GlaxoSmithKline, Bredtfred, England). Four point five percent hypertonic saline solution was used to induce sputum production by using an ultrasonic nebulizer with output set at 4.5 mL/min (Omron Co., Tokyo, Japan), and this administration repeated up to 4 times at an interval of 5 minutes. After each nebulization, the subjects rinsed their mouths and spitted the sputum carefully into a Petri dish.

The induced sputum sample was mixed with the same amount of 10% dithiothreitol. The mixture was shaken, centrifuged, and crushed. In this study, we did not separate the sputum samples to the fluid phase or the cellular phase. DNA was then extracted from 400 μL of the whole sputum samples using Mixer Mill MM 300 (Retsch GmbH, Haan, Germany) and DNeasy Kit (Qiagen, Hilden, Germany) with 0.2-mm glass beads (250 mg) and two 5-mm metal beads.

3. Polymerase chain reaction amplification

The 16S rRNA gene from the bacterial DNA extracted from the whole sputum samples was amplified using polymerase chain reaction (PCR). The extracted DNA (100 ng) was used to obtain a PCR amplified product about 920 bp in size using the following primers: forward, 5'-AGAGTTTGATCMTG-GCTCAG-3' labeled with 6-carboxyfluorescein at the 5' end; reverse, 5'-CCGTCAATTCMTTTRAGTTT-3'. A GeneAmp

9700 system (Applied Biosystems, Foster City, CA, USA) was used; a 50- μ L reaction mixture was prepared for each sample using 5 μ L Taq buffer (20 mM MgCl₂ is contained), 0.5 μ M forward primer, 0.6 μ M reverse primer, 0.2 mM dNTP, 0.7 mg/mL bovine serum albumin (New England Biolabs, Beverly, MA, USA), 2 U ExTaq HS (TaKaRa, Tokyo, Japan), 100 ng template DNA, and sterile distilled water. The cycling conditions were 1 minute at 95°C, 30 seconds at 97°C, 32 cycles of 30 seconds at 97°C, 20 seconds at 58°C, and 90 seconds at 72°C. The amount of amplified DNA through three-time PCR repetition using identical conditions was measured with the help of an EPOCH spectrophotometer (BioTek Instruments Inc., Winoski, VT, USA). The formation of the specific product was verified by gel electrophoresis.

4. Terminal restriction fragment length polymorphism profiling

Terminal restriction fragment length polymorphism (T-RFLP) was performed as described previously 15-18. Briefly, four restriction enzymes (Apol, Hhal, NlaIV, and Hpy166II), suitable for the analysis of respiratory bacterial pathogens, and the Restriction Enzyme Picker (REPK) Online version 1.3 (http:// rocaplab.ocean.washington.edu/tools/repk) were used for the profiling. The PCR products (200 ng) were added to the four restriction endonucleases at 37°C for 2 hours (HhaI, NlaIV, and Hpy166II) or at 50°C for 2 hours (ApoI). The mixture was then incubated at 65°C (HhaI, NlaIV, and Hpy166II) or at 80°C for 20 minutes and then stopped. The size of the fragmented T-RF gene was analyzed using capillary electrophoresis and an ABI PRISM 3130 genetic analyzer (Applied Biosystems). The fragment sizes were estimated using the Local Southern Method in GENESCAN 3.1 software (Applied Biosystems Japan, Tokyo, Japan). Based on the positive control bacterial DNAs (12 DNAs), GeneScan LIZ 1200 Size Standards were recalibrated. An operational taxonomic unit (OTU) was determined using T-REX software (http://trex.biohpc.org/, Cornell University), and the relative amounts of each OTUs were measured. The representative bacterial species or taxonomically related groups in the sputum were predicted using the T-RF size information from the T-RFLP database (http://mededu. cau.ac.kr/micro/FRETS.asp) created in this study.

5. Statistical analysis

The OTUs obtained from whole induced sputum samples were analyzed using multivariate statistical techniques with the software PRIMER (Plymouth Routines In Multivariate Ecological Research) version 6 (PRIMER-E Ltd., Plymouth, UK). T-RF data were subjected to multivariate analysis, including biodiversity analysis, analysis of similarity (ANOSIM), cluster analysis including hierarchal clustering analysis and non-metric multidimensional scaling (nMDS) and similarity

percentage (SIMPER).

Biodiversity of each sample was measured by the Shannon-Weiner diversity index, while differences in biodiversity between asthma group and normal control group were assessed by a nonparametric Mann-Whitney test. Using ANOSIM, we assessed the similarities between the healthy controls group and the asthma patients group. The R statistic value describes the extent of similarity between each groups in the ANOSIM, which R=1 indicates that the two groups are entirely separate, while R=0 signifies that there is no difference between the groups. The contribution of specific T-RFs to differences in bacterial composition between groups was assessed. These statistical analyses were performed to determine whether there was a difference in the bacterial composition of the airways between the healthy controls and the asthma patients by SIMPER, and quantitative difference analysis was performed for the identification of specific T-RFs contributing to differences in bacterial composition between groups. Statistical significance was defined as a p-value of less than 0.05.

6. Ethics statement

This study was approved by the Institutional Review Board of Chung-Ang University Hospital and was performed with patients' consents (C2014056(1252)).

Results

1. Characteristics of the study subjects

The mean age of the asthma patients was 62.0 ± 14.8 years; 47.2% of the subjects were males, 60.7% were non-smokers, and 39.3% were ex-smokers (Table 1). The mean FEV₁ predicted was $82.0\pm17.2\%$, and PC₂₀ was 10.9 ± 8.1 mg/mL. In the control group, the mean age of the 36 subjects was 32.2 ± 10.3 years, and 47.2% were male, 88.9% were non-smokers, and 11.1% ex-smokers. The mean FEV₁ predicted was $100.8\pm10.3\%$.

2. Comparison of microbiota in the sputum samples

The Shannon index was used to evaluate the intra-group diversity in the sputum samples of the two groups; it was 3.5 ± 0.3 in the control group and 3.5 ± 0.3 in the asthma group, showing no significant difference in diversity between the two groups (p=0.807) (Figure 1).

There was no marked difference between the compositions of the microbial communities in the sputum samples of the two groups, when analyzed at the phylum level (Table 2). For the phylum *Firmicutes*, the prevalence was 46.4% in the control group and 44.2% in the asthma group. Further, *Proteobacteria* and *Bacteroides* showed a prevalence of 14.5% and 10.8% in the control group and 16.0% and 11.6% in the asthma

Table 1. Characteristics of the study subjects with asthma

Characteristic	Asthma patients (n=89)
Male	47 (52.8)
Age, yr	62.0±14.8
BMI, kg/m²	24.9±3.7
Non-smoker:ex-smoker	54 (60.7):35 (39.3)
Blood WBC counts, /μL	4,993.7±3,977.5
Blood eosinophil counts, /μL	279.3±281.8
Serum total IgE, kU/L	331.5±546.5
Sputum neutrophil, %	2.1±2.1
Sputum eosinophil, %	10.6±8.7
Atopy, %	44.3
FVC % predicted	92.9±15.6
$\mathrm{FEV}_{\scriptscriptstyle 1}\%$ predicted	82.0±17.2
FEV ₁ /FVC, %	68.9±12.2
PC ₂₀ , mg/mL	10.9±8.1

Values are presented as number (%) or mean \pm standard deviation. BMI: body mass index; WBC: white blood cell; IgE: immunoglobulin E; FVC: forced vital capacity; FEV $_1$: forced expiratory volume in one second; PC $_2$ 0: provocative concentration of methacholine causing a 20% fall in FEV $_1$.

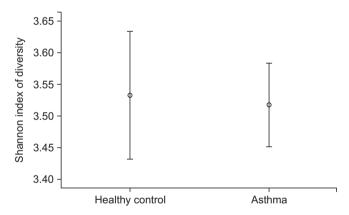


Figure 1. Diversity of microbiota profile in the healthy control and asthma groups. Diversity of microbiota did not show a distinction between healthy controls and asthma patients (p=0.937).

group, respectively.

3. Dissimilarity between the controls and asthma patients

The results of the ANOSIM test indicated that the bacterial composition in the two groups was different as the global R-value was 0.148, with a p-value of 0.002 (Figure 2). The SIM-PER of the bacterial communities in the sputum samples of the control and the asthma groups was calculated using SIM-PER, and the average dissimilarity between the bacterial com-

Table 2. Microbial composition at the phylum level

Phylum	Healthy controls (%)	Asthma patients (%)
Firmicutes	46.4	44.2
Proteobacteria	14.5	16.0
Bacteroidetes	10.8	11.6
Firmicutes/Proteobacteria	4.2	5.4
Actinobacteria	4.0	4.0
Firmicutes/Bacteroidetes	4.0	2.7
Fusobacteria	3.8	3.6
Unknown	12.3	12.4

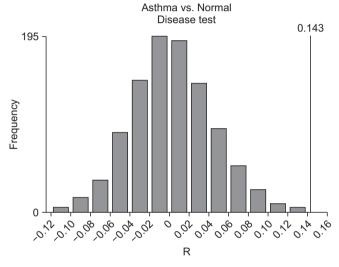


Figure 2. Analysis of similarity test for global community composition (ANOSIM test). Microbial composition showed a significant difference between healthy controls and asthma patients (Global R value=0.148, p=0.002).

munities of the two groups was 28.74%. Hierarchal clustering analysis (Figure 3) or nMDS (Figure 4) indicated no distinct difference in the OTUs for the sputum samples of the two groups.

4. OTUs contributing to the dissimilarity between the control and asthma groups

OTUs accounting for the top 50% contribution to the dissimilarities in the bacterial communities between the two groups are presented in Appendix 1. OTUs 492, 805, 361, 375, 198, 661, 789, 743, 658, and 598 accounted for the top 10%.

As shown in Figure 5, nine OTUs showed more than 1.5 times the amount of bacteria in the induced sputum samples of the asthma group as compared to that in the control group. Particularly, OTU 789 showed 3.1 times more bacteria in the asthma group than in the control group, and belonged to

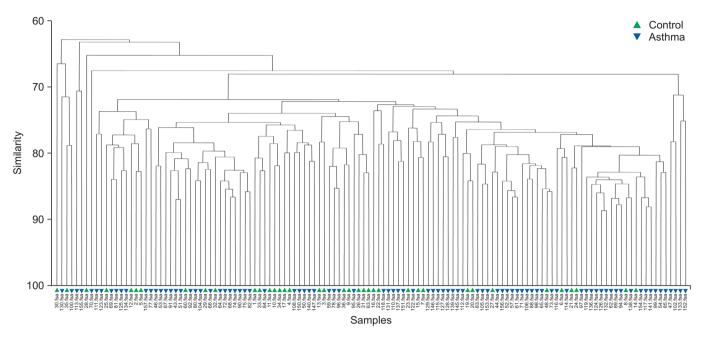


Figure 3. Cluster analysis. Hierarchical cluster analysis indicated no distinct difference in the operational taxonomic units in the sputum samples between the two groups.

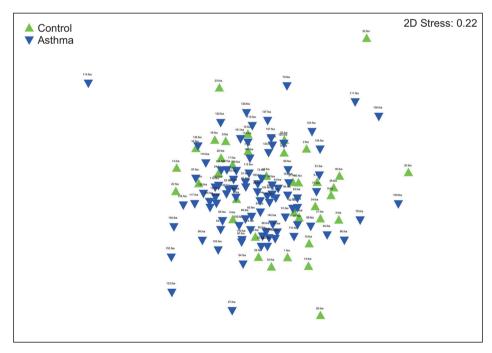


Figure 4. Nonmetric multidimensional scaling (nMDS). nMDS indicated no distinct difference in the operational taxonomic units in the sputum samples between the two groups.

Lachnospiraceae, Alteromonadaceae, Methylococcaceae, Moraxellaceae, and Pseudomonadaceae. Both OTU 517 (Rhizobiales, Rhodospirillales, and Comamonadaceae) and OTU 633 (Prevotella) showed 1.9 times more abundant in the asthma group than in the control group. In addition, OTU 645 (Actinobacteria, Propionibacterium acnes; 1.8-fold), 607 (Lactobacillus buchneri, Lactobacillus otakiensis, Lactobacil

lus sunkii, Rhodobacteraceae; 1.8-fold), 661 (Acinetobacter, Pseudomonas, Leptotrichiaceae; 1.8-fold), 650 (Brevibacteriaceae, Lachnospiraceae, Fusobacterium; 1.5-fold), 821 (Lachnospiraceae, Acinetobacter, Actinomyces; 1.5-fold), 415 (Comamonadaceae, Streptomycetaceae; 1.5-fold) were significantly more prevalent in the sputum of asthma patients than normal control. On the contrary, only OTU 598, belonging to

OTUs	Phylum	Family
598	Firmicutes Bacteroidetes	Lactobacillaceae, Bacteroidales
415	Proteobacteira Actinobacteria	Burkholderiales Actinomycetales
821	Firmicutes Proteobacteria	Lachnospiraceae Gammaproteobacteria
650	Bacteroidetes Fusobacteria	Clostridiales Fusobacteriales
661	Proteobacteria	Moraxellaceae Pseudomonadaceae
607	Firmicutes	Lactobacillaceae Rhodobacteraceae
645	Actinobacteria	Actinobacteria Propionibacterium acnes
633	Bacteroidetes	Prevotella
517	Proteobacteria	Rhizobiales, Rhodospirillales, Comamonadaceae
789	Firmicutes Proteobacteria	Lachnospiraceae, Alteromonadaceae, Methylococcaceae Moraxellaceae, Pseudomonadaceae

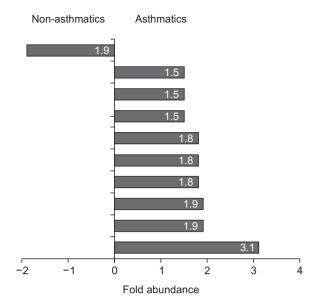


Figure 5. Operational taxonomic units (OTUs) with at least 1.5-fold differences in sputum microbiota between the controls and asthma patients and fold abundance.

Lactobacillaceae and *Bacteroidales*, showed 1.9 times more abundant in the control group than in the asthma group.

As shown in Table 3, 17 OTUs including OTUs 789, 420, 661, 664, 808, 633, 337, 222, 523, 517, 821, 415, 263, 110, 467, 152, and 298, contributed to 90% of the induced sputum microbiota in the asthma patients but not in the healthy controls. In addition, OTUs 528 and 144 contributed to 90% of the induced sputum microbiota in healthy controls but not in the asthma patients (Table 4).

Discussion

The application of culture-independent tools for identification of microbes has been changing concepts of pathogenesis in many diseases. The interests of microbiome research have expanded to the role of the airway microbiome in asthma which is still largely unknown about its predisposition and development. There are some evidence suggesting the microbiome of the airways differs distinctly in patients with asthma from healthy subjects ^{11,12}. However, evidences from recent studies are still inconclusive.

In this study, we compared the microbiome in induced whole sputum samples representing the airways of steroid naïve asthma patients with those of healthy subjects using T-RFLP and we expanded sample size compared to previous studies. The diversity and the bacterial composition at the phylum level in the asthma group were not significantly different from that in the control group. However, there was a slight but significant difference of OTUs in the composition of microbiome community between the two groups. In addition,

OTUs showed differences in the relative amounts of certain bacteria between the two groups; there were also some OTUs representing bacteria found only in one group, indicating a difference in the microbiome of the airway between asthma patients and healthy controls.

In previous metagenomics studies on asthma, the small number of study subjects fewer than 30 was the main obstacle to reach a conclusion. To overcome this, we expanded the sample size to 135 subjects, larger than any previous studies conducted. T-RFLP is a fingerprinting method and has a limitation in the range of microbiota. However, T-RFLP technique can be useful clustering communities according to changes in the dominant members across large number of samples. There are several microbiome studies using T-RFLP in humans and those mainly focused on inflammatory bowel disease and irritable bowel syndrome with fecal microbiota but some studies dealt with cystic fibrosis 22,23 and asthma 10.

In this study, we used whole sputum obtained through induction. Previously, researchers investigated the microbial composition in the BAL fluid, bronchial epithelial brushings, and the lung tissue. The limitation of these studies was that these samples could not be easily obtained. Thus, induced sputum samples, relatively less invasive and easy to acquire, were used in this study. Only a few previous metagenomics studies using inducted sputum samples have already shown significant results in asthma patients ¹⁰⁻¹². Moreover, studies involving an investigation of the microbiome of the oropharynx in asthma patients have shown significant results ^{24,25}. Another study reported that similar bacteria are distributed from the oral cavity to the lungs²⁶. It is postulated that the induced sputum samples in this study can sufficiently reflect the airway of

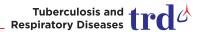


Table 3. OTUs existing in 90% of the induced sputum microbiota in asthma patients, but not found in healthy controls

OTU	Contribution (%)	Predicted bact	erial group
010	Contribution (%)	Family	Phylum
789	0.63	Lachnospiraceae	<i>Firmicutes</i>
		Alteromonadaceae	Proteobacteria
		Methylococcaceae	
		Moraxellaceae	
		Pseudomonadaceae	
420	0.61	Veillonellaceae	<i>Firmicutes</i>
		Lachnospiraceae	
661	0.61	Moraxellaceae	Proteobacteria
		Pseudomonadaceae	
664	0.56	NBG	NBG
800	0.47	Gammaproteobacteria	Proteobacteria
633	0.46	Prevotella	Bacteroidetes
337	0.45	Acidobacteria	Acidobacteria
222	0.44	Streptococcaceae	Firmicutes
		Peptococcaceae	
		Clostridiaceae	
523	0.44	Rhizobiales	Proteobacteria
		Kordiimonadales	
517	0.41	Rhizobiales	Proteobacteria
		Rhodospirillales	
		Comamonadaceae	
821	0.38	Lachnospiraceae	<i>Firmicutes</i>
		Gammaproteobacteria	Proteobacteria
415	0.36	Burkholderiales	Proteobacteira
		Actinomycetales	Actinobacteria
263	0.32	Flavobacteriaceae	Bacteroidetes
110	0.31	Veillonellaceae	<i>Firmicutes</i>
467	0.31	Peptostreptococcaceae	Firmicutes
152	0.31	Rhizobiaceae	Proteobacteria
		Rhodobacteraceae	Bacteroidetes
		Flavobacteriaceae	
298	0.3	Bacteroidales	Bacteroidetes

OTU: operational taxonomic unit; NBG: numerous bacterial groups.

asthma patients. However, in previous studies, microbiome studies have been performed only with induced sputum cell pellets or supernatants after in cytospin preparations ¹⁰⁻¹² and an appropriate sputum sample for studying lung microbiome in the airway of asthmatics is not established yet. A study on feces in the colon reported that there is a difference between the bacterial flora in the intestinal epithelium and the bacterial

flora in the feces¹⁸. Furthermore, quantitative differences exist in specific bacterial groups between luminal and mucosal-associated intestinal microbiota using two independent techniques such as culture analysis and DNA study²⁷. Therefore, microbiome analysis using the whole sputum sample including both the cellular and the fluid phases of the sputum should be compared that from supernatants of sputum. In addition,

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ОТИ	Contribution (07)	Predicted l	oacterial group
010	Contribution (%)	Family	Phylum
598	0.46	Lactobacillaceae	Firmicutes
		Bacteroidales	Bacteroidetes
144	0.35	Lachnospiraceae	<i>Firmicutes</i>
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Table 4. OTUs existing in 90% of the induced sputum microbiota in healthy controls, but not found in asthma patients

OTU: operational taxonomic unit.

genetic, dietary factors and socioeconomic status can affect microbiota in human²⁸. In this study, considering that all study subjects were Koreans, genetic and dietary factors would be relatively homogeneous but may contribute and alter lung microbiota compared to other ethnic groups.

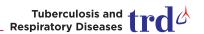
With regard to the diversity of the bacteria, researchers have reported that the diversity of the gut microbiome decreases immediately after birth and that it is associated with the development of asthma and its severity²⁹⁻³¹. The decreased diversity of the microbiome can increase the risk of asthma and allergic disease(s) and it is consisting with hygiene hypothesis. In addition, the reduced diversity of the microbiome in neutrophilic asthma among poorly controlled asthmas has been reported¹². However, some researchers have reported the diversity to be rather significantly higher in asthma patients than in controls, and that it is associated with AHR^{11,32}, while it has been reported that there was no difference in the diversity based on the absence or presence of asthma^{25,33}. In this study, no distinctive difference was found in the diversity between the control group and the asthma group.

Several metagenomics studies have reported that the airway of asthma patients is represented with an increase of the number of proteobacterial species, which is a different characteristic from that observed in healthy people $^{11,25,34,35}. \ \mbox{In}$ addition, it has been reported that *Bacteroidetes* species are more common in healthy controls than in asthma patients ^{25,33} and other studies have reported that Acinetobacteria and Firmicutes exist at a greater proportion in the control group¹¹. This discrepancy may be attributed from the limited number of study subjects included in the studies in part. In addition, the difference in the bacterial numbers at the phylum level, consisted of a number of species, can result in opposing findings. This study was performed in a large number of subjects but we could not find significant difference in the bacterial community composition at the phylum level between the two groups. However, considering the 12% of unidentified bacteria and 8% of OTUs representing two phyla as a partial limitation of the T-RFLP method used in this study, the chance of false negative results cannot be ruled out. However, for the OTUs showing a significant difference between the two groups in this study, the number of *Proteobacteria* was much greater in the asthma group than in the control group, and the proportion of *Bacteroidetes* was higher in the control group. In particular, among bacteria found only in the asthma group, the OTUs representing *Proteobacteria* tended to be relatively greater and the OTUs observed only in the healthy controls all belonged to *Bacteroidetes*.

Green et al.¹⁰ reported that *Moraxella catarrhalis*, and members of the *Haemophilus* or *Streptococcus* genera are the major bacteria in treatment-resistant patients with severe asthma, and a greater ratio of these bacteria is associated with reduced FEV₁, increased neutrophil count, and increased interleukin 8 expression, indicating that the composition of the bacterial community is associated with the severity of asthma. In addition, one previous study has also reported that *Comamonadaceae*, *Sphingomonadaceae*, and *Oxalobacteraceae*, which belong to the phylum *Proteobacteria*, are associated with bronchial hypersensitivity³². In agreement with this result, we found that OTU 789 in *Moraxellaceae* showed a 3.1-times higher prevalence in the asthma group than in the control group, OTU 517 representing *Comamonadaceae* showed a 1.5 times higher prevalence in the asthma group.

A major limitation of this study is that the age of healthy control group is much lower than that of the asthma group. This was caused by recruitment of subjects without age matching. Study about aging and lung microbial change is rare³⁶. Instead, studies from gut microbiome showed that age affect the microbiome profile and age related changes to the intestinal microbiome influence asthma^{37,38}. The environment of the lung microbiome can also change according to aging³⁶. Therefore, further studies would need to be performed with age matching. In other limitation, there were many unidentified bacteria as this study used T-RFLP, a fingerprint method. However, T-RFLP method is still being used before applying to next generation sequencing (NGS) owing to its several advantages over sequencing methods. T-RFLP method is a highly robust, easy to handle and a cheap alternative to the computationally demanding NGS analysis. In addition, T-RFLP method used in this study was tailored for the respiratory tract bacteria by in-silico testing the common bacteria residents in the respiratory tract and the four best restriction enzymes was chosen prior to performing the T-RFLP procedures in order to overcome the limitation of T-RFLP.

In conclusion, this study, unlike previous studies, involved



the investigation of the microbiome in whole sputum samples of asthma patients using T-RFLP in a relatively larger sample size and the microbial composition in airway could not be clearly divided according to the presence of asthma.

Conflicts of Interest

No potential conflict of interest relevant to this article was reported.

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References

- 1. Wenzel SE. Asthma: defining of the persistent adult phenotypes. Lancet 2006;368:804-13.
- 2. Cosentini R, Tarsia P, Canetta C, Graziadei G, Brambilla AM, Aliberti S, et al. Severe asthma exacerbation: role of acute *Chlamydophila pneumoniae* and *Mycoplasma pneumoniae* infection. Respir Res 2008;9:48.
- Bisgaard H, Hermansen MN, Buchvald F, Loland L, Halkjaer LB, Bonnelykke K, et al. Childhood asthma after bacterial colonization of the airway in neonates. N Engl J Med 2007;357: 1487-95.
- Huang YJ, Boushey HA. The microbiome in asthma. J Allergy Clin Immunol 2015;135:25-30.
- Holt PG. The mechanism or mechanisms driving atopic asthma initiation: the infant respiratory microbiome moves to center stage. J Allergy Clin Immunol 2015;136:15-22.
- Panzer AR, Lynch SV. Influence and effect of the human microbiome in allergy and asthma. Curr Opin Rheumatol 2015; 27:373-80.
- 7. Edwards MR, Bartlett NW, Hussell T, Openshaw P, Johnston SL. The microbiology of asthma. Nat Rev Microbiol 2012;10: 459-71.
- 8. Huang YJ. The respiratory microbiome and innate immunity in asthma. Curr Opin Pulm Med 2015;21:27-32.
- Althani AA, Marei HE, Hamdi WS, Nasrallah GK, El Zowalaty ME, Al Khodor S, et al. Human microbiome and its association with health and diseases. J Cell Physiol 2016;231:1688-94.
- Green BJ, Wiriyachaiporn S, Grainge C, Rogers GB, Kehagia V, Lau L, et al. Potentially pathogenic airway bacteria and neutrophilic inflammation in treatment resistant severe asthma. PLoS One 2014;9:e100645.
- 11. Marri PR, Stern DA, Wright AL, Billheimer D, Martinez FD. Asthma-associated differences in microbial composition of

- induced sputum. J Allergy Clin Immunol 2013;131:346-52.
- 12. Simpson JL, Daly J, Baines KJ, Yang IA, Upham JW, Reynolds PN, et al. Airway dysbiosis: *Haemophilus influenzae* and *Tropheryma* in poorly controlled asthma. Eur Respir J 2016;47: 792-800.
- National Asthma Education and Prevention Program. Expert Panel Report 3 (EPR-3): guidelines for the diagnosis and management of asthma: summary report 2007. J Allergy Clin Immunol 2007;120(5 Suppl):S94-138.
- Sohn SW, Lee HS, Park HW, Chang YS, Kim YK, Cho SH, et al. Evaluation of cytokine mRNA in induced sputum from patients with allergic rhinitis: relationship to airway hyperresponsiveness. Allergy 2008;63:268-73.
- 15. Marsh TL. Terminal restriction fragment length polymorphism (T-RFLP): an emerging method for characterizing diversity among homologous populations of amplification products. Curr Opin Microbiol 1999;2:323-7.
- Schutte UM, Abdo Z, Bent SJ, Shyu C, Williams CJ, Pierson JD, et al. Advances in the use of terminal restriction fragment length polymorphism (T-RFLP) analysis of 16S rRNA genes to characterize microbial communities. Appl Microbiol Biotechnol 2008;80:365-80.
- Andoh A, Imaeda H, Aomatsu T, Inatomi O, Bamba S, Sasaki M, et al. Comparison of the fecal microbiota profiles between ulcerative colitis and Crohn's disease using terminal restriction fragment length polymorphism analysis. J Gastroenterol 2011;46:479-86.
- Carroll IM, Ringel-Kulka T, Keku TO, Chang YH, Packey CD, Sartor RB, et al. Molecular analysis of the luminal- and mucosal-associated intestinal microbiota in diarrhea-predominant irritable bowel syndrome. Am J Physiol Gastrointest Liver Physiol 2011;301:G799-807.
- 19. Andoh A, Sakata S, Koizumi Y, Mitsuyama K, Fujiyama Y, Benno Y. Terminal restriction fragment length polymorphism analysis of the diversity of fecal microbiota in patients with ulcerative colitis. Inflamm Bowel Dis 2007;13:955-62.
- Andoh A, Tsujikawa T, Sasaki M, Mitsuyama K, Suzuki Y, Matsui T, et al. Faecal microbiota profile of Crohn's disease determined by terminal restriction fragment length polymorphism analysis. Aliment Pharmacol Ther 2009;29:75-82.
- 21. Matsumoto M, Sakamoto M, Hayashi H, Benno Y. Novel phylogenetic assignment database for terminal-restriction fragment length polymorphism analysis of human colonic microbiota. J Microbiol Methods 2005;61:305-19.
- 22. Rogers GB, Carroll MP, Serisier DJ, Hockey PM, Jones G, Bruce KD. characterization of bacterial community diversity in cystic fibrosis lung infections by use of 16s ribosomal DNA terminal restriction fragment length polymorphism profiling. J Clin Microbiol 2004;42:5176-83.
- 23. Rogers GB, Hart CA, Mason JR, Hughes M, Walshaw MJ, Bruce KD. Bacterial diversity in cases of lung infection in cystic fibrosis patients: 16S ribosomal DNA (rDNA) length heterogeneity PCR and 16S rDNA terminal restriction fragment length

- polymorphism profiling. J Clin Microbiol 2003;41:3548-58.
- 24. Cardenas PA, Cooper PJ, Cox MJ, Chico M, Arias C, Moffatt MF, et al. Upper airways microbiota in antibiotic-naive wheezing and healthy infants from the tropics of rural Ecuador. PLoS One 2012;7:e46803.
- 25. Park H, Shin JW, Park SG, Kim W. Microbial communities in the upper respiratory tract of patients with asthma and chronic obstructive pulmonary disease. PLoS One 2014;9:e109710.
- 26. Venkataraman A, Bassis CM, Beck JM, Young VB, Curtis JL, Huffnagle GB, et al. Application of a neutral community model to assess structuring of the human lung microbiome. MBio 2015;6:e02284-14.
- 27. Russell SL, Gold MJ, Hartmann M, Willing BP, Thorson L, Wlodarska M, et al. Early life antibiotic-driven changes in microbiota enhance susceptibility to allergic asthma. EMBO Rep 2012;13:440-7.
- 28. Fortenberry JD. The uses of race and ethnicity in human microbiome research. Trends Microbiol 2013;21:165-6.
- 29. Carroll IM, Chang YH, Park J, Sartor RB, Ringel Y. Luminal and mucosal-associated intestinal microbiota in patients with diarrhea-predominant irritable bowel syndrome. Gut Pathog 2010;2:19.
- Abrahamsson TR, Jakobsson HE, Andersson AF, Bjorksten B, Engstrand L, Jenmalm MC. Low gut microbiota diversity in early infancy precedes asthma at school age. Clin Exp Allergy 2014;44:842-50.
- 31. Webley WC, Aldridge KL. Infectious asthma triggers: time to

- revise the hygiene hypothesis? Trends Microbiol 2015;23:389-91
- 32. Huang YJ, Nelson CE, Brodie EL, Desantis TZ, Baek MS, Liu J, et al. Airway microbiota and bronchial hyperresponsiveness in patients with suboptimally controlled asthma. J Allergy Clin Immunol 2011;127:372-81.
- 33. Goleva E, Jackson LP, Harris JK, Robertson CE, Sutherland ER, Hall CF, et al. The effects of airway microbiome on corticosteroid responsiveness in asthma. Am J Respir Crit Care Med 2013;188:1193-201.
- 34. Hilty M, Burke C, Pedro H, Cardenas P, Bush A, Bossley C, et al. Disordered microbial communities in asthmatic airways. PLoS One 2010;5:e8578.
- Perez-Losada M, Castro-Nallar E, Bendall ML, Freishtat RJ, Crandall KA. Dual transcriptomic profiling of host and microbiota during health and disease in pediatric asthma. PLoS One 2015;10:e0131819.
- 36. Chotirmall SH, Burke CM. Aging and the microbiome: implications for asthma in the elderly? Expert Rev Respir Med 2015;9:125-8.
- 37. Trompette A, Gollwitzer ES, Yadava K, Sichelstiel AK, Sprenger N, Ngom-Bru C, et al. Gut microbiota metabolism of dietary fiber influences allergic airway disease and hematopoiesis. Nat Med 2014;20:159-66.
- 38. Vital M, Harkema JR, Rizzo M, Tiedje J, Brandenberger C. Alterations of the murine gut microbiome with age and allergic airway disease. J Immunol Res 2015;2015:892568.

Appendix 1. Contribution of OTUs to the dissimilarity in microbial communities in the induced sputum between controls and asthma patients

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	Averagea	Average abundance	Avorono	Disconiation	Individual	Cumulotivo	Fold abundance	ndance	Predicted bacterial group	al group
OTO	Controls	Asthma patients	dissociation	/SD	contribution (%)	contribution (%)	Controls	Asthma patients	Family	Phylum
492	0.59	0.5	0.29	1.28	1	1	1.2	0.8	Syntrophaceae	Proteobacteria
805	1.13	1.1	0.29	1.31	1	2.01	1.0	1.0	Bacillaceae	Firmicutes
									Staphylococcaceae	
361	1.02	0.92	0.28	1.36	0.99	က	1.1	6.0	Actinomycetales	Actinobacteria
									Pasteurellaceae	Proteobacteria
									Thermoactinomycetaceae	Firmictues
375	1.14	1.12	0.28	1.38	26:0	3.97	1.0	1.0	Haemonphilus	Proteobacteria
198	1.14	1.07	0.28	1.39	96.0	4.93	1.1	6.0	Enterobacteriaceae	Proteobacteria
									Pasteurellaceae	
199	0.35	0.62	0.27	1.32	0.95	5.88	9.0	1.8*	Moraxellaceae	Proteobacteria
									Pseudomonadaceae	
789	0.18	0.55	0.26	1.54	6:0	6.78	0.3	3.1*	Lachnospiraceae	Firmicutes
									Alteromonadaceae	Proteobacteria
									Methylococcaceae	
									Moraxellaceae	
									Pseudomonadaceae	
743	0.34	0.4	0.26	1.13	0.89	2.67	6.0	1.2	Acetobacteraceae	Proteobacteria
									Bradyrhizobiaceae	
658	0.47	0.63	0.25	1.26	0.87	8.55	0.7	1.3	Clostridiales	Firmicutes
									Gammaproteobacteria	Proteobacteria
298	0.5	0.27	0.24	1.27	0.84	9:39	1.9^{*}	0.5	Lactobacillaceae	Firmicutes
									Bacteroidales	Bacteroidetes
459	9.0	0.51	0.24	1.25	0.83	10.21	1.2	6.0	Carnobacteriaceae	Firmicutes
633	0.26	0.49	0.24	1.28	0.82	11.03	0.5	1.9^{*}	Prevotella	Bacteroidetes
871	0.61	0.82	0.23	1.18	0.8	11.84	0.7	1.3	Alphaproteobacteria	Proteobacteria
356	0.45	0.48	0.23	1.28	0.8	12.63	6.0	1.1	Neisseria	Proteobacteria
699	0.78	0.98	0.22	1.18	0.77	13.4	8.0	1.3	Prevotella	Bacteroidetes
105	0.92	1.01	0.22	1.33	0.76	14.16	6.0	1.1	Prevotella	Bacteroidetes
266	0.38	0.28	0.22	1.16	0.75	14.91	1.4	0.7	Lachnospiraceae	Firmicutes
431	0.27	0.38	0.21	1.2	0.75	15.65	0.7	1.4	Prevotella	Bacteroidetes

ppena	ppendix 1. Confined	nan								
	Average abundance	oundance	Avorago	Discociation	Individual	Cumulativa	Fold abundance	ndance	Predicted bacterial group	al group
OTO	Controls	Asthma patients	dissociation	/SD	contribution (%)	contribution (%)	Controls	Asthma patients	Family	Phylum
808	0.28	0.36	0.21	1.15	0.74	16.39	0.8	1.3	Bacillales	Firmicutes
									Clostridiales	Proteobacteria
									Gammaproteobacteria	
179	0.87	0.95	0.21	1.26	0.73	17.13	6.0	1.1	NBG	NBG
547	0.56	0.44	0.21	1.23	0.73	17.86	1.3	8.0	Streptococcus	Firmicutes
800	0.41	0.47	0.21	1.21	0.73	18.59	6.0	1.1	Gammaproteobacteria	Proteobacteria
821	0.29	0.43	0.2	1.13	0.71	19.3	0.7	1.5^{*}	Lachnospiraceae	Firmicutes
									Gammaproteobacteria	Proteobacteria
453	0.43	9.0	0.2	1.21	0.71	20	0.7	1.4	Corynebacteriaceae	Actinobacteria
467	0.38	0.37	0.2	1.22	2.0	20.7	1.0	1.0	Peptostreptococcaceae	Firmicutes
865	0.34	0.35	0.2	1.17	69.0	21.4	1.0	1.0	Alphaproteobacteria	Proteobacteria
									Gammaproteobacteria	
186	89.0	29.0	0.2	1.14	69.0	22.08	1.0	1.0	Carnobacteriaceae	Firmicutes
551	0.45	0.49	0.2	1.19	69'0	22.77	6.0	1.1	Streptococcus	Firmicutes
395	0.46	0.64	0.2	1.16	0.68	23.45	0.7	1.4	Erysipelotrichaceae	Firmicutes
144	0.37	0.31	0.2	1.21	0.68	24.13	1.2	8.0	Lachnospiraceae	Firmicutes
									Flavobacteriaceae	Bacteroidetes
602	0.21	0.25	0.2	0.92	0.68	24.81	0.8	1.2	Prevotella	Bacteroidetes
400	0.31	0.37	0.19	1.22	29.0	25.49	0.8	1.2	Clostridiales	Firmicutes
									Erysipelotrichaceae	
282	0.35	0.3	0.19	1.15	29.0	26.16	1.2	6.0	Corynebacterium	Actinobacteria
835	0.37	0.5	0.19	1.15	99'0	26.82	0.7	1.4	Lachnospiraceae	Firmicutes
415	0.28	0.41	0.19	1.21	99.0	27.47	2.0	1.5^{*}	Burkholderiales	Proteobacteira
									Actinomycetales	Actinobacteria
022	0.26	0.31	0.19	1.1	0.65	28.13	0.8	1.2	Clostridiales	Firmicutes
352	0.42	0.37	0.19	1.16	0.65	28.78	1.1	6.0	Coriobacteriaceae	Actinobacteria
									Burkholderiaceae	Proteobacteria
222	0.32	0.45	0.19	1.19	0.65	29.42	0.7	1.4	Streptococcaceae	Firmicutes
									Peptococcaceae	
									Clostridiaceae	
307	0.81	6.0	0.18	1.27	0.64	30.06	6.0	1.1	Prevotella	Bacteroidetes

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	Averagea	Average abundance	Аургаор	Dissociation	Individual	Cumulative	Fold abundance	ndance	Predicted bacterial group	al group
OTO	Controls	Asthma patients	dissociation	/SD	contribution (%)	contribution (%)	Controls	Asthma patients	Family	Phylum
195	0.38	0.43	0.18	1.19	0.64	30.7	6.0	1.1	Enterobacteriaceae	Proteobacteria
785	0.26	0.25	0.18	1.02	0.64	31.34	1.0	1.0	Lachnospiraceae	Firmicutes
									Peptostreptococcaceae	
650	0.21	0.32	0.18	1.15	0.63	31.97	0.7	1.5^{*}	Clostridiales	Bacteroidetes
									Fusobacteriales	Fusobacteria
664	0.37	0.5	0.18	1.15	0.63	32.61	0.7	1.4	NBG	NBG
445	0.62	0.57	0.18	1.14	0.63	33.24	1.1	6.0	Fusobacteriaceae	Fusobacteria
									Acidithiobacillaceae	Proteobacteria
									Clostridiales	Firmicutes
645	0.17	0.31	0.18	1.09	0.63	33.88	0.5	1.8*	Actinobacteria	Actinobacteria
									Propionibacterium acnes	
212	0.21	0.4	0.18	1.25	0.63	34.51	0.5	1.9*	Rhizobiales	Proteobacteria
									Rhodospirillales	
									Comamonadaceae	
372	0.28	0.28	0.18	1.07	0.63	35.14	1.0	1.0	Nesseria/E. coli	Proteobacteria
523	0.32	0.43	0.18	1.13	0.63	35.76	0.7	1.3	Rhizobiales	Proteobacteria
									Kordiimonadales	
285	0.28	0.26	0.18	1.09	0.63	36.39	1.1	6.0	Streptococcaceae	Firmicutes
242	0.37	0.3	0.18	1.21	0.63	37.02	1.2	0.8	Bacillales	Firmicutes
									Lachnospiraceae	
									Peptococcaceae	
674	1.58	1.64	0.18	1.28	0.63	37.64	1.0	1.0	Staphylococcaceae	Firmicutes
									Streptococcaceae	
209	0.17	0.31	0.18	1.09	0.62	38.27	0.5	1.8*	Lactobacillaceae	Firmicutes
									Rhodobacteraceae	
176	0.77	0.77	0.18	1.26	0.62	38.89	1.0	1.0	Acetobacteraceae	Proteobacteria
									Thiotrichaceae	Actinobacteria
									Methylobacteriaceae	Tenericutes
									Coriobacteriaceae	
									Mycoplasmataceae	
298	0.29	0.35	0.18	1.19	0.62	39.52	0.8	1.2	Bacteroidales	Bacteroidetes

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	America	Avorage abundance					Fold shiindance	ndanco	Dradicted hacterial groun	ol group
	Avelage a	Duiluailce	Average	Dissociation	Individual	Cumulative	roid abu	linalice	riemered bacter	ai gi oup
OTO	Controls	Asthma patients	dissociation	/SD	contribution (%)	contribution (%)	Controls	Asthma patients	Family	Phylum
684	0.72	0.7	0.18	1.11	0.62	40.14	1.0	1.0	Staphylococcus /Veillonellaceae	Firmicutes
72	0.39	0.53	0.18	1.11	0.62	40.76	0.7	1.4	Lachnospiraceae	Firmicutes
535	0.57	0.51	0.18	1.08	0.62	41.38	1.1	6.0	Rikenellaceae	Bacteroidetes
									Alteromonadales	Proteobacteria
386	0.42	0.45	0.18	1.11	0.62	41.99	6:0	1.1	Clostridiales	Firmicutes
									Gammaproteobacteria	Proteobacteria
302	0.75	92.0	0.18	1.11	0.61	42.61	1.0	1.0	Lachnospiraceae	Firmicutes
									Porphyromonadaceae	Bacteroidetes
574	0.56	99.0	0.18	1.14	0.61	43.22	0.8	1.2	Corynebacterium	Actinobacteria
390	0.79	0.72	0.17	1.14	0.61	43.83	1.1	6.0	Cyanobacteria	Cyanobacteria
									Clostridiales	Firmicutes
									Caulobacteraceae	Proteobacteria
102	1.14	1.14	0.17	1.34	9.0	44.43	1.0	1.0	Prevotella	Bacteroidetes
83	0.25	0.26	0.17	1.05	9.0	45.03	1.0	1.0	Fusobacteriaceae	Fusobacteria
182	0.27	0.33	0.17	1.19	9.0	45.63	0.8	1.2	Staphylococcus	Firmicutes
347	0.3	0.33	0.17	1.19	9.0	46.23	6.0	1.1	Burkholderiaceae	Proteobacteria
									Comamonadaceae	
									Oxalobacteraceae	
436	0.55	0.62	0.17	1.05	9.0	46.83	6.0	1.1	Actinomyces	Actinobacteria
									Lachnospiraceae	Firmicutes
365	0.81	0.88	0.17	1.14	0.59	47.42	6.0	1.1	Haemonphilus	Proteobacteria
110	0.36	0.35	0.17	1.13	0.59	48.01	1.0	1.0	Veillonellaceae	Firmicutes
251	0.42	0.47	0.17	1.09	0.59	48.6	6.0	1.1	Staphylococcus	Firmicutes
379	0.58	9.0	0.17	1.09	0.59	49.18	1.0	1.0	Haemonphilus	Proteobacteria
148	0.32	0.31	0.17	1.15	0.58	49.77	1.0	1.0	Flavobacteriaceae	Bacteroidetes
									Comamonadaceae	Proteobacteria
*More th	an twice the	difference h	*More than twice the difference between the two groun) groups						

*More than twice the difference between the two groups. OTU: operational taxonomic unit; SD: standard deviation; NBG: numerous bacterial groups.