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Type 2–high asthma is associated with a specific indoor mycobiome and microbiome

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GRAPHICAL ABSTRACT

Background: The links between microbial environmental exposures and asthma are well documented, but no study has combined deep sequencing results from pulmonary and indoor microbiomes of patients with asthma with spirometry, clinical, and endotype parameters.

Objective: The goal of this study was to investigate the links between indoor microbial exposures and pulmonary microbiomes and to document the role of microbial exposures on inflammatory and clinical outcomes of patients with severe asthma (SA).

Methods: A total of 55 patients with SA from the national Cohort of Bronchial Obstruction and Asthma cohort were enrolled for analyzing their indoor microbial flora through the use of electrostatic dust collectors (EDCs). Among these communities and to document the role of microbial exposures on inflammatory and clinical outcomes of patients with severe asthma (SA).

Results: The analysis were based on bacterial (V3-V4 locus of 16S gene) and fungal (ITS2 region of rDNA) amplification using a 250-bp paired-end technology on MiSeq (Illumina) platform.

Main findings

(1) FeNO level is correlated with indoor fungal Chao1 indexes

(2) Beta-diversity of indoor fungal communities is clustered according to T2 endotypes

(3) Indoor and Respiratory microbiomes: More fungal taxa are significantly shared with indoor microbiome during exacerbation periods, compared to clinical stability periods

Abbreviations:

FeNO: Fraction of exhaled nitric oxide
Chao1 index: Alpha-diversity index estimating the number of species in a community

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patients, 22 were able to produce sputum during “stable” or pulmonary “exacerbation” periods and had complete pairs of EDC and sputum samples, both collected and analyzed. We used amplicon targeted metagenomics to compare microbial communities from EDC and sputum samples of patients according to type 2 (T2)-asthma endotypes.

Results: Compared with patients with T2-low SA, patients with T2-high SA exhibited an increase in bacterial α-diversity and a decrease in fungal α-diversity of their indoor microbial florae, the latter being significantly correlated with fraction of exhaled nitric oxide levels. The β-diversity of the EDC mycobiome clustered significantly according to T2 endotypes. Moreover, the proportion of fungal taxa in common between the sputum and EDC samples was significantly higher when patients exhibited acute exacerbation.

Conclusion: These results illustrated, for the first time, a potential association between the indoor mycobiome and clinical features of patients with SA, which should renew interest in deciphering the interactions between indoor environment, fungi, and host in asthma. (J Allergy Clin Immunol 2021;147:1296-1305.)

Key words: Indoor environment, mycobiome, microbiome, severe asthma, FENO, type 2 asthma

Asthma is a complex, chronic inflammatory disease of the airways that affects 339 million people worldwide.1,2 Despite recent advances in new treatments and notable efforts to elucidate asthma pathophysiology, patients with severe asthma (SA) remain at high risk for complications, exacerbations, poor quality of life, and increased mortality and morbidity.3 According to the definition of SA proposed by the international European Respiratory Society/American Thoracic Society Task Force, SA is still recognized as a major unmet need, having an overall prevalence estimated at 5% to 10%1-3 and accounting for approximately 50% of asthma-associated health care costs.4

Similar to nonsevere asthma, SA is currently accepted as a heterogeneous disease comprising multiple endotypes that combine clinical characteristics with several identifiable mechanistic pathways.5 Developments in biologic medications to treat asthma have allowed for targeting of specific endotypes, such as allergic asthma, eosinophilic asthma, and more recently, type 2 (T2)-high asthma. It is now standard practice to differentiate patients with SA in terms of having T2-high or T2-low asthma based on the levels of both fraction of exhaled nitric oxide (FENO) and blood eosinophil count.4

SA is also associated with microbial lung dysbiosis, which activates the inflammasome and other induced pathways.5,7 In agreement with the hygiene hypothesis, recent findings have demonstrated that exposure to environmental microbes significantly decreases the incidence of wheezing illnesses in young children with a genetic susceptibility at chromosome 17q21.8 Moreover, the ubiquitous distribution of fungal spores leads to their regular inhalation as an inevitable consequence of breathing.9,10 Exposure to environmental microbes, especially the fungal indoor environment, is associated with a wide range of adverse health effects, including asthma, and it can influence asthma severity.9,11,12

In the present pilot study, our goal was to evaluate the relationships between bacteria and fungi in sputum samples and dust particles from patients with T2-high and T2-low SA. To this end, we examined whether the indoor fungal and bacterial flora (exogenous mycobiome and microbiome, respectively) are associated with clinical parameters in patients with SA and how the exogenous and endogenous mycobiomes and/or microbiomes are connected.

METHODS

Patients

A total of 55 patients from the National COhort of BRonchial obstruction and Asthma (COBRA [ethics committee approval no. 2008-A00294-51/1])16 who had SA based on the European Respiratory Society/American Thoracic Society guidelines and were followed at the University Hospital of Bordeaux were included in the COBRA-ENV (Indoor environment analysis of COBRA cohort) study. The COBRA-ENV protocol (no. 2017/68) was approved by the ethics committee of CPP du Sud-Ouest et Outre Mer III.

Demographic features, clinical characteristics, lung function, and induced sputum test results were determined at COBRA visits as previously described.16 Patients with T2-high SA were defined according to either FENO level (>25 ppb) or blood eosinophil count (>300 cells/mm³).16 In contrast, patients with T2-low SA were defined on the basis of an FENO level lower than 25 ppb and a blood eosinophil count lower than 300 cell/mm³.

Sputum and indoor dust samples

Induced sputa were successfully collected from 22 patients, as previously described.16 Indoor dust samples were collected by using an electrostatic dust collector (EDC). Briefly, each EDC consisted of a 20 × 17-cm textile surface used to catch fungi and bacteria during the exposure period.18 This textile device was mounted in a plastic folder that was left open for 10 weeks in a horizontal position with the textile exposed to the room air such that the settling dust was passively collected.18,19 With use of the same protocol, EDCs were placed in patient bedrooms, opened on the same date (ie, December 21, 2017) and exposed for the same duration (ie, 10 weeks), as described previously.19

DNA extraction, library preparation, deep sequencing, and taxonomic assignment

DNA was extracted from the EDC and sputum samples by using a DNeasy PowerSoil Kit (QIAGEN, Les Ulis, France). Negative extraction controls (250 µL of DNA-free water, as an extraction blank) were processed using the same protocol. The microbial diversity and taxonomic composition of samples were assessed by using the variable 3–variable 4 of the 16S rRNA encoding gene (V3-V4) region of the bacterial 16S rRNA gene and the internal transcribed spacer 2 (ITS2) region of the fungal rDNA, with adherence to the optimized and standardized library preparation protocols from Metabiote (Genoscreen, Lille,
France). The respective primers used to amplify the V3-V4 and ITS2 loci were as follows: 16S-forward, TACGGRAGGCAGCAG; 16S-reverse, CTACCNNGGGATCTAAT; ITS2-forward, GTGARTCATGAATCTTT; and ITS2-reverse, GATATGCTTAAGTTCAGGGGT. In addition to the extraction blanks, 2 negatives controls (1 library blank and 1 unexposed EDC), but analyzed following the same process as the exposed EDCs) and 2 positive controls (artificial bacterial and fungal communities) were used to validate the experimental procedures. Briefly, PCR amplification was performed by using barcoded primers (at a final concentration of 0.2 μM), with an annealing temperature of 50°C for 30 cycles. PCR products were purified by using magnetic beads, quantified according to the protocol provided by GenoScreen, and mixed in equimolar amounts. Next-generation sequencing (NGS) was performed by using 250-bp paired-end technology on the MiSeq platform (Illumina, San Diego, Calif) at GenoScreen (GenoScreen, Lille, France).

The resulting raw sequences were subjected to a cleaning process as follows: (1) sorting of the sequences according to the indexes and the 16S and ITS2 primers using CutAdapt, with no mismatch allowed within the primers sequences; (2) quality filtration using the PRINSEQ-lite PERL script by truncating bases at the 3’ end with Phred quality scores less than 30; and (3) generation of a paired-end read assembly using FLASH with a minimum overlap of 30 bases and greater than 97% identity. The bioinformatic analysis was performed on a fully automated (Metabiota OnLine, version 2.0) pipeline using QIIME, version 1.9.1, software. Following the preprocessing steps, the full-length 16S and ITS2 amplicons were assessed for chimeric sequences by using an in-house method based on Usearch 6.1. Then, a clustering step was performed to group similar sequences with a defined nucleic identity threshold (97% identity for an affiliation at least at the genus level for the V3-V4 locus and at the species level for the ITS2 locus) by using Uclust version 1.2.22q through an open-reference operational taxonomic unit (OTU) picking process and a complete-linkage method to generate groups of sequences or OTUs. An OTU cleaning step was performed to eliminate singletones. For each OTU, the most abundant sequence was used as a reference sequence and was taxonomically compared with reference databases, including the Greengenes database (release 13_8) for bacterial OTUs and the Unite database (release 7.2) for fungal OTUs using the RDP classifier method, version 2.2. A final table with all of the identified OTUs was generated; it comprised the number of sequences for each sample related to each OTU and their respective taxonomic affiliation. α-Diversity metrics within samples (Chao1, Simpson, and Shannon indexes) were generated from this OTU table by using QIIME version 1. The sequence information has been deposited in the National Center for Biotechnology Information Sequence Read Archive under the accession number PRJNA635002.

Estimation of the EDC fungal load using a qPCR assay

The fungal load for each EDC was estimated with a universal quantitative PCR (qPCR) assay that quantifies rDNA (in pg per EDC). Briefly, previously published primers and probes, Tagman Gene Expression Master Mix (Applied Biosystems, Life Technologies, Carlsbad, Calif) and a 7900HT Sequence Detection System were used to perform qPCR. The protocol is detailed in the Methods section of the Online Repository (at www.jacionline.org).

Data analysis

The R programming environment and packages were used for statistical analyses. Nonparametric tests (the Mann-Whitney test for quantitative variables) and the Fisher exact test (for qualitative variables) were used when appropriate. Linear relationships between Feso levels and Chao1 fungal diversity were estimated by using the Pearson correlation coefficient. Differences were considered statistically significant at P values less than .05.

Both α-diversity and β-diversity were determined to identify changes in microbial community structures between EDCs and sputa, according to the clinical data. The principal coordinates analysis (after a Bray-Curtis distance analysis) method was used to visualize overall microbial differences among groups. The microbial core was defined by using the Venn diagram function in the RAM package.

To compare α-diversity estimates across patients’ clinical parameters, Mann-Whitney tests were performed; Wilcoxon tests were used when sputum and EDC pairs were compared. After a cumulative sum scaling normalization, sample clustering hypotheses were tested by using permutational multivariate ANOVA, with a dispersion among groups assessed by using the Betadisper function to complete the β-diversity analysis. Finally, differentially abundant OTUs between groups were identified by using linear discriminant analysis effect size (LEfSe) analysis, as previously described. According to Segata et al., we applied LEfSe analysis with the default parameters (the α value for Wilcoxon tests was set at 0.05, the logarithmic linear discriminant analysis score threshold was set at 2.0, and bootstrapping was performed over 30 cycles) to identify indoor taxonomic biomarkers that characterize the differences between patients with normal and elevated Feso levels by using T2 endotypes as covariables.

Additional details of patient characteristics and sampling, as well as methodologic procedures are provided in the Methods section of the Online Repository.

RESULTS

Study population

A total of 55 patients with SA were enrolled, but only 22 patients with SA had complete pairs of EDC and sputum samples collected and analyzed (see Fig E1 in this article’s Online Repository at www.jacionline.org). Patient characteristics from the T2-high and T2-low groups are summarized in Table I. Except for Feno levels, the characteristics of patients with T2-high SA were not significantly different from those of patients with T2-low SA. Similarly, sputum microbial cultures and indoor fungal loads measured by qPCR were not significantly different between the 2 SA groups (Table I).

Indoor mycobiomes and microbiomes differed significantly according to T2 endotypes

Patients with T2-high SA showed an increase in the 3 α-diversity metrics for the EDC bacterial communities compared with those observed for the patients with T2-low SA, but only the Shannon and Simpson indexes were statistically significant (Fig 1, A). In contrast, the same 3 α-diversity metrics for the EDC fungal communities were lower for the patients with T2-high SA than those observed for the patients with T2-low SA, but only the Shannon indexes were statistically significant (Fig 1, B). Feno levels were significantly correlated with the α-diversity of the mycobiome as measured by Chao1 index values (Fig 1, C). The β-diversity of the EDC fungal communities clustered significantly according to the T2 endotypes (Fig 1, D).

Indoor mycobiomes and microbiomes of patients with T2-high SA were significantly enriched for medically relevant fungi and bacteria

The bacterial composition of the indoor microbiome was dominated at the phylum level by Actinobacteria, Bacteroidetes, Firmicutes, and Proteobacteria, with the most abundant OTUs annotated as Paracoccus spp, Acinetobacter spp, and Streptococcus spp (see Fig E2 in this article’s Online Repository at www.jacionline.org). The fungal community was represented by the 2 major phyla, Ascomycota and Basidiomycota, with the most abundant OTUs annotated as Epicoccum nigrum.
indoor and sputum samples of patients with asthma exacerbation share a mycobiome core composed of medically relevant fungi

In all, 22 patients were able to produce induced sputum during clinically stable periods (n = 10, including 9 patients with T2-high SA and 1 patient with T2-low SA) or exacerbated periods (n = 12, including 7 patients with T2-high SA and 5 patients with T2-low SA) defined at COBRA visit. Both the bacterial and fungal diversities of these sputa were lower than those of the corresponding EDCs (Fig 3, A). The taxonomic composition of these 2 types of samples was significantly different (Fig 3, B-D, and see Figs E2-E5 in this article’s Online Repository at www.jacionline.org).

**TABLE I. Clinical and biologic characteristics of patients with SA and complete NGS analysis**

| Variables                  | Overall (n = 22) | Type 2-low SA (n = 6) | Type 2-high SA (n = 16) | P values |
|----------------------------|------------------|-----------------------|-------------------------|----------|
| Demographic features       |                  |                       |                         |          |
| Age                        | 59.09 (16.84)    | 52.33 (19.69)         | 61.62 (15.58)           | .259     |
| Male sex, no. (%)          | 10 (45.5)        | 2 (33.3)              | 8 (50.0)                | .827     |
| BMI (kg/m²), mean (SD)     | 26.96 (4.73)     | 27.31 (2.98)          | 26.83 (5.32)            | .838     |
| Respiratory data, mean (SD)|                  |                       |                         |          |
| FEV₁ (% predicted)         | 75.36 (27.60)    | 79.32 (33.83)         | 73.77 (25.89)           | .689     |
| FVC (% predicted)          | 91.20 (21.99)    | 89.18 (28.75)         | 92.01 (19.83)           | .798     |
| FEV₁/FVC (% predicted)     | 68.25 (17.04)    | 74.84 (15.82)         | 65.62 (17.31)           | .274     |
| FEF₂₅₋₇₅ (% predicted)     | 49.24 (31.30)    | 61.83 (36.35)         | 44.21 (28.85)           | .254     |
| FENO (ppb)                 | 43.32 (32.74)    | 15.86 (6.26)          | 53.62 (32.75)           | .012     |
| Clinical features at COBRA visit (no.) |          |                       |                         |          |
| Very poorly controlled asthma† | 13             | 4                     | 9                       | .788     |
| SAFS                       | 10               | 3                     | 7                       | 1.000    |
| ABPA                       | 3                | 0                     | 3                       | .657     |
| Asthma medication used during the last 6 mo† |                  |                       |                         |          |
| ICS                        | 22               | 6                     | 16                      | NA1      |
| LABA                       | 22               | 6                     | 16                      | NA       |
| LAMA                       | 16               | 6                     | 10                      | 222      |
| Continuous OCS             | 9                | 3                     | 6                       | .965     |
| Biotherapy                 | 9                | 3                     | 6                       | .965     |
| Gastroesophageal reflux treatment | 8              | 3                     | 5                       | .752     |
| Azole treatment            | 1                | 0                     | 1 (VRC)                 | 1.000    |
| Exacerbations during COBRA-ENV period, mean (SD), no (%) |          |                       |                         |          |
| OCS courses during COBRA-ENV period, mean (SD), no (%) |          |                       |                         |          |
| ATB courses during COBRA-ENV period, mean (SD), no (%) |          |                       |                         |          |
| Biologic data at COBRA visit |                  |                       |                         |          |
| Blood eosinophil count (/mm³), mean (SD) | 526.82 (1184.46) | 12.00 (90.77) | 679.38 (1367.61) | .336     |
| Total serum IgE (kUI/L), mean (SD) | 308.23 (264.80) | 236.67 (268.59) | 335.06 (267.03) | .451     |
| Sputum microbial cultures (no.) |                  |                       |                         |          |
| Pseudomonas aeruginosa     | 1                | 1                     | 0                       | .601     |
| Escherichia coli           | 1                | 1                     | 0                       | .601     |
| Haemophilus influenzae, parainfluenzae | 2              | 1                     | 1                       | 1.000    |
| Non-pneumoniae Streptococcus | 2             | 1                     | 1                       | 1.000    |
| Candida albicans           | 14               | 6                     | 8                       | .094     |
| Other yeast                | 5                | 1                     | 4                       | 1.000    |
| Aspergillus fumigatus      | 5                | 3                     | 2                       | .194     |
| Other Aspergillus          | 3                | 2                     | 1                       | .342     |
| Other mold                 | 3                | 1                     | 2                       | 1.000    |
| Indoor fungal contamination |                  |                       |                         |          |
| Total fungal load (pg/EDC), mean (SD) | 11.39 (9.82) | 11.68 (12.76) | 11.28 (8.98) | .934     |

ARPA: Allergic bronchopulmonary aspergillosis; ATB, antibiotic; BMI, body mass index; COBRA-ENV, indoor environment analysis of COhort of BRonchial obstruction and Asthma; FEF₂₅₋₇₅, forced expiratory flow between the 25% and 75% of the FVC; FVC, forced vital capacity; ICS, inhaled corticosteroid; LABA, long-acting β₂ agonist; LAMA, long-acting muscarinic antagonist; NA, not available; OCS, oral corticosteroid; SAFS, SA with fungal sensitization; VRC, Voriconazole.

†P value with the Mann-Whitney test used for quantitative variables analysis and the Fisher exact test used for qualitative variables.

†NA because all samples were in the same condition regarding these variables.

Cladosporium cladosporioides, Cladosporium sphaerospermum, Malassezia restricta, and Trametes versicolor (see Fig E3 in this article’s Online Repository at www.jacionline.org). All bacterial and fungal OTUs were then subjected to the LEfSe analysis method with T2 endotypes used as covariables.

The LEfSe analysis results indicated significant community enrichment of EDCs from patients with an FENO level higher than 25 ppb and T2-high SA as a covariable; this group of patients consisted of 7 families and 10 genera of bacterial OTUs (Fig 2, A) as well as 2 genera plus 5 species of fungal OTUs (Fig 2, B). Whereas enrichments of EDCs from patients with low FENO levels were primarily composed of environmental Basidiomycota OTUs (Fig 2, B), those from patients with an FENO level greater than 25 ppb and T2-high SA as a covariable were composed of fungal and bacterial OTUs known to be medically relevant in allergies or asthma (ie, Cladosporium, Aspergillus, Epicoccum, Candida, Rhodotorula, Pseudomonas, Pedobacter, Erwinia, Sphingomonas, Sphingomonadaceae, or Methylocystaceae).

Indoor and sputum samples of patients with asthma exacerbation share a mycobiome core composed of medically relevant fungi

In all, 22 patients were able to produce induced sputum during clinically stable periods (n = 10, including 9 patients with T2-high SA and 1 patient with T2-low SA) or exacerbated periods (n = 12, including 7 patients with T2-high SA and 5 patients with T2-low SA) defined at COBRA visit. Both the bacterial and fungal diversities of these sputa were lower than those of the corresponding EDCs (Fig 3, A). The taxonomic composition of these 2 types of samples was significantly different (Fig 3, B-D, and see Figs E2-E5 in this article’s Online Repository at www.jacionline.org).
These results revealed the highly individual signature of sputum mycobiomes and microbiomes compared with that observed for the EDCs, which exhibited a very closely spaced clustering of samples (Fig 3, C and D). For instance, the proportion of Basidiomycota, a phylum primarily composed of environmental fungi, compared with the proportion of Ascomycota, which includes medically relevant fungi, was higher in the EDCs than in the sputa (Fig 3, B), highlighting the complex microbial ecology of the human respiratory tract.

We did not observe any differences in the bacterial and fungal diversities of these sputum samples according to T2 endotypes (data not shown). In contrast, the proportion of common bacterial taxa between the EDCs and sputa was similar when sputa were collected during exacerbation or at a stable state (Fig 3, E), whereas the proportion of common fungal taxa between EDCs and sputa was significantly higher when sputa were collected during the exacerbation period than the stable period (Fig 3, F).

The Venn diagram results confirmed a limited bacterial core composed of 5 OTUs (Fig 4, A) and a large fungal core composed of 27 OTUs (Fig 4, B).

We identified the core microbes between pairs of sputa and EDCs according to the absence or presence of an asthma exacerbation and confirmed that only 1 OTU was shared between the bacterial core and the clinical states (ie, Sphingomonas [Fig 4, A]). Surprisingly, 9 fungal OTUs (ie, Hypoderntia radula, Resinicium bicolor, Phlebia radiate, Penicillium brevicomactum, Vishniacozyma victoriae [synonym Cryptococcus victoriae], Sistotremastrum spp, Cryptococcus magnus, Bjerkandera spp, and Botryotinia fuckeliana [Fig 4, B]) were identified as the fungal core and were specifically shared between EDCs and sputa collected during exacerbation. In addition, 17 other OTUs were identified as members of the fungal core independent of the clinical status, which included medically relevant yeasts (ie, Saccharomyces cerevisiae, Debaryomyce hansenii, and Malassezia species) and molds (Aspergillus and Cladosporium species) or environmental Basidiomycota and Ascomycota. Only 1 Alternaria brassicae OTU was shared with sputa collected at a stable state (Fig 4, B). Among these OTUs, several were both microbial core members and significantly enriched OTUs in EDC microbial communities, including C sphaerospermum, T versicolor, and Sphingomonadaceae (Figs 2, B and 4, B).

**DISCUSSION**

To the best of our knowledge, this pilot study is the first to report that the indoor mycobiomes and microbiomes of patients with SA exhibit distinct signatures according to T2 endotypes. On the basis of pairs of EDC and sputum samples, we compared exogenous and endogenous microbial communities and identified a fungal core comprising medically relevant fungi that was.

[FIG 1. The α- and β-diversities of the EDCs from patients with SA according to T2 endotypes (n = 22). The α-diversity of bacterial (A) or fungal (B) communities from EDCs is shown using Chao1, Shannon, and Simpson indexes. C, Correlation between Feno levels and Chao1 fungal diversities. D, The β-diversity of fungal communities. Patients were split according to whether than had T2-high SA (dark gray, circles) or T2-low SA (light gray, triangles).]
significantly more pronounced when sputa were collected during asthma exacerbations.

Regarding the indoor environment, we clearly demonstrated that a higher bacterial α-diversity together with a lower fungal α-diversity was associated with patients with T2-high SA and that the fungal β-diversity of EDC communities was clustered according to T2 endotypes. Moreover, the indoor mycobiomes and microbiomes of patients with an FENO value higher than 25 ppb and T2-high SA were significantly enriched for medically relevant fungi and bacteria. Among them, we identified changes in the relative abundances of Sphingomonadaceae, Methylocystaceae, Erwinia, Sphingomonas, Pseudomonas, and Candida, which is in agreement with previously published data on the gut microbiota and allergy and/or asthma. Although not exclusively focused on T2 endotypes, several studies have previously investigated the respiratory microbiomes or mycobiomes of patients with varying degrees of asthma severity. In addition, an excess of Proteobacteria was shown to be associated with moderately severe to severe asthmatics, whereas an excess of Streptococcus was shown to be associated with eosinophilic asthma. However, little is known regarding the lung mycobiome in chronic respiratory diseases. Malassezia yeasts have been shown to be significantly associated with asthma and more recently with exacerbation in cystic fibrosis. The mycobiomes of patients with SA were shown to harbor higher loads of fungi than were patients with nonsevere asthma or healthy individuals, with a lower diversity observed in patients with T2-high asthma. A low indoor fungal diversity has been proposed to be a causal factor in the development of childhood asthma, and although a few studies considered adult patients with asthma with respect to its severity and control, none of them referred to T2 endotypes and combined exogenous to endogenous mycobiome and microbiome analyses.

The first studies of indoor microbial ecology used culture-based techniques, but such studies have more recently been performed by using NGS-based approaches. Overall, sampling the indoor microbiome by a passive collector such as an EDC allows for the recovery of house dust particles, which are a complex mixture of inorganic and organic substances. The indoor mycobiome appears to be a large fungal community assembly that is primarily derived from the outdoor environment but also depends on the inhabitants’ mycobiomes under selective pressures. As previously reported, we sampled patient bedrooms, which are known to accurately estimate the level of indoor fungal pollution. Our NGS results for the EDC mycobiome and microbiome were congruent with these previous studies. Briefly, we identified Actinobacteria, Bacteroidetes, Firmicutes, and Proteobacteria as the most abundant bacterial phyla in agreement with recently published data. Within these phyla, Pseudomonas and Sphingomonas were significantly enriched in the EDCs of patients with T2-high SA. Regarding the mycobiome, Aspergillus versicolor...
...and *C. sphaerospermum* have been identified as indoor species correlated to asthma. Using NGS, Dannemiller et al identified numerous fungi that were significantly associated with SA. Our results confirmed that several *Cryptococcus* species and *E. nigrum* were associated with SA and that the genus *Epicoccum* was significantly enriched in the indoor mycobiome of patients with T2-high SA. Moreover, our indoor data are congruent with the recent identification of *Cladosporium* and *Aspergillus* as dominant enriched genera of respiratory samples from patients with T2-high asthma.

In addition to viral infections, fungal sensitization and fungal exposure have been shown to be highly associated with asthma exacerbations and were recently reviewed. We confirmed that fungal exposure may be a key player during exacerbation, as the fungal OTU proportion shared by EDCs and sputa was significantly higher when sputa were collected during asthma exacerbation. These results support the idea that microbial analysis should not be restricted to a specific type of microorganism (viruses, bacteria, or fungi) but rather should involve analyses of interkingdom interactions that may be involved in promoting exacerbations.

**FENO** is an endogenous gaseous molecule incorporated into the clinical management of chronic respiratory diseases and is currently recognized as a biomarker of T2 airway inflammation. Through use of conventional mycologic methods (ie, cultures and PCR), FENO levels of patients with asthma have been related to environmental exposure whereas the indoor isolation of *A. versicolor* and *Cladosporium* have been associated with higher FENO levels, which is in agreement with our results. FENO levels were shown to be positively correlated with T2 cytokine levels in patients with asthma and with serum IL-17A levels in patients with SA. As adaptive antifungal immunity includes both Th2- and Th17-type CD4 T cells, we hypothesize that a number of the fungi isolated from the indoor mycobiome of patients with SA (such as *Aspergillus* and *Cladosporium* species) should contribute to a detrimental immunopathology in patients with SA. Once fungi are inhaled and colonize the respiratory airways of patients, they participate in complex microbial-host interactions by producing secondary metabolites that are well known to be involved in this immune response, such as cell wall components (β-glucan and/or chitin) or secreted enzymes (proteases and/or glycosidases, or toxins).

Although our work is novel, it does have several limitations, including its cross-sectional design and the limited number of fully analyzed pairs of EDC plus sputum, which reflects...
difficulties in sampling the respiratory tract. Indeed, the small number of patients limits the generalizability of our results, especially regarding LEfSe analysis results. For instance, the absence of difference between the group with T2-low SA and that with T2-high SA with respect to the exacerbation rate is probably underpowered. We used induced sputa to analyze the endogenous microbiomes and mycobiomes, a sampling method recently proposed as an acceptable and less invasive alternative compared with bronchoalveolar lavage samples but with several biases.\textsuperscript{59-61} This approach can be considered to be a bias of selection as only patients with higher than 50% predicted postbronchodilator FEV\textsubscript{1} value were allowed to achieve induced sputum. As most of the published studies on this topic have focused on asthma-associated microbiomes and mycobiomes, this study suffers from small size limitations and a lack of longitudinal data. Furthermore, amplicon deep sequencing represents a promising method, but the protocols need to be improved and standardized. Indeed, there are many potential biases, ranging from the primer, amplification protocol, and NGS machine used to the pipeline and databases selected for analysis.\textsuperscript{10,62} Thus, further studies are warranted to confidently determine the relationships between the indoor environment, the lung mycobiome and microbiome, the inflammatory response, and the development and severity of asthma.

In summary, in the present study, we identified a correlation between the indoor mycobiome and FENO levels of patients with SA (Fig. 1, C), which should renew interest in deciphering the interactions between the indoor environment, fungi, and the host to facilitate the development of novel therapeutic management.\textsuperscript{14} Given the role of fungi as potent inducers of airway inflammation,\textsuperscript{72} an increase in FENO level may serve as an indicator of respiratory inflammation related to exposure to specific molds that the indoor mycobiome analysis will characterize.
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Key messages
- The indoor mycobiomes and microbiomes of patients with SA harbor a distinct signature according to T2 endotypes, with lower fungal diversity but higher bacterial diversity associated with T2-high SA.
- The indoor environments of patients with T2-high have significant enrichment of medically relevant fungi and bacteria, such as Aspergillus, Candida, Sphingomonas, and Pseudomonas.
- The respiratory mycobiome of patients with SA shares more fungal taxa with the indoor mycobiome during exacerbation than during stable periods.

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METHODS

Study patients

A diagnosis of SA with fungal sensitization was considered in patients with SA who were sensitized to 1 or more fungi (as demonstrated by skin prick tests or serum specific IgE) but had a normal serum total IgE level (<150 kU/L) and were negative for allergic bronchopulmonary aspergillosis. Asthma control was established by using the 7-Item Asthma Control Questionnaire, with a score higher than 1.5 reflecting very poorly controlled asthma.

The numbers of moderate asthma exacerbations (ie, loss of asthma control requiring oral steroid and/or antibiotic courses) were recorded within the year preceding the EDC exposure (ie, from December 2016 to December 2017, during the EDC exposure time (ie, from December 2017 to March 2018), and during the year following EDC exposure (ie, from March 2018 to March 2019). Spirometry was performed according to American Thoracic Society standard. The Fino level (in ppb) was assessed, as previously described.

Sputum samples and EDC

Induced sputa were successfully collected from 22 patients as previously described. Briefly, after oral rinse to prevent excessive salivary contamination, inhalation of increasing concentrations of nebulized saline solutions at 3%, 4%, and 5%, respectively, for a 7-minute session each was used. For patients with an FEV1 value of at least 50% predicted but less than 60% predicted, induction was performed with normal/isotonic saline (0.9%). Sputa from subjects with a postbronchodilator FEV1 value less than 50% predicted were not induced. Sputa treated with dithiothreitol were subjected to microbial cultures and stored at –20°C until DNA extraction.

All the patients included into COBRA-ENV (indoor environment analysis of CONhor of BRonchial obstruction and Asthma) study received guidelines during their COBRA visit to deploy an EDC at home (in the patient’s bedroom) according to the same protocol. The EDCs were then sent to the mycology laboratory, where each EDC was processed as described previously. Each collected solution was stored at –20°C until DNA extraction. As a negative control, 1 unexposed EDC was subjected to the same laboratory protocol.

DNA extraction, library preparation, deep sequencing, and taxonomic assignment

DNA extraction was carried out on the EDC and sputum samples by using a DNeasy PowerSoil Kit (Qiagen). Negative extraction controls (250 µL of DNA-free water) were processed using the same protocol. The microbial diversity and taxonomic composition of samples were assessed by using the V3-V4 of bacterial 16S gene and the ITS2 region of the fungal rDNA, according to an optimized and standardized library preparation protocol Metabiote (Genoscreen, Lille, France). Two positive qualitative controls comprising an artificial bacterial community and a fungal mock community and 2 negatives controls were used to validate the experimental processes. NGS sequencing was performed by using a 250-bp paired-end sequencing protocol on the Illumina MiSeq platform, as described previously. Taxonomic analysis was performed with the Metabiote Online version 2.0 pipeline (GenoScreen, Lille, France), as previously described. Each OTU was then taxonomically compared with reference databases as follows: the Greengenes database (release 13_8) for bacterial OTUs and the Unite database (release 7.2) for fungal OTUs.

Estimation of EDC fungal load using qPCR assay

The fungal load of each EDC was estimated with a universal qPCR assay quantifying 18S rDNA (in pg per EDC). Briefly, the Taqman Gene Expression Master Mix (Applied Biosystem, Life Technologies) kit was used according to the manufacturer’s instructions. qPCR was performed using the 7900HT Sequence Detection System and the following published primers: FungiQuantF: 5-GGAAATCTACACCAGTGTCG-3; FungiQuantR: 5-GSWCTATCCCCAKCACGA-3; and FungiQuantPhr. (6FAM)-TGGTGT CATGAGCGCTT-3(MGBNFQ). The cycling conditions were 15 seconds at 95°C and 40 repeats of 15 seconds at 92°C (denaturation step) and 1 minute at 60°C (annealing and extension steps). Each cycling condition was repeated under the basis of a standard comprising 11 genomic DNAs extracted from different fungal strains (the artificial fungal community MetaBiote Control) by using the ΔΔCt method. Each sample was analyzed in triplicate; the Ct mean was finally compared with the standard curve to deduce the quantity of fungal 18S DNA of each EDC.

Data analysis

Measurement of α-diversity (within-sample) included computation of the Shannon index, Simpson index, and number of observed OTUs and fungal OTUs (Chao1 index). For β-diversity (between-sample) measurements, we used the vegan package to compute the Bray-Curtis distances. The principal coordinates analysis method was used to visualize group overall microbial differences with use of the phyloSeq package. To determine and analyze the microbial core, OTUs not contained in at least 20% of spota per group and 90% of EDCs were removed and the Venn diagram function of the RAM package was used.

To compare α-diversity estimates across patients’ clinical parameters, Mann-Whitney tests were performed; the Wilcoxon test was used when pairs of sputum and EDC samples were compared. To analyze β-diversity between samples, OTU tables normalized via cumulative sum scaling by metagenomeSeq package were used. Sample clustering hypotheses were tested by using permutation multivariate ANOVA. Homogeneity of dispersion among sample groups was assessed by using the Betadisper function (vegan package). To detect differentially abundant OTUs between groups, LEfSe analysis was performed.

It induced the ranking of biomarker relevance with a substantial false positive rate (mainly in <0.5% of the cases) and a high false-negative rate.

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FIG E1. Flowchart of the COBRA-ENV (indoor environment analysis of COhort of BRonchial obstruction and Asthma) study and samples distribution across time.
FIG E2. Histograms representing the relative abundance at the genus level of the EDC microbiome.
FIG E3. Histograms representing the relative abundance at the genus or species level of the EDC mycobiome.
FIG E4. Histograms representing the relative abundance at the genus level of the sputum microbiome.
FIG E5. Histograms representing the relative abundance at the genus or species level of the sputum mycobiome.