Increased urinary L-histidine in patients with asthma–COPD overlap: a pilot study

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Purpose: Asthma–COPD overlap (ACO) is heterogeneous in nature and requires a unified diagnostic approach. We investigated the urinary levels of L-histidine, a precursor of histamine related to inflammatory responses, as a new candidate biomarker for diagnosing this condition.

Patients and methods: We performed a prospective multicenter cohort study with retrospective analysis of 107 patients, who were divided into three groups: asthma, COPD, and ACO, according to the Spanish guidelines algorithm. Urinary L-histidine levels were measured using liquid chromatography-mass spectrometry. High-resolution metabolomic analysis, coupled with liquid chromatography-mass spectrometry and followed by multivariate statistical analysis, was performed on urine samples to discriminate between the metabolic profiles of the groups.

Results: Urinary L-histidine levels were significantly higher in patients with ACO than in those with asthma or COPD, but the subgroups of ACO, classified according to disease origin, did not differ significantly. High urinary L-histidine level was a significant factor for the diagnosis of ACO even after adjusting for age, sex, and smoking amount. Among patients with airflow obstruction, the urinary L-histidine levels were elevated in patients with a documented history of asthma before the age of 40 years or bronchodilator responsiveness ≥400 mL; bronchodilator responsiveness ≥200 mL of forced expiratory volume in 1 second and exceeding baseline values by 12% on two or more visits; blood eosinophil count ≥300 cells·μL⁻¹; and frequent exacerbations (P < 0.05).

Conclusion: Urinary L-histidine could be a potential biomarker for ACO, regardless of the diversity of diagnostic definitions used.

Keywords: asthma, COPD, ACO, urinary L-histidine, metabolomics, inhaled corticosteroid

Introduction

Asthma–COPD overlap (ACO) was initially described as a condition characterized by an increased variability in airflow associated with an incompletely reversible airflow obstruction. However, there is no universally accepted definition for this condition, and this leads to diverse epidemiology and outcomes of ACO. Indeed, ACO is not a single disease, but a combination of syndromes, consisting of several different endotypes and phenotypes with diverse pathogeneses. Classification based on the inflammatory profiles and treatable traits, irrespective of taxonomy, would help provide more personalized treatment. However, the identification of new biomarkers for treatable symptoms is needed.

The main reason for diagnosing ACO in clinical practice is to identify patients with COPD who are likely to have a better response to inhaled corticosteroids (ICSs). Recently, the use of a Th-2 inflammatory signature (defined by a blood eosinophil count ≥300 cells·μL⁻¹ and/or a spum eosinophil count ≥3%) and genomic signatures...
of Th-2 inflammation, instead of historical classifications of chronic obstructive airway disease (COAD), was suggested to
guide ICS therapy.6,7 However, the cutoff points of blood
eosinophil counts are arbitrary, and other Th-2 signatures
are difficult to measure. Thus, more markers that represent
Th-2 features are required to distinguish ACO for clinical
applications.

Histamine is important in the inflammatory response and
airway smooth muscle contraction.8,9 Patients with ACO
might secrete higher levels of histamine than patients with
asthma or COPD, because of airway hyper-responsiveness.10,11
repetitive Th-2-related inflammation, and smooth muscle
contraction.11,12 Moreover, urine is easy to collect and has
a high concentration of metabolites,13 which are important
indicators of pathological states.14,15

Thus, we investigated urinary L-histidine, a precursor of
histamine, as a new candidate for the diagnosis of ACO
by using high-resolution metabolomics (HRM) and liquid
chromatography-mass spectrometry (LC-MS), which is an
emerging technique to measure small molecules <1,000 Da
in given samples simultaneously and to determine the impor-
tant features associated with diseases.

**Patients and methods**

**Study design**

We performed this prospective cohort study for developing
optimal clinical and biological markers for the diagnosis and
monitoring of treatment response in patients with chronic
airway disease. This multicenter study was performed in
Korea University Guro Hospital, Seoul St Mary’s Hospital,
and Chonbuk National University Hospital and used the
airway disease registry from June 2015 to May 2016. The
inclusion criteria for the cohort were as follows: adult
patients aged ≥19 years and patients clinically diagnosed
with asthma, COPD, or ACO who had been in a stable state
without exacerbation and worsening of symptoms for at least
3 months. Patients with all other diseases that could cause
airflow obstruction, including bronchiectasis, tuberculosis-
destroyed lungs, or interstitial lung disease, were excluded.

The patients were divided into three groups based on age,
medical history, smoking history, pulmonary function test
(PFT) results, and blood eosinophil count. COPD (n = 38)
was diagnosed on the basis of the American Thoracic Society
and Global Initiative for Chronic Obstructive Lung Disease
guidelines,16 ie, airflow limitation confirmed by post-broncho-
dilator spirometry, with forced expiratory volume in 1 second
(_FEV_1)/forced vital capacity (FVC) <70% and a history of
smoking or biomass exposure. Asthma (n = 32) was defined
according to the Global Initiative for Asthma guidelines.17

The diagnosis of ACO (n = 37) was based on the GesEPOC
(the Spanish COPD guidelines) and GEMA (the Spanish
Guidelines for the Management of Asthma) algorithms.18

Patients aged >35 years with significant exposure to smoking
(≥10 pack-years), persistent airflow limitation, and with a
current diagnosis of asthma or with a significantly positive
bronchodilator response (BDR ≥ 400 mL and FEV_1 ≥ 15%) and/or
significant blood eosinophilia (≥300 cells mm^-3) were
included in the ACO group.

The following variables were evaluated: smoking history
and amount, body mass index (BMI), modified Medical
Research Council (mMRC) dyspnea score, St George
Respiratory Questionnaire (SGRQ) score, COPD assessment
test (CAT) score, asthma control test (ACT) score, PFT
(spirometry, lung volume, and diffusing capacity of the lung
for carbon monoxide [DL_co] results, fractionated exhaled
nitric oxide (FeNO; ppb), and exacerbation history for 1 year.
Blood tests, including eosinophil counts, were performed
routinely on the day of enrollment. Urine samples in the
overnight-fasted and medication-free state were collected.

FeNO measurement was performed online by using the
NIOX MINO® instrument (Aerocrine AB, Solna, Sweden)
at 50 mL min⁻¹. An exacerbation was defined as a sustained
worsening of the patient’s condition, from the stable state
and beyond normal day-to-day variations, that is acute in
onset and necessitates a change in regular medication with
an unscheduled Dr’s visit because of the worsened respira-
tory symptoms. Moderate exacerbation was defined as an
increase in symptoms that requires treatment with antibiotics
and/or corticosteroids, and severe exacerbation was defined
as one that requires hospitalization. Patients with two or more
moderate-to-severe exacerbations within 1 year and with one
or more hospitalizations for severe exacerbation were defined
as frequent exacerbators. A significantly positive BDR was
defined as ≥400 mL and high eosinophilia was defined as
blood eosinophil count ≥300 cells·µL⁻¹.

The study protocol was approved by the institutional
review boards (Korea University Guro Hospital: KUGH
13246; Seoul St Mary’s Hospital: KC15OIM0553; and
Chonbuk National University Hospital: 2015-01-018-005).
Written informed consent was provided by the patients,
and the study conformed to the tenets of the Declaration of
Helsinki.

**HRM coupled with LC-MS**

For HRM analyses, 50 µL of sample aliquots was treated
with acetonitrile (Burdick & Jackson, Muskegon, MI, USA)
(1:4, v/v) and centrifuged at 13,000 × g for 10 min at 4°C
to separate proteins.19,20 Urine samples were analyzed using
Agilent 1290 Infinity ultra performance liquid chromatography (UPLC) (Agilent Technologies, Santa Clara, CA, USA) connected to a 6550 iFunnel Q-ToF LC-MS with an electrospray ionization (ESI) source (Agilent Technologies). Samples were analyzed in the ESI-positive ion mode. Mobile phases were high-performance liquid chromatography (HPLC)-grade water (Tedia, Fairfield, OH, USA) with 0.1% formic acid (Fluka, St Louis, MO, USA) (A) and HPLC-grade acetonitrile (Burdick & Jackson) with 0.1% formic acid (Fluka) (B). UPLC gradient conditions were as follows: 0.0–1.0 min 95% A; 1.0–9.0 min 55% A; 9.0–12.0 min 10% A; 12.0–13.5 min 10% A; and 13.5–13.6 min 95% A at a flow rate of 0.4 mL min⁻¹. The injection volume was 3 μL, and each sample was randomly analyzed in triplicate. The analytical column was a Hypersil GOLD aQ (Thermo Fisher Scientific, Waltham, MA, USA; 1.9 μm, 100 × 2.1 mm²) maintained at 45°C. The gas temperature was 250°C, and the nebulizer pressure was 35 psig. The flow rates of the drying gas and sheath gas were 14 L min⁻¹ and 11 L min⁻¹, respectively. Mass spectra, ranging from a mass-to-charge ratio (m/z) of 50–1,000, were collected with 20,000 resolutions over 15 min LC runs.²¹

Data preprocessing and extraction of mass spectral data
Mass spectral data acquired were converted to mzXML by using ProteoWizard (http://proteowizard.sourceforge.net/) followed by adaptive processing of high-resolution LC-MS data (apLCMS) to obtain ion intensities.²¹ The apLCMS provided 5,213 m/z within a range of ions set from 50 to 1,000. Data were further processed with normalization, scaling, filtering, and statistical analysis using MetaboAnalyst 3.0 (www.metaboanalyst.ca), which is a web server designed to analyze metabolomic data comprehensively for visualization and interpretation. MetaboAnalyst improves the statistical analysis by filtering the non-informative variables that are unlikely to be of use for further modeling of data, by applying an interquartile range (IQR). In preprocessing mass spectral data, data sets were quantile normalized, log transformed, and pareto-scaled to obtain more comparable individual features before carrying out statistical analyses.

Determination of metabolic profiles and identification of potential metabolites of ACO from asthma and COPD
Metabolites altered between each pair of groups were considered important toward the identification of potential metabolites related to the metabolic effects of ACO that were different from those of asthma and COPD. Accurate masses of significant metabolites were searched against the online biochemical database service The Human Metabolome Database (http://www.hmdb.ca).²²

Statistical analysis
Clinical data were presented as the median and IQR for continuous variables and as a percentage for categorical variables. Data were compared using the Kruskal–Wallis test for continuous variables, comparing each two groups using the Mann–Whitney U-test with Bonferroni corrections (α < 0.017; significant). For categorical variables, Pearson’s χ² test or Fisher’s exact test was used. Correlations between plasma biomarkers and parameters were analyzed using Spearman’s correlation analysis. Significance was defined as P < 0.05. All analyses were conducted using IBM SPSS Statistics for Windows, Version 20.0 (IBM Corporation, Armonk, NY, USA). Student’s t-test and orthogonal signal correction partial least squares-discriminant analysis (OPLS-DA) were performed using MetaboAnalyst 3.0, to identify the differentially expressed metabolites among the comparison groups. The intensity of L-histidine in the compared groups was analyzed using GraphPad Prism v 5.03 (GraphPad Software, Inc., La Jolla, CA, USA).

Results
Patient characteristics
A total of 107 patients were enrolled. Their baseline characteristics are described in Table 1. Patients with asthma were significantly younger than patients with COPD. Most patients in the COPD and ACO groups were male. All of the patients with COPD and ACO were smokers by definition. Comorbidities other than allergic rhinitis were not significantly different between the groups. The mMRC and SGRQ scores of patients with COPD were higher than those of patients with asthma. In the PFTs, BDR was the highest in patients with ACO. Median serum creatinine levels (asthma, 0.8 [0.6–0.9] mg dL⁻¹; COPD, 0.8 [0.6–0.9] mg dL⁻¹; ACO, 0.7 [0.6–0.8] mg dL⁻¹; P = 0.167) and creatinine clearance rates (asthma, 100.5 [71.5–120.9] mL min⁻¹; COPD, 95.7 [66.4–123.0] mL min⁻¹; ACO, 95.7 [66.4–122.9] mL min⁻¹; P = 0.606) were similar among the groups.

Urine metabolomics
To observe the urinary metabolic alterations among asthma, COPD, and ACO, we compared the three groups through multiple testing, by inserting the apLCMS feature table containing 5,213 features obtained from 107 urine samples. Student’s t-test was first utilized to observe the difference in
Table 1  Comparison of the baseline characteristics of chronic inflammatory airway diseases

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Asthma n = 32</th>
<th>COPD n = 38</th>
<th>ACO n = 37</th>
<th>P-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age, years</td>
<td>57.0 (47.0–66.8)</td>
<td>67.0 (61.5–74.5)</td>
<td>64.0 (57.0–71.5)</td>
<td>0.001</td>
</tr>
<tr>
<td>Sex, male, n (%)</td>
<td>12 (37.5)</td>
<td>19 (50.0)</td>
<td>13 (35.1)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>BMI (kg m⁻²)</td>
<td>23.2 (21.5–26.9)</td>
<td>22.3 (20.6–23.9)</td>
<td>24.6 (21.3–27.3)</td>
<td>0.039</td>
</tr>
<tr>
<td>Smoker, n (%)</td>
<td>3 (9.4)</td>
<td>38 (100.0)</td>
<td>37 (100.0)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>Ex-smoker, n (%)</td>
<td>1 (3.1)</td>
<td>17 (44.7)</td>
<td>23 (62.2)</td>
<td></td>
</tr>
<tr>
<td>Smoking amount (pack-years)</td>
<td>0.0 (0.0–0.0)</td>
<td>40.0 (33.8–54.8)</td>
<td>35.0 (20.0–44.5)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>Allergic rhinitis</td>
<td>19 (59.4)</td>
<td>2 (5.3)</td>
<td>17 (45.9)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>Heart diseases</td>
<td>12 (37.5)</td>
<td>16 (42.1)</td>
<td>16 (43.2)</td>
<td>0.585</td>
</tr>
<tr>
<td>Symptom scores</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>mMRC</td>
<td>0.0 (0.0–1.0)</td>
<td>1.0 (0.0–2.0)</td>
<td>0.0 (0.0–1.0)</td>
<td>0.015</td>
</tr>
<tr>
<td>CAT</td>
<td>–</td>
<td>14.0 (10.0–22.0)</td>
<td>12.0 (9.5–16.8)</td>
<td>0.050</td>
</tr>
<tr>
<td>ACT</td>
<td>23.0 (21.0–24.0)</td>
<td>–</td>
<td>20.0 (14.0–22.3)</td>
<td>0.001</td>
</tr>
<tr>
<td>SGRQ</td>
<td>10.5 (5.1–15.0)</td>
<td>16.9 (11.0–32.6)</td>
<td>15.3 (8.1–23.6)</td>
<td>0.008</td>
</tr>
<tr>
<td>PFTs</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>FEV₁, L</td>
<td>2.6 (2.0–2.8)</td>
<td>1.7 (1.0–2.5)</td>
<td>1.9 (1.6–2.4)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>FEV₁, %</td>
<td>91.0 (78.3–102.8)</td>
<td>66.5 (35.8–76.3)</td>
<td>65.0 (49.0–71.5)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>FEV₁/FVC, %</td>
<td>76.0 (71.3–80.8)</td>
<td>54.0 (34.8–58.3)</td>
<td>56.0 (44.5–64.5)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>FEF 25−75%, %</td>
<td>64.5 (47.0–92.8)</td>
<td>21.5 (10.8–28.5)</td>
<td>27.0 (16.5–37.0)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>BDR, mL</td>
<td>105.0 (32.5–150.0)</td>
<td>50.0 (17.5–142.5)</td>
<td>290.0 (200.0–400.0)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>TLC, %</td>
<td>101.5 (94.3–108.5)</td>
<td>108.5 (102.8–117.0)</td>
<td>110.0 (103.5–116.0)</td>
<td>0.013</td>
</tr>
<tr>
<td>FRC, %</td>
<td>107.0 (92.3–123.0)</td>
<td>141.5 (128.8–157.0)</td>
<td>130.0 (110.0–149.5)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>RV, %</td>
<td>103.5 (86.3–125.8)</td>
<td>142.0 (132.8–155.8)</td>
<td>138.0 (120.5–153.0)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>DLco, %</td>
<td>97.0 (90.3–110.8)</td>
<td>69.5 (47.5–98.8)</td>
<td>90.0 (67.5–98.5)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>FeNO, ppb</td>
<td>22.0 (17.0–35.0)</td>
<td>18.0 (11.5–25.5)</td>
<td>20.0 (10.3–35.8)</td>
<td>0.198</td>
</tr>
<tr>
<td>Frequent exacerbator, n (%)</td>
<td>1 (3.1)</td>
<td>5 (13.2)</td>
<td>6 (16.2)</td>
<td>0.204</td>
</tr>
</tbody>
</table>

Notes: Data are presented as median (IQR) for continuous variables and percentage (number) for categorical variables. *Significantly different between asthma and COPD; †significantly different between COPD and ACO; ‡significantly different between asthma and ACO.

Abbreviations: ACO, asthma–COPD overlap; ACT, asthma control test; BDR, bronchodilator response; BMI, body mass index; CAT, COPD assessment test; DLco, diffusing capacity of the lung for carbon monoxide; FEF, forced expiratory flow; FeNO, fractionated exhaled nitric oxide; FEV₁, forced expiratory volume in 1 second; FRC, functional residual capacity; FVC, forced vital capacity; IQR, interquartile range; mMRC, modified Medical Research Council; PFTs, pulmonary function tests; RV, residual volume; SGRQ, St George Respiratory Questionnaire; TLC, total lung capacity.

the urinary metabolic phenotype of ACO in comparison with asthma and COPD. Of the 5,213 features, 223 were found significant in the t-test (P < 0.05) between the ACO and asthma group urine samples, whereas 215 features were significantly different between the ACO and COPD group urine samples. Furthermore, to better define the metabolic alteration among the three groups and to visualize the variation among groups, we performed OPLS-DA. In OPLS-DA, the entire sample was separated in a clear separation from the clusters of asthma (Figure 1A) and COPD (Figure 1B). Such an apparent separation of ACO in comparison with asthma and COPD indicated that the metabolic profile of ACO was highly differentiated from that of asthma and COPD. Moreover, the results of the cross-validation procedure indicated that neither models were overfitting and that both models were reliable, as shown in Figure 1A (intercepts: $R^2 = 0.964$ and $Q^2 = 0.153$) and Figure 1B (intercepts: $R^2 = 0.971$ and $Q^2 = 0.046$).

Asthma and COPD urine samples also showed a similar separation pattern in OPLS-DA, as shown in Figure 1C (intercepts: $R^2 = 0.963$ and $Q^2 = 0.163$).

After observing such variations in the metabolome of ACO, we desired to analyze the urinary level of L-histidine in ACO compared with that in asthma and COPD. Interestingly, as shown in Figure 2A, the urinary level of L-histidine (m/z 178.0571, [M+Na]⁺) was significantly higher in patients with ACO than in those with asthma and COPD. Furthermore, to confirm that L-histidine was specifically elevated among patients with ACO, we analyzed the urine samples of patients with asthma and compared them with those of patients with COPD. Although the OPLS-DA showed a clear separation between the two groups (Figure 1C), the urinary level of L-histidine was not significantly different among patients.
Increased urinary L-histidine in ACO

with asthma and COPD (Figure 2A). These results suggested that L-histidine was specifically elevated among patients with ACO.

Urinary L-histidine in chronic airway disease

Although the median urinary L-histidine levels were significantly higher in patients with ACO than in those with asthma or COPD (Figure 2A), no significant difference was observed in the levels of urinary L-histidine between smoking-related obstructive asthma (n = 27) and COPD with a significantly positive BDR and/or blood eosinophilia (BDR ≥ 400 mL and/or blood eosinophil count ≥ 300 cells·mm⁻³, n = 10) (Figure 2B).

The overall sensitivity and specificity of ACO diagnosis were 75.7% and 63.0%, respectively, at the urinary

Figure 1 Differential urinary metabolic profiles in patients with asthma–COPD overlap (ACO), asthma, and COPD.

Notes: (A) Discrimination of the urine metabolome between patients with ACO and those with asthma by using orthogonal signal correction partial least squares-discriminant analysis (OPLS-DA). Intercepts: $R^2 = 0.964$ and $Q^2 = 0.153$. (B) Discrimination of the urine metabolome between patients with ACO and those with COPD by using OPLS-DA. Intercepts: $R^2 = 0.971$ and $Q^2 = 0.0466$. (C) Discrimination of the urine metabolome between patients with asthma and those with COPD by using OPLS-DA. Intercepts: $R^2 = 0.963$ and $Q^2 = 0.163$. 
l-histidine cutoff of 13,860. Multivariate logistic regression revealed that a high urinary l-histidine level was a significant parameter of ACO even after adjusting for age, sex, and smoking amount (urinary l-histidine > 13,860; OR, 5.254; 95% CI, 2.085–13.236; \( P < 0.001 \)).

Urinary l-histidine levels and parameters
In the total study population, urinary l-histidine levels were positively correlated with SGRQ scores (\( R = 0.222; P = 0.033 \)) and negatively correlated with ACT scores (\( R = -0.284; P = 0.024 \)). In the PFTs, urinary l-histidine level was positively associated with BDR (\( R = 0.300; P = 0.002 \)), total lung capacity (\( R = 0.242; P = 0.012 \)), and residual volume (\( R = 0.263; P = 0.006 \)), and negatively correlated with FEV\(_1\) (\( R = -0.220; P = 0.022 \)) and FEV\(_1\)/FVC (\( R = -0.228, P = 0.018 \)) (Figure 3). There was no association between smoking amount and urinary l-histidine levels (\( R = 0.001, P = 0.990 \)).

Urinary l-histidine in COAD
To evaluate the role of urinary l-histidine as a Th-2 inflammatory marker in patients with COAD without diagnostic labeling, we compared the urinary l-histidine levels between the following groups of patients: patients with and without a documented history of asthma before 40 years of age or a BDR ≥ 400 mL of FEV\(_1\); patients with and without a BDR ≥ 200 mL of FEV\(_1\) and exceeding the baseline values by 12% on two or more visits; and patients with high and low eosinophil counts (according to the experts’ agreement regarding the Th-2 features of ACO).\(^{23}\) The levels of urinary l-histidine were significantly higher in patients with the previously mentioned characteristics than in those without these characteristics (Figure 4). Moreover, urinary l-histidine levels were significantly higher in frequent exacerbators than in non-frequent exacerbators (Table 2).

Discussion
This study investigated urinary l-histidine levels as a biomarker of ACO. Urinary l-histidine levels were elevated in patients with severe airflow obstruction, high BDR, severe symptoms, and exacerbation. Moreover, the levels were significantly higher in patients with key Th-2 features of ACO agreed upon by experts.\(^{23}\) Histidine is an essential amino acid that is a potential inflammatory metabolite,\(^{24}\) and it has been correlated with inflammation and oxidative stress in previous studies.\(^{25}\) Histidine is a precursor of histamine, which is important in the Th-2 inflammatory response and airway smooth muscle contraction.\(^{8,12,26}\) Thus, the current results imply that urinary l-histidine increases as smooth muscle contraction with reversibility increases and that urinary l-histidine could be an indicator of Th-2-related inflammation.

ACO has both features of asthma and COPD and is regarded to stem from either asthma or COPD.\(^{27}\) Miravitlles
has suggested that smoking-related obstructive asthma and COPD with a Th-2 genetic background or with significantly positive BDR could be classified as ACO.\textsuperscript{28} The former is regarded to originate from asthma, while the latter originates from COPD. However, when we divided patients with ACO into those with smoking-related obstructive asthma and COPD with a significantly positive BDR and/or blood eosinophilia groups, no significant differences were observed in urinary L-histidine levels between the two groups, even though other clinical characteristics, such as lung functions, were significantly different. The two subgroups of ACO are divided on the basis of different historic classifications, but this classification does not reflect inflammatory heterogeneity. ACO might be the different expressions of a single disease entity; this is because we frequently found difficulties in differentiating the origin between asthma and COPD.\textsuperscript{5} A recent study demonstrated that Th-2-related inflammatory gene expression was also up-regulated in COPD that could not be distinguished using a clinical history of asthma,\textsuperscript{7} and the Th-2-related signature provided a more clear distinction than the classical diagnosis.\textsuperscript{6} Moreover, in this study, if we applied the ACO criteria proposed by experts,\textsuperscript{23} we could classify only 48.1\% of patients with smoking-related obstructive asthma and 40.0\% of the patients with COPD with significantly positive BDR and/or high blood eosinophilia as having ACO. Therefore, the term “ACO” may be replaced when new phenotypes or endotypes are identified.\textsuperscript{2,27,29}

Recently, emphasis has shifted from diagnostic labels to “treatable traits” among chronic airway inflammatory diseases.\textsuperscript{4} A primary goal in patients with COPD is to identify patients who may benefit from ICSs. ICS therapy can prevent exacerbation and improve the symptoms and health status of patients with COPD.\textsuperscript{30,31} However, the use of ICSs also carries the risks of osteoporosis, diabetes, and pneumonia.\textsuperscript{31,32} Therefore, it is important to find biomarkers that could guide the use of ICSs to reduce the risk–benefit ratio of ICS therapy.

In this study, urinary L-histidine levels were higher in patients identified using the recently proposed algorithm of ACO\textsuperscript{18} than in patients with asthma or COPD. However, the definition of and diagnostic criteria for ACO are variable, and thus, we removed the classical diagnosis in patients with airflow obstruction and compared the patients according to the characteristics – agreed upon by experts – related to Th-2 features in ACO (ie, a documented history of asthma before 40 years of age or a BDR ≥ 400 mL of FEV\textsubscript{1}, a
Oh et al

BDR ≥ 200 mL of FEV₁ and exceeding baseline values by 12% on two or more visits, and a peripheral blood eosinophil count ≥300 cells µL⁻¹.²³ Urinary L-histidine levels were significantly higher in patients with these characteristics. Thus, urinary L-histidine levels could be a biomarker for ACO, regardless of the diversity of ACO definitions.

In addition to considering Th-2 features, ICS use in patients with frequent exacerbation is still needed despite the recent trend of withdrawing ICS in patients with moderate-to-severe COPD without a history of exacerbation.²⁵,³³ A history of previous exacerbation is the most potent predictor of future exacerbation.³⁴–³⁶ Thus, there is a need to perform ICS therapy in patients with previous frequent exacerbations despite treatment with bronchodilators. In this study, urinary L-histidine levels were also higher in frequent exacerbators, which supports the evidence indicating that urinary L-histidine is a candidate biomarker for guiding the decision to initiate ICS therapy.

Figure 4 Urinary L-histidine levels for each category of patients with Th-2 features of ACO as agreed upon by experts.

Notes: (A) Urinary L-histidine levels in patients with and without a documented history of asthma before 40 years of age or a BDR ≥ 200 mL of FEV₁. (B) Urinary L-histidine levels in patients with and without BDR ≥ 200 mL of FEV₁ and exceeding the baseline values by 12% on 2 or more visits. (C) Urinary L-histidine levels in patients with blood eosinophil count ≥300 cells µL⁻¹ and <300 cells µL⁻¹.

Abbreviations: ACO, asthma–COPD overlap; BDR, bronchodilator response; FEV₁, forced expiratory volume in 1 second.
Despite its strengths, this study had some limitations. First, because this study was based on urine analysis, the results could be affected by the time of urine collection. However, we collected urine samples after overnight fasting and in a medication-free state when patients were stable without exacerbation for 3 months. Moreover, we checked the metabolites three times and used the mean results. No patients had renal diseases, and the creatinine clearance rates of all enrolled patients were within the normal range. Furthermore, different from other parameters such as blood markers, urine has an advantage that its sampling is non-invasive and easy. Second, a quantitative analysis was not performed for urine metabolites. Further studies including quantitative analysis and proteomics/genomics studies are necessary to confirm the diagnostic value of urinary L-histidine and to elucidate the underlying mechanism.

Conclusion

Urinary L-histidine could be a potential biomarker for ACO and could be an adjunctive biomarker for ICS guidance, regardless of the diagnosis.

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Author contributions

All authors contributed toward data analysis, drafting and revising the paper and agree to be accountable for all aspects of the work.

Disclosure

The authors report no conflicts of interest in this work.

Table 2 Comparison between frequent exacerbaror and non-frequent exacerbaror

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Frequent exacerbaror</th>
<th>Non-frequent exacerbaror</th>
<th>P-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age, years</td>
<td>61.5 (54.3–72.3)</td>
<td>64.0 (57.0–70.0)</td>
<td>0.667</td>
</tr>
<tr>
<td>Sex, male</td>
<td>10 (83.3%)</td>
<td>72 (75.8%)</td>
<td>0.547</td>
</tr>
<tr>
<td>Smoker</td>
<td>10 (83.3%)</td>
<td>64 (67.3%)</td>
<td>0.010</td>
</tr>
<tr>
<td>Ex-smoker</td>
<td>9 (75.0%)</td>
<td>35 (36.8%)</td>
<td></td>
</tr>
<tr>
<td>Current smoker</td>
<td>1 (8.3%)</td>
<td>29 (30.5%)</td>
<td></td>
</tr>
<tr>
<td>Smoking amount (pack-years)</td>
<td>37.5 (12.5–48.8)</td>
<td>40.0 (33.8–54.8)</td>
<td>0.301</td>
</tr>
<tr>
<td>Urine L-histidine level</td>
<td>61,917.8</td>
<td>12,226.3</td>
<td>0.032</td>
</tr>
</tbody>
</table>

Note: Data are presented as median interquartile range for continuous variables and percentage (number) for categorical variables.

References


