Development of a Flavor Fingerprint by Gas Chromatography Ion Mobility Spectrometry with Principal Component Analysis for Volatile Compounds from *Eucommia ulmoides* Oliv. Leaves and its Fermentation Products

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Data obtained with gas chromatography coupled with ion mobility spectrometry (GC-IMS) was explored to investigate the characteristics of volatile compounds from edible fungus, from Eucommia ulmoides Oliv. leaves (EUI) that served as growth medium, and from their fermentation products. A total of 162 signal peaks were found, of which 68 compounds were identified, including alcohols, aldehydes, ketones, acids, and esters. There were differences in the volatile constituents of the edible fungi. EUI also contained special volatile components. The volatile components in the fermentation product were different compared to the raw material, and the difference in composition and content of the characteristic compounds was also obvious. The best classification performance was obtained by principal component analysis (PCA) based on the signal intensity of the characteristic volatile compounds. The results clearly showed that the samples (edible fungi, EUI and fermentation products) in a relatively independent space would be well distinguished. This further illustrated that the composition and content of volatile components of EUI could be changed by different microbial strains through biofermentation technology. Combining the signal intensity of the flavor substance, the difference was also clearly observed. This result suggested that the flavor compounds fingerprint could be established by GC-IMS and PCA.

Keywords: Eucommia ulmoides Oliv. leaves; Gas chromatography-ion mobility spectrometry; Fermentation product; Characteristic volatile compounds; Principal component analysis

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INTRODUCTION

Ion mobility spectrometry (IMS) technology was developed in recent years and was initially used for rapid detection of drugs, explosives, and chemical agents (Shvartsburg 2010; Armenta *et al.* 2011). The working principle of IMS is that the sample to be tested is vaporized by the ion source and becomes a gaseous molecule, which is chemically ionized and carries a certain amount of electric charge. Then, different target ions will produce corresponding ion spectra in the electric field. This method has the advantages of fast detection speed and high sensitivity (Karpas 2013). However, for complex samples systems in food and agricultural products, the analysis process is often limited (Arce *et al.* 2014). The combination with gas chromatography (GC) technology will overcome the

limitations of IMS separation efficiency, and at the same time give full play to the advantages of different instruments (Bunkowski et al. 2010). Under this condition, the ion mobility spectrum enriches the chemical information obtained by chromatographic separation by drift time information; meantime, the ion mobility spectrum signal response is significantly improved in mass and quantity after pre-separation by GC (Zhang et al. 2016; Garrido-Delgado et al. 2018). The three-dimensional matrix (migration time, retention time, and signal strength) obtained by gas chromatography-ion mobility spectrometry (GC-IMS) provides richer chemical information for more comprehensive data processing (Garrido-Delgado et al. 2012; Hajialigol et al. 2012; Zhang et al. 2016; Garrido-Delgado et al. 2018). Research results show that GC-IMS technology combined with chemometric methods is being gradually applied in the field of food testing and natural active ingredient analysis (Fink et al. 2014; Gallegos et al. 2015; Garrido-Delgado et al. 2015a, 2015b; Gallegos et al. 2017; Gerhardt et al. 2017, 2018; Mochalski et al. 2018). In this process, principal component analysis (PCA) is a commonly used feature for extraction and application of data dimensionality reduction in chemometrics (Jourdren et al. 2017; Pu et al. 2019).

For the GC-IMS technology, the complex sample is initially separated by GC technology, and then it is analyzed by an IMS detector. This combination technology can greatly improve the accuracy of mixture detection. Additionally, because GC separation is completed in seconds to minutes, while IMS detection time is measured in milliseconds, the detection time is greatly reduced compared with conventional chromatography, so it can meet the needs of the field of rapid analysis. (Politis *et al.* 2010; Jafari *et al.* 2012; Liedtke *et al.* 2018). The sample can be treated by GC prior to testing, which also effectively reduces the effect of humidity on IMS. Finally, a three-dimensional spectrum containing retention time, drift time, and signal strength can be obtained, which also makes the qualitative analysis more accurate. And both GC and IMS can operate under atmospheric pressure, they are easy to operate, and low in cost (Jafari *et al.* 2012). So the combined technology of GC and IMS has been widely considered.

At present, GC-IMS is considered an important technology for detecting volatile components from complex samples. Meanwhile, volatile components (flavor substances) are an important factor in the popularity of food, consumer acceptance, and are a vital indicator of the difference between different types of food (Cohen *et al.* 2015; Fang *et al.* 2017). Microbial conversion is a method with the most potential to improve taste and flavor. Compared with common chemical synthesis technology, this technology has the advantages of high chemical specificity, positional specificity, and stereospecificity. And the simple operation process makes it more economical and environmentally friendly (Akacha and Gargouri 2015). In addition, edible fungi can produce characteristic volatile components through their own metabolism (Vajpeyi and Chandran 2015).

Modern biofermentation technology is based on the fermentation method of traditional Chinese medicine processing and combined with micro-ecological research results and modern microbial engineering technology to form a new sample processing method (Liese and Filho 1999). On the basis of solid fermentation, the bi-directional solid-state fermentation technology of medicinal fungi is studied and developed, this method mainly refers to the use of medicinal plants or residues with active ingredients as a matrix of active ingredients instead of traditional nutrient bases, and the preferred strains are added for microbial transformation, which will form a special fermentation product. In this process, fermentation matrix provides the nutrients required by the fungus and is also affected by the enzymes from the fungus to change its own tissues and components, and to

produce new flavor substances and active ingredients. It is of great significance to increase the utilization of biomass resources and broaden the scope of its application (Bel-Rhlid *et al.* 2018).

As is well known, Eucommia ulmoides Oliv. is one of the oldest nourishing herbs in traditional Chinese medicine (He et al. 2014). Eucommia ulmoides leaves (EUI) contain many active ingredients, such as flavonoids, iridoids, lignans, phenylpropanoids, and polysaccharides, which have the effects of lowering blood pressure, regulating blood lipids, preventing osteoporosis, lowering blood sugar, calming nerves, and resisting fatigue. The resources are rich and also have high utilization value (He et al. 2014; Hirata et al. 2014; Zhu and Sun 2018). In recent years, the chemical composition, activity, and bioavailability of EUI have continually been the focus of attention, but there have been relatively few studies on the characteristic volatile components of EUI, especially with the use of GC-IMS technology (Hirata et al. 2014). Further, the investigation of volatile components from fermentation product has rarely been reported. It is worth noting that Ganoderma lucidum (GL) strain, Hericium erinaceus (HE) strain, and Griflola frondosa (GF) strain are important edible fungi (Xu et al. 2010; He et al. 2017; Zhao et al. 2017) and have obvious health benefits and medicinal value. Based on the above mentioned, solid-state fermentation of different edible fungi and EUl may produce some interesting results, this phenomenon is worth exploring, and it is also necessary to analyze the characteristic volatile components by GC-IMS technology.

The objective of this study was to first develop a simple and rapid method for the investigation of the characteristic volatile components of EUI, different edible fungi, and their fermentation products using GC-IMS technology. Differences were compared by the fingerprinting of different sample compounds obtained and PCA techniques. Furthermore, some of the marked compounds were identified throughout the spectrum, and the composition and relative content in different samples were analyzed. This would provide a theoretical basis for the development of new fermentation products with special activity.

EXPERIMENTAL

Materials

EUI were obtained from Cili Du-zhong Forestry Centre (Zhangjiajie, China). The fresh leaves were dried at 60 $^{\circ}$ C, and then the sample was prepared and stored at 4 $^{\circ}$ C until use.

Ganoderma lucidum (GL) preservation strain (strain number GDMCC5.250), *Hericium erinaceus* (HE) preservation strain (strain number GDMCC5.66), and *Griflola frondosa* (GF) preservation strain (strain number GDMCC5.63) were purchased from Guangdong Institute of Microbiology Culture Collection (Guangzhou, China).

All the reagents used in the experiment were of analytical grade. Ultrapure water (Milli-Q Plus system, Millipore, Bedford, MA, USA) was used throughout the work.

Preparation of Fermentation Samples

The sample of EUI prepared was selected, and then an appropriate amount of water was added until the sample was wetted, and the sample was placed in the cultivation bag after being uniformly stirred. These samples needed to be sterilized at 121 °C. After the sample was cooled to room temperature, under aseptic conditions, GL strain, HE strain, GF strain, or GL-GF complex strain were inoculated into the fermentation medium (EUI),

and the moisture content of the substrate was about 65%. The mixed fermentation system was cultured in the dark at 25 ± 3 °C until the mycelium was overgrown with the cultivation bag to stop the fermentation, and the sample was taken out to obtain different fermented fungus substance. These samples were stored in low temperature conditions until analyzed. There were eight kinds of samples in the experiment, which were *Ganoderma lucidum* microbial strain (GL-M), *Hericium erinaceus* microbial strain (HE-M), *Griflola frondosa* microbial strain (GF-M), *Eucommia ulmoides* leaves (EUI), the *Ganoderma lucidum* and *Eucommia ulmoides* leaves fermentation group (GL-EUI-F), the *Hericium erinaceus* and *Eucommia ulmoides* leaves fermentation group (GF-EUI-F), and the *Ganoderma lucidum Griflola frondosa* and *Eucommia ulmoides* leaves fermentation group (GL-GF-EUI-F).

GC-IMS Instrumentation and Analysis Parameters

The experiments were performed on a GC-IMS prototype manufactured by G.A.S. (Gesellschaft für Analytische Sensorsysteme mbH, Dortmund, Germany) based on an Agilent 6890N gas chromatograph (Agilent Technologies, Palo Alto, CA, USA), coupled to a drift time IMS cell. Analyses for the identification of characteristic volatile compounds of the samples were performed on an IMS commercial instrument (FlavourSpec) from Gesellschaft für Analytische Sensorysteme mbH (G.A.S., Dortmund, Germany) fitted with a non-polar column (FS-SE-54-CB) constituted by 94% methyl-5% phenyl-1% vinylsilicone with a 30 m length \times 0.32 mm and 0.5 µm film thickness. The injection rate was 100 µL/s, and the carrier flow rate was 5 mL/s.

For analysis, 1.0 g of different samples that needed to be analyzed (edible fungi, EUI, and its different fermentation products) were placed in a 20-mL vial that was closed with magnetic caps. After 20 min of incubation at 80 °C, 200 μ L of sample headspace was automatically injected by means of a heated syringe (80 °C) into the heated injector (80 °C) of the GC-IMS equipment. After injection, the nitrogen gas (99.999%) used as carrier gas, passed through the injector inserting the sample into the gas column, which was heated at 40 °C for timely separation. Then, the analytes were eluted in the isothermal mode and driven into the ionization chamber for ionization, prior to spectrometric detection. Molecules were ionized using a tritium source (6.5 keV), and the resulting ions were driven to the drift region *via* a shutter grid (Bradbury and Nielson design). The drift tube was 5 cm long and operated at a constant voltage of 400 V/cm, a temperature of 45 °C, and a drift gas flow rate of 250 mL/min (nitrogen). Data were acquired *via* the spectrometer's built-in computer. Each sample spectra had an average of 32 scans.

Data Analysis

The study of specific volatile compounds to identify them was realized by the software LAV version 2.0.0 from G.A.S. (Dortmund, Germany). Based on the use of the information included in the whole spectral fingerprint, raw IMS data were converted to .csv format using LAV software. Moreover, GC-IMS Library Search software supplied by G.A.S. (Dortmund, Germany) was employed to identify unknown compounds. Chemometric processing of the IMS data was performed with SIMCA-P 14.0 (Umetrics, Umea, Sweden). The processing technique mainly included principal component analysis (PCA). Data were initially subjected to PCA to reduce their dimensions and apply the classifying procedure to a smaller subspace (Garrido-Delgado *et al.* 2011), and the minmax normalization method was used to perform PCA.

RESULTS AND DISCUSSION

GC-IMS Topographic Plots from Different Samples

In this study, a simpler and quicker GC-IMS technology was proposed for the discrimination of the composition of EUl according to the volatile components. The five batches of Eucommia ulmoides samples mentioned above (EUI, GL-EUI-F, HE-EUI-F, GF-EUI-F, and GL-GF-EUI-F) were investigated using this method, and the signal intensity of some representative peaks were observed and analyzed. Simultaneously, corresponding different edible fungi (GL-M, HE-M, and GF-M) were also analyzed by GC-IMS under the same conditions (Li et al. 2015). All signal peaks determined in this study were consecutively numbered and in the following either termed by their names or by a number from 1 to 162, which summarized the GC-IMS results. These compounds could be expected to distinguish the differences of the samples (Jünger et al. 2012). Notice that one compound can result in more than one signal or spot (monomer or dimer), depending on the concentration. Spectra at different retention times can be obtained in a topographic plot. For example, the analysis results of the EUI sample is shown in Fig. 1. Different peaks are shown and marked. It was worth nothing that each peak was represented by a spot in the topographic plot (Arroyo-Manzanaresa et al. 2018). There were significant differences in the volatile components of the different samples. Furthermore, this was the first time that some of these target compounds had been studied for EUl samples by Headspace-GC-IMS (HS-GC-IMS).



Fig. 1. Imaging of volatile compounds represented by GC-IMS for EUI



Drift Time (ms)





Fig. 3. Gallery plot of the signal peak areas obtained from different sample (edible fungi, and *Eucommia ulmoides* Oliv. leaves, and its fermentation products)

Figures 2 and 3 show the topographic plot for different samples. As shown, the different samples analyzed exhibited significant differences in some monomer or dimer ions.

The differences of volatile compounds in EUI and its fermentation products were analyzed by GC-IMS. The data were presented by topographical visualization, where the Y-axis represented the retention time of the gas chromatograph, the X-axis represented the ion migration time for identification, and signal strength represented the relative content of the compound. As shown in Figs. 2 and 3, the composition of the volatile compounds from the different samples was similar, but the signal intensity was slightly different. The ion migration time and the position of the reactive ion peak (RIP) were normalized (Wang et al. 2019). The whole spectrum represented the total headspace volatile compounds of the different samples, including edible fungi and EUl samples. Each point in the spectrum represented a volatile compound. It could be seen that most of the signals appeared in the retention time of 100 to 900 s and the drift time of 1.0 to 1.8 ms. Color represented the signal intensity of the target substance. Black indicated lower intensity and red indicated higher intensity (Li et al. 2019). It can be seen from the figure that a total of 162 characteristic peaks were found from the three edible fungi and different EUI samples by GC-IMS technology. In terms of three edible fungi (GL, HE, and GF), although they had some similar components, the relative content varied greatly, and the difference in the main volatile components was obvious. The results from Fig. 3 showed that different edible fungi had their own unique composition of volatile components and were also well represented in the figure, and it also shown that this detection method could be fully applied to find the differences between different samples. In view of the different chemical composition and flavor differences of different edible fungi, applying them to fermentation treatment, some interesting results may be produced that have important research value. It can also be clearly seen from the figure that for the 162 peaks that were found, the content of some of the three edible fungi was low or even absent. Some of these compounds were found in three edible fungi and EUI materials, but the difference in composition and content of the overall volatile compounds was significant. These results further indicate that through the biotransformation technology, the composition and relative content of edible fungi and EUl could be affected. This method would have an important impact on the new target components, characteristic volatile components, and flavor improvement of the fermentation products. Additionally, it was also meaningful to study the composition and identification of the target volatile compounds.

GC-IMS Topographic Plots from Different Samples Fermented

Biotransformation technology is currently the most promising method for discovering novel compounds (Liese and Filho 1999). EUI were fermented with different strains, such as GL, HE, GF, and GL-GF complexes. The volatile components of the fermentation product were analyzed by the GC-IMS. As shown in Figs. 2 and 3, the composition and relative content of the volatile compounds in the corresponding fermentation products were changed compared to the raw materials of EUI. It could also be seen from the topographic plots that the relative content of some compounds in edible fungi could be significantly reduced or even disappear through fermentation. The EUI samples also had similar trends, but the number of compounds were less than that of edible fungi. Meanwhile, different novel compounds were produced in the fermentation products based on the different edible fungi, and the composition and relative content of the compounds in the final fermentation product were also significantly different. Compared

with EUl raw materials, based on the relative peak intensities derived from instrumental software analysis, among them, the content of 10 compounds was reduced, and the content of 23 compounds was increased; the relative content of 57 compounds was comparable to or slightly increased with the raw materials, and 15 compounds were similar to or slightly reduced from the raw materials. In addition, there were 49 compounds in the corresponding fermentation products with a relatively large fluctuation (increased or decreased) under the fermentation conditions of different edible fungi. Obvious changes had not been discovered. This also indicated that different edible fungi could significantly change the composition characteristics of volatile components in EUI samples through solid state fermentation. Choosing the right strain, sample, and fermentation conditions is of great significance for discovering new target compounds and improving the original flavor of the sample. It is worth noting that although the relative content of most of the compounds in the fermentation product changes after fermentation, compared with the EUI, and the changes of some compounds were very obvious, the basic composition of the whole compound was nearly similar to the raw materials. However, compared with the three edible fungi, the volatile components in the fermentation product changed obviously. It can be seen from the figure that a large number of compounds are significantly reduced or disappeared, which further indicates that microorganisms affect their own flavor components during the fermentation process, and that the fermentation substrate also plays an important role in this process. Figures 2 and 3 show that the volatile components in the product after co-fermentation with edible fungi and EUl were not the superposition of the flavors of the two substances but had obvious adjustment and improvement on the original characteristic flavor. The experimental results showed that the IMS technology could efficiently and quickly detect and analyze volatile components in samples. In the process of fermentation, first, the microorganisms pass through a specific fermentation substrate, and their flavor composition would change significantly. Secondly, by using microbial metabolism, the structure and flavor of EUI would also change significantly. Therefore, the selection of suitable microbial strains and fermentation conditions can effectively improve the appearance, functional ingredients, nutritional value, and taste and flavor of the raw material samples.

Analysis of Volatile Compounds Identified from Different Samples

Flavor quality plays an important role in directing consumers' preference. Therefore, it is necessary to separate and identify volatile components by different chemical methods for the application and development of new food raw materials (Cohen *et al.* 2015). Based on the analysis results of this study, for the EUI samples with and without fermentation, Fig. 3 shows that a total of 162 signal peaks were discovered by GC-IMS. Simultaneously, from the results shown in Table 1, 68 of the characteristic compounds were identified, which mainly contained alcohols, aldehydes, ketones, and esters.

Based on a comprehensive analysis of the different samples by the HS-GC-MS, the volatile composition characteristics of all samples can be obtained according to fingerprint analysis techniques. At the same time, according to the migration time of the IMS, the retention index, and the corresponding database, some compounds can be identified. Due to differences in concentration, some compounds exhibit multiple spots or signals (monomers or dimers) in the spectrum.

Table 1. GC–IMS Integration Parameters of Volatile Compounds Identified to Distinguish Different Samples

#	Gallery Plot-area Label	Compound	Formula	RI	Rt (s)	Dt (RIP Relative)	Comment
1	1	Nonanal	C ₉ H ₁₈ O	1105.9	945.36	1.474	Monomer
2	2	Nonanal	C ₉ H ₁₈ O	1105.1	943.02	1.941	Dimer
3	7	(E)-2-Octenal	C ₈ H ₁₄ O	1059.3	815.10	1.333	
4	82	Limonene	C ₁₀ H ₁₆	1043.2	771.03	1.222	
5	8	Phenylacet- aldehyde	C ₈ H ₈ O	1037.1	754.65	1.255	
6	9	1,8-Cineol	C ₁₀ H ₁₈ O	1024.9	721.89	1.3	Monomer
7	4	(E, E)-2,4- Heptadienal	C7H10O	1017.5	702.39	1.201	Monomer
8	120	(E, E)-2,4- Heptadienal	C7H10O	1017.2	701.61	1.615	Dimer
9	52	1,8-Cineol	C ₁₀ H ₁₈ O	1025.5	723.45	1.732	Dimer
10	15	Octanal	C ₁₀ H ₁₈ O	1006.3	673.53	1.408	Monomer
11	16	Octanal	C ₁₀ H ₁₈ O	1005.1	670.41	1.822	Dimer
12	17	2-Pentylfuran	C ₉ H ₁₄ O	990.7	634.53	1.251	Monomer
13	18	2-Octanone	C ₈ H ₁₆ O	997.3	650.91	1.335	
14	19	1-Octen-3-one	C ₈ H ₁₄ O	983.4	616.59	1.277	
15	20	1-Octen-3-ol	C ₈ H ₁₆ O	982.1	613.47	1.606	
16	53	5-Methyl-2- furfural	$C_6H_6O_2$	967.8	579.93	1.128	Monomer
17	54	Benzaldehyde	C7H6O	956.1	553.41	1.149	Monomer
18	121	Benzaldehyde	C7H6O	954	548.73	1.466	Dimer
19	21	2-Furanmeth- anol, 5-methyl-	$C_6H_8O_2$	960.3	562.77	1.254	Monomer
20	26	2-Furanmeth- anol, 5-methyl-	C ₆ H ₈ O ₂	957.2	555.75	1.566	Dimer
21	22	(E)-2-Heptenal	C7H12O	959.2	560.43	1.667	
22	55	Linalool	C ₁₀ H ₁₈ O	1086	889.59	1.218	
23	85	Linalool oxide	C ₁₀ H ₁₈ O ₂	1069	842.21	1.257	
24	23	Methyl hexanoate	C7H14O2	927.2	492.96	1.286	Monomer
25	24	Methyl hexanoate	C7H14O2	926.8	492.18	1.677	Dimer
26	41	Heptanal	C7H14O	900.9	444.60	1.335	Monomer
27	45	Heptanal	C7H14O	900	443.04	1.693	Dimer
28	42	2-Heptanone	C7H14O	892.1	429.78	1.265	Monomer
29	32	2-Heptanone	C7H14O	891.1	428.22	1.631	Dimer
30	47	1-Hexanol	C ₆ H ₁₄ O	876.8	405.99	1.326	
31	48	3-Methylbut- anoic acid	C5H10O2	868.7	394.29	1.223	Monomer
32	81	3-Methylbut- anoic acid	$C_5H_{10}O_2$	867.6	392.73	1.492	Dimer

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33	3	2-Hexen-1-ol	C ₆ H ₁₂ O	851.8	371.28	1.181	
34	123	2-Furfural	$C_5H_4O_2$	831.1	345.54	1.084	Monomer
35	88	2-Furfural	$C_5H_4O_2$	828.8	342.81	1.33	Dimer
36	43	Butyl acetate	$C_6H_{12}O_2$	811.1	322.72	1.239	Monomer
37	33	Butyl acetate	$C_6H_{12}O_2$	810.1	321.56	1.62	Dimer
38	49	Hexanal	$C_6H_{12}O$	793.6	303.81	1.562	
39	60	2-Methylprop- anoic acid	$C_4H_8O_2$	775.4	285.09	1.156	Monomer
40	124	2-Methylprop- anoic acid	$C_4H_8O_2$	773.1	282.75	1.369	Dimer
41	77	1-Pentanol	$C_5H_{12}O$	766.7	276.32	1.512	Dimer
42	50	1-Pentanol	$C_5H_{12}O$	764.9	274.56	1.255	Monomer
43	64	Propanoic acid	$C_3H_6O_2$	751.6	261.69	1.105	Monomer
44	125	2-Methyl-1- butanol	$C_5H_{12}O$	753.7	263.64	1.481	
45	101	Propanoic acid	$C_3H_6O_2$	750.8	260.91	1.356	Dimer
46	57	Acetoin	$C_4H_8O_2$	713.1	228.34	1.33	
47	65	Pentanal	$C_5H_{10}O$	696.3	216.06	1.425	
48	80	2-Pentanone	$C_5H_{10}O$	688	210.60	1.37	
49	69	1-Butanol	$C_4H_{10}O$	658.4	194.22	1.371	
50	5	3- Methylbutanal	$C_5H_{10}O$	645.9	188.37	1.408	Dimer
51	108	3-Methyl- butanal	$C_5H_{10}O$	645.9	188.37	1.198	Monomer
52	6	Ethyl Acetate	$C_4H_8O_2$	608.9	173.36	1.337	
53	70	Butanal	C_4H_8O	598.9	169.65	1.291	
54	71	2-Butanone	C ₄ H ₈ O	585.6	164.78	1.246	
55	73	Methylpropanal	C_4H_8O	552.7	152.69	1.285	
56	67	Acetone	C ₃ H ₆ O	512.3	137.87	1.119	
57	68	Ethanol	C_2H_6O	464.5	120.31	1.051	
58	91	Propyl acetate	$C_5H_{10}O_2$	714.1	229.12	1.478	
59	92	2-Hexanol	$C_6H_{14}O$	811.1	322.72	1.57	
60	93	Ethyl 2- ydroxypro- panoate	$C_5H_{10}O_3$	812.6	324.28	1.534	
61	132	2-Pentylfuran	C9H14O	989.6	631.80	1.549	Dimer
62	133	5-Methyl-2- furfural	$C_6H_6O_2$	966.2	576.03	1.418	Dimer
63	142	Butanoic acid	$C_4H_8O_2$	798	308.49	1.391	
64	38	3-Octanone	C ₈ H ₁₆ O	990.7	634.52	1.307	
65	148	2-Acetylfuran	$C_6H_6O_2$	912	464.10	1.118	Monomer
66	149	2-Acetylfuran	$C_6H_6O_2$	912.2	464.49	1.448	Dimer
67	150	1-Propene-3- methylthio	C_4H_8S	697.4	216.84	1.045	
68	161	Propylsulfide	$C_6H_{14}S$	881.9	413.59	1.159	

A total of 68 typical target compounds from topographic plots were identified by the GC-IMS database (Table 1). The appreciated visual plots were chosen and shown together by gallery plot for intuitive comparison of different samples. The differences of volatile compounds identified from different fermentation products and raw materials could be observed from the figure, and the fingerprint of each group of samples was also established. In addition to the identified compounds, the differences and composition of non-target volatile compounds can also be presented. As shown in Table 1, a total of 68 characteristic compounds were identified, of which aldehydes were the most and 23 compounds were identified. There were 13 kinds of alcohol compounds and eight kinds of ketones. In addition, seven kinds of acid compounds and ester compounds were found. Besides, 10 other types of compounds were found.

Compared with EUl raw materials, after fermentation treatment, the relative content of (E, E)-2,4-heptadienal, (E)-2-octenal, 2-furanmethanol, 5-methyl-, methyl hexanoate, 2heptanone (monomer and dimer), 1-hexanol, 1-pentanol, limonene, 2-furfural, ethyl 2hydroxypropanoate, and 5-methyl-2-furfural had different degrees of increase. However, the relative content of 3-methylbutanoic acid, 2-methylpropanoic acid, and (E, E)-2,4heptadienal (dimer) had a decreasing trend. More so, 2-pentylfuran was only found in the GF fermentation group. In addition to the GF fermentation group, the compounds 5methyl-2-furanmethanol (dimer) and acetoin showed a significant increase compared with the EUl. However, in the GF fermentation group, the increase of propanoic acid was obviously higher than that of other groups. The peak intensities of acetone, ethanol, butanal, and 2-butanone in each group were substantially similar. In the HE fermentation group, the relative contents of (E)-2-heptenal and linalool were the highest; linalool oxide was a unique component in HE. The signal intensity of 2-pentanone in GL and GL-GF fermentation products was more obvious. The 3-methylbutanal (dimer) was almost absent in the raw material and GF fermentation group, and ethyl acetate was relatively high in the three edible fungi and GL fermentation groups. In the HE group, the signal intensity of (E)-2-octenal was more pronounced, and the relative content of phenylacetaldehyde was in the GL and the GL-GF group was higher than in the other groups. Octanal (monomer and dimer), 2-pentylfuran, 2-octanone, 1-octen-3-one, and 1-octen-3-ol were the major volatile components in GL, and the peak intensities in EUl and fermentation products were relatively weak. Propyl acetate, 2-hexanol, and ethyl 2-hydroxypropanoate were relatively more abundant in the GL fermentation group than the other groups. Benzaldehyde (dimer), 2-furfural, 2-methyl-1-butanol, 2-pentylfuran (dimer), and 5-methyl-2-furfural (dimer) were also observed only in the GF group, which could be considered as the main characteristic component of the sample and was of great significance for qualitative sample types. Butanoic acid and 2-acetylfuran were identified in the GF fermentation group, while the signal intensity of 1-propene-3-methylthio was the highest in the raw material and GF fermentation groups, which also indicated that GF does not affect the content of the compound compared to other species.

After the *Eucomnia* leaves were fermented by the GL-GF complex, the relative content of the propylsulfide was the highest. Among the identified compounds, some compounds could be observed in edible fungi, raw materials, and fermentation products, but the signal intensity corresponding to edible fungi was higher than other groups, such as 2-heptanone (dimer), butyl acetate (dimer), and so on. According to the analysis results, although some compounds were not the highest relative content, they could also be considered as representative compounds in EUI, which was basically consistent with previous reports in the literature. At the same time, according to the identified compounds,

the composition and content of the compounds in different strains were also different. These results were similar to those reported in the literature (Gallegos *et al.* 2015; Gallegos *et al.* 2017).

Similarity Analysis of Fingerprint-based on PCA

All data from GC-IMS were pre-processed independently before the study of ion mobility profiling. To further interrogate the data, a statistical method was required to emphasize variation and to visualize any patterns within the dataset; on that basis, PCA was selected; it is a multivariate statistical analysis model. By determining a few principal component factors to represent many complex and difficult-to-find variables in the original samples, the regularity and difference among the samples could be evaluated according to the contribution rate of the principal component factors in the different samples (Sebzalli and Wang 2001; Li *et al.* 2019).

In this study, PCA was established using signal intensity to highlight the differences in the volatile compounds from the different samples. The differences of these samples (with or without the three edible fungi) were analyzed by the PCA model. The PCA of the volatile compounds from the different samples is presented in Fig. 4. According to Fig. 4A, all the samples could be categorized into five different groups. In these figures, a good discrimination was obtained between the edible fungus and EUI samples. It showed the distribution map for the first two principal components determined by PCA, the first two components explained 66% of variance, component PC1 contributing 43% and component PC2 contributing 23%, and a visualization of the data was obtained. These components were thought to show the similarity between the different EUI samples. This result also further confirmed that different samples contain their own unique chemical components

For the samples from EUI and its fermentation products, according to Fig. 4B, all the samples could also be divided into three different groups. The first two components explained 74% of variance; component PC1 contributing 59% and component PC2 contributing 15%. Firstly, in terms of the volatile components, the difference between EUI and its fermentation products was obvious. Moreover, EUI raw materials and GF groups did not show a clear separation. It was important to mention that GL was similar to the GL-GF group in terms of the volatile components, and the HE group was different from the other groups. This also explained that compared with other fermentation groups, GF had a relatively small effect on EUI, while the HE, GL, and GF-GL groups could significantly change the characteristic volatile components for EUI. It is worth noting that the effects of GL and GL-GF on the EUI sample were similar.

The figure of PCA showed that consistent with the previous analysis results, after the biological fermentation, the characteristic volatile components of EUI would obviously change, and the results produced by different edible fungi also had differences. The volatile components of EUI were significantly changed through fermentation process. Different fermentation strains had different effects on the composition and content of the volatile components.

In this study, GC coupled with IMS equipment was utilized to determine volatile compounds from the fermentation product of EUl with different edible fungus. The importance of using IMS technology in the flavor analysis of the fermentation product of EUl was demonstrated for the first time. The results confirmed the usefulness of IMS as a screening system for distinguishing different samples. The significance of the present work is that it compared the organic volatile components of different fermentation products' abilities to evaluated different fermentation systems, based on the followed two aspects:

(1) all the information obtained by ion mobility spectra; and (2) the differences in different fermentation systems.



Fig. 4. Principal component analysis based on the signal intensity obtained of volatile compounds from different samples (edible fungi and *Eucommia ulmoides* Oliv. leaves and its fermentation products). a: all the sample of raw materials and fermentation product; b: EUI sample and its different fermentation products, except for edible fungi

The target compounds, which had been identified using the GC-IMS method, were used to construct a database containing the retention index and drift time. The database would help to quickly identify volatile constituents and efficiently identify new volatile compounds of *Eucommia ulmoides* fermentation products. Therefore, further work would seek to apply the developed methodology for the detection of *E. ulmoides* raw materials and new fermentation products would enrich their corresponding databases so that the composition of the volatile compounds from different samples could be more clearly understood.

CONCLUSIONS

- 1. In this study, a simple, fast and reliable method was developed to evaluate the characteristic volatile compounds of edible fungus, *Eucommia ulmoides* Oliv. leaves (EUI), and its fermentation products by establishing the fingerprint with gas chromatography-ion mobility spectrometry (GC-IMS) and principal component analysis (PCA), which required minimal sample preparation steps and reduced the time required for analysis.
- 2. The spectral fingerprint approach provided better investigation and classification for the different samples. According to the signal peaks in the spectrum and analysis results, a total of 162 signal peaks were found, of which 68 compounds were identified, including alcohols, aldehydes, ketones, acids, and esters. Meanwhile, fingerprints of characteristic volatile components of different samples were obtained by GC-IMS.
- 3. Different samples have their own characteristic volatile components. The volatile components of EUl can be changed through biological fermentation. Based on the data of the signal intensity from all the peaks, the best classification performance was obtained by PCA, the results clearly showed that the samples in a relatively independent space would be well distinguished. Additionally, the volatile components of the fermentation products also distributed in the respective characteristic regions, which had differences. This further illustrated that the composition and content of the volatile components of EUl could be significantly changed by different edible fungus through bio-fermentation technology.

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