Coupling solid phase microextraction to complementary separation platforms for metabotyping of $E.\ coli$ metabolome in response to natural antibacterial agents

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Introduction. Essential oils are known to possess antimicrobial activity; thus, their use has played

an important role over the years in medicine and for food preservation purposes.

Objective. The effect of clove oil and its major constituents as bactericidal agents on the global

metabolic profiling of E. coli bacteria was assessed by means of metabolic alterations, using solid

phase microextraction (SPME) as a sample preparation method coupled to complementary

analytical platforms.

Method. E. Coli cultures treated with clove oil and its major individual components were sampled

by HS-SPME-GCxGC-ToF/MS and SPME-UPLC-MS. Full factorial design was applied in order

to estimate the most effective antibacterial agent towards E. coli. Central composite design

(CCD) and factorial design were applied to investigate parameters influencing metabolite

coverage and efficiency by SPME.

Results. The metabolic profile, including 500 metabolites identified by LC-MS and 789

components detected by GCxGC-ToF/MS, 125 of which were identified as dysregulated

metabolites, revealed changes in the metabolome provoked by the antibacterial activity of clove

oil, and in particular its major constituent eugenol. Analyses of individual components selected

using Orthogonal Projections to Latent Structures Discriminant Analysis (OPLS-DA) showed a

neat differentiation between control samples in comparison to treated samples in various sets of

metabolic pathways.

Conclusions. The combination of a sample preparation method capable of providing cleaner

extracts coupled to different analytical platforms was successful in uncovering changes in

metabolic pathways associated with lipids biodegradation, changes in the TCA cycle, amino

acids, and enzyme inhibitors in response to antibacterial treatment.

Keywords: E.Coli, Solid Phase Microextraction, clove oil, metabolomics, LC-MS,

GCxGC-ToF/MS

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1. Introduction

Owing to their antimicrobial characteristics, essential oils are widely used in medicine as well as in the food and fragrance industries. Their antibacterial, antiviral, insecticidal, antiparasitic, antifungal, and antioxidant properties are extensively documented (Astani et al. 2011; Brenes and Roura 2010; S. Burt 2004; Deans and Ritchie 1987; Kalemba and Kunicka 2003; Kim et al. 2003). Among the essential oils constituents, terpenes, terpenoids, phenol-derived aromatic components, and aliphatic compounds have been shown to diminish or interrupt the growth of bacteria, yeast, and mold through disruption of cell membrane and cytoplasm (S. A. Burt and Reinders 2003; Chorianopoulos et al. 2008; De Martino et al. 2009). Especially naturallyoccurring oxygenated terpenoids such as alcohols and phenolic terpenes have demonstrated the highest antimicrobial potential (S. Burt 2004; Delaquis 2002; Koroch et al. 2007; Smith-Palmer et al. 1998). In related work, 21 plant essential oils were applied as antibacterial agents against five food-borne pathogens (Campylobacter jejuni, Salmonella enteridis, E.coli, Staphylococcus aureus, and Listeria monocytogenes); their findings indicated that oils derived from bay, cinnamon, clove, and thyme yielded the highest bactericidal effect (Smith-Palmer et al. 1998). The interactions between different constituents of essential oils may lead to synergistic, antagonistic, indifference, or additive effects (Davidson and Parish 1989; K. Gill and Brown 2002). The mechanisms of action of these naturally-occurring compounds against pathogens are still not fully understood, especially at the metabolic level. As the metabolism of living systems alters in response to environmental stress, metabolomics platforms aim to provide complementary information to genomics, transcriptomics, and proteomics (Jozefczuk et al. 2010). Within this context, global analysis of all metabolites in a given system can be employed to discover potential biomarkers of specific effects or reactions (Dettmer et al.) attributable to specific stimuli. The global metabolomics platform includes different steps: sample preparation, instrumental analysis, and data analysis, where determinations of significant metabolite changes can be made using bioinformatics software. Due to the broad chemical and physical characteristics of metabolites, no single analytical platform could provide identification of all metabolites. Usually the most used instrumentation platforms for metabolomics are nuclear magnetic resonance (NMR) and mass spectrometry (MS), the latter coupled with both gas- (GC) and liquid-chromatography (LC). In addition, to increase method sensitivity and provide wider metabolome coverage, a proper sample preparation strategy, able to obtain the most representative, yet clean extract possible, needs to be used. Solid phase microextraction (SPME), as one of the recently emerging techniques utilized in sample preparation for metabolomics studies, is capable of fulfilling many of the criteria for ideal

sample preparation in metabolomic investigations such us non-selectivity, reproducibility, simplicity and possibility for automation (Bojko et al. 2014; Vuckovic et al. 2011).

A crucial aspect of the metabolomics workflow, unfortunately often overlooked, is experimental planning. Considering that, untargeted metabolomics experiments involve the analysis of a consistent number metabolic features, often characterized by different physical-chemical properties, the optimization of each factor affecting their extraction, separation and detection needs to be carefully carried out, in order to find the proper operating conditions for the obtainment of representative sets of data. In particular, the currently presented work was undertaken using a statistical experimental design approach comprising careful design of experiments, including multivariate designs, that offer simultaneous optimization of several control variables, consequently taking into account interactions between factors and requiring fewer experiments than the traditional univariate approaches (Sousa et al. 2006).

The primary purpose of this study was to employ SPME coupled to GCxGC-ToF/MS and UPLC-MS and bioinformatics tools to study changes in metabolic pathways of *E. coli* metabolome submitted to treatment with clove bud oil and its major constituents. Multivariate experimental design was applied to optimize factors that impact extraction and to evaluate the type of interactions occurring between the major active components of clove oil. Individual components of the clove oil were characterized and identified. Finally, the metabolic profiles of antibacterial agent-treated cells and control cells were generated by both optimized platforms and subjected to multivariate data analysis. These metabolic patterns produced clear separation between controls and treated samples on an OPLS-DA analysis due to up-regulated and down-regulated metabolites.

2. Experimental section

2.1 Chemical and materials and Metabolite Standard Mixture Preparation

Details regarding chemicals, materials and mixture preparation can be found in Supplementary Info (section 1.1)

2.2 Bacterial strain and antibacterial activity evaluation

Frozen cultures of *E. coli* BL21 were streaked on an LB agar media plate (10 g trypton; 5 g yeast extract; 5 g NaCl; 15 g Agar in 1 L nanopure water), and incubated overnight at 37 °C. One isolated colony was re-streaked on LB Agar media and incubated at 37 °C for 24 h. Following, one isolated colony was inoculated into 5 mL of LB media (10 g trypton; 5 g yeast extract; 5 g NaCl in 1 L nanopure water), then incubated at 37 °C for 24 h under agitation at 125

rpm. The microbial broth was then serially diluted in fresh media. The resulting 10⁷ colony-forming units (CFU/mL) were utilized for the whole study.

LB agar media was used to count the number of CFU/mL in bacterial media. Cultures were serially diluted with sterile media, and 10 μ L of diluted culture is spread uniformly over the surface of warm agar plates. After keeping plates in 37 °C overnight, the colonies were counted. In order to get precise results, diluted cultures forming 30 to 100 colonies were applied for calculation of CFU/L.

To investigate the antibacterial activity of the most important constituent of clove bud oil, a two-level full factorial design, 2³, was carried out. Table SI.2 shows the number of different possible conditions in this full factorial design experiment.

2.3 Coating preparation, multivariate optimization of the 96-blade SPME method, and HS-SPME sampling conditions

Prior to LC analyses, extractions automated by the robotic concept 96-autosampler were performed using a 96-blade SPME system consisted of PS-DVB-WAX:HLB 50:50 [w/w] as extraction phase coated onto stainless steel blades. Choice of coating and coating preparation procedure, information related to the concept 96-autosampler are reported elsewhere (Mousavi et al. 2015). Other details can be found in Supplementary Information (section 1.2)

2.4 UHPLC-ESI-MS, GC-IT-MS and GCxGC-ToF-MS conditions

For analyses an UPLC-Exactive (Thermo, San Jose, California, USA), a Varian CP-3800 GC system coupled to a Varian 4000 Ion-Trap Mass Spectrometer (Varian, Palo Alto, California, USA) and a GCxGC-ToF/MS Pegasus 4D (METLIN) were used, further details concerning operational parameters and separation conditions can be found in Supplementary Information (section 1.3).

2.5 Metabolite identification, data mining, and statistical analysis

For LC based experiments, Raw data (.raw) obtained by Xcalibur software (version 2.1, Thermo, San Jose, California, USA) were converted to (mzXML) with MS conversion software. The converted data were then processed with a web-based platform, XCMS online (Scripps Center for Metabolomics, California, USA). The output consists in a table containing retention times, m/z values, and intensity of features. XCMS identifies features whose relative intensity differs between sample groups, determining p-values as well as fold changes. The software carried out mass spectral peak deconvolution, alignment, peak picking (feature detection), and ion annotation on the list of features in order to identify the detected isotopes, adducts, and in-source fragment

ions. Multivariate data analysis was carried out by SIMCA-P+ software (Umetrics, NJ, USA). Unsupervised principal component analysis (PCA) was used to evaluate grouping trends between the different antibacterial agent treatments. Orthogonal Projections to Latent Structures Discriminant Analysis (OPLS-DA) was performed on obtained data, and the resulting S-plots were inspected to investigate discriminant biomarkers produced in bacteria culture treated by antibacterial agents. Metabolite identities were specified based on their accurate mass, retention time, comparison of fragmentation data with authentic standards, and METLIN databases within 5 ppm. By the help of metabolite databases, it is applicable to find the m/z ratio of peaks of interests manually. The KEGG database was used to identify important metabolic pathways and for biological interpretation. Identification can be confirmed applying MS² fragmentation analysis. It is not practical to get standards of all compounds to do independent comparison. Therefore, quantification can be done through a relative comparison between groups applying statistics.

All acquired GCxGC-ToF/MS chromatograms were processed by ChromaTOF version 4.5 (LECO Corporation). Baseline correction, noise reduction, deconvolution, library matching, and peak area calculation were performed. Peak area calculations were performed using the unique mass of the detected analyte. Only peaks bearing S/N ratio ≥ 100 were selected, and a spectral match threshold of 700 was selected for further data interpretation. Identification of selected entries was performed using the National Institute of Standards and Technology (NIST) reference library for matching of mass spectra, and by comparison of calculated retention indices with data present in literature.

The Statistical Compare feature of ChromaTOF was successively used for alignment of multidimensional peaks obtained from different chromatograms. Alignment parameters are specified in Supporting Information (Section 1.3). The obtained data tables (960 entries) were further manually scrutinized to remove peaks related to the extraction phase or chromatographic column bleeding (mainly siloxane-derived compound peaks), and to merge double entries, finally resulting in 789 entries.

The obtained data sets were then processed by to SIMCA (version 14.0, Umetrics, Umea, Sweden) by performing PCA, and OPLS-DA in a similar manner as for the LC-based data.

Considering the large amount of intra- and inter- sample variation on metabolomics data, Hotelling T2 eclipse and DmodX were applied for outlier evaluation. With two component plots, T2 was easily visualized in score plots. Moderate outliers investigated in DmodX plots based on residuals observation.

3 Results and discussion

3.1 Optimization of SPME parameters

Optimization of desorption solvent was carried out using the model analytes presented in Table SI.1. Using the 96-blade SPME method, a triangular design was carried out to determine the effectiveness of different desorption solvents and their interactions on analyte recovery. The experimental design followed and the results obtained by CCD are reported in Supplementary Information (Section 2.1).

3.2 Interaction between clove oil components using full factorial design

Once the 96-blade SPME method was optimized, a full factorial design was applied to investigate the effectiveness of constituents of clove bud in preventing *E. coli* metabolism, as well as to explore potential interactions between these major components.

Using GC-MS, the major components of clove bud oil were identified as eugenol (76.8%), eugenyl acetate (9.5%), and β -caryophyllene (6.0%).

The minimum inhibitory concentration (MIC) threshold for clove oil was identified through addition of different amounts of clove oil to the same $E.\ coli$ culture, followed by incubation at 37 °C for 16 hours. Afterwards 10 Γ of treatment and sample solutions from each cell tube was plated all over the LB agar plates. Plates were incubated at 37 °C overnight and colonies were counted in each case. Results indicated that 10 μ L was the MIC for clove oil for a 10^7 CFU/mL $E.\ coli$ culture.

To evaluate the antibacterial activity of each constituent of clove oil against *E.coli*, different amounts of eugenol (8 μ L), eugenyl acetate (1 μ L) and β -caryophyllene (0.6 μ L) were separately added to the same concentration of bacteria in different cell tubes based on the percent of each component present in clove bud oil. No bacteria growth was observed for cultures treated with eugenol, while growth was observed for cultures treated with eugenyl acetate and β -caryophyllene, indicating eugenol as the most effective antibacterial agent in the whole oil.

Metabolic profiling for each set of experiments was acquired; the variable effects and interactions between variables are summarized in the Pareto charts illustrated in Figure SI.3, SI.4 for SPME-LC-MS and SI.5 for GC-IT/MS. The Pareto charts obtained for LC amenable metabolites (Figures SI.3 and SI.4) reveal that eugenol, the major compound of clove bud oil, was found to possess the strongest antibacterial activity at the 95% confidence level, whereas the other constituents, when applied individually, did not indicate an antibacterial effect on *E. coli* growth at the 95% confidence level.

By studying the volatile metabolic profiling, the same effect was monitored. Pareto chart plots based on sums of peak areas obtained from HS-SPME-GC-IT/MS for two metabolites, indol (downregulated) and octanoic acid (upregulated), are shown in Figure SI.5. Based on the results obtained from the Pareto charts, no significant interactions between the constituents for both platforms (at the 95% confidence level) were observed.

3.3 Metabolic profiling of *E. coli* under different compositions of antibacterial agents using 96-blade SPME LC-MS

The obtained results indicated variations in the amounts (dysregulations) of some of the identified metabolites as a function of application of different compositions of antibacterial agents. Figures SI.6 and SI.7 demonstrates the LC-MS chromatograms of *E.coli* extract by the SPME coatings in positive and negative modes, respectively. Clove oil treatment sample extract chromatograms for both positive and negative ESI modes are shown by Figures SI.8 and SI.9, respectively. The output given by XCMS Online software is provided in Figure 1. Figure 1-D is a cloud plot demonstrating dysregulated features to represent ions whose intensities are varied between sample groups according to statistical thresholds. Down-regulated features in these treated samples are presented as red bubble-plots (bottom plot), while up-regulated features are represented by green bubble-plots (top plot).

About 10,000 features were detected in this study in both electrospray ionization modes, of which almost 60% of the peaks showed statistically significant changes due to treatment with antibacterial agents (clove oil and eugenol).

This discrimination in the metabolic profile of *E. coli* was not observed when eugenyl acetate, β -caryophellene, or a mixture of both was applied as antibacterial agents. As a proof of concept, a lethal concentration of clove oil was applied to 10^7 CFU mL⁻¹ of bacteria, and then we verified that the addition of eugenol alone in the same amount present in clove oil also prohibited bacteria growth. Conversely, individual additions of eugenyl acetate and β -caryophyllene at comparable

amount of what added through clove oil, did not suffice to stop bacteria metabolism, and no interactions were observed between eugenol and the other two constituents towards E. coli growth. Table 1 demonstrates the dysregulated metabolites found in samples treated with clove oil and eugenol. In fact, in case of eugenol, belonging to the phenylpropene class, it has been demonstrated that the hydroxyl group present in the molecule plays an effective role in the prevention of bacterial growth (A. O. Gill et al. 2002; Hyldgaard et al. 2012). As a consequence, the antibacterial activity of eugenyl acetate is lower than that of eugenol, as there are no free hydroxyl groups in its structure. Phenylpropenes, which are synthesized by plants, can disrupt the metabolism of bacterial cells by forming hydrogen bonds through interactions with proteins (Laekeman et al. 1990). On the other hand, β-caryophyllene, a natural bicyclic sesquiterpene, showed low or absent antibacterial activity against E. coli (Koutsoudaki et al. 2005). In other studies, clove oil and eugenol indicated strong antibacterial activities in comparison to βcaryophyllene towards periodontopathogenic bacteria (Moon et al. 2011). In addition, information related to the effect of clove oil on the fatty acid profile of the E. coli cell membrane, on proteins and anti-quorum sensive activity are elucidated in Supplementary Information (Sections 2.1, 2.2, and 2.3)

3.4 Effect of eugenol on VOC profile of E.coli

Bacterial production of VOCs has been largely investigated (Schulz and Dickschat 2007). Bacterial volatile emissions are mainly related to by-products of metabolic pathways; for instance, the emission of hydrocarbons, aliphatic alcohols, and ketones derived from fatty acid biosynthesis. The VOC profile of *E. coli* was investigated by HS-SPME-GCxGC-ToF/MS, with particular emphasis on metabolic perturbations induced by the treatment of bacterial culture with clove oil and its major constituent eugenol. The bacterial VOC profile was determined *in-vivo*, with the SPME coating exposed to the headspace of *E. Coli* cultures incubated at 37 °C, under agitation at 500 rpm, and during the stationary phase of the bacteria growth cycle (8 to 16 hours after sample preparation). The chromatograms obtained from bacteria culture before and after treatment with eugenol and clove oil are presented in Supplementary Information (Figure SI.10). The results obtained were processed by SIMCA software; PCA (Figure SI.11) and OPLS-DA (Figure 2) revealed a neat differentiation between non-treated samples, and samples treated with whole clove oil or eugenol only. From the S-plots, presented in Figures 2-c and 2-d, the features that mostly contributed to the separation of the investigated sets of samples were selected and further investigated. While the production of terpenoids by bacteria has already been reported

(Schulz and Dickschat 2007), in this study, the complex composition of the essential oil used to treat the bacterial culture made the process of distinguishing between bacterial-produced terpenoids and compounds deriving from the essential oil unreliable, as well as assessing the extent to which changes in their concentrations were exclusively related to the alteration of bacterial metabolism due to the antibacterial agents. Thus this class of compounds was not considered. Table 2 shows a list of the tentatively identified bacterial VOCs that most contributed to discrimination between cultures of E. coli in the presence or absence of antibacterial agents. Results demonstrated that variations in the amounts of metabolites (a decrease or increase in chromatographic areas) were dependent on the application of different compositions of antibacterial agents. As evidenced in Table 2, the compounds most affected by treatment with antibacterial agents were derivatives formed along the fatty acid biosynthetic pathway (Schulz and Dickschat 2007), such as alcohols, ketones, and esters. In particular, methyl ketones with an odd number of carbon atoms were reported to derive from the decarboxylation of even-numbered β-keto acids. The production of methyl ketones bearing an even number of atoms rarely occurs, since they derive from odd numbered fatty acids; however, production of 2-butanone has also been reported as a component of VOC emissions from several bacteria (DeMilo et al. 1996; Dickschat, Martens, et al. 2005; Elgaali et al. 2002; Lee et al. 1995; Robacker and Bartelt 1997; Sunesson et al. 1997). Esters also constitute a very important class of bacterial VOCs, and their enhanced volatility compared to their precursor carboxylic acids facilitates their determination by headspace sampling and gas-chromatographic analysis. Among the different esters detected, ethyl octanoate has already been reported as one of the constituents of E. coli VOC emissions (Elgaali et al. 2002). On the other hand, emission of aldehydes, especially unbranched aldehydes, rarely occurs owing to their reactivity. In this work, the aldehydes whose concentrations were mostly affected by treatment with clove oil and eugenol were furfural, benzaldehyde, and nonanal. Benzaldehyde production can be related to two different pathways, leading to the conversion of L-phenylalanine to benzoyl-CoA, whereas nonanal derives from oxidation of oleic acid. Other compounds deriving from L-tryptophan detected in this work were indole and 2,5dimethylpyrazine. In particular, the presence of pyrazines bearing one to four methyl substituents is widespread, and generally reported in the literature (Dickschat, Bode, et al. 2005; Dickschat, Reichenbach, et al. 2005; Schulz et al. 2004). Dysregulation of various VOC metabolites in treated samples demonstrated the effect of eugenol on several metabolic pathways of E. coli due to its interaction on different biological compartments, such as the cellular membrane. Considering this, alterations in the respiratory pathway of E. coli could be monitored by

investigating metabolites with a high vapor pressure escaping from liquid media with the use of headspace SPME. Previously, metabolome analysis by Cox et al. demonstrated that *E.coli* treatment by carvacrol changed respiration to fermentation, and caused K+ leakage (Cox et al. 1998). For example, decreases in indol, thiazol, methanethiol, and butanone levels demonstrate disruptions in tryptophan, thiamine, cysteine, methionine (and other amino acid-containing sulfurs), and riboflavin metabolism, respectively. Additionally, the levels of esters were also noticed to increase in treated samples. All the above-mentioned alterations in the volatomic profile could be attributed to the inhibition effect of eugenol on enzymes such as tryptophanase and lipoxygenase, due to deamination and termination or reduction of decarboxylation and.

Further interpretation of the observed changes would require a more comprehensive investigation and interpretation of biochemical pathways of *E. coli* growing under clove oil treatment, including integration of other 'omics' branches, such as transcriptomics and proteomics.

4. Conclusion

The current work demonstrated the suitability of Solid-Phase Microextraction as a reliable tool for the capture of variations of E. Coli metabolome and volatilome in response to natural antibacterial agents such as clove oil and its major component eugenol. Here, multivariate data analysis tools were successfully employed to investigate the E. coli metabolome profile under mild and severe stress conditions induced by these naturally-occurring antibacterial agents. SPME applied as a sample preparation technique provides clean extracts from complex biological samples, while allowing for its coupling to complementary separation platforms, as herein proposed for untargeted metabolomics strategies. The use of SPME coupled to GCxGC-ToF/MS and UPLC-MS provides a comprehensive metabolome snapshot of metabolites with a wide variety of physical and chemical characteristics, including volatiles, polar, and nonpolar metabolites, thus enhancing the amount of chemical information retrievable from the system under investigation in comparison to conventional extraction techniques. In the current work, shifts in different pathways involved in lipid, carbohydrate, and amino acid metabolisms, as well as cell signaling variations were monitored for treated and control samples. New alterations in the E. coli metabolic pathways as a function of treatment by clove oil have been suggested in this study, such as changes in proline and glycerol levels attributed to cell osmoregulation. Evaluation of the discriminant metabolites in treated samples confirmed that eugenol is a lead candidate for further development as an active agent in anti-cancer treatments, owing to its ability to cause glycolysis inhibition of E. coli as a model organism. Changes in different metabolite levels and pathways identified from our approach confirms the interaction of clove oil with various enzymes

in the model system biology, as well as the capacity of metabolomics to identify the influence of stressors on microorganisms. This approach offers a valid option for future analyses seeking to establish a comprehensive mechanistic understanding of *E. coli* response to clove oil.

Acknowledgments

The authors thank the Natural Sciences and Engineering Research Council (NSERC) of Canada for financial support, PAS Technology for the Concept 96-blade device and autosampler, and Prof. Mark Servos from the University of Waterloo for his permission to use the facilities in his laboratory at the Department of Biology. The authors also wish to acknowledge Dr. Richard Smith, Nathaly Reyes-Garces, Dr. Barbara Bojko, Dr. Angel Rodriguez Lafuente, and Dr. Selenia De Grazia for their kind help. E.G. and E.A.S.S. greatly appreciate the technical support of Olivier Niquette and KC Walbank with the GCxGC-ToF/MS system. E. C. thanks CNPq (Conselho Nacional de Desenvolvimento Cientifico e Tecnologico) for the post-doctoral fellowship.

Compliance with Ethical Standards:

Funding: This study was funded by Natural Sciences and Engineering Research Council (NSERC) of Canada. E. C. received his postdoctoral fellowship from CNPq (Conselho Nacional de Desenvolvimento Cientifico e Tecnologico), Brazil.

Conflict of interests: F. Mousavi declares that she has no conflict of interest. E. Gionfriddo declares that she has no conflict of interest. E. Carasek declares that he has no conflict of interest. E. A. Souza-Silva declares that she has no conflict of interest. J. Pawliszyn declares that he has no conflict of interest.

Ethical approval: This article does not contain any studies with human participants or animals performed by any of the authors.

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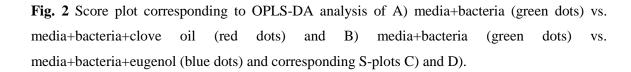
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Figure Captions

Fig. 1 XCMS online output. total ion chromatograms (TICs) before (A) and after (B) retention time correction. C: retention time deviation versus retention time for different analyzed samples. D: cloud plot: down-regulated (red bubbles) and up-regulated (green bubbles) features of *E.coli* sample treated by eugenol



Tables

Table 1 Dysregulated metabolites in *E.coli* samples treated by clove oil and eugenol, obtained with 96-blade SPME-LC/MS. Up-regulated metabolites are shown in red, while down-regulated metabolites are demonstrated in blue.

Metabolite	Adduct	mz	METLIN ID	Mass accuracy	Chemical Formula	Regulation p-value		
Amino Acids and Derivatives								
Proline	[M+H] ⁺	116.0707	58150	0.8	C ₅ H ₉ NO ₂	6.64574e-6		
Asparagine	[M+H] ⁺	133.0608	65674	0	C ₄ H ₈ N ₂ O ₃	7.37334e-6		
Tryptophan	[M+H] ⁺	205.0966	33	2	$C_{11}H_{12}N_2O$	0.00345		
N-Hydroxy-L-tryptophan	[M+H] ⁺	221.0920	73314	0	$C_{11}H_{12}N_2O$	9.57623e-6		
Valine	[M+H] ⁺	118.0864	71199	0.8	C ₅ H ₁₁ NO ₂	0.00008		
Tyrosine	[M+NH4] ⁺	199.1078	58353	0	C ₉ H ₁₁ NO ₃	0.00019		
Homoarginine	[M+H] ⁺	189.1343	5640	-1.6	C ₇ H ₁₆ N ₄ O ₂	0.00019		
Homoglutamine	[M+H] ⁺	161.0920	3281	0	C ₆ H ₁₂ N ₂ O ₃	0.00021		
Serine	[M+H] ⁺	106.0501	63419	2.8	C ₃ H ₇ NO ₃	0.00001		
Glutamate	[M+H] ⁺	148.0604	3761	0	C ₅ H ₉ NO ₄	0.00014		
Citrulline Nα-Acetyl-L-arginine	[M+ACN+ H] ⁺ [M+H] ⁺	217.1295	16	0	C ₆ H ₁₃ N ₃ O ₃	0.00014		
Histidine	[M+2Na- H] ⁺	200.0403	65529	-1.5	C ₆ H ₉ N ₃ O ₂	0.00005		
Aspartic acid	[M+H] ⁺	134.0448 63097		0	C ₄ H ₇ NO ₄	0.00003		
N6-Acetyl-N6-hydroxy-L- lysine	[M+H] ⁺	205.1182	63465	0	C ₈ H ₁₆ N ₂ O ₄	0.00007		
Threonine	[M+H] ⁺	120.0656	32	0	C ₄ H ₉ NO ₃	0.00007		
		Azolir	nes					
Creatinine	[M+ACN+ H] ⁺	155.092	<mark>7</mark> 8	0	C ₄ H ₇ N ₃ O	9.74675e-6		
	Benzene	and Substit		vatives				
Phoxim	[M+2Na- H] ⁺	343.025	4 725 30	1.6	$C_{12}H_{15}N_2O$ ₃ PS	0.00227		
	Carbohydrat	es and carb	ohydrate	conjugates				
Acetylmannosamine	[M+H] ⁺	222.097	$\begin{bmatrix} 335 \\ 7 \end{bmatrix}$	0.5	C ₈ H ₁₅ NO ₆	0.00002		
Glucose	[M+Na] ⁺	203.052	7 631	0.5	C ₆ H ₁₂ O6	0.00004		
Sucrose	[M+H- H2O] ⁺	325.112	137	-1.5	C ₁₂ H ₂₂ O ₁₁	1.17474e-6		

Glucosaminide	[M+H-	466.2041	704	1.1	C ₁₈ H ₃₅ N ₃ O	0.00003
	2H2O] ⁺				13	0.00000
	Carbox	ylic Acids and	Derivat	ives		
N-Acetylcystathionine	[M+H- H2O] ⁺	247.0748	665	1.6	C ₉ H ₁₆ N ₂ O ₅ S	1.48854e-6
Lysopine	[M+H] ⁺	219.1340	894 68	0	C ₉ H ₁₈ N ₂ O ₄	0.00368
Adipic acid	[M+Na] ⁺	169.0468	115	-1.7	C ₆ H ₁₀ O ₄	0.00055
Cystine	[M+H] ⁺	241.0310	636 35	0.4	$C_6H_{12}N_2O_4 \\ S_2$	0.00153
Glutathione	[M+H] ⁺	308.0909	44	0.3	C ₁₀ H ₁₇ N ₃ O ₆ S	0.00033
N-Acetylcadaverine	[M+H] ⁺	145.1336	659	0.7	C ₇ H ₁₆ N ₂ O	0.00026
γ-Glutamyl-γ- aminobutyraldehyde	[M+H] ⁺	217.1183	634 81	0	C ₉ H ₁₆ N ₂ O ₄	0.00119
Diaminopimelic acid	[M+H] ⁺	191.1027	352	0.5	C ₇ H ₁₄ N ₂ O ₄	0.00342
Homocitrulline	[M] +	189.1116	46	1.5	C ₇ H ₁₅ N ₃ O ₃	3.84520e-6
		Diazines				
Cytosine	[M+H] ⁺	112.0507	283	1.7	C ₄ H ₅ N ₃ O	0.00016
	Fatty	Acids and Co		S		
2-keto valeric acid	[M+ACN+ H] ⁺	158.0811	324	0	C ₅ H ₈ O ₃	0.00245
2E,4E-dodecadienoic acid (unsaturated)	[M+H] ⁺	197.1538	348 95	-1.0	$C_{12}H_{20}O_2$	0.00011
4,7,10-hexadecatrienoic acid (unsaturated)	[M+H] ⁺	251.2007	348 09	0.8	$C_{16}H_{26}O_2$	0.00010
dihydroxy-stearic acid (saturated)	[M+Na] ⁺	339.2509	<u>458</u> <u>32</u>	1.2	C ₁₈ H ₃₆ O ₄	0.00001
nonadienoic acid (unsaturated)	[M+NH4] ⁺	172.1332	351 03	0	C ₉ H ₁₄ O ₂	0.00024
hydroxypalmitic acid (saturated)	[M-H] ⁻	271.2265	458 29	-4.7	C16H32O 3	0.0001
Isodecenoic acid (unsaturated)	[M+H] ⁺	171.1380	347 22	0	C ₁₀ H ₁₈ O ₂	0.00015
tridecadienoic acid (unsaturated)	[M+H- H2O] ⁺	193.1588	349 09	2.1	C ₁₃ H ₂₂ O ₂	193.1588
dodecadienoic acid (unsaturated)	[M+H] ⁺	197.1538	348 95	-1.3	C ₁₂ H ₂₀ O ₂	0.00004

Myristoleic acid (unsaturated)	[M+H] ⁺	227.2007	642	0.8	C ₁₄ H ₂₆ O ₂	0.00496		
Myristic Acid Alkyne	[M+H- H2O] ⁺	207.1745	352 48	-1.9	C ₁₄ H ₂₄ O ₂	0.00190		
Decatetraenedioic acid	[M+NH4] ⁺	212.0918	748 98	0.5	C ₁₀ H ₁₀ O ₄	0.00168		
		Fatty acyls	3					
Octyl acetate	[M+K] ⁺	211.1095	461 96	0	$C_{10}H_{20}O_2$	0.00001		
Caproic acid	[M+K] ⁺	155.0469	111	0	$C_6H_{12}O_2$	7.38651e-6		
2-Tridecene-4,6,8-triyn-1-ol	[M+H] ⁺	187.1119	872 96	1.1	C ₁₃ H ₁₄ O	0.00306		
		Fatty aldehyo	des					
2,4,7-tridecatrienal	[M+H] ⁺	193.1589	753 53	1.5	C ₁₃ H ₂₀ O	0.00006		
Tetradecadienal	[M+H] ⁺	209.1902	464 61	1.4	C ₁₄ H ₂₄ O	0.00007		
		Glycerolipid	lc					
1-Octylglycerol	[M+K] ⁺	243.1357	466 20	0	C ₁₁ H ₂₄ O ₃	0.00065		
Glyceryl 5- hydroxydecanoate	[M+Na] ⁺	285.1669	883 41	-1.1	C ₁₃ H ₂₆ O ₅	0.00301		
nyuronyuccunoucc	Glv	cerophosphog						
PE(P-		cerophosphog	467		C ₄₅ H ₇₈ NO ₇	Down		
18:0/22:6(4Z,7Z,10Z,13Z,1 6Z,19Z))	[M+H- H2O] ⁺	758.5489	08	0	P	0.00220		
PA(13:0/20:5(5Z,8Z,11Z,14 Z,17Z))	[M+]	652.4114	812 51	-1.5	C ₃₆ H ₆₁ O ₈ P	0.00001		
PE(O-20:0/0:0)	M+2Na-H	540.3392	777 00	-1.5	C ₂₅ H ₅₄ NO ₆ P	0.00219		
	I	midazopyrimi	dines					
Adenine	$[M+H]^+$	136.0618	85	0	C ₅ H ₅ N ₅	0.00005		
Guanine	[M+H] ⁺	152.0568	315	1.3	C ₅ H ₅ N ₅ O	4.68891e-6		
6-Thiouric acid	[M+NH4] ⁺	202.0393	718	0	$C_5H_4N_4O_2$ S	0.00197		
Hypoxanthine	$[M+H]^+$	137.0458	83	0	C ₅ H ₄ N ₄ O	0.00110		
Indoles and Derivatives								
3-Methylindole	[M+H] ⁺	132.0809	545 3	1.5	C ₉ H ₉ N	8.91649e-7		
Indole	[M+H] ⁺	118.0653	286	1.6	C ₈ H ₇ N	0.00145		
Lactams								
Homoserine lactone	[M+K] ⁺	140.0108	658 63	0	C ₄ H ₇ NO ₂	8.96060e-6		
Lineolic acid and derivatives								
octadecadienoic acid	[M] ⁺	312.2300	353	0	C ₁₈ H ₃₂ O ₄	0.00002		

			40			
			48			
9-hydroperoxy-10E,12- octadecadienoic acid	[M] ⁺	312.2300	353 48	0	C ₁₈ H ₃₂ O ₄	0.0000054
		Prenol lipid	S			
2-Hexaprenyl-3-methyl-6-methoxy-1,4-benzoquinol	[M] ⁺	562.4379	628 13	-1.2	C ₃₈ H ₅₈ O ₃	0.00028
	Purine N	Nucleosides and	d Analo	gues		
Adenosine	[M+H] ⁺	268.1039	86	-0.4	$C_{10}H_{13}N_5O$	0.00029
Guanosine	[M+H] ⁺	284.0988	87	-0.4	C ₁₀ H ₁₃ N ₅ O	0.00006
Deoxyguanosine Adenosine	[M+ACN+ H] ⁺	309.1304	339	-0.6	C ₁₀ H ₁₃ N ₅ O	0.00241
cGMP	[M+H] ⁺	346.0541	348 5	1.7	C ₁₀ H ₁₂ N ₅ O ₇ P	0.00106
Isopentenyl adenosine	[M+H] ⁺	336.1666	640 86	0	C ₁₅ H ₂₁ N ₅ O	0.00103
Xanthine	[M+H]+	153.0407	82	0	C ₅ H ₄ N ₄ O ₂	0.0000053
1-Methyladenosine	[M+H] ⁺	282.1196	688 8	0	C ₁₁ H ₁₅ N ₅ O	0.00027
		Pyrenes				
Pyrene	[M+ACN+ H] ⁺	244.1121	699 74	0	C ₁₆ H ₁₀	0.00048
	Pyri	dines and Der	ivatives			
Nicotinic acid (Niacin)	[M+H] ⁺	124.0394	240	0.7	C ₆ H ₅ NO ₂	2.46054e-6
Pyridoxamine	[M+ACN+ H] ⁺	210.1238	238	0.4	C ₈ H ₁₂ N ₂ O ₂	0.00210
Pyridoxal (Vitamin B6)	[M+ACN+ H] ⁺	209.0921	220	0	C ₈ H ₉ NO ₃	0.00266
	Pyrimidine	e Nucleosides a	and Ana	logues		
Cytidine diphosphate choline	[M+H] ⁺	489.1142	358 1	-0.8	$C_{14}H_{26}N_4O$ $_{11}P_2$	0.00004
2',3' cyclic CMP	[M+H] ⁺	306.0481	624 29	1.3	C ₉ H ₁₂ N ₃ O ₇ P	1.92863e-7

Table 2 Dysregulated metabolites in *E.coli* samples treated by clove oil and eugenol, obtained by HS-SPME GCxGC-ToFMS. Up-regulated metabolites are shown in red, while down-regulated metabolites are demonstrated in blue.

		I		I	
Compound name	1 st dimention r.t., 2 nd dimension r.t.	Chemical formula	RI exp (RI lit*)	Molecular weight (Quantitation ion)	Regulation P(corr)
		Alcohols	·		
1-butanol	315,1.055	C ₄ H ₁₀ O	(675)	74.1 (56)	0.0753525
2-methyl-1-butanol	440,0.98	C ₅ H ₁₂ O	(739)	88.1 (56)	0.01631
Benzyl alcohol	1005,1.495	C ₇ H ₈ O	1037 (1037)	108(79)	0.001539
2-Heptanol	765,0.795	C ₇ H ₁₆ O	903 (905)	116.2 (45)	0.001413
2-nonanol	1115,0.75	C ₉ H ₂₀ O	1100 (1098)	144.3 (45)	0.072198
1-decanol	1370,0.74	C ₁₀ H ₂₂ O	1273 (1281)	158.3 (55)	0.04252
2-undecanol	1410,0.72	C ₁₁ H ₂₄ O	1300 (1309)	172.3 (45)	0.053873
2-tridecanol	1670,0.695	C ₁₃ H ₂₈ O	1500 (1510)	200.4 (45)	0.006879
		Ketones			
2-Butanone	240,0.695	C ₄ H ₈ O	(601)	72.1 (72)	0.009902
3-hydroxy-2-butanone	385,1.34	C ₄ H ₈ O ₂	(712)	88.1 (45)	0.033523
2-heptanone	735,0.715	C ₇ H ₁₄ O	889 (889)	114.2 (58)	0.008249
2-undecanone	1405,0.7	C ₁₁ H ₂₂ O	1297 (1294)	170.3 (58)	0.120224
2-tridecanone	1660,0.69	C ₁₃ H ₂₆ O	1492 (1496)	198.3 (198.3)	0.000535
		Esters			
butyl acetate	585,0.69	$C_6H_{12}O_2$	814 (813)	116.2 (56)	0.006677
methyl hexanoate	800,0.675	C ₇ H ₁₄ O ₂	923 (916)	130.2 (74)	0.006925
isoamyl acetate	715,0.7	C ₇ H ₁₄ O ₂	880 (877)	130.2 (55)	0.001604
ethyl hexanoate	940,0.695	C ₈ H ₁₆ O ₂	997 (999)	144.2 (88)	0.001018
methyl benzoate	1105,0.82	C ₈ H ₈ O ₂	1094 (1103)	136.1 (136)	4.33E-05
benzyl acetate	1205,0.82	$C_9H_{10}O_2$	1160	150.2 (108)	0.053059

			(1164)		
ethyl benzoate	1215,0.825	C ₉ H ₁₀ O ₂	1166 (1171)	150.2 (105)	0.00083
2-heptyl acetate	1010,0.715	C ₉ H ₁₈ O ₂	1040 (1034)	158.2 (87)	0.00147
methyl octanoate	1150,0.665	C ₉ H ₁₈ O ₂	1124 (1120)	158.2 (74)	0.009092
ethyl octanoate	1255,0.645	$C_{10}H_{20}O_2$	1191 (1190)	172.3 (88)	0.01690
9-decen-1-yl acetate	1540,0.685	$C_{12}H_{22}O_2$	1396 (1399)	198.3 (55)	0.02780
n-decyl acetate	1550,0.715	$C_{12}H_{24}O_2$	1404 (1407)	200.3 (70)	0.07923
	A	ldehydes			
furfural	625,1.305	C ₅ H ₄ O ₂	(895)	96.1 (96)	0.05318
benzaldeyde	875,0.935	C ₇ H ₆ O	964 (961)	106.1 (106)	0.04350
nonanal	1115,0.675	C ₉ H ₁₈ O	1094 (1103)	142.2 (57)	0.00761
	Other o	hemical classe	es		
4-(2,5-dihydro-3-methoxyphenyl)butylamine	1505,1.395	C ₁₁ H ₁₉ NO	1473 (1480)	181.2 (137)	0.01609
1-(4- hydroxybenzylidene)acetone	1660,1.095	$C_{10}H_{10}O_2$	1493 (na)	162.1 (147)	0.00115
methanethiol	170,3.14	CH ₄ S	(500)	48.1 (47)	0.02805
2-methoxy-2-methyl-propane	220,0.62	C ₅ H ₁₂ O	(554)	88.1 (73)	0.15244
thiazole	440,1.035	C ₃ H ₃ NS	(743)	85.1 (58)	0.00713
dimethyldisulfide	440,0.76	$C_2H_6S_2$	(747)	94.2 (94)	0.01649
2,5-dimethyl-pyrazine	785,0.795	C ₆ H ₈ N ₂	915 (911)	108.1 (108)	0.15859
indole	1415,1.85	C ₈ H ₇ N	1303 (1303)	117.2 (107)	0.27859
trans-1,3-dimethyl-2- methylenecyclohexane	830,0.7	C ₉ H ₁₆	940 (na)	124.2 (83)	3.55E-0
2-pentyl-furan	925,0.645	C ₉ H ₁₄ O	990 (na)	138.2 (81)	0.00299
2-ethenyl-1,1-dimethyl-3- methylidenecyclohexane	1130,0.64	$C_{11}H_{18}$	1110 (na)	150.3 (69)	0.12919
2,3,5,8-tetramethyl-decane	1365,0.65	C ₁₄ H ₃₀	1269 (na)	198.4 (57)	0.00747

^{*}Retention index data were retrieved from the following on-line databases: www.pherobase.com/database/kovats, www.chemspider.com, www.flavornet.org, webbook.nist.gov, http://www.vcf-online.nl, pubchem.ncbi.nlm.nih.gov/