

This is the author's manuscript



AperTO - Archivio Istituzionale Open Access dell'Università di Torino

Strategies for Accurate Quantitation of Volatiles from Foods and Plant-Origin Materials: A Challenging Task

Original Citation:	
Availability:	
This version is available http://hdl.handle.net/2318/1693441	since 2019-03-12T11:13:47Z
Published version:	
DOI:10.1021/acs.jafc.8b06601	
Terms of use:	
Open Access	
Anyone can freely access the full text of works made available as under a Creative Commons license can be used according to the t of all other works requires consent of the right holder (author or protection by the applicable law.	terms and conditions of said license. Use

(Article begins on next page)

Strategies for Accurate Quantitation of Volatiles from Foods and

Plant-Origin Materials: A Challenging Task

Barbara Sgorbini1*,	Cecilia	Cagliero ¹ ,	Erica	Liberto ¹ ,	Patrizia	Rubiolo ¹ ,	Carlo	Bicchi ¹	and	Chiara
Cordero ¹ *										

Authors' affiliation:

1 Dipartimento di Scienza e Tecnologia del Farmaco, Università di Torino, Turin, Italy

* Address for correspondence:

Prof. Chiara Emilia Cordero and Prof. Barbara Sgorbini - Dipartimento di Scienza e Tecnologia del Farmaco, Università di Torino, Via Pietro Giuria 9, I-10125 Torino, Italy –

e-mail addresses: chiara.cordero@unito.it; barbara.sgorbini@unito.it;

Phone: +39 011 6707172

Abstract

The volatile fraction of foods and of plant-origin materials provides functional information on sample-related variables and gas-phase extractions are ideal approaches for its accurate chemical characterization. However, for gas-phase sampling, the usual procedures adopted to standardize results from solvent extraction methods are not appropriate: headspace (HS) composition depends on the intrinsic physico-chemical analyte properties (volatility, polarity, partition coefficient(s)) and matrix effect. Method development, design and expression of the results are therefore challenging. This review article focuses on volatile vapour-phase quantitation methods (Internal Standard normalization, Standard Addition, Stable Isotope Dilution Assay, Multiple Headspace Extraction) and their suitability in different applications. Because of the analyte informative role, the different ways of expressing results (Normalized Chromatographic Area, Percent Normalized Chromatographic Areas and Absolute Concentrations) are discussed and critically evaluated with examples on quality markers in chamomile, process contaminants (furan and 2-methyl furan) in roasted coffee and keyaroma compounds from high-quality cocoa.

Key-words:

- Food volatiles; gas chromatography-mass spectrometry; accurate quantitation of volatiles; multiple
- 19 headspace solid phase microextraction

Introduction

The volatile fraction of foods and of plant-origin materials is a mine of functional information concerning sample-related variables: plant genotype and phenotype expression, pedoclimatic and harvest conditions, post-harvest processing and treatments, shelf-life storage conditions, and the effects of transformation/processing technologies. This fraction mainly consists of organic compounds with molecular weight generally below 350 u, characterized by medium-to-high Log Ko/w; they are readily released or vaporized from the condensed phase (solid or liquid) under suitable conditions (temperature, pressure, solubility in the medium, ion strength, etc.¹

Thanks to their physical-chemical properties, volatiles can easily be extracted or sampled, through gas-phase extraction approaches, also known as headspace (HS) sampling. Sampling from the vapour phase can be carried out under equilibrium or non-equilibrium conditions, and provides information about analytes distribution and concentration in the original sample on the basis of compound-specific partition coefficients K_{hs} . 1 K_{hs} is a constant, temperature dependent, related to the ratio of the analyte concentration in the sample condensed phase (solid or liquid) vs. that in the vapour phase – Equation 1.

$$K_{hs} = \frac{c_0}{c_a}$$

where: C_0 is the analyte concentration in the sample and C_g is the analyte concentration in the vapour phase or headspace.

Headspace sampling can be achieved through different approaches, mainly static (S-HS) or dynamic (D-HS), as well as in fully-automated systems combining analyte extraction, separation by gas chromatography (GC), and detection, either by mass spectrometry (MS) or by other suitable detectors (flame ionization detector FID, electron capture detector ECD or other element-specific detectors).

To achieve suitable sensitivity and, therefore, to access the highest possible level of information, provided by an almost comprehensive mapping of the food or plant volatilome ^{2–4}, high concentration capacity (HCC) techniques are of great help. In this scenario, adequate selectivity and extraction capability, to meet the required method sensitivity and specificity, can be achieved through static or dynamic accumulation of volatiles on polymers, operating in sorption and/or adsorption modes.

Headspace solid phase microextraction (HS-SPME) is undoubtedly the most popular HCC approach $^{3,5-10}$, since it is easy to standardize and can be fully integrated within the analysis system, thus making it an ideal solution for high-throughput screenings and comparative studies.

An HS-SPME system (and more in general an HCC-HS system) implies the distribution of analytes across three separate physical phases: the sample, the headspace, and the layer of polymer coating the fiber. Analyte recovery from the HS of a sample depends on two closely related but distinct equilibria: the first is the sample/HS equilibrium, which influences HS composition (measured by its distribution coefficient K_{hs}), and the second is the HS/fiber equilibrium (measured by its distribution coefficient, K_{fh}). The amount of analyte extracted from the fiber (n) at equilibrium is expressed as:

61 Eq. 2
$$n = \frac{K_{hs}K_{fh}V_{f}V_{s}C_{0}}{K_{hs}K_{fh}V_{f} + K_{hs}V_{h} + V_{s}}$$

where C_0 is the initial concentration of the analyte in the sample, K_{fh} is the fiber/headspace distribution coefficient; K_{hs} is the headspace/sample distribution coefficient, V_s is the sample volume, V_f is the fiber coating volume, V_h is the headspace volume

The equation shows that the amount of an analyte extracted by the SPME device and its initial concentration in the sample are in direct proportion, thus making HS-SPME appropriate for reliable quantitative analysis. The highest sensitivity (i.e. recovery) for an analyte is achieved when

sampling is carried out in equilibrium conditions. However, the dynamics of adsorption during the SPME process, as described in the theoretical model 11 , refers to the linear relation between n and C_0 ; thus, even in non-equilibrium conditions, reliable quantitation is possible. Theory also suggests which are the influential parameters to be tuned to achieve analyte recovery maximization: nature of SPME polymeric coating; coating volume, extraction temperature, sample agitation, pH and ionic strength, phase ratio, and extraction time.

Headspace sampling approaches: challenges for accurate quantitation

One of the most common practices related to volatile fraction characterization, also referred to as profiling ¹², is cross-sample comparison. Analytes and/or informative markers are identified in each sample headspace profile, and compared through quantitative indicators generally based on chromatographic peak areas (Raw Areas, Percentage Area), peak volume percentage for comprehensive two-dimensional GC (GC×GC) (Raw Volume, Percentage Volume), or internal standard (IS) normalization (Normalized Area, Normalized Volume). These approaches are based on analyte relative/normalized responses and, although accepted by the scientific community for several application fields ¹³, may be inaccurate or misleading if considered as analyte(s) concentration indicators.

These indicators are suitable for the cross-sample analysis of solid or liquid samples, provided that the effect of the condensed phase (matrix effect) on analyte release into the HS is comparable or known. In many cases, the matrix effect may have a dramatic influence on analyte recovery, and lead to erroneous conclusions ^{7,11–14}. In solid or heterogeneous samples multiphasic equilibria in the condensed phases are established.

Solid samples are characterized by a heterogeneous composition and structure; thus native volatiles may be partitioned (absorbed) or adsorbed into the solid particles network, conditioning

their release and equilibration with the HS. This is a very important factor when the focus of the investigation (profiling or fingerprinting) is generalized to all detectable compounds, since analytes may show widely different K_{hs} values and adsorption behaviour.

All parameters concurring to produce the phenomenon known as the matrix effect on volatiles release must be clearly characterized, or at least verified, in order to achieve reproducible, accurate and, above all, meaningful results. Suitable procedures capable of compensating/modelling the matrix effect are known as quantitation approaches and are generally based on external or internal calibration with authentic standards or stable isotopologues of target analytes. The most widely

- ✓ external calibration in matrix-matched blank samples suitable for liquid samples;
- ✓ standard addition (SA) by spiking the sample with known incremental amounts of analyte(s);
- ✓ stable isotope dilution assay (SIDA), which is a specific application of the SA;
- ✓ multiple headspace extraction (MHE).

used techniques are:

Whichever approach is applied, HS linearity conditions must be established ¹ during sampling for an accurate quantitation and to avoid errors. This condition is verified when the amount of condensed sample (liquid or solid), under the applied sampling conditions is enough to release the minimal analyte amount that matches method sensitivity and, at the same time, does not saturate the HS. Linearity is easily achievable for trace and sub-trace target analytes, or at least in those studies where the focus is on a few compounds, but it becomes challenging in multi-analyte quantitation.

The following paragraphs illustrate the most common quantitation approaches adopted in HS-SPME sampling, while discussing their suitability in studies where analyte concentration in the sample must be assessed with appropriate accuracy.

The application section presents case-studies selected from the authors' research activity, in which headspace quantitation was implemented through different approaches depending on the aims of the study, and the results will be critically analysed for their information potential.

Normalized responses: normalized peak areas and area percent responses

The use of an internal standard (IS), or multiple internal standards, is a recommended practice in GC ¹³ and, for HS-SPME sampling, it enables chromatographic response data to be normalized and recoveries of the accumulating polymer(s) compared, compensating for differences due to random instrumental errors. The IS must be an exogenous compound, with certified purity and stability, not present in the samples under study. It should mimic analyte(s) physico-chemical properties (e.g. volatility and polarity) and share chemical functionality(ies) with the analytes of interest. Co-elution issues must be taken into consideration, especially in those applications in which MS is not adopted for detection, and analyte response cannot be isolated by interferents through specific *m/z* fragments.

The IS must be added to the sample HS before sampling and in a concentration that falls within its linearity range of response.

An interesting alternative consists of the standard-in-fibre procedure introduced in 2005 by Wang et al. ¹⁴; this entails pre-loading the IS into the polymer coating before sampling. It informs about the reliability and efficiency of the accumulating polymer over time, and better exploits the isotropy of absorption and desorption of an analyte into and from the SPME device. This approach is useful in particular for heterogeneous and solid samples and is also suitable for on-site or *in vivo* applications.

Normalized areas and area percent responses, the latter calculated after IS normalization, are suitable analyte indicators for comparative purposes provided that the matrix effect is

comparable if not identical. These indicators are not, *per se* quantitative, since they simply relate to the relative amount of analytes transferred to the analytical system. Each analyte is therefore characterized by its own K_{hs} , K_{fh} and detector response factor.

Practical examples of the correct use of normalized responses are given concerning chamomile phenotyping and cocoa potent odorants (see below).

Standard Addition: single-point or multiple-point calibrations

Standard addition was one of the first approaches introduced for HS quantitation that was able to compensate the sample matrix effect at least for liquid samples. It consists of a series of experiments in which the original sample, and sample aliquots spiked with increasing and known amounts of reference compounds, are submitted to the analytical process.

With the approach known as "single-point" calibration, the analyte concentration in the sample is estimated with Equation 3:

154
$$A_{(0+a)} = (A_0/C_0) \times C_a + A_0$$
 Eq.3

where: C_0 is the amount of analyte in the sample, C_a the amount of analyte added to the sample, A_0 the instrumental response obtained from analysis of the original sample, and $A_{(0+a)}$ the instrumental response of the analyte obtained from analysis of the spiked sample.

Multiple standard additions, up to 6 levels, are recommended to improve accuracy. In linear regression analysis, C_a and C_0 are recorded so that the amount of analyte in the original sample C_0 is given by the ratio between the intercept C_0 and the slope C_0 , Equation 4:

163
$$b/a = A_0/(A_0/C_0)$$
 Eq. 4

Standard addition can be implemented by: (*a*) directly spiking the target analyte(s), in a gaseous state, into the sample headspace (gas phase addition - GPA); (*b*) spiking the analyte(s) in solution directly onto the sample (sample phase addition - SPA) or (*c*) spiking stable-isotope-labeled analyte(s) dissolved in a suitable solvent (stable isotope dilution analysis - SIDA) onto the sample. Stable isotope dilution assay, introduced by Grosh and Schieberle in 1987 ¹⁵ and extended to HS applications by Steinhaus et al. in 2003 ¹⁶ for the accurate quantitation of linalool enantiomers in beer after HS-SPME sampling, is also suitable for solid samples, although some limitations may affect the accuracy of the results. The heterogeneous nature of solids requires longer equilibration times for isotopologues to reach the full multiple partition equilibria with the analyte in all sample compartments/phases. When this condition is not achieved, since the labeled standard is generally spiked to the sample as a solution, it may be more recovered than the native analyte, which may be "trapped" in compartments of the sample, from where its recovery is more difficult.

Conversely, SIDA has some unquestionable advantages given by the very close chemical nature of the analyte and its isotopically labelled standard, thereby eliminating some sources of errors due to sample manipulation or sampling dynamics. These advantages are lacking when using IS with a different chemical nature. Quantitation by SIDA is achieved by spiking the sample with a known amount of labeled standard, provided that the relation between isotopologue ratio and intensity ratio is known, either by a multi-point calibration curve or by a relative response factor (RF) as described by Equation 5.

184
$$RF = \frac{(C_{analyte}/C_{labeled})}{(A_{analyte}/A_{labeled})}$$
 Eq. 5

where Canalyte and Clabeled are the concentrations and Aanalyte and Alabeled the instrumental responses.

The unavailability of labeled certified standards or, when available, their relatively high cost, are, to date, the most significant limitations affecting the widespread use of SIDA as quantitative approach in routine applications. However, it is fully compliant with EU guidelines and performance requirements ¹⁷ for quantitative methods on food contaminants, as here discussed below for furan and 2-methyl furan in roasted coffee.

Multiple headspace extraction

Multiple headspace extraction consists in a dynamic, stepwise gas extraction carried out to quantitate accurately volatiles in solid or heterogeneous samples where the matrix effect is known to play a relevant role in analyte release. The approach was introduced for S-HS applications and later adapted to HS-SPME ^{1,18–32}. It can be run as an external standard calibration and consists of three experimental steps:

- Step 1. Exhaustive extraction of analytes from calibration solutions within a range of concentrations/amounts, matching real-sample concentrations.
- Step 2. Exhaustive extraction of analytes from representative samples (solid, liquid or heterogeneous) to define HS linearity boundaries.
- Step 3. Application of the MHE procedure to samples of interest.

With the first two steps, a cumulative instrumental response function is built after repeated consecutive extractions from the HS of the same aliquot of calibration solutions or representative samples. Extractions must reach complete (exhaustive) analyte removal from the sample. Four to six consecutive extractions are recommended to validate the exhaustiveness of the extraction process for all analytes of interest. The decrease of the chromatographic peak area, after consecutive extractions, is exponential with HS linearity 1,22 . This condition refers to the linear function between the analyte concentration in the sample (C_0) and its concentration in the gas phase

 (C_g) , or between C_0 and the instrumental response (A) obtained when analysing an aliquot of the HS.

The actual linear range depends on the analyte partition coefficient (K_{hs}) and its activity coefficient.

In general, it ranges between 0.1 and 1% in the sample and can be tuned by modifying sampling temperature, equilibration times, and the ratio between HS (V_h) and condensed phase volume (V_s).

The sum of the instrumental response (As) from each step of HS extraction corresponds to the total response (A_T) as it is generated by the analyte originally present in the sample. Equation 6 is applied to obtain the cumulative instrumental response (A_T):

219
$$A_T = \sum_{i=1}^{-\infty} A_i = A_1 \frac{1}{(1-e^{-q})} = \frac{A_1}{(1-\beta)}$$
 Eq. 6

where A_T is the total estimated response (chromatographic area), A_1 is the analyte response after the first extraction, and q is a constant associated with the response exponential decay (β) through consecutive extractions.

The sequential steps corresponding to the exhaustive extraction of an analyte from a sample in HS linearity conditions are illustrated in Figure 1 for a HS-SPME approach.

The term q can be obtained by plotting the natural logarithm of the chromatographic peak areas as a function of the number of extractions. From this, a linear regression equation (Equation 7) can be calculated as follows:

230
$$\ln A_i = a (i-1) + b$$
 Eq. 7

where i is the number of extraction steps, b is the intercept on the y axis, and a is the slope.

 β (e^{-q}) is analyte and matrix dependent, and is a constant for those samples exerting comparable matrix effects ^{21,24}. β values can be adopted to confirm, or otherwise, HS linearity; its dependence

on K_{hs} provides information on matrix retention behaviour and on the actual release of target analyte under specified conditions.

When applied to calibration solutions, MHE provides data for external calibration; curves can be adopted to estimate analyte amount in the sample with a simplified procedure, where the analyte response after the first extraction step (A_1) is sufficient for accurate quantitation ³³.

The potential of MHE in combination with HS-SPME, also known as MHS-SPME, is illustrated here in two case studies and emphasized for accurate quantitation of process contaminants in roasted coffee samples, and for a multi-analyte quantitation, targeted to potent odorants in cocoa nibs and mass. This latter example also discusses the additional information provided by matrix effect characterization through the β parameter, and the advantages deriving from parallel detection by MS/FID.

Materials and Methods

See associated contents.

Applications

Comparative studies: chamomile chemotyping

Chamomile (*Matricaria recutita* L.) is a medicinal plant widely used in folk medicine for its well-known spasmolitic, sedative, anti-inflammatory and antiseptic effects 34,35 . The drug consists of the flower-heads while its biological activity is due to the presence of several secondary metabolites, including sesquiterpenoids, coumarins and flavonoids (apigenine and related glucosides) 34 . In this perspective, chemical composition of chamomile is fundamental for quality control and safety assessment; it is generally analyzed by qualifying the essential oil (EO) composition. The EO mainly consists of sesquiterpenoids, namely *trans*- β -farnesene, α -bisabolol, bisabolol oxide A, bisabolol oxide B, α -bisabolone oxide A, chamazulene, and spiroether. To facilitate qualification and to

rationalize chamomile classification according to compositional characteristics, Schilcher 36 defined six different chemotypes of chamomile in function of the sesquiterpenoids distribution in the EO (Table 1). Type A is characterized by bisabolol oxide A as main component, Type B by bisabolol oxide B, Type C by α -bisabolol, while Type D contains comparable amounts of bisabolol oxide A and B and α -bisabolol. The two other chemotypes were not included in the study

The conventional approach to distinguish the chamomile chemotypes entails EO profiling by GC-FID (hydrodistillation + GC-FID). Analysis is preceded by analyte identity confirmation through reference spectra comparison and Linear Retention Index (I_s) confirmation. The evaluation of target analytes relative distribution is by percent responses (chromatographic area %) from FID detection. FID shows very stable and reliable Response Factors (RF) over a wide linear range of concentrations 37 .

The conventional approach for chamomile EO chemotyping, although reliable and accurate, is time-consuming since, by applying the Pharmacopoeia protocol, it takes at least 4 hours for chamomile dried material hydrodistillation and not less than 30 minutes for a GC-FID/MS EO profiling.

An alternative approach to discriminate chamomile chemotypes has been proposed: the chamomile flower-heads' volatile fraction is characterized by HS-SPME combined with GC-FID and principal component analysis (PCA) targeted to EO markers ^{38–41}. For this specific application, chemotyping can be considered a comparative cross-sample analysis; thus quantitative data are not strictly necessary, the normalized responses (chromatographic area %) from the FID detector being sufficient for consistent comparisons between samples.

Experimental results, based on markers relative distribution (chromatographic areas %) validated the feasibility of applying a more time effective profiling approach by HS-SPME instead of hydrodistillation at the sample preparation level. The cross-validation was based on 127 samples

processed in parallel for EO direct profiling by GC-FID vs. automated HS-SPME-GC-FID of dried flower heads. The results are in Figure 2, which shows the score plots resulting from PCA on marker relative distribution (chromatographic areas %) resulting from EO GC-FID profiling (Figure 2A) or from direct analysis of chamomile dried plant material by HS-SPME-GC-FID (Figure 2B). Elaborations was on a sub-set of samples from the original study ³⁸. The total variability explained by the two principal components accounted for 71.34% of variability for EO and 75.16% for HS-SPME. In both cases the four chemotypes form coherent sub-clusters according to their different relative distribution of markers and the three well defined chemotypes (A, B and C) are independently clusterized while chemotype D occupies an intermediate position; this reflects Schilcher's classification, in which chemotype D is not definitively characterized by one prevailing component.

These results support the adoption of normalized responses (chromatographic area %) for comparative and discriminatory purposes. The use of the FID detector strengthens the reliability of the results, because RF values are almost identical for analytes with the same formulae ³⁷, while also offering linear responses within a wide range, which is useful when analyte amounts in the sample span over different orders of magnitude.

Accurate quantitation of target analytes: process contaminants in roasted coffee

Furan and its homologous are oxygenated heterocycles present in the volatile fraction after thermal treatment of different food crops and drinks, as one of the Maillard reaction products ⁴². These compounds, in particular furan, are considered food process contaminants, and thus their presence in food is the object of a constant attention by the U.S. Food and Drug Administration (FDA) and the European Food Safety Authority (EFSA) because of carcinogenic and cytotoxic activity in animals and harmful effects on human health ⁴³. In fact, the International Agency for Research on

Cancer has classified furan as a possible human carcinogen (Group 2B), therefore it must be quantified in food to guarantee safe consumption.

In 2004, the FDA proposed a method based on static headspace combined with gas chromatography-mass spectrometry analysis (S-HS-GC-MS) to quantitate furan in different food crops, following the SA approach ⁴⁵. The method was updated in two steps, in June 2005, and in October 2006 to extend its application range. Based on this protocol, and to overcome the relatively low sensitivity of S-HS, an enrichment step by HS-SPME with a Carboxen/PDMS fiber was proposed as alternative technique ^{46–48}, further using d4-furan as IS with external calibration as quantitation approach.

The authors' group ²¹ studied the accuracy and precision of the available quantitation approaches for determining furan and 2-methylfuran in roasted coffee and compared them in view of their possible application to on-line monitoring of these process contaminants during the roasting process. The quantitation approaches explored for their performance parameters where SA, SIDA and MHS-SPME; the results were compared and validated *versus* the FDA method taken as reference.

The sample set consisted of 150 coffee samples of different varieties (Arabica and Robusta) and origins (Costa Rica, Nicaragua, Colombia, Brazil and Kenya) submitted to different technological processing (roasting, cooling, grinding and degassing). Commercial blends were also included to cover sample variability during shelf-life storage.

To improve sampling repeatability, to facilitate the addition of the internal standard and to increase method sensitivity for furan and 2-methyl-furan, both hydrophobic analytes, sampling was on ground coffee suspended in ultrapure water. Details of all procedural steps followed for sampling and calibration are described in the associated contents and reported elsewhere ²¹.

The average concentrations (expressed as mg/kg) and related coefficient of variations (CV%) of furan and 2-methyl-furan for all analyzed samples, resulting from the application of the three quantitation approaches (SA, SIDA and MHS-SPME), were in close agreement with the FDA method, most of them giving CV% values well below 15%, chosen as limit of acceptance in agreement with the EU Decision on analytical procedures for food safety assessment ¹⁷.

The experimental results from MHS-SPME also suggested that accurate determination of analytes could be achieved by adopting an average β value (see Equation 7), instead of the specific value for each sample, thanks to the homogeneity of the matrix effect of the coffee samples under study. In addition, the possibility of quantifying analytes after a single-step extraction makes the MHS-SPME approach rapid and highly competitive with SA and SIDA.

To satisfy the ever increasing demand for rapid analyses, for use in routine control when a large number of samples must be screened, MHS-SPME sampling was also combined with direct MS detection (MS-nose) ^{21,23}, skipping the separation step by GC. Furan and 2-methylfuran quantitative results, as obtained by MHS-SPME-MS were comparable to those from conventional separative methods ²¹ giving satisfactory coefficients of variation (CV%). Figure 3 shows the very good correlation between quantitative data obtained by MHS-SPME and those from the FDA method (taken as reference) with a regression coefficient value of 0.974.

The determination of process contaminants on roasted coffee requires accurate and reliable quantitation ¹⁷. Due to the nature of the analytes (highly volatile and hydrophobic) and to the matrix effect exerted by ground coffee, HS sampling is the route of choice, whereas MHS-SPME, among available techniques, combines lower determination errors, low relative uncertainty (due to high repeatability of the sampling approach) and, once optimized, single shot analysis.

Extending method capabilities by MHS-SPME and parallel detection by MS/FID for high-quality cocoa odorants

The cocoa volatilome is very complex ^{49–53} and, among the several hundred volatiles identified in it, the potent odorants are the analytes that contribute to the characteristic aroma signature, or *aroma blueprint* ⁵⁴. The *sensomic* approach enabled key-aroma compounds of some cocoa varieties and chocolate products to be defined ^{55–57} by adopting the well-established workflow, which includes, as a fundamental step, the accurate quantitation of odorants in the reference food sample. Quantitation of potent odorants, revealed by GC-olfactometry (GC-O) after Aroma Extract Dilution Analysis (AEDA), provides a more consistent evaluation of the actual role played by single odorants and affords efficient aroma recombination studies ⁵⁸.

High-throughput profiling approaches are required when the investigation embraces the entire volatilome as informative fraction of sample functional characteristics (origin/phenotype, harvest and climate conditions, post-harvest practices, processing). Analytical automation on the entire process from sample preparation to separation and detection in combination with appropriate data processing, makes it possible to screen larger sample sets to achieve representative and consistent results.

In this study, accurate quantitation by MHS-SPME combined with GC-MS/FID targeted to several potent odorants, including key-odorants ^{55–57} and process indicators was addressed. Method flexibility was also discussed, since the quantitation of uncalibrated analytes by FID predicted relative response factors (RRFs). Thanks to the key characteristics of MHS-SPME, accurate quantitative results are achievable with just few analyses per sample and additional information on the matrix effect is obtained, describing the odorants released from the condensed phase (cocoa nibs or mass) ²². The parallel detection by MS/FID is complementary in nature: (*a*) analyte identity is confirmed by MS fragmentation pattern, and the amount is accurately assessed through

diagnostic ions (MS target ions – Ti profiles) and external calibration; (*b*) the FID response on the parallel detection channel provides additional confirmation on the analyte amount, through its specific response factor, which can also be predicted by combustion enthalpies and molecular formulae ³⁷. In this way, external calibration can be avoided.

The study focused on the key aroma compounds described by Schieberle and co-workers ^{55–57}; these include alkyl pyrazines (2,3,5-trimethylpyrazine - TMP, 2-ethyl-3,5-dimethylpyrazine, and 3,5-diethyl-2-methylpyrazine), with *earthy, roasted* notes; short-chain and branched fatty acids (acetic acid, butanoic acid, 2-methylpropanoic acid, and 3-methylbutanoic acid), with *rancid, sour*, and *sweaty* notes; Strecker aldehydes (2- and 3-methylbutanal) with *malty, cocoa* and *buttery* notes; phenylacetaldehyde, with *flowery honey-like* note. Additional investigated components are some esters (ethyl-2-methylbutanoate – *fruity*; 2-phenylethyl acetate – *flowery*; ethyl phenylacetate – *honey-like*), linear alcohols (2-heptanol – *green, fatty*), phenyl propanoid derivatives (2-phenylethanol – *flowery*), sulfur-derived compounds (dimethyl trisulfide - *sulfury*), and phenols (guaiacol – *phenolic*), benzaldehyde (*bitter almond-like*), 3-hydroxy-2-butanone/acetoin (*buttery*), ethyl octanoate (*green, fruity*) and (*E*)-2-phenyl-2-butenal discriminant for processing stage.

Performance parameters of the MHS-SPME-GC-MS/FID method are reported in the reference paper ⁵⁹ while the quantitation results, based on MS external calibration, are illustrated as a heatmap in Figure 4. The results, which are rendered in a relative colour scale (white to brown), correspond to the mean values of three replicated measurements from two sample batches. Hierarchical clustering (HC), based on Euclidean distances, informs about samples compositional similarities and on the analytes common trends within samples. On the basis of these results, odor activity values (OAVs) can be estimated, and the odorant ranking compiled.

To extend the method quantitation potential while keeping the results accurate and reliable, FID response factors can be exploited. The rationale behind the applicability of this quantitation

approach is related to HS linearity, the mandatory condition for MHE quantitation. By means of the analyte characteristic β value, A_T is predictable (ref. Equation 6) and thus the actual absolute amount of that analyte in the sample can be estimated through FID RRF. In case of liquid injections, the area ratio between the analyte and the IS added to the sample in known amounts, is normalized/corrected to the predicted RRF estimated from the molecular formula 60,61 .

The reference equation (Equation 9) to calculate analyte RRFs is as follows:

$$RRF = 10^{3} \left(\frac{MW_{i}}{MW_{IS}} \right) \left(\frac{-61.3 + 88.8n_{C} + 18.7n_{H} - 41.3n_{O} + 6.4n_{N} + 64.0n_{S}}{-20.2n_{F} - 23.5n_{Cl} - 10.2n_{Br} - 1.75n_{I} + 127n_{benz}} \right)^{-1} \text{Eq. 9}$$

where n_C , n_H , n_O , n_N , n_S , n_F , n_{Cl} , n_{Br} , n_I , and n_{benz} are, respectively, the number of carbon, hydrogen, oxygen, nitrogen, sulfur, fluorine, chlorine, bromine, and iodine atoms and the number of benzene rings. MW_i and MW_{IS} are the molecular weights of the analyte i and the IS (methyl octanoate), adopted by de Saint Laumer et al. to the develop the model 60 .

In the study cited, the analyte-specific RRF was corrected to the TMP/methyl octanoate ratio (i.e. RRF_{i,TMP}=0.7028/RRF_{i,methyl octanoate}) to adapt the model to TMP.

Table 4 reports the RRF values calculated for all analytes externally calibrated and for additional compounds of interest: 3-methylbutyl acetate (isoamyl acetate - banana-like), γ -butyrolactone (creamy), octanoic acid (sweaty), 2-ethyl-5(6)-methylpyrazine (roasty, nutty), phenylacetic acid (honey-like), phenol (phenolic), 2-acetyl pyrrole (musty), and 2,3-dihydro-3,5-dihydroxy-6- methyl(4H)-pyran-4-one (DDMP – informative of processing stages ⁴⁹). Other analytes 2-heptanone, 2-nonanone, and heptanal are known secondary products of lipid oxidation. For these additional compounds, the total chromatographic peak area (A_T) is estimated from four consecutive extractions of the same sample, from which the analyte characteristic β value can be calculated. MHS-SPME quantitation by FID RRFs results are validated for accuracy by cross-matching results

between external calibration on MS signals and FID RRFs. Regression analysis, reported in Figure 5, shows good correlation between approaches, and confirms the consistency of FID predicted RRFs for accurate quantitation.

At this stage, it is interesting to note that other descriptors, based on analyte responses and not on absolute amounts, would give a different picture concerning the samples volatile fingerprint. Graphs of Figures 6A and 6B, referred to Ecuador cocoa nibs (6A) and mass (6B), show analytes distribution based on relative responses (bars - normalized on IS) or by quantitative results (areas – true amount). The different trends followed by indicators clearly shows that the matrix effect influences HS composition, but if normalized indicators alone are used to derive information about analyte presence in the sample, this leads to erroneous conclusions. For cocoa mass, the intermediate exerting the strongest matrix effect on odorant release, relative indicators reliably inform about odorant distribution in the sample headspace. This could be of relevance for orthonasal perception of odorants.

Figure 7 shows the average β values for a selection of key-odorants, determined in cocoa samples of five different origins and at different stages of processing (nibs and mass); cocoa powders are also included to illustrate how this intermediate exerts its matrix effect. As the data clearly shows, cocoa mass strongly retains 2-heptanol, 2,3,5-trimethylpyrazine and benzaldehyde, which possess β values twice as high as those of nibs.

In this context, MHS-SPME is mandatory for accurate quantitation of key-odorants; of additional interest are data on actual analyte release from the condensed phase (mass, nibs and powder). β values, between gas-phase or calibration solution and real sample multiple extractions, give access to this phenomenon and may be of help in interpreting sensory analysis data.

446	Acknowledgment
447	The study on chamomile chemotyping and on process contaminant in coffee were supported by the
448	Ricerca Locale - Ex60% of University of Turin, Turin (Italy).
449	The cocoa study was carried out thanks to the financial support of Firmenich S.A. Geneva,
450	Switzerland.
451	

452 **References**

- 453 (1) Kolb, B.; Ettre, L. S. *Static Headspace-Gas Chromatography : Theory and Practice*; Wiley-VCH:
- 454 New York, 2006.
- 455 (2) Cordero, C.; Kiefl, J.; Reichenbach;, S. E.; Bicchi, C. Characterization of Odorant Patterns by
- 456 Comprehensive Two-Dimensional Gas Chromatography: A Challenge in Omic Studies. *Trends*
- 457 *Anal. Chem.* **2018**. https://doi.org/10.1016/j.trac.2018.06.005
- 458 (3) Cordero, C.; Schmarr, H. G.; Reichenbach, S. E.; Bicchi, C. Current Developments in Analyzing
- 459 Food Volatiles by Multidimensional Gas Chromatographic Techniques. J. Agric. Food Chem.
- **2018**, *66* (10), 2226–2236.
- 461 (4) Bicchi, C.; Maffei, M. The Plant Volatilome: Methods of Analysis. in High-Throughput
- 462 Phenotyping in Plants. Methods in Molecular Biology (Methods and Protocols); Normanly, J.,
- 463 Ed.; Humana Press: Totowa, NJ, 2012.
- 464 (5) Risticevic, S.; Vuckovic, D.; Lord, H. L.; Pawliszyn, J. 2.21 Solid-Phase Microextraction. In
- 465 *Comprehensive Sampling and Sample Preparation*; 2012; pp 419–460.
- 466 (6) Lord, H. L.; Pfannkoch, E. A. Sample Preparation Automation for GC Injection; Elsevier, 2012;
- 467 Vol. 2.
- 468 (7) Ross, C. F. *Headspace Analysis*; Elsevier, 2012; Vol. 2.
- 469 (8) Bicchi, C.; Cordero, C.; Liberto, E.; Sgorbini, B.; Rubiolo, P. Headspace Sampling of the Volatile
- 470 Fraction of Vegetable Matrices. *J. Chromatogr. A* **2008**, *1184* (1–2), 220–233.
- 471 (9) Bicchi, C.; Cordero, C.; Rubiolo, P. A Survey on High-Concentration-Capability Headspace
- Sampling Techniques in the Analysis of Flavors and Fragrances. J. Chromatogr. Sci. 2004, 42
- 473 (8).
- 474 (10) Yang, C.; Wang, J.; Li, D. Microextraction Techniques for the Determination of Volatile and
- 475 Semivolatile Organic Compounds from Plants: A Review. *Anal. Chim. Acta* **2013**, *799*, 8–22.

- 476 (11) Jiang, R.; Xu, J.; Lin, W.; Wen, S.; Zhu, F.; Luan, T.; Ouyang, G. Investigation of the Kinetic
- 477 Process of Solid Phase Microextraction in Complex Sample. Anal. Chim. Acta 2015, 900, 111–
- 478 116.
- 479 (12) Cordero, C.; Liberto, E.; Bicchi, C.; Rubiolo, P.; Schieberle, P.; Reichenbach, S. E.; Tao, Q.
- 480 Profiling Food Volatiles by Comprehensive Two-Dimensional Gas Chromatography Coupled
- with Mass Spectrometry: Advanced Fingerprinting Approaches for Comparative Analysis of
- the Volatile Fraction of Roasted Hazelnuts (Corylus Avellana L.) from Different Origins. J.
- 483 *Chromatogr. A* **2010**, *1217* (37), 5848-5858.
- 484 (13) Brevard, H.; Cantergiani, E.; Cachet, T.; Chaintreau, A.; Demyttenaere, J.; French, L.;
- Gassenmeier, K.; Joulain, D.; Koenig, T.; Leijs, H.; et al. Guidelines for the Quantitative Gas
- 486 Chromatography of Volatile Flavouring Substances, from the Working Group on Methods of
- 487 Analysis of the International Organization of the Flavor Industry (IOFI). Flavour Fragr. J. 2011,
- 488 *26* (5), 297–299.
- 489 (14) Wang, Y.; O'Reilly, J.; Chen, Y.; Pawliszyn, J. Equilibrium In-Fibre Standardisation Technique
- 490 for Solid-Phase Microextraction. J. Chromatogr. A 2005, 1072 (1), 13–17.
- 491 (15) Schieberle, P.; Grosch, W. Quantitative Analysis of Aroma Compounds in Wheat and Rye
- 492 Bread Crusts Using a Stable Isotope Dilution Assay. J. Agric. Food Chem. 1987, 35 (2), 252–
- 493 257.
- 494 (16) Steinhaus, M.; Fritsch, H. T.; Schieberle, P. Quantitation of (R)- and (S)-Linalool in Beer Using
- Solid Phase Microextraction (SPME) in Combination with a Stable Isotope Dilution Assay
- 496 (SIDA). J. Agric. Food Chem. **2003**, *51* (24), 7100–7105.
- 497 (17) Commission, E. Commission Decision 2002/657/EC Implementing Council Directive 96/23/EC
- 498 Concerning the Performance of Analytical Methods and the Interpretation of Results. Off. J.
- 499 Eur. Union **2002**, L221 (23 May 1996), 8–36.

- 500 (18) Penton, Z. E. B. T.-C. A. C. Chapter 10 Headspace Gas Chromatography. In *Sampling and*
- 501 Sample Preparation for Field and Laboratory; Elsevier, 2002; Vol. 37, pp 279–296.
- 502 (19) Marquez, A.; Serratosa, M. P.; Merida, J.; Zea, L.; Moyano, L. Optimization and Validation of
- an Automated DHS-TD-GC-MS Method for the Determination of Aromatic Esters in Sweet
- 504 Wines. *Talanta* **2014**, *123*, 32–38.
- 505 (20) Zapata, J.; Lopez, R.; Herrero, P.; Ferreira, V. Multiple Automated Headspace In-Tube
- 506 Extraction for the Accurate Analysis of Relevant Wine Aroma Compounds and for the
- 507 Estimation of Their Relative Liquid-Gas Transfer Rates. J. Chromatogr. A 2012, 1266, 1–9.
- 508 (21) Bicchi, C.; Ruosi, M. R.; Cagliero, C.; Cordero, C.; Liberto, E.; Rubiolo, P.; Sgorbini, B.
- Quantitative Analysis of Volatiles from Solid Matrices of Vegetable Origin by High
- 510 Concentration Capacity Headspace Techniques: Determination of Furan in Roasted Coffee. J.
- 511 *Chromatogr. A* **2011**, *1218* (6), 753–762.
- 512 (22) Nicolotti, L.; Cordero, C.; Cagliero, C.; Liberto, E.; Sgorbini, B.; Rubiolo, P.; Bicchi, C.
- Quantitative Fingerprinting by Headspace-Two-Dimensional Comprehensive Gas
- 514 Chromatography-Mass Spectrometry of Solid Matrices: Some Challenging Aspects of the
- 515 Exhaustive Assessment of Food Volatiles. *Anal. Chim. Acta* **2013**, *798*, 115–125.
- 516 (23) Sgorbini, B.; Bicchi, C.; Cagliero, C.; Cordero, C.; Liberto, E.; Rubiolo, P. Herbs and Spices:
- 517 Characterization and Quantitation of Biologically-Active Markers for Routine Quality Control
- 518 by Multiple Headspace Solid-Phase Microextraction Combined with Separative or Non-
- 519 Separative Analysis. *J. Chromatogr. A* **2015**, *1376*, 9–17.
- 520 (24) Griglione, A.; Liberto, E.; Cordero, C.; Bressanello, D.; Cagliero, C.; Rubiolo, P.; Bicchi, C.;
- 521 Sgorbini, B. High-Quality Italian Rice Cultivars: Chemical Indices of Ageing and Aroma Quality.
- 522 *Food Chem.* **2015**, *172*, 305–313.
- 523 (25) Brachet, A.; Chaintreau, A. Determination of Air-to-Water Partition Coefficients Using

- Automated Multiple Headspace Extractions. *Anal. Chem.* **2005**, *77* (10), 3045–3052.
- 525 (26) Brachet, A.; De Saint Laumer, J.-Y.; Chaintreau, A. Automated Multiple Headspace Extraction
- 526 Procedure: Adsorption Modeling and Determination of Air-to-Water Partition Coefficients.
- 527 Anal. Chem. **2005**, 77 (10), 3053–3059.
- 528 (27) Wenzl, T.; Lankmayr, E. P. Comparative Studies of the Static and Dynamic Headspace
- 529 Extraction of Saturated Short Chain Aldehydes from Cellulose-Based Packaging Materials.
- 530 *Anal. Bioanal. Chem.* **2002**, *372* (5–6), 649–653.
- 531 (28) Costa, R.; Albergamo, A.; Bua, G. D.; Saija, E.; Dugo, G. Determination of Flavor Constituents
- in Particular Types of Flour and Derived Pasta by Heart-Cutting Multidimensional Gas
- Chromatography Coupled with Mass Spectrometry and Multiple Headspace Solid-Phase
- Microextraction. LWT Food Sci. Technol. 2017, 86, 99–107.
- 535 (29) Costa, R.; Tedone, L.; De Grazia, S.; Dugo, P.; Mondello, L. Multiple Headspace-Solid-Phase
- Microextraction: An Application to Quantification of Mushroom Volatiles. *Anal. Chim. Acta*
- **2013**, *770*, 1–6.
- 538 (30) Birch, A. N.; Petersen, M. A.; Hansen, Å. S. The Aroma Profile of Wheat Bread Crumb
- Influenced by Yeast Concentration and Fermentation Temperature. LWT Food Sci. Technol.
- **2013**, *50* (2), 480–488.
- 541 (31) Slaghenaufi, D.; Tonidandel, L.; Moser, S.; Román Villegas, T.; Larcher, R. Rapid Analysis of 27
- Volatile Sulfur Compounds in Wine by Headspace Solid-Phase Microextraction Gas
- 543 Chromatography Tandem Mass Spectrometry. *Food Anal. Methods* **2017**, *10* (11), 3706–3715.
- 544 (32) Feng, H.; Skinkis, P. A.; Qian, M. C. Pinot Noir Wine Volatile and Anthocyanin Composition
- 545 under Different Levels of Vine Fruit Zone Leaf Removal. *Food Chem.* **2017**, *214*, 736–744.
- 546 (33) Kolb, B.; Ettre, L. S. Theory and Practice of Multiple Headspace Extraction. *Chromatographia*
- **1991**, *32* (11–12), 505–513.

- 548 (34) Bruneton, J. Pharmacognosie-Phytochimie, Plantes Médicinales; DOC, T. &, Ed.; Parix Cedex,
- 549 1999.
- 550 (35) Bisset, N. G. Herbal Drugs and Phytopharmaceuticals: A Handbook for Practice on a Scientific
- Basis. Boca Raton, Fla. New York Bot. Gard. **1994**, *59* (1), 102–103.
- 552 (36) Schilcher, H. Die Kamille. In Handbuch Für Ärtze, Apotheker Und Andere Naturwissenschaftler;
- 553 Wiss. Verl-Jess., Ed.; Stuttgard, 1987.
- 554 (37) De Saint Laumer, J. Y.; Leocata, S.; Tissot, E.; Baroux, L.; Kampf, D. M.; Merle, P.; Boschung,
- A.; Seyfried, M.; Chaintreau, A. Prediction of Response Factors for Gas Chromatography with
- Flame Ionization Detection: Algorithm Improvement, Extension to Silylated Compounds, and
- Application to the Quantification of Metabolites. J. Sep. Sci. 2015, 38 (18), 3209–3217.
- 558 (38) Rubiolo, P.; Belliardo, F.; Cordero, C.; Liberto, E.; Sgorbini, B.; Bicchi, C. Headspace-solid-
- Phase Microextraction Fast GC in Combination with Principal Component Analysis as a Tool
- to Classify Different Chemotypes of Chamomile Flower-Heads (Matricaria Recutita L.).
- 561 Phytochem. Anal. **2006**, 17 (4), 217–225.
- 562 (39) Marengo, E.; Baiocchi, C.; Gennaro, M. C.; Bertolo, P. L.; Lanteri, S.; Garrone, W. Classification
- of Essential Mint Oils of Different Geographic Origin by Applying Pattern Recognition
- Methods to Gas Chromatographic Data. Chemom. Intell. Lab. Syst. 1991, 11 (1), 75–88.
- 565 (40) Bicchi, C. P.; Binello, A. E.; Legovich, M. M.; Pellegrino, G. M.; Vanni, A. C. Characterization of
- Roasted Coffee by S-HSGC and HPLC-UV and Principal Component Analysis. J. Agric. Food
- 567 *Chem.* **1993**, *41* (12), 2324–2328.
- 568 (41) Ziegel, E. R. Statistics and Chemometrics for Analytical Chemistry. *Technometrics* **2004**, *46* (4),
- 569 498–499.
- 570 (42) Maga, J. A.; Katz, I. Furans in Foods. C R C Crit. Rev. Food Sci. Nutr. 1979, 11 (4), 355–400.
- 571 (43) Crews, C.; Castle, L. A Review of the Occurrence, Formation and Analysis of Furan in Heat-

- 572 Processed Foods. *Trends Food Sci. Technol.* **2007**, *18* (7), 365–372.
- 573 (44) IARC Monograph; **1995**, Vol. 63.
- 574 (45) FDA. Determination of Furan in Foods. Online 2006, Oct, 1–6.
- 575 (46) Goldmann, T.; Périsset, A.; Scanlan, F.; Stadler, R. H. Rapid Determination of Furan in Heated
- Foodstuffs by Isotope Dilution Solid Phase Micro-Extraction-Gas Chromatography Mass
- 577 Spectrometry (SPME-GC-MS). *Analyst* **2005**, *130* (6), 878–883.
- 578 (47) Bianchi, F.; Careri, M.; Mangia, A.; Musci, M. Development and Validation of a Solid Phase
- 579 Micro-Extraction-Gas Chromatography-Mass Spectrometry Method for the Determination of
- 580 Furan in Baby-Food. *J. Chromatogr. A* **2006**, *1102* (1–2), 268–272.
- 581 (48) M.S. Altaki; F.J. Santos; M.T. Galceran. Automated Headspace Solid-Phase Microextraction
- versus Headspace for the Analysis of Furan in Foods by Gas Chromatography-mass
- 583 Spectrometry. *Talanta* **2009**, *78* (4–5), 1315–1320.
- 584 (49) Magagna, F.; Liberto, E.; Reichenbach, S. E.; Tao, Q.; Carretta, A.; Cobelli, L.; Giardina, M.;
- Bicchi, C.; Cordero, C. Advanced Fingerprinting of High-Quality Cocoa: Challenges in
- Transferring Methods from Thermal to Differential-Flow Modulated Comprehensive Two
- 587 Dimensional Gas Chromatography. J. Chromatogr. A 2018, 1535, 122-136.
- 588 (50) Oliveira, L. F.; Braga, S. C. G. N.; Augusto, F.; Hashimoto, J. C.; Efraim, P.; Poppi, R. J.
- Differentiation of Cocoa Nibs from Distinct Origins Using Comprehensive Two-Dimensional
- 590 Gas Chromatography and Multivariate Analysis. *Food Res. Int.* **2016**, *90*, 133–138.
- 591 (51) Magagna, F.; Guglielmetti, A.; Liberto, E.; Reichenbach, S. E.; Allegrucci, E.; Gobino, G.; Bicchi,
- 592 C.; Cordero, C. Comprehensive Chemical Fingerprinting of High-Quality Cocoa at Early Stages
- of Processing: Effectiveness of Combined Untargeted and Targeted Approaches for
- Classification and Discrimination. J. Agric. Food Chem. **2017**, 65 (30), 6329–6341.
- 595 (52) Nazaruddin, R.; Osman, H.; Mamot, S.; Wahid, S.; Nor, A. Influence of Roasting Conditions on

- Volatile Flavour of Roasted Malaysian Cocoa Beans. *J. Food Process. Preserv.* **2006**, *30*, 280–
- 597 298.
- 598 (53) Aprotosoaie, A. C.; Luca, S. V.; Miron, A. Flavor Chemistry of Cocoa and Cocoa Products-An
- 599 Overview. *Compr. Rev. Food Sci. Food Saf.* **2016**, *15* (1), 73–91.
- 600 (54) Purcaro, G.; Cordero, C.; Liberto, E.; Bicchi, C.; Conte, L. S. Toward a Definition of Blueprint of
- Virgin Olive Oil by Comprehensive Two-Dimensional Gas Chromatography. J. Chromatogr. A
- **2014**, *1334*, 101–111.
- 603 (55) Schnermann, P.; Schieberle, P. Evaluation of Key Odorants in Milk Chocolate and Cocoa Mass
- by Aroma Extract Dilution Analyses. J. Agric. Food Chem. 1997, 45 (3), 867–872.
- 605 (56) Chieberle, P. E. S. Frauendorfer 2006 Identification of the Key Aroma Compounds in Cocoa
- Powder Based on Molecular Sensory Correlations.Pdf Unknown.Pdf. J. Agric. Food Chem.
- **2006**, *54*, 5521–5529.
- 608 (57) Frauendorfer, F.; Schieberle, P.; Chieberle, P. E. S. Changes in Key Aroma Compounds of
- 609 Criollo Cocoa Beans During Roasting Changes in Key Aroma Compounds of Criollo Cocoa
- Beans During Roasting. J. Agric. Food Chem. **2008**, 56, 10244–10251.
- 611 (58) Dunkel, A.; Steinhaus, M.; Kotthoff, M.; Nowak, B.; Krautwurst, D.; Schieberle, P.; Hofmann,
- T. Nature's Chemical Signatures in Human Olfaction: A Foodborne Perspective for Future
- 613 Biotechnology. *Angew. Chemie Int. Ed.* **2014**, *53* (28), 7124–7143.
- 614 (59) Cordero, C.; Guglielmetti, A.; Sgorbini, B.; Bicchi, C.; Allegrucci, E.; Gobino, G.; Baroux, L.;
- Merle, P. Odorants Quantitation in High-Quality Cocoa by Multiple Headspace Solid Phase
- Micro-Extraction: Adoption of FID-Predicted Response Factors to Extend Method Capabilities
- and Information Potential. Anal. Chim. Acta 2018. https://doi.org/10.1016/j.aca.2018.11.043
- 618 (60) De Saint Laumer, J. Y.; Cicchetti, E.; Merle, P.; Egger, J.; Chaintreau, A. Quantification in Gas
- 619 Chromatography: Prediction of Flame Ionization Detector Response Factors from Combustion

620		Enthalpies and Molecular Structures. Anal. Chem. 2010, 82 (15), 6457–6462.
621	(61)	Jean-Yves, de S. L.; Sabine, L.; Emeline, T.; Lucie, B.; M., K. D.; Philippe, M.; Alain, B.; Markus,
622		S.; Alain, C. Prediction of Response Factors for Gas Chromatography with Flame Ionization
623		Detection: Algorithm Improvement, Extension to Silylated Compounds, and Application to
624		the Quantification of Metabolites. J. Sep. Sci. 2015, 38 (18), 3209–3217.
625		

626 **Figures captions** 627 Figure 1. Sequential steps corresponding to the exhaustive extraction of an analyte from a sample 628 in headspace (HS) linearity conditions for a HS solid phase micro extraction (HS-SPME) approach; 629 from Cordero *et al* ⁵⁹. 630 Figure 2. Scores plot resulting from the Principal Component Analysis (PCA) on chamomile marker 631 analytes relative distribution (chromatographic areas %) resulting from essential oil (EO) GC-FID profiling (Figure 2A) or from direct analysis of dried plant material by HS solid phase micro extraction 632 (HS-SPME) GC-FID (Figure 2B). Modified from Rubiolo et al 38. 633 Figure 3. Linear regression analysis on a sub-set of coffee samples analyzed by multiple headspace 634 635 solid phase micro extraction (MHS-SPME) versus FDA method (taken as reference). 636 Figure 4. Heatmap illustrating the cocoa quantitation results based on MS external calibration; from Cordero *et al.* ⁵⁹. 637 638 Figure 5. Linear regression analysis on quantitation results obtained by multiple headspace solid 639 phase micro extraction (MHS-SPME)-GC-MS/FID: external calibration on MS signal vs. FID predicted 640 relative response factors (RRFs) results; from Cordero et al. 59. 641 Figure 6. Analyte distribution based on relative responses - bars (normalized on IS) or on 642 quantitative results – areas for cocoa nibs (6A) and mass (6B). 643 **Figure 7.** Average β values estimated in cocoa samples of five different origins and at different stages

of processing (nibs and mass) for a selection of key-odorants; from Cordero et al. 59.

644

Table 1. Chamomile chemotypes according di Schilcher ³⁶.

Chemotype	Compositional characteristics	Number of samples*
1 - Type A	bisabolol oxide A as main component	17
2 - Type B	bisabolol oxide B as main component	19
3 - Type C	lpha-bisabolol as main component	69
4 - Type D	comparable amount of $\alpha\text{-bisabolol}$ and bisabolol oxide A e B	22
5	lpha-bisabolone oxide A as main component	-
6	green essential oil, low amount of matricine	-

^{*}Sub-set of those from the original study [39]

Table 2. Cocoa samples, modified from ref 59 .

Origin	Commercial description	Supplier - Trader	Harvest year
Mexico	Chontalpa Cacao fermentado seco calidad Baluarte	"Mercados alternativos y solidarios para productos del campo S. de RL. de CV" Calle Exterior Manzana 17 Lote 18 Colonia Fracc. Lomas de Ocuiltzapotlan localidad Villa de Ocuiltzapotlan referencia Tabasco Mexico http://www.lacoperacha.org.mx	2016
Colombia	Fino de Aroma Colombia Premium 1	Newchem Srl, Via M.F. Quintiliano 30 20138 Milan, Italy http://www.newchem.it	2016
Sao Tomè	Superior Cacau Fino, good fermented	Satocao LDA -Morro Peixe, Distrito de Lobata São Tomé e Príncipe - CP 762 http://www.satocao.com	2016
Venezuela	Venezuela Superior fermented Carenero	Daarnhouwer & Co. B.V., Korte Hogendijk 18 1506 MA Zaandam, The Netherlands http://www.daarnhouwer.com/	2016
Ecuador	Ecuador ASS (Arriba Superior Selecto)	Domori S.r.l Via Pinerolo 72-74 10060 None (Torino), Italy	2016
Powder	Alkalized cocoa powder 22-24%	Gobino srl, Turin, Italy	

Table 3. Cocoa targeted analytes together with their experimental odor quality, odor threshold (ng/g orthonasal from oily matrix), I^{T}_{S} , Ti adopted for quantitation, and calibration range covered (absolute amount of analyte, ng); modified from ref ⁵⁹.

Target analyte	Odor quality	OT (ng/g)	Exp I ^T s	Ti (<i>m/z</i>)	Range (ng)
2 Hudrowy 2 hutonono	Dutton	800	1250	88	20-5000
3-Hydroxy-2-butanone 2-Heptanol	Buttery Fatty green	263	1295	80	1-100
2,3,5-Trimethylpyrazine	Nutty green	290	1365	122	1-50
2-Ethyl-3,5(6)-dimethylpyrazine	Nutty earthy	57	1406	135	1-50
Ethyl octanoate	Green, fruity	16	1411	88	1-50
Benzaldehyde	Almond	350	1478	77	1-50
2-Methylpropanoic acid	Cheesy	190	1590	88	20-5000
3-Methylbutanoic acid	Cheesy	22	1641	87	20-5000
Ethyl phenyl acetate	Honey-like	650	1695	91	1-50
2-Phenylethyl acetate	Flowery	233	1767	104	1-50
Guaiacol	Phenol	16	1808	109	1-50
2-Phenylethanol	Sweet, floral	211	1857	91	1-50
2-r nenyiethanor	Jweet, jiorui	211	1037		50-500
(E)-2-Phenyl-2-butenal	-	-	1955	115	1-50

OT – odor threshold; Exp I_{S}^{T} – experimental Linear Retention Indices; Ti, target ion.

Table 4. extended list of cocoa informative volatiles together with their experiments ITS, information for predicted FID relative response factors estimation and RRF values adopted for their quantitation; modified from ref ⁵⁹.

Target analyte	Exp <i>I</i> ⁷ s	MW	Formula	nc	n _H	no	n _{Arom}	n _N	RRF
Isoamyl acetate	1104	130.19	C7H14O2	7	14	2	0	0	0.63
2-Heptanone	1156	114.180	C7H14O2	7	14	1	0	0	0.76
Heptanal	1184	100.160	C6H12O	6	12	1	0	0	0.73
3-Hydroxy-2-butanone	1250	88.105	C4H8O2	4	8	2	0	0	0.46
2-Heptanol	1295	116.201	C7H16O	7	16	1	0	0	0.78
2-Ethyl-5-methylpyrazine	1353	122.171	C7H10N2	7	10	0	0	2	0.69
2-Nonanone	1360	142.242	C9H18O	9	18	1	0	0	0.81
2,3,5-Trimethylpyrazine (REF)	1365	122.170	C7H10N2	7	10	0	0	2	0.69
3-Ethyl-2,5-dimethylpyrazine	1406	136.198	C8H12N2	8	12	0	1	2	0.82
Ethyl octanoate	1411	172.268	C10H20O2	10	20	2	0	0	0.72
2-Ethyl-3,6-dimethylpyrazine	1425	136.198	C8H12N2	8	12	0	1	2	0.82
Benzaldehyde	1478	106.121	C7H6O	7	6	1	1	0	0.79
2-Methylpropanoic acid	1590	88.110	C4H8O2	4	8	2	0	0	0.46
γ-Butyrolactone	1574	86.090	C4H6O2	4	6	2	0	0	0.42
3-Methylbutanoic acid	1641	102.132	C5H10O2	5	10	2	0	0	0.53
Ethyl phenyl acetate	1695	164.204	C10H12O2	10	12	2	1	0	0.74
2-Phenylethyl acetate	1767	164.200	C10H12O2	10	12	2	1	0	0.74
Guaiacol	1808	124.140	C7H8O2	7	8	2	1	0	0.68
2-Phenylethanol	1857	122.160	C8H10O	8	10	1	1	0	0.84
(E)-2-Phenyl-2-butenal	1955	146.189	C10H10O	10	10	1	1	0	0.84
2-Acetyl pyrrole	1913	109.13	C6H7NO	6	7	1	0	1	0.58
Phenol	1955	94.11	C6H6O	6	6	1	1	0	0.79
Octanoic acid	2065	144.21	C8H16O	8	16	1	0	0	0.70
DDMP	2278	144.13	C6H8O4	6	8	4	0	0	0.35
Phenylacetic acid	2580	136.15	C8H8O2	8	8	2	0	0	0.58

Figure 1

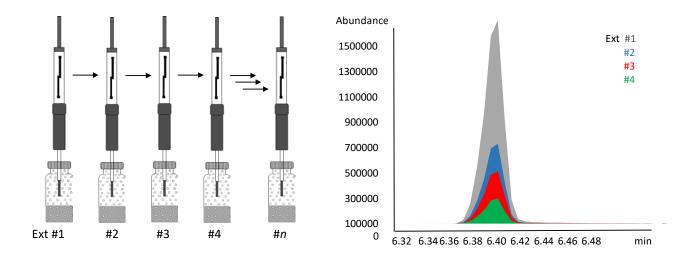


Figure 2

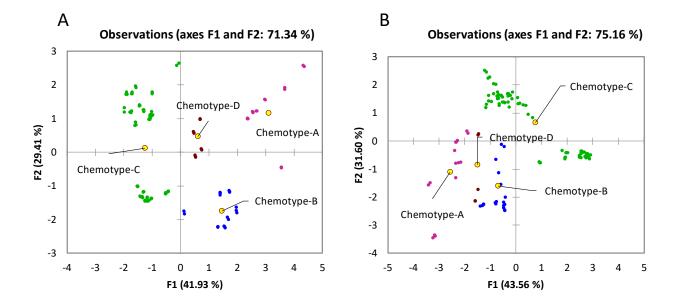
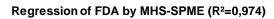


Figure 3



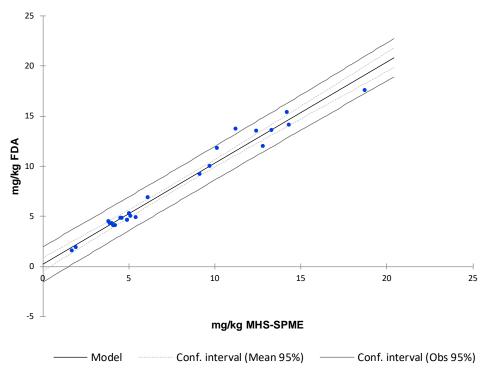


Figure 4

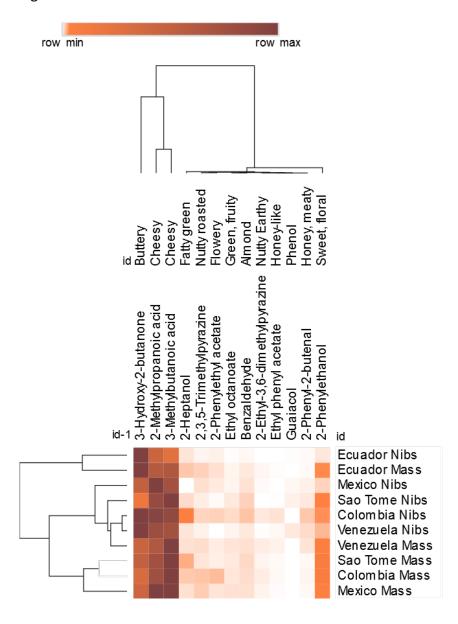


Figure 5

Regression of FID by MS (R²=0.9809)

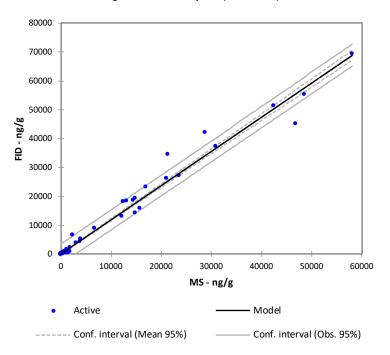


Figure 6

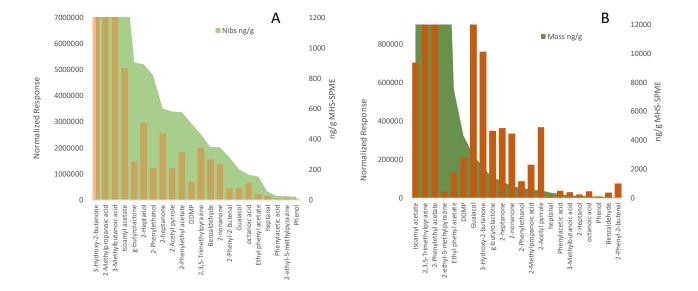
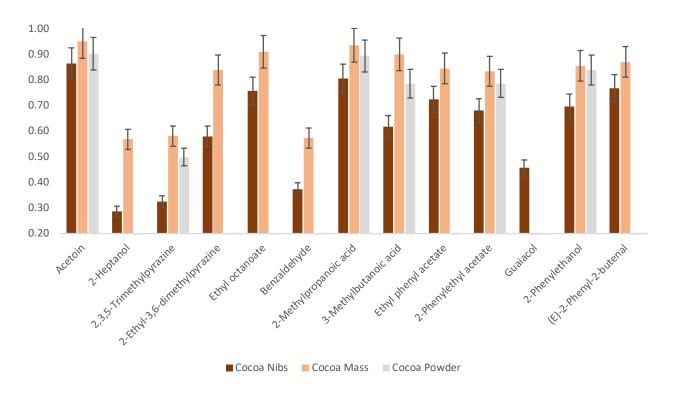


Figure 7



For Table of Contents Only

