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Performance evaluation of non-targeted peak-based cross-sample analysis for comprehensive two-dimensional gas chromatography-mass spectrometry data and application to processed hazelnut profiling

This is the author's manuscript

Original Citation:

Availability:

This version is available http://hdl.handle.net/2318/104646

since 2016-12-01T14:16:12Z

Published version:

DOI:10.1016/j.chroma.2012.04.048

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1	Performance evaluation of non-targeted peak-based cross-sample analysis for
2	Comprehensive Two-Dimensional Gas Chromatography-Mass Spectrometry data and
3	application to processed hazelnut profiling.
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23 Abstract

The continuous interest in non-targeted profiling induced the development of tools for 24 automated cross-sample analysis. Such tools were found to be selective or not 25 26 comprehensive thus delivering a biased view on the qualitative/quantitative peak distribution across 2D sample chromatograms. Therefore, the performance of non-targeted 27 28 approaches needs to be critically evaluated. This study focused on the development of a validation procedure for non-targeted, peak-based, GC×GC-MS data profiling. The 29 procedure introduced performance parameters such as specificity, precision, accuracy, and 30 uncertainty for a profiling method known as Comprehensive Template Matching. The 31 performance was assessed by applying a three-week validation protocol based on 32 CITAC/EURACHEM guidelines. Optimized ¹D and ²D retention times search windows, MS 33 match factor threshold, detection threshold, and template threshold were evolved from two 34 35 training sets by a semi-automated learning process. The effectiveness of proposed settings to consistently match 2D peak patterns was established by evaluating the rate of 36 mismatched peaks and was expressed in terms of results accuracy. The study utilized 23 37 38 different 2D peak patterns providing the chemical fingerprints of raw and roasted hazelnuts (Corylus avellana L.) from different geographical origins, of diverse varieties and different 39 roasting degrees. The validation results show that non-targeted peak-based profiling can be 40 reliable with error rates lower than 10 % independent of the degree of analytical variance. 41

The optimized *Comprehensive Template Matching* procedure was employed to study hazelnut roasting profiles and in particular to find marker compounds strongly dependent on the thermal treatment, and to establish the correlation of potential marker compounds to geographical origin and variety/cultivar and finally to reveal the characteristic release of aroma active compounds.

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Keywords: GC×GC-MS; validation; comprehensive template matching fingerprinting;
roasted hazelnut; *Corylus avellana* L. volatile fraction; key aroma markers.

51

52 **1. Introduction**

53 Comprehensive two-dimensional gas chromatography coupled with mass spectrometry (GC×GC-MS) is a powerful tool for targeted and non-targeted analysis of 54 55 complex mixtures of volatile compounds due to the enhanced peak capacity compared to GC [1-3]. 56 one-dimensional Non-targeted fingerprint analysis can reveal 57 qualitative/quantitative differences in chemical compositions facilitating the identification of potential marker compounds [4,5] and grouping or classification of samples [6,7]. Recent 58 59 publications have described the development of non-targeted, peak-based fingerprinting tools to exploit the informative content of three dimensional GC×GC-MS data sets 60 characterized by first (¹D) and second (²D) dimension retention times and mass 61 fragmentation patterns [5,6,8-12]. Such tools (name of the tool is set in brackets) were 62 developed by Oh et al. (MSort [8]), Wang et al. (DISCO [9]), Kim et al. (mSPA [12]), Castillo et 63 al. (Guineu [11]), Almstetter et al. (INCA [5]), and Leco Corporation (Statistical Compare, St. 64 65 Joseph, USA). The latter provides commercially available software that was tested by Almstetter et al. [10]. In a recent review, Reichenbach et al. [13] characterized such 66 67 approaches as peak feature analysis. Peak features collate the response data of individual analytes across chromatograms; for comparative analysis, analyte peaks must therefore be 68 consistently matched across 2D chromatograms. However, consistent peak feature analysis 69 remains challenging because retention times and mass spectra are subject to run-to-run 70 71 variations due to random or systematic errors depending on sample preparation, injection, 72 chromatographic and mass spectrometric conditions. This process of peak matching is a critical step of data elaboration, because matching errors produce false qualitative and 73 74 quantitative differences [13], thus complicating the ongoing data interpretation.

75 The extent of variability, generated during analysis, can be minimized by adopting automated sample preparation techniques, validated protocols for sample processing, and 76 robust instruments. On the other side, software for data processing should reliably match 77 78 corresponding analyte peaks, within a set of sample chromatograms, by accounting for the remaining variation on, for example, ¹D and ²D retention times and fragmentation pattern 79 80 intensities. This variability can be propagated during data processing, if the *feature* content 81 (i.e., response values of one analyte matched across many chromatograms) is erroneously 82 computed and/or elaborated by the software. Data processing errors can occur at different 83 stages and have partially been addressed: (a) removal of background [14], (b) peak detection

84 [15], (c) recognition of retention time shifts [16], and (d) peak alignment [5,8-12,16-18]. For 85 example, peak detection thresholds are used to detect and integrate peaks above a certain signal-to-noise ratio (S/N). Therefore, trace peaks may be detected in some samples, but not 86 in others [13]. This inconsistent peak detection may be propagated by the peak alignment 87 88 routine by matching just detected peaks while ignoring corresponding analyte peaks with lower S/N. Approaches to eliminate such propagation of errors have scarcely been 89 90 evaluated, but this issue is critical for the challenging task of a consistent and unbiased *peak* feature analysis. A further example is given by Castillo et al. [11]; the authors compared sixty 91 92 2D chromatograms of serum samples getting an aligned data file with 14756 compounds. However, only 1540 compounds were aligned across minimum six 2D chromatograms [11]. 93 This indicates that 90 % of the detected peaks were lost during data processing and these 94 might have included diagnostic analyte peaks. 95

96 Non-targeted, peak-based fingerprint analysis should produce reliable data on the qualitative/quantitative distribution of compounds within a sample set. Then, results can be 97 interpreted and research can provide compound identifications, structure elucidation of 98 99 unknowns, or quantitative information. It is proposed that the quality of such investigations can be assured by applying an appropriate validation procedure. The analytical protocol 100 101 adopted in the present study was designed and validated in accordance with general protocols (Eurachem/Citac guidelines [19,20]) to systematically investigate method 102 103 performance parameters (i.e., specificity, repeatability and intermediate precision, accuracy 104 and uncertainty) as a function of the most critical variables (data acquisition and data 105 elaboration levels). An existing alignment tool was adopted using Comprehensive Template 106 Matching as introduced by Reichenbach et al. [21]. The validated Comprehensive Template 107 Matching procedure was then applied to a food-type intervention study designed to identify 108 marker peaks highly informative for the hazelnuts roasting process. Various raw hazelnuts 109 were subjected to a roasting protocol to induce chemical changes which were recorded with 110 the help of the alignment routine. Hazelnuts are an ideal model system, because different known and unknown formation pathways are generating odorless volatiles and odorants 111 112 with different kinetics [22,23]. Therefore, non-targeted analysis is useful for studying the 113 biological and/or chemical response after perturbation or technological intervention and provides significant and valid information about interaction-relationships in a food system. 114

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116 **2. Materials and Methods**

117 2.1 Reference compounds and solvents

Pure reference compounds for identity confirmation and n-alkanes (n-C9 to n-C25) 118 were supplied by Sigma-Aldrich (Taufkirchen, Germany) except α/β -thujone (α/β -1-119 isopropyl-4-methylbicyclo[3.1.0]hexan-3-one, 95/5 weight ratio) supplied by Fluka (Milan, 120 Italy). A standard stock solution of α/β -thujone diluted to 45 ng/mL was prepared in 121 ultrapure water and the solution was stored in a sealed vial at 5 °C. Both, α - and β -thujone, 122 were used as internal standards for peak response normalization (ISTD) adopting the ISTD 123 124 loading procedure [24,25]. Solvents (cyclohexane, n-hexane, dichloromethane) were all HPLC-grade from Riedel-de Haen (Seelze, Germany). 3-Methyl-4-heptanone was synthesized 125 126 according to [22].

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128 **2.2 Hazelnut samples and roasting**

Raw and shelled hazelnuts of Corylus avellana L. (harvest 2009, supplied by 129 Marchisio, Cortemilia- CN, Italy) from different geographical origins and varieties/cultivars 130 131 were roasted at 160°C in a ventilated oven for 7, 12, 17 and 23 minutes. Tonda Gentile Romana (Romana, Lazio, Italy), Tonda Gentile delle Langhe (Gentile, Piedmont, Italy), Tonda 132 133 di Giffoni (Giffoni, Campania, Italy) were monovarieties and Azerbajian hazelnuts were a blend of different locally grown cultivars. Roasting was conducted every day and the 134 hazelnuts of a uniform dimension (caliber within 12-13 mm) were left at room temperature 135 136 to cool down. No storage of manually roasted hazelnuts was necessary, thereby avoiding an 137 alteration of the volatile fraction. The hazelnut samples were manually ground prior to vial 138 filling, and the particle size was compared to a ground reference sample.

Standard roasted hazelnuts (harvest 2009, supplied by Soremartec Italia SpA, Alba-CN, Italy) of Romana, Gentile and Giffoni were submitted to roasting in an industrial plant at different time/temperature ratios consistent with their desirable final sensory characteristics. These samples were hermetically sealed under vacuum in non-permeable polypropylene/aluminium/polyethylene packages and stored at -20°C prior to analysis.

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148 **2.3** Headspace Solid Phase Microextraction (HS-SPME) devices and sampling conditions

The SPME device and fibers were from Supelco (Bellefonte, PA, USA). A 149 Divinylbenzene/Carboxen/Polydimethylsiloxane (DVB/CAR/PDMS) df 50/30 μm, 2 cm length 150 fiber was chosen and conditioned before use as recommended by the manufacturer. 151 Roasted hazelnuts (1.5 g) were ground, sealed in a 20 mL vial, and equilibrated for 20 min at 152 50°C before sampling. The internal standard (ISTD) loading procedure onto the SPME fiber 153 [24,25] was as follows: the SPME device was manually inserted into a 20 mL sealed vial 154 containing 1 mL of ISTD (α/β -thujone) standard solution at 45 ng/mL, then the fiber was 155 exposed to the headspace at 50°C for 20 min. After ISTD loading, the fiber was exposed to 156 the matrix headspace at 50°C for another 20 min. Just the lower part of the vial, filled with 157 the solid sample, was heated to improve the vapor phase/fiber coating distribution 158 coefficient. The SPME device was directly introduced into the GC injector for thermal 159 desorption for 10 min at 250 °C. Samples for validation purposes were analyzed in triplicate 160 each week. Relative standard deviations (RSD %) for first and second retention times and 2D-161 peak normalized volume (i.e., cumulative 2D peak area) for 24 identified target analytes are 162 163 given (Table 1 and S1).

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5 **2.4 GC×GC-qMS instrument set-up**

GC×GC analyses were performed on an Agilent 6890 GC unit coupled with an Agilent 166 167 5975 MS detector operating in the EI mode at 70 eV (Agilent, Little Falls, DE, USA). The transfer line was set at 280°C. A Standard Tune was used and the scan range was set at m/z 168 35-250 with a scan rate of 10,000 amu/s to obtain a suitable number of data points for each 169 chromatographic peak for reliable identification and quantitation. The system was equipped 170 171 with a two-stage KT 2004 loop thermal modulator (Zoex Corporation, Houston, TX) cooled with liquid nitrogen and with the hot jet pulse time set at 250 ms with a modulation time of 172 4 s adopted for all experiments. Fused silica capillary loop dimensions were 1.0 m length and 173 100 µm inner diameter. The column set was configured as follows: ¹D Carbowax CW20M 174 column (100% polyethylene glycol)(30 m × 0.25 mm i.d., 0.25 μ m df) coupled with a ²D 175 OV1701 column (86% polydimethylsiloxane, 7% phenyl, 7% cyanopropyl) (1 m × 0.1 mmi.d., 176 177 0.10 µm df). Columns were from Mega (Legnano, Milan, Italy).

178 One microliter of the *n*-alkane sample solution was automatically injected with an 179 Agilent ALS 7683B injection system under the following conditions: split/splitless injector, 180 split mode, split ratio 1:50, injector temperature 280°C. The HS-SPME sampled analytes were 181 thermally desorbed from the fiber for 10 min into the GC injector using the following conditions: split/splitless in split mode, split ratio 1:20, injector temperature 250°C. The 182 carrier gas was helium at a constant flow of 0.7 mL/min (initial head pressure 260 KPa). The 183 temperature program was 40°C (2 min) to 180°C at 2.5°C/min and to 250°C at 20°C/min (5 184 min). Data were acquired by Agilent MSD ChemStation ver D.02.00.275 and processed using 185 GC Image GC×GC Software version 2.1b1 (GC Image, LLC Lincoln NE, USA). Statistical 186 analysis was performed with SPSS 14.0 (SPSS Inc. Chicago, Illinois, USA). 187

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189 **2.5 Validation protocol and data elaboration**

The validation protocol was designed and applied to characterize the following performance parameters: specificity, repeatability and intermediate precision of the analytical method (sample preparation and separation), and accuracy of results related to the *Comprehensive Template Matching* methodology. The five-days/three-weeks validation scheme is summarized in **Table S2**.

195 Raw data files were loaded into the software and were background corrected according to [14]. The 2D peaks were automatically detected by fixing a peak signal-to-noise 196 197 ratio (S/N) threshold of 10 and a footprint-area threshold of 10. The peak detection uses the watershed algorithm [15]. A set of 24 target peaks (Table 1), selected over the GC×GC 198 199 patterns to cover homogeneously the chromatographic run, were matched over nine 200 replicates of standard roasted Romana samples and over nine replicates of self-roasted 201 Romana samples to evaluate the inter- and intra-week variability of retention times, 2D peak 202 normalized volumes (normalization was done over the ISTDs 2D peak volume), and mass 203 spectra match factors. Target analysis was performed as described in [3]. Normality of the 204 distribution of normalized 2D peak volumes and normalized S/N values was tested with 205 Kolmogorov-Smirnov and Shapiro-Wilk test with p = 0.05 (Origin 6.1, OriginLab Corporation 206 MA, USA).

To determine optimized peak matching parameters, ¹D and ²D retention times search windows were fixed on the basis of the average standard deviations of retention-times data (**Table 1**). The thresholds were set to three times the standard deviation for each dimension giving a retention time window of 5 modulations for the ¹D and 0.17 seconds for the ²D dimension. Therefore, the peaks could be expected to elute within this time window with a high probability (here 99.7 %) considering run-to-run retention time shifts. Based on the
reported MS match factors (**Table 1**), a match factor threshold of 600 was used (see below)
to confirm target identity, accepting again a run-to-run match factor variability of three
times the standard deviation.

Validation was performed *(a)* to assess the fingerprinting method specificity by determining false-positive and false-negative error rates of the peak matching process; *(b)* to assess precision (repeatability and intermediate precision of standard roasted Romana samples) and to estimate the contribution of manual roasting to data dispersion (selfroasted Romana samples); *(c)* to evaluate fingerprinting method accuracy; and finally *(d)* to assess the fingerprinting method uncertainty.

After the validation step, Comprehensive Template Matching fingerprinting was 222 223 adopted to investigate chemical changes on extended data set samples of different geographical origins and roasting conditions. A schematic work-flow of the procedure is 224 reported in Figure 1 and details are discussed in the next paragraphs. The 23 sample 225 226 patterns were processed to build a consensus template by adding all the detected peaks, 227 exceeding fixed S/N and area thresholds from one randomly selected chromatogram within the set to an empty template. Then, this template was matched with a second 228 229 chromatogram to add unmatched peaks, and the resulting "updated template" was matched with a third chromatogram of the set and this was repeated with all patterns [6]. An 230 automatic retention time alignment was here used to compensate for pattern shifts due to 231 232 the intrinsic variation of chromatographic performance [16].

Each non-target peak included in the *consensus template* was saved with its chromatographic descriptors (see above) and every entry was labeled with a unique number to unequivocally identify each non-target peak.

236

237 **3. Results and Discussion**

This section is divided into two parts. The first part describes the routine adopted to assess performance parameters of the fingerprinting method. The second part is dedicated to a practical example where advanced fingerprinting is applied to profile the roasting process of hazelnuts of different origin and variety/cultivar altered by increasing roasting time at constant temperature.

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3.1 Assessment of method performance parameters

245 **3.1.1 Specificity**

Specificity is strictly related to the chromatographic separation and requires that a 246 peak is correctly assigned to an analyte [19]. From a chromatographic point of view, the 247 separation of hazelnut volatiles presented in this study is the result of an optimization 248 249 procedure based on systematic columns selection aimed at finding the best orthogonal column combination that minimizes the number of coelutions and maximizes the 2D peak 250 spreading over the chromatographic space [7]. With a view on data elaboration,, peak-based 251 252 features evaluate the analyte responses across many chromatograms, where one feature can contain information of spuriously matched peaks. Specificity was evaluated in three 253 steps: (a) optimization of peak matching parameters (i.e., ¹D and ²D retention times search 254 windows and MS match factors threshold); (b) matching of selected targets reported in 255 256 **Table 1** across nine replicate samples to establish peak correspondences and to verify peak 257 identities; and (c) analysis of blank samples to locate and remove interfering peaks.

258 The optimum peak matching involves low levels of false-positive and false-negative 259 hits, so that an analyte is consistently matched across several chromatograms and not mismatched with impurity/blank peaks or other interfering analytes. Figure 2 illustrates the 260 261 effect of different matching criteria on the peak matching specificity. When a template with 8 peaks (thin circles) is matched by defining just ¹D and ²D retention times constraints 262 (Figure 2a), analyte peaks are positively matched if the Euclidean Distance between sample 263 and template peaks does not exceed the user-defined retention time window threshold 264 (here of 5 modulations for the ¹D and 0.17 seconds for the ²D dimension). In this case, false-265 positive matches are denoted for peaks 1, 2 and 5-8. When a MS constraint with 600 match 266 factor is added, peak matching gives consistent assignment (Figure 2b). However, the 267 268 average MS match factors and the corresponding relative standard deviations (Table 1) indicate that an increasing MS threshold causes an increasing number of analyte peaks to fail
the matching criterion, thus producing false-negative hits. Figure 2c illustrates the effect of a
too restrictive peak matching obtained by increasing the MS threshold to 774 (Table 1): only
3 peaks of 8 template peaks are now correctly matched.

273 Results reported in Table 1 confirm that the method is specific for matching an extended set of target analytes containing resolved – unresolved peaks as well as high and 274 275 low abundant peaks. Peak 15 (nonanal), for example, showed a higher variation of 276 normalized volume, because of its long-tail and consequent difficult peak integration. False-277 positive and false-negative error rates were zero percent and the true-positive rate was 100 % for 24 target analytes matched across nine replicate chromatograms analyzed over 278 279 three weeks, supporting the effectiveness of the settings for the selected retention time 280 search window and minimum match factor.

On average, 179 2D peaks were detected above S/N 10 in blank sample runs (**Table S2**) and were associated to fiber bleeding, column bleeding or impurities derived from solvents and ISTDs reference material. Templates were matched against the blank runs and matched peaks were removed to obtain pruned templates which were then matched against sample chromatograms.

286 Literature dealing with peak-based comparative analysis also aimed at improving the specificity of the data elaboration. For example, Oh et al. [8] determined true-positive and 287 288 false-positive error rates by pair-wise matching the mass spectra of 46 derivatized standards 289 in 16 samples. A standard was positively matched when the Pearson correlation coefficient 290 was near one; a true-positive rate of 92 % and a false-positive error incidence of 11 % was 291 obtained. Retention times, however, were not used as constraints. Wang et al. [9] spiked 6 analytes and correctly matched 5 analytes across 5 replicate chromatograms. Kim et al. [12] 292 293 used mass spectra and retention times to match corresponding peaks across metabolite 294 samples from rat plasma; a true-positive rate of 70 % was determined. Almstetter et al. [5] 295 spiked 20 standard compounds to E. coli extracts and optimized the peak matching procedure until true-positive rate was 100 %. This is in accordance with results reported 296 297 here, except that our peak matching optimization calculates threshold values using simple 298 and intuitive peak descriptors, i.e., retention time standard deviations and mass spectrum match factor standard deviations, while other approaches adopted iteratively measured 299 300 threshold values requiring extensive computational work [5,8,9,12]. Thus, optimization of 301 matching parameters can be done routinely using a training set of sample chromatograms 302 and relative standard deviations of matching parameters calculated from peak features.

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304 **3.1.2 Precision**

305 Two levels of precision [26], repeatability and intermediate precision, were evaluated. The *repeatability* of the separation method was assessed by analyzing standard roasted 306 Romana and self-roasted Romana samples in single weeks, with the same instrument, 307 308 laboratory, and operator, and the intermediate precision was calculated over a period of 309 three weeks. Repeatability and Intermediate precision were calculated as relative standard deviations (RSD %) for chromatographic peak volumes (normalized volume vs. ISTD) and 310 retention times (¹D and ²D t_R) for each analyte in each validation week and over all weeks. 311 Results showed a good intermediate precision for retention times, with maximum averaged 312 RSD values of 1.0 % and 3.2 % for 1 D and 2 D respectively (**Table 1**). 313

A higher dispersion is evident for quantitative data (normalized volumes). The 314 normalized volumes for standard roasted Romana samples were submitted to the Analysis 315 316 of Variance (ANOVA) to evaluate the comparability of data between weeks. The One-Way ANOVA on the nine sample replicates collected over the three-weeks revealed that the null 317 318 hypothesis, "there is no difference between normalized volume values measured in different weeks" could be accepted for each of the 22 target compounds (excluding the ISTD) with 319 p <= 0.05. The RSD % on normalized volumes showed an average value of 12.4 %, which is an 320 321 acceptable intermediate precision (**Table 1**). The highest RSD % value was 47.6 % for peak 15 322 (nonanal) probably related to its long tailing, as already mentioned above.

The average RSD % of normalized volumes of the self-roasted Romana samples is, however, larger (i.e. 46 %). This increase of dispersion is remarkable and originates from sample processing, and not from the analytical procedure. With the designed validation protocol, it has been possible: (*a*) to estimate the magnitude of this external source of variation and (*b*) to evaluate better fingerprinting accuracy in the cross-comparison of realworld samples. The dispersion registered for the self-roasted Romana samples was considered by assigning a suitable uncertainty interval to fingerprinting results.

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331 **3.1.3 Accuracy**

332 The accuracy of the Comprehensive Template Matching fingerprinting data was verified [20] on nine standard roasted Romana samples and nine self-roasted Romana 333 samples acquired over the entire validation period and considering all separated and 334 detected 2D peaks. The extension of the validation procedure to the entire chemical pattern 335 (i.e., all 2D peaks above the fixed threshold) is required to evaluate fingerprinting accuracy 336 as a function of the chromatographic performance (specificity and precision) and data 337 elaboration parameters (specificity). Therefore, chromatograms were processed with an 338 optimized procedure to match consistently all detected analyte peaks across GC×GC 339 340 patterns collected on different days, within three weeks.

First, a *Sample Template* for each chromatogram of the set was created by including all detected peaks exceeding the fixed S/N and area threshold (see section *Validation protocol and data elaboration*). Each non-targeted peak included in each *Sample Template* was saved with its ¹D and ²D retention times, detector response, mass fragmentation pattern, match factor threshold and with a unique number. This number was assigned to a non-targeted analyte peak just after successful matching (**Figure 1**).

347 However, small variations of overall peak intensities were observed within the set. Figure 3 shows the distribution of S/N values of detected peaks across replicate 348 349 chromatograms. The histogram evidences that: (a) the S/N-threshold for peak detection should be sufficiently low, otherwise most of the peak information is discarded, and (b) the 350 351 S/N values, like 2D peak volumes, are normally dispersed. This means that S/N values of low 352 abundant peaks can scatter around the peak detection limit and hence can sometimes be 353 detected/matched and sometimes not (Figure 4a). Thus, only analytes with S/N values 354 always above the peak-detection limit were uploaded into the Sample Template to keep a consistent peak-matching. This was automatically done using internal standardization and 355 356 template thresholds that are higher than the peak detection limit (Figure 4b). This procedure enables reduction of the rate of mismatches resulting from template peaks that cannot be 357 matched with the corresponding analyte peaks, because they were not detected. The 358 template threshold was calculated for every chromatogram adopting the S/N of α - and β -359 360 thujone (ISTDs). The value was fixed as three times the standard deviation of S/N values 361 from the least intense chromatogram peaks (i.e., those peaks just above the peak detection limit) multiplied by the ISTD S/N abundance. The resulting thresholds are reported in Table 362

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2.

The closeness of agreement (i.e., accuracy) for the fingerprinting was verified through a *pair-wise comparison* showing that the method achieves a high similarity rate, which is always within a fixed interval and always above a certain value (**Table 2**) for samples of the same origin and roasting conditions acquired over the entire validation period. The similarity rate is here expressed as percentage of matched peaks between sample pairs: a maximum of 100% is expected for patterns obtained by analyzing the same sample even over an extended time period.

371 The results show that arbitrarily selected Sample Templates reliably match the 372 chromatograms of all replicates (Table 2). Accuracy results, expressed as percentage of matched peaks, ranged from a minimum of 91 % (for the sample acquired in week-one, day-373 two) to a maximum of 95 % with 2 % RSD (standard roasted hazelnut samples). The 374 375 percentages of matched peaks of self-roasted hazelnut samples ranged from a minimum of 376 85 % (for the sample acquired in week-three, day-three) to a maximum of 95 % with 3 % RSD. As a consequence, the peak matching performance proved to be robust for samples 377 378 with high dispersive quantitative values (standard roasted samples 12.4 RSD % and self-379 roasted samples 46.2 RSD % on normalized peak volumes).

The qualitative differences in peak patterns are important and should consistently be 380 381 "extractable" by compensating retention time shifts (cf. 3.1.1) and dispersive response values through appropriate template threshold values. On the contrary, problems with 382 383 consistent peak matching are often reported in literature. Oh et al. [8] compared eight 384 replicate chromatograms of a derivatized fatty acid/organic acid mixture with eight replicate 385 chromatograms of a derivatized fatty acid/amino acid mixture and resulted in a list with 386 8683 features, while only 46 were expected to be generated from standard compounds. 387 Performing a metabolite profiling on wild-type vs. double mutant E. coli strains, Almstetter 388 et al. [5] obtained a list of 2259 features from nine replicates per sample group (a peak 389 detection threshold of 500 was fixed). To limit the number of entries, the authors excluded 390 those features that could not be matched in at least 9 of 18 samples resulting in 398 peaks. Castillo et al. [11] studied the metabolic profile of 60 human serum samples, from which 391 392 14756 features were extracted, but only 1013 features were found to be useful. Kim et al. 393 [12] aligned 5 replicate samples of rat plasma with an average of 446 analyte peaks per 2D chromatogram, but just 146 peaks could be matched throughout all replicates. Although 394 395 these works might have handled differently complex peak patterns (depending on sample composition, sample preparation and injection technique), the evaluation and optimizationof fingerprinting accuracy could help to reliably extract their true qualitative differences.

In our study, analyte peaks (on average 166 for each standard roasted sample 398 chromatogram) could successfully be extracted from the larger number of detected peaks (in 399 average 387 for standard roasted samples) after a systematic optimization of data 400 elaboration parameters. As a consequence, analyte peaks could reliably be matched across 401 replicate chromatograms acquired over three weeks with an average of 94 % of true-positive 402 403 peak matches. Manual investigation of the missing peak matches revealed that especially coeluting compounds were not reliably detected by the software (Figure 2). This limit might 404 405 be overcome by adopting a suitable deconvolution/unmixing algorithm for unresolved peaks. 406

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408 **3.1.4 Uncertainty**

Equation 1

The goal of the fingerprinting is to reveal qualitative and quantitative differences within a set of samples. As a consequence, the uncertainty [20,21] should account for the dispersion of the quantitative data, mainly influenced by sample preparation and chromatographic separation, and the consistency of the qualitative data (% of matched peaks) mostly influenced by method specificity and accuracy.

The combined standard uncertainty can be calculated through the classical equation [20]:

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$$5 \quad [u(y)/y]^2 = [u(y)_A/y]^2 + [u(c)_B/c]^2 + [u(d)_B/d]^2 + [u(f)_B/f]^2 + [u(p)_B/p]^2$$

where u(y)/y is the combined standard uncertainty for the measurand y; $u(y)_A/y$ is the uncertainty referred to as repeatability and intermediate precision data; $u(c)_B/c$ is the uncertainty derived from calibration data; $u(d)_B/d$ is the uncertainty derived from dilution; $u(f)_B/f$ is the uncertainty derived from the efficiency of the method; and $u(p)_B/p$ is the uncertainty derived from errors on weight.

The contribution to the combined uncertainty of the method of sample weighing and dilution were negligible being respectively 1.0E-6 and 1.0E-4. Calibration was not included in this analytical procedure; thus, the uncertainty range, to be associated to normalized volumes after successful peak matching, was expressed through $u(y)_A/y$ (i.e., *Repeatability* of the method). The relative uncertainty associated with the quantitative results was calculated
as +/-12.4 %.

On the other hand, the standard uncertainty for the *Comprehensive Template Matching* fingerprinting to be associated with the result of the cross-comparison of samples and expressed as percent of matched peaks, was referred to as the accuracy of the data (i.e., % error) of the peak-based fingerprinting. The relative uncertainty associated with the fingerprinting results was +/-6 %.

434

435 **3.2** Application of the validated pair-wise Comprehensive Template Matching 436 fingerprinting in profiling the roasting process of hazelnuts.

437 *Comprehensive Template Matching* fingerprinting has been demonstrated to be a 438 general tool to compare sample fingerprints without any knowledge of sample composition. 439 To be truly comprehensive, the fingerprinting procedure should evaluate the complete peak 440 information and, as a consequence, all 2D analyte peaks of all sample chromatograms should 441 accurately be aligned.

442 This is demonstrated by studying the effect of roasting at different time-temperature profiles on hazelnut volatile development. In view of this, the validated pair-wise 443 444 comparison method was applied to extract temperature-sensitive *features* from a representative sample subset (here Tonda Gentile). The resulting consensus template then 445 446 was used to extract qualitative/quantitative data on analytes in all 23 chromatograms. This 447 was done to verify the independence of the potential marker compounds from the 448 geographical origin and variety/cultivar of the hazelnuts. The consensus template was 449 created according to the procedure described in validation protocol and data elaboration 450 section (Figure 1).

451 Marker compounds indicating roasting should be detectable at an early stage of heat treatment. The potential marker compounds are detectable within 7 minutes of roasting, 452 453 making it reasonable to create a consensus template for the Tonda Gentile with samples of 0 and 7 minutes roasting time. The consensus template of the sample pair was next matched 454 455 with all chromatograms of the Tonda Gentile subset. The average matching rate was 88 % 456 (n=5, i.e., 0, 7, 12, 17 and 23 minutes roasting) across all chromatograms. Peak data were automatically aligned. The resulting list, containing the respective normalized peak volume 457 458 values, was sieved to define significant *features* (Table 3). The fold changes of increasing 459 response values were calculated relative to the peak responses of the 7 min roasted 460 hazelnut sample. A two-fold increase (corresponds to minimum +200%; Table 1) was required to define a significant feature for self-roasted hazelnut samples. Finally, 24 of the 461 83 features were regarded as "significant indicators" for the roasting process of Tonda 462 463 Gentile samples. Ongoing data reduction addressed potential roasting markers that were independent of geographical origin or variety. Thus, normalized peak volume values of the 464 selected 24 supposed markers were extracted from all chromatograms of the entire sample 465 466 set, using specific quantifier ions (Table S3), and submitted to a Principal Component 467 Analysis (PCA)

PCA was used for an unsupervised analysis and was performed initially on each 468 variety/origin independently (Figure S1) to show the degree of correlation between 469 470 potential marker compounds and roasting time. The first principal component (roasting degree) explained on average 71 % of the total variance. Several marker features had 471 472 loadings >+/-0.6 on component 2, indicating dependence on an unknown factor. Features 473 with loadings of more than 0.6 for component 1 and loadings of <+/-0.6 for component 2 474 were deemed strong markers for "roasting degree" thus enabling the removal of visually recognizable "outliers". The number of potential markers was thereby reduced from 24 to 475 476 11 (Table 3). Linear regression analysis on normalized volume values of the proposed 11 marker compounds for all hazelnut samples (Figure 5) revealed a nearly proportional 477 478 relationship between increased roasting time and increased normalized peak volume values, with R² of 0.8147. A look backward into raw data (Figure 3) revealed that normalized 479 480 volumes of markers, e.g., 2,3-pentandione and unknown compound 81, steadily increased 481 with roasting time, whereas 5-methyl-(E)-2-hepten-4-one and unknown compound 79, 482 excluded on the basis of the PCA screening, reached their maximum after 12 minutes (Figure 483 **S1, S3**).

This *Comprehensive Template Matching* fingerprinting method combined with a food-type intervention study has proven to be reliable and straightforward to select significant peak data from a set of 23 unique sample files. The simple linear model derived from marker compounds, for example, allows an interpolation and prediction of roasting time, thereby facilitating automated food processing. With manual roasting, the marker compound responses considerably vary (Figure 5) limiting the possibility to distinguish between smaller time intervals. These results also provide deep insights into the formation 491 of volatile compounds during roasting and differences emerging from different varieties of different origin. Alasalvar et al. reported on the formation of 71 volatile compounds from 492 hazelnuts from Tombul (Turkey) [27]. Burdack-Freitag and Schieberle [22] recently 493 investigated the formation of key aroma compounds during roasting of Tonda Romana (Italy) 494 hazelnuts, showing that 2-methyl-butanal (8, malty), 2,3-pentandione (19, buttery), 495 phenylacetaldehyde (77, flowery), and 5-methyl-(E)-2-hepten-4-one (52, filbertone, nutty) 496 have an important impact on the aroma of roasted hazelnuts. However, GC×GC-qMS has 497 498 been used in this study to explore systematically the effect of roasting at different time intervals on the formation of odor-active as well as odorless volatile compounds. 499 Concentrations of identified odorants 8, 19, 77 (2-methylbutanal, 2,3-pentandione, 500 phenylacetaldehyde) increase almost linearly with roasting time independently on variety 501 502 suggesting them as suitable marker compounds to predict the degree of roasting. Future studies could investigate whether these odorants can evoke different distinct aromas 503 504 despite their similar release profiles across different hazelnuts. 1-Methyl-pyrrole (29), 505 pyridine (35), 3-hydroxy-2-butanone (50), 2-ethyl-pyrazine (54), furfural (66) and the three 506 unknowns 23, 73, 81 belong to the group of proposed roasting markers (Table 3). Some of them, e.g., furfural and pyrazines are known Mailliard reaction products [28,29]. It is 507 508 assumed that the reaction of amino compounds and reducing sugars is mediated by heat and fits well with our observation of almost linear increase with increasing roasting time. In 509 510 contrast, a second group of temperature-sensitive compounds has been identified (13, 22, 511 36, 41, 45, 52, 58, 63, 71, 74, 78, 79, 80). They are not roasting marker compounds, because 512 they lack in the typical release profile at least in one variety (Table 3). For example, 5-513 methyl-(E)-2-hepten-4-one is similarly released in the Gentile and Giffoni samples, whereas its level in Azerbaijan sample remains low (Figure S3). According to this, compound 79 tracks 514 515 approximately the same profile for the Gentile, Giffoni and Romana samples, but not for the 516 Azerbaijan hazelnuts (Figure S3). The release profiles of these components show that their 517 normalized volume values rapidly increase and remain constant or decrease with longer 518 roasting times (17-23 min). The formation of these ketones is still unclear, although 52 was 519 suggested to originate from a yet unknown precursor [23].

520

521 4.Conclusions

This study presents a systematic approach to evaluate the *fitness for purpose* of a peak-based fingerprinting method, Comprehensive Template Matching fingerprinting. The reliability of the proposed method was confirmed by employing performance parameters such as specificity, precision, accuracy, and uncertainty [30], and following a general validation protocol based on Eurachem/Citac guidelines [19,20].

The results of the comparative 2D data analysis were improved by properly compensating the dispersion of detector response values through an appropriate tuning of the main data elaboration parameters (i.e., ¹D and ²D retention times search windows, MS match factor threshold, detection threshold, and template threshold) with two sets of training samples showing different degrees of analytical variance (i.e., industrially and manually prepared sample material). Optimization was done by non-iterative standardized procedures that could be fully automated by the GC-Image software.

The validated non-targeted, peak-based fingerprinting method has successfully been applied to elucidate the generation of volatile compounds during roasting in a set of 23 hazelnut samples, where 11 roasting markers were identified, and to study the release of key aroma compounds showing specific profiles as a function of variety/origin of hazelnut samples.

539 Further investigations are under way: a) to study the effect of data elaboration 540 settings on the result quality when, for example, high density data from high frequency MS 541 detectors or high resolution MS are studied and b) to investigate a possible correlation 542 between concentration changes of character-impact compounds with overall odor 543 impressions, to identify further unknown significant *features* and to clarify formation 544 pathways explaining observed trends.

545

546 Acknowledgments

The results reported in this chapter regarding work in the authors' laboratory were obtained within the "ITACA" and "ECOFOOD" projects of the POR-FESR "Competitività regionale e occupazione" 2007/2013, Asse 1, Misura I.1.1, "Piattaforme innovative" of the Piedmont Region (Italy).

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Figure legends

Figure 1: Work flow adopted for the *Comprehensive Template Matching* fingerprinting of hazelnut samples under study.

Figure 2: Effect of user-defined retention time window and MS match factor thresholds on consistent peak matching. 2a) Only a retention time window constraint was defined: 2 of 8 overlaid template peaks (thin circles) were correctly matched and 6 were erroneously matched; 2b) The retention time window and a MS match factor threshold (600) were defined: all analyte peaks were correctly matched; 2c) Restrictive rules have been applied (retention time window and MS match factor threshold of 774): only 3 of 8 peaks were correctly matched.

Figure 3: Histogram reporting the dispersion of signal-to-noise ratios (S/N) of all detected peaks of six replicate sample chromatograms of Standard Roasted Romana hazelnuts acquired over the three weeks of validation.

Figure 4: a) Signal-to-noise-ratio (S/N) of a low abundant peak measured on nine days in three weeks. A peak can be consistently matched across the nine 2D chromatograms, if the peak is always detected, which means that the peak's S/N is always above the user-defined peak detection limit (here S/N 10 indicated by the lower line). Such peaks are recognized using template thresholds for Sample Template development (indicated by upper line at S/N 30). Asterisks label peaks that would be detected and matched, if the peak detection limit would be set to S/N 30 (inconsistent peak matching indicating false marker compounds); b) S/N of the internal standard α -thujone was used for calculating the template threshold for each 2D chromatogram.

Figure 5: Linear correlation of 11 potential marker compounds with roasting time (0-7-12-17-23 min, 11x4 data points) from four different hazelnut varieties/origins.

Captions to Tables

Table 1: Target analytes adopted for method performance evaluation on Standard Roastedand Self-roasted Romana samples over a period of three weeks*.

Table 2: Results of the pair-wise comparison of replicate analyses by *Comprehensive Template Matching*. Replicates were acquired during the three-week long validation period at different days in the week (1-3).

Table 3: Primary output of the non-targeted analysis of differently roasted hazelnuts (7-12-17-23 minutes) using *Comprehensive Template Matching* on Gentile sample subset.

Supplementary files

Table S1: detailed RSD % data (week 1, week 2, week 3) according to table 1.

Table S2: Experimental design giving the number of replicates per sample and total number of analyses runs. Standard and Self-roasted Romana samples as well as blank and alkane runs were acquired for validation purposes.

Table S3: secondary data output from non-targeted analysis. Peak features were identified and markers were confirmed using selective quantifier ions.

Figure S1: PCA-plots of four different hazelnut samples a) Azerbaijan b) Tonda Gentile from Piedmont c) Tonda Romana from Lazio and d) Tonda Giffoni from Campania. The 2D Normalized Volume values of compounds (numbered rectangular) showing increasing concentrations during hazelnut roasting were analyzed. Some compounds show consistently high correlation with the factor "roasting degree" (component 1) despite the variety/origin.

Figure S2: Ten most temperature-sensitive roasting markers (cf. table 3 for peak numbering) of Gentile hazelnuts. Fold changes were calculated relative to 7 min normalized volume value (7 min = 1).

Figure S3: Release profiles of potential roasting markers 2,3-pentanedione and unknown 81. 5-Methyl-(E)-2-hepten-4-one and unknown compound 79 were excluded from marker screening, because releases were inconsistent across all hazelnut samples (cf. 3.2).

			absolute ret	ention times	6		normalized res	oonse	match factor		
				week 1-3		week 1-3		week 1-3		week 1-3	
ID	compound	ľs	¹ D t _r (min)	(RSD %)	² D t _r (s)	(RSD %)	peak volume	(RSD %)	MS spectra	(RSD %)	
1	2,3-pentanedione [#]	1041	8.6	1.2	1.1	2	0.4	9.1	767	5	
2	2,2-dimethyl-3-hexanone	1107	10.5	1	1.8	1.9	0.09	17.2	771	14	
3	(E)-3-penten-2-one	1115	10.8	0.9	1.3	2.4	1.29	7.6	765	17	
4	pyridine	1169	13	0.9	1.2	3.2	0.17	15.6	781	3	
5	5-methyl-(Z)-2-hepten-4-one [#]	1182	13.5	0.8	2.2	1.4	0.13	11.7	848	6	
6	3-methyl-3-penten-2-one	1185	13.7	0.8	1.6	1.6	0.14	13.9	771	12	
7	1-pentanol	1238	16	0.8	1.2	3.7	0.11	9	839	2	
8	2-methylpyrazine	1252	16.6	0.7	1.3	2.9	0.96	7.6	871	2	
9	5-methyl-(E)- 2-hepten-4-one [#]	1280	17.9	0.6	2.3	2.1	1.44	12	867	6	
10	2,5-dimethylpyrazine [#]	1311	19.3	0.7	1.6	2.8	0.55	8.1	816	13	
11	2-ethylpyrazine	1321	19.8	0.6	1.6	3	0.37	10.2	829	2	
12	2,3-dimethylpyrazine	1334	20.4	0.6	1.6	3.7	0.08	7.5	813	3	
13	1-hydroxy-2-butanone	1362	21.8	0.6	1.2	4	0.09	8.2	861	5	
14	2-ethyl-5-methyl-pyrazine	1373	22.3	0.6	1.9	2.9	0.15	11.9	838	2	
15	nonanal [#]	1383	22.8	0.5	2.9	1.9	0.28	47.6	779	11	
16	2,3,5-trimethylpyrazine [#]	1393	23.3	0.5	1.8	3.2	0.15	7.6	680	3	
17	3-ethyl-2,5-dimethyl-pyrazine [#]	1434	25.3	0.5	2.1	2.9	0.11	11.1	804	3	
18	furan-2-carbaldehyde	1451	26.1	0.5	1.2	5	1.6	12.5	901	6	
19	benzaldehyde [#]	1509	28.9	0.4	1.5	3.9	0.18	17.7	680	10	
20	2-phenylacetaldehyde [#]	1628	34.4	0.5	1.6	4.1	0.14	17.4	730	3	
21	2-furanmethanol	1649	35.4	0.4	1.1	5.4	0.78	13.6	766	10	
22	phenylmethanol	1859	44.5	0.4	1.4	7.4	0.03	12.3	722	5	
23	α-thujone [#]	1409	24	0.5	3	2.7	1	0	880	2	
24	β-thujone [#]	1428	24.9	0.5	2.9	2.2	0.07	7.5	848	8	
	α-thujone raw peak volume							23.5			
	β-thujone raw peak volume							29.7			
	average of standard roasted Romana			0.6		3.2		12.4	801	6	
	average of self roasted Romana			1		2.8		46.2	774	9	

Table 1: Target analytes[#] adopted for method performance evaluation on Standard Roasted and Self Roasted Romana samples over a period of three weeks^{*}.

[#]: Analytes were identified with authentic standards (linear retention index (I_{S}^{T}), ²D absolute retention time, EI mass spectrum); all other analytes were tentatively identified (linear retention indices and mass spectra identical to data reported in literature).

* Detailed RSD % data on each single week (w1, w2 and w3) are shown in Supplementary Information Table 1.

Table 2: Results of the pair-wise comparison of replicate analyses by *Comprehensive Template Matching*. Replicates were acquired during the three-week long validation period at different days in the week (1-3).

Validation Week	1	1	1	2	2	2	3	3	3	
Validation Day	1	2	3	1	2	3	1	2	3	RSD %
Replicate analysis of standard roasted Romana sam	ples									
Number of peaks above peak detection limit ^a	487	395	470	431	321	325	357	342	354	
Template threshold [S/N] ^b	54	30	51	31	33	25	30	33	30	29
Number of template peaks ^c	172	168	162	205	150	159	156	155	165	10
Percentage of matched peaks [%] ^d	95	91	95	94	92	93	95	94	93	2
Error [%] ^e	5	9	5	6	8	7	5	6	7	
Replicate analysis of self roasted Romana samples										
Number of peaks above peak detection limit ^a	478	640	436	579	552	624	605	618	501	
Template threshold [S/N] ^b	47	144	30	157	85	93	132	90	98	43
Number of template peaks ^c	68	50	67	64	72	72	55	66	48	15
Percentage of matched peaks [%] ^d	92	92	86	95	89	89	90	91	85	3
Error [%] ^e	8	8	14	5	11	11	11	9	15	

a: Peak detection limit was fixed at S/N > 10.

b: Template threshold values were calculated relative to S/N of the ISTDs.

c: Number of template peaks, i.e. peaks exceeding the template threshold were loaded into the template after blank sample removal.

d: Percentage of matched peaks is the arithmetic mean of the pair-wise comparison of arbitrarily chosen templates with the other eight 2D chromatograms of the set (n=3). e: Error % represents fingerprinting accuracy. **Table 3**: Primary output of the non-targeted analysis of differently roasted hazelnuts (7-12-17-23 minutes) using *Comprehensive Template Matching* on Gentile sample subset.

	1D t _r	2D t _r		h				1D t _r	2D t _r		b				1D t _r	2D t _r			h		
no.ª	(min)	(s)	fold c	hange [®]			no."	(min)	(s)	fold c	fold change [®]			fold change" no." (min) (s) fold			fold	iold change [®]			
			7	12	17	23				7	12	17	23				7	12	17	23	
1	4.2	0.97	1.0	1.2	0.5	0.4	30	11.4	1.47	1.0	1.2	1.4	1.7	59	22.2	1.47	1.0	0.7	0.8	1.2	
2	4.7	1.47	1.0	1.3	0.6	0.6	31	11.5	1.94	1.0	1.0	0.6	1.2	60	22.6	1.73	1.0	0.7	0.6	0.8	
3	4.8	0.67	1.0	0.4	0.9	1.1	32	11.8	2.44	1.0	1.2	1.1	1.2	61	23.7	1.73	1.0	0.8	0.6	0.7	
4	5.1	1.81	1.0	1.4	0.6	0.6	33	12.4	1.56	1.0	1.8	1.1	1.4	62	25.3	1.81	1.0	2.6	2.0	2.0	
5	5.2	0.76	1.0	1.2	2.6	2.7	34	12.5	1.81	1.0	0.9	1.4	1.2	63	25.6	1.81	1.0	2.4	2.0	3.5	
6	5.2	1.05	1.0	0.7	1.4	2.6	*35	13.1	1.18	1.0	2.3	3.2	5.5	64	25.7	0.88	1.0	2.9	2.3	2.8	
7	5.4	1.64	1.0	1.4	0.7	0.6	36	13.6	2.15	1.0	11.4	4.5	8.9	65	25.8	1.60	1.0	1.0	2.3	1.7	
*8	5.8	0.97	1.0	5.2	4.3	6.5	37	13.8	1.60	1.0	2.3	1.4	2.3	*66	26.0	1.22	1.0	6.5	8.4	11.6	
9	6.3	1.01	1.0	2.1	2.8	2.0	38	13.8	2.23	1.0	1.9	2.7	2.6	67	26.8	1.81	1.0	1.5	1.3	2.6	
10	6.6	1.56	1.0	1.7	1.2	1.6	39	14.0	1.85	1.0	1.5	2.0	2.2	68	27.1	1.98	1.0	1.3	1.4	1.2	
11	6.8	1.14	1.0	0.5	0.4	0.5	40	14.1	1.14	1.0	1.2	0.5	1.2	69	27.4	1.94	1.0	0.3	0.7	0.9	
12	7.4	1.47	1.0	1.3	1.6	2.4	41	14.4	2.15	1.0	6.8	2.6	4.3	70	27.6	1.73	1.0	1.4	0.9	1.1	
13	7.6	1.05	1.0	1.5	3.7	3.8	42	14.5	1.64	1.0	0.6	0.2	0.2	71	27.8	1.98	1.0	2.8	3.4	2.6	
14	7.6	1.39	1.0	1.7	1.5	2.6	43	15.8	1.94	1.0	1.8	1.7	2.3	72	29.2	1.64	1.0	1.0	1.0	0.6	
15	7.7	0.93	1.0	0.6	0.5	0.3	44	15.9	1.09	1.0	1.8	1.4	1.4	*73	29.5	1.89	1.0	5.6	3.9	3.9	
16	7.9	1.39	1.0	1.3	1.8	1.1	45	16.0	1.22	1.0	1.7	1.8	3.0	74	32.1	1.98	1.0	1.4	1.4	3.2	
17	8.2	0.63	1.0	1.8	0.8	0.8	46	16.2	2.44	1.0	1.4	0.7	1.8	75	32.8	1.68	1.0	1.8	2.7	2.9	
18	8.6	1.43	1.0	0.3	0.2	0.3	47	16.3	1.85	1.0	2.0	1.7	2.9	76	33.4	2.06	1.0	1.8	0.8	2.1	
*19	8.7	1.05	1.0	2.3	3.3	3.5	48	17.0	1.98	1.0	1.1	1.5	1.4	*77	34.2	1.64	1.0	22.8	27.1	38.5	
20	9.0	1.85	1.0	0.7	0.7	0.9	49	17.4	1.52	1.0	0.9	0.5	2.0	78	34.9	1.39	1.0	3.5	2.9	2.5	
21	9.0	1.35	1.0	0.9	0.9	0.9	*50	17.6	1.14	1.0	2.7	2.9	3.9	79	37.1	1.52	1.0	4.9	1.4	4.0	
22	9.8	1.26	1.0	2.0	2.0	3.8	51	17.6	1.64	1.0	0.8	0.8	1.1	80	37.2	1.22	1.0	1.4	2.1	3.9	
*23	10.4	1.64	1.0	2.1	3.2	3.9	52	17.9	2.31	1.0	4.0	2.9	3.9	*81	38.0	1.26	1.0	2.0	0.9	6.2	
24	10.6	1.73	1.0	1.2	1.7	1.6	53	18.4	2.02	1.0	2.2	2.4	2.2	82	44.6	1.09	1.0	0.6	0.6	0.8	
25	10.7	1.05	1.0	0.3	0.3	0.5	*54	19.8	1.60	1.0	6.8	16.2	26.7	83	45.8	1.56	1.0	2.3	1.7	2.5	
26	10.9	2.02	1.0	0.8	0.5	0.8	55	20.2	1.26	1.0	0.7	1.0	1.3								
27	11.0	1.26	1.0	1.5	1.5	1.6	56	20.9	1.43	1.0	1.4	2.1	1.5								
28	11.0	1.64	1.0	1.9	1.7	1.7	57	21.0	1.26	1.0	0.8	0.9	1.3								
*29	11.3	1.18	1.0	1.5	3.5	10.7	58	21.2	1.77	1.0	2.2	1.7	3.5								

Table 3 continued:

a: Features are reported together with the template peak numbering. A feature characterizes an analyte peak with ¹D and ²D retention times and EI mass spectra.

b: Relative increase of Normalized Peak Volumes of 7 minutes roasted sample versus 12, 17 and 23 minutes roasted samples.

* An asterisk indicates analytes which were identified as roasting markers.

Figure 1

GC Image[™] Background substraction Blob detection above certain S/N ratio

Consensus template development [3,6], every template peak gets an unique number

-> remove template peaks of blank samples -> remove multiple entries



GC Project™

Consensus template is consecutivley matched with all chromatograms, template peak number is assigned to positive matched analyte

-> apply correct 1st, 2nd rentention time window for peak pattern matching -> apply correct MS match factor threshold

-> apply correct template threshold

Peak tables are automatically aligned according to the unique numbers, response data of same analytes are sorted in one row

Figure 2







sule 5











roasting time [min]

Supplementary Figure 3

