- 1 Quality classification of Spanish olive oils by untargeted gas chromatography coupled
- 2 to hybrid quadrupole-time of flight mass spectrometry with atmospheric pressure
- 3 chemical ionization and metabolomics-based statistical approach
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# 8 ABSTRACT

9 The novel atmospheric pressure chemical ionization (APCI) source has been used in combination with gas chromatography (GC) coupled to hybrid quadrupole time-of-flight 10 11 (QTOF) mass spectrometry (MS) for determination of volatile components of olive oil, enhancing its potential for classification of olive oil samples according to their quality 12 using a metabolomics-based approach. The full-spectrum acquisition has allowed the 13 14 detection of volatile organic compounds (VOCs) in olive oil samples, including extra virgin, virgin and lampante qualities. A dynamic headspace extraction with cartridge 15 16 solvent elution was applied. The metabolomics strategy consisted of three different steps: a full mass spectral alignment of GC-MS data using MzMine 2.0, a multivariate analysis 17 using Ez-Info and the creation of the statistical model with combinations of responses for 18 19 molecular fragments. The model was finally validated using blind samples, obtaining an accuracy in oil classification of 70 % taking the official established method, "PANEL 20 TEST" as reference. 21

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# 27 1. INTRODUCTION

Food quality is a matter of concern for consumers and producers. This quality is, in most cases, connected to the origin of the products, as it is the case for French wine (Arvanitoyannis, Katsota, Psarra, Soufleros, & Kallithraka, 1999) or Spanish ham (Sánchez-Peña, Luna, García-González, & Aparicio, 2005), and contributes to rise their prizes. For this reason and to avoid fraud, guarantee of the genuine quality is a critical step from the economical point of view. Thus, the characterization of this kind of products is an important aim where analytical chemists can be of great support.

35 In vegetable products, qualitative and quantitative analysis of volatile organic compounds (VOCs) has been an important issue of scientific interest for their organoleptic 36 characterization. Thus, the determination of VOCs and characterization of volatile profiles 37 in agricultural or feeding products, among others, are interesting work fields involved in 38 39 assuring food quality (Dutta, Kashwan, Bhuyan, Hines, & Gardner, 2003; Ribes, Carrera, 40 Gallego, Roca, Berenguer & Guardino, 2007). One of the most interesting studies is related to the determination of olive oil quality, as its classification is typically performed by 41 42 testers who establish if an olive oil must be labelled as extra virgin, virgin or lampante (not recommended for consumption) (Council, 2016). This strategy is known as "PANEL 43 TEST", which classifies the oils according to two main properties: defects and goods. The 44 classification requirements are established by the International Council of Olive Oil 45 46 (Council, 2016), regarding their characteristics. These defects can be fusty, mouldy, sour and woody, and the goods (positive factors) can be fruity (specifying green attribute), bitter 47 and spicy. According to the literature (Kalua, Allen, Bedgood Jr, Bishop, Prenzler & 48 Robards, 2007; Luna, Morales, & Aparicio, 2006), the organic compounds responsible of 49

these flavours are typically esters, ketones, aldehydes, alcohols, terpenes, phenols and their 50 derivatives, in concentrations ranging from few ng  $L^{-1}$  to hundreds of mg  $L^{-1}$  and with 51 different odour thresholds. PANEL TEST methodology could be considered slightly 52 53 subjective, as the opinion of testers may vary in such a significant way that the same sample can sometimes be designated as belonging to two different qualities at a time. These 54 55 discrepancies can lead to a quality misclassification of an oil, causing considerable 56 economical loses, commercial problems and fraud. A more objective alternative could be based in the use of chromatographic techniques coupled to mass spectrometry (MS), which 57 could allow the determination of chemical composition of the volatile fraction of olive oil 58 59 samples, even at really low concentration levels.

In this sense, the most adequate analysis technique is gas chromatography coupled to mass
spectrometry (GC-MS) (Angerosa, Servili, Selvaggini, Taticchi, Esposto, Montedoro,
2004; Flath, Forrey, & Guadagni, 1973).

According to the volatile characteristics of the compounds of interest, a specific extraction 63 64 technique is also an important issue in order to perform a suitable separation from the matrix. From several studies on determination of VOCs in different matrices (Barco-65 Bonilla, Plaza-bolaños, Fernández-Moreno, Romero-González, Garrido, Martínez, 2011; 66 67 Jiménez, Aguilera, Beltrán, & Uceda, 2006; Lam & Proctor, 2003; Salemi, Lacorte, 68 Bagheri, & Barceló, 2006; Serrano, Beltrán, & Hernández, 2009), there is an evidence of 69 the extended use of trapping processes of the compounds in some kind of sorbent; either by 70 forcing them to pass through the sorbent bed (P&T) (Barco-Bonilla et al., 2011; Salemi et al., 2006) or letting them to establish an equilibrium between the vapour phase and the 71 adsorbent in a closed place (SPME) (Pouliarekou, Badeka, Tasioula-margari, Kontakos, 72

Longobardi, & Kontominas, 2011; Serrano et al., 2009). Other techniques also used for
VOCs extraction include: direct headspace (HS) injection (Hu et al., 2014), stir bar sorptive
extraction (SBSE) (Bicchi, Iori, Rubiolo, & Sandra, 2002), or liquid phase micro extraction
(LPME) (Lee, Lee, Rasmussen, & Pedersen-Bjergaard, 2008).

77 The determination of the chemical fingerprint of food samples is an interesting well-known approach for characterization of food products, for example in oil (Reboredo-Rodríguez, et 78 al 2016, Reboredo-Rodríguez, González-Barreiro, Cancho-Grande, & Simal-Gándara, 79 2012) or tomato samples (Vaz-Freire, Da Silva, & Freitas, 2009). Chemical fingerprint, 80 81 normally obtained using targeted analyses, can be further used to classify unknown 82 samples, but due to the fact that a limited number of compounds must be selected a priori, 83 some information regarding the samples is lost. In this context, metabolomics, which is defined as "the unbiased, global screening approach to classify samples based on metabolite 84 patterns or fingerprints that change in response to disease, environmental or genetic 85 perturbations with the ultimate goal to identify discriminating metabolites" (Cevallos-86 87 Cevallos, Reyes-De-Corcuera, Etxeberria, Danyluk, & Rodrick, 2009), can be an 88 interesting approach to solve complex classification problems. One of the main drawbacks of metabolomics occurs when the relations or discrepancies between samples are just 89 determined by a compound or group of compounds which are present at very low levels in 90 91 complex matrices and can be easily dismissed. High resolution (HR) MS systems, such as 92 time of flight (TOF)(Kind, Tolstikov, Fiehn, & Weiss, 2007) or magnetic sector (Kieken et al., 2009), enhance the detection of molecules in complex matrices at very low levels 93 94 (Salihovic, Nilsson, Hagberg, & Lindström, 2013) due to their accurate mass measurements and extraordinary sensitivity, providing good results for the determination of non-target 95

compounds. Furthermore, the use of atmospheric pressure chemical ionization (APCI), a 96 very promising ionization source in GC, which is softer than the common electron 97 ionization (EI), allows to obtain high intensity peaks for the molecular ion  $(M^{+})$  and/or the 98 99 protonated molecule  $([M+H]^+)$ , depending on the nature of the compounds (Portolés, Sancho, Hernández, Newton, & Hancock, 2010). This information is really convenient as in 100 101 some cases the molecular peak is absent in EI due to its high fragmentation and in terms of 102 sensitivity, the peak intensity can be reduced. GC coupled to hybrid quadrupole time of flight (QTOF) MS equipped with APCI allows the acquisition of the accurate-mass full 103 scan spectra at low and at high collision energy (MS<sup>E</sup> mode) being a very useful tool for 104 105 elucidation purposes.

Data processing, together with data acquisition, are the mainstays of metabolomics. The direct observation of sample chromatograms does not give significant information about the difference between samples quality, and thus specialized software is required to obtain chromatographic peaks and masses from raw data. In literature, metabolomics studies use different software to get the information needed from the chromatograms, such as XCMS package of R (Díaz, Pozo, Sancho, & Hernández, 2014), MetAlign (Tikunov et al., 2005), or MzMine 2.0 (Kind et al., 2007).

The aim of this work has been the development of a GC-(APCI)QTOF MS methodology to obtain chemical profile/fingerprint of olive oil volatile compounds in order to establish differences between virgin olive oil qualities using P&T extraction and through the use of metabolomics techniques in order to give a more objective decision in their classification compared with that provided by the "PANEL TEST". internacional)
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# 119 2. MATERIALS AND METHODS

### 120 2.1. Chemicals and reagents

121 Internal standard triphenyl phosphate (TPP) ≥99% was purchased from Sigma Aldrich
122 (Germany). Diethyl ether (residue analysis quality GC) and hexane (trace analysis quality
123 (AT) GC) were provided by Scharlau (Barcelona, Spain).

Supelclean ENVI-Carb® SPE tubes 500 mg, volume 6 mL, 120-400 mesh, surface area 100
 m<sup>2</sup> g<sup>-1</sup>, used as traps were purchased from Supelco (Barcelona, Spain).

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# 127 **2.2. Olive oil samples**

A total of 425 olive oil samples were provided by the "Interprofesional del Aceite de Oliva Español" Organization (INTERPRO, Spain), the "Agencia para el Aceite de oliva del Ministerio de Agricultura, Alimentación y Medio Ambiente" and the oficial control services from the "Consejería de Agricultura, Pesca y Desarrollo Rural de la Junta de Andalucía".

Oil samples were taken from different regions of Spain and included 300 quality characterized samples (120 extra virgin, 120 virgin and 60 lampante) and 125 blind samples (the quality was unknown during analysis). Samples were stored in freezer at -22 °C until their use. Samples were characterized by means of pH measurements and physicochemical and organoleptic properties by the official participating laboratories (Laboratorio Arbitral Agroalimentario del MAGRAMA, Laboratorios Agroalimentarios de Córdoba y Atarfe de la Junta de Andalucía) and their corresponding certified "PANEL TESTs".

### 141 **2.3. Sample treatment**

Olive oil samples were allowed to defrost at room temperature before analysis. Then, they
were aliquoted in 4 different 10 mL vials. One aliquot was used to perform the extraction
and the remaining ones were stored at 4 °C.

5 g of olive oil were weighted directly into a 150 mL flask before inserting a magnetic 145 146 stirrer. The flask was rapidly closed with a glass tap with a nitrogen entrance and the exit 147 connected to the sorbent trap (Envi-Carb cartridge). The cartridge was conditioned with 2 x 148 5 mL of a mixture of hexane: diethyl ether (50/50; v/v) and vacuum dried for 10 min. Sample extraction was then carried out for 60 min at 40 °C with a nitrogen flow of 1 L 149 min<sup>-1</sup> and with stirring at 300 rpm. After extraction, Envi-carb® cartridges were eluted by 150 gravity with 5 mL of the hexane:diethyl ether mixture (50/50; v/v), into a glass tube 151 previously weighted. 50  $\mu$ L of TPP solution at 5 mg L<sup>-1</sup> (in hexane) were added as internal 152 standard. The entire extract was then concentrated until total removal of diethyl ether, *i.e.* 153 154 until approximately a final volume of 0.5 mL, under vacuum conditions with a MiVac Duo 155 Concentrator (Genevac, Italy). Finally, in order to adjust the final volume of the extract (0.5 mL of hexane), several drops of hexane were added until adjusted mass (0.3274 g). An 156 aliquot of 100 µL of this extract was transferred to a 20 mL vial in order to generate a pull 157 158 of extracts to prepare the quality control (QC) sample. The remaining extract was divided 159 into two different vials with 200 µL inserts, sealed and stored in freezer at -20 °C until their 160 analysis by GC-QTOF MS. In each extraction batch, 2 extra virgin, 2 virgin and 2 lampante oil samples were processed simultaneously. 161

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### 164 2.4. GC-(APCI)QTOF MS

The chromatographic analysis were performed using an Agilent 7890A gas chromatograph, 165 166 equipped with an Agilent 7693 autosampler, coupled to a quadrupole-TOF mass spectrometer, Xevo G2 QTOF MS (Waters Corporation, Manchester, UK), with APCI 167 168 source. The GC separation was performed using a fused silica HP-5MS capillary column with a length of 30 m x 0.25 mm ID and a film thickness of 0.25 µm (J&W Scientific, 169 Folson, CA, USA). The oven program was set as follows: 40 °C (3 min); 5 °C/min to 160 170 °C (1 min); 50 °C/min to 300 °C (2.2 min) (total runtime 33 min). Injections of 1 µL of 171 172 sample extracts were performed using pulsed splitless mode (50 psi) at a temperature of 173 270 °C with a pulse time of 0.90 min. Helium (99.999 %; Praxair, Valencia, Spain) was used as the carrier gas at a constant flow rate of 3 mL min<sup>-1</sup>. 174

The interface temperature was set to 310 °C using N<sub>2</sub> as auxiliary gas at a flow rate of 150 175 L h<sup>-1</sup>, the make-up gas flow rate was set at 300 mL min<sup>-1</sup> and the cone gas flow rate at 16 L 176  $h^{-1}$ . The APCI corona pin was operated at 1.7  $\mu$ A with a cone voltage of 20 V. The 177 178 ionization process occurred within an enclosed ion volume, which enabled control over the protonation/charge transfer processes. Water placed in an uncapped vial, which was located 179 180 within a special designed holder placed in the source door, was used as modifier. Xevo QTOF MS was operated with a scan time of 0.4 s, acquiring the mass range m/z 50–650. 181 TOF MS resolution was approximately 18,000 (FWHM) at m/z 614. For MS<sup>E</sup> 182 183 measurements, two alternating acquisition functions were used applying different collision energies; 4 eV for low energy function (LE), and a collision energy ramp (10-40 eV) for 184 high energy function (HE). Heptacose (PFTBA) was used for the daily mass calibration. 185 Mass accuracy calibration was continuously performed using a background ion coming 186

187 from the GC-column bleed as lock mass (protonated molecule of octamethylcyclotetrasiloxane, m/z 297.0830).

In order to avoid bias in the methodology, samples were analyzed in batches of 60 (20 pertaining to each class, extra virgin, virgin and lampante) randomly distributed. In addition, in each analysis batch, 10 QCs were injected along the sequence (one QC every 10 samples and 2 QCs at the beginning and at the end of the sequence).

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# 194 2.5. Data processing

195	GC-QTOF MS data were converted from ".raw" to ".cdf" format using Databridge, a
196	software provided by MassLynx v4.1 (Waters Corporation). Data mining was carried out
197	using MzMine 2.0 software with the conditions indicated in Figure 1. The first part of the
198	procedure (including cropping, mass detection, building chromatograms and baseline cut-
199	off) led to a data matrix of areas corresponding to the markers detected in the form of
200	combination of m/z value and retention time $(m(m/z \text{ in } Da))t(\text{time in sec}))$ . Data mining was
201	performed by MzMine 2.0 software following the procedure described in Figure 1.
202	Data were saved in a ".csv" file, exporting the row identification, the retention time, the
202 203	Data were saved in a ".csv" file, exporting the row identification, the retention time, the $m/z$ , the number of detected peaks and the peak area. This matrix was then filtered to
203	m/z, the number of detected peaks and the peak area. This matrix was then filtered to
203 204	m/z, the number of detected peaks and the peak area. This matrix was then filtered to remove too low intensity peaks (could be considered noise) and unpaired markers (which
203 204 205	m/z, the number of detected peaks and the peak area. This matrix was then filtered to remove too low intensity peaks (could be considered noise) and unpaired markers (which were only present in a low proportion -less than 20%- of samples). The last step consisted

209	normalization step which led to best results. Finally a last filter was applied to eliminate
210	duplicated signals that corresponds to the same marker.
211	Data matrix obtained after applying MzMine 2.0 was exported to a text file (.csv). Using
212	Microsoft Excel absolute areas were converted to relative areas using the area of one of the
213	TPP m/z ions (m/z 327), which was quantified in each sample using TargetLynx®
214	software. The relative areas were corrected with the nearest QC and then pareto scaled to
215	minimize the impact of the peaks with higher standard deviation. In order to carry out the
216	statistical analysis the EzInfo software (U-Metrics) was applied as it is easy to use and,
217	from our previous experience, it allows to obtain well founded results (Díaz et al., 2014).
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220	

# 221 3. RESULTS AND DISCUSSION

## 222 **3.1. GC-(APCI)QTOF MS**

Chromatographic analysis of sample extracts was carried out by GC-(APCI)QTOF MS 223 working in MS<sup>E</sup> mode in order to get both maximum sensitivity and full information. From 224 225 previous experiences in our laboratory (Cervera, Portolés, López, Beltrán, & Hernández, 2014) it was expected that the soft ionization provided by the APCI source combined with 226 low energy acquisition function, would allow to see the molecular ion or the protonated 227 228 molecule (M<sup>+-</sup> or [M+H]<sup>+</sup>), as only low fragmentation occurs. Additionally, the information acquired at high energy function (collision energy ramp of 10-40 eV) allowed obtaining the 229 230 fragmentation pattern, which resulted very useful for elucidation purposes.

As an example, Figure <u>S12</u> shows the accurate-mass full spectrum obtained for a 231 sesquiterpene, the  $\alpha$ -farnesene, acquired by GC-APCI-QTOF MS (at low and high energy) 232 233 comparing them with its corresponding EI (70 eV) spectrum from a commercial library 234 (NIST). The information about the protonated molecule in the low energy function together with the fragmentation obtained in the high energy one was of great help in the elucidation 235 236 process, which followed the steps detailed in a previous work (Portolés, Pitarch, López, 237 Hernández, & Niessen, 2011). All instrumental parameters were selected according to previous screening research works (Cervera et al., 2014; Portolés, Mol, Sancho, López, & 238 Hernández, 2014), which have been found to be reproducible and robust, basic 239 requirements of metabolomics. An example of the instrument performance is shown in Fig 240 241 S2, a total ion chromatogram obtained for the injection of a QC using the detailed 242 parameters.

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## 244 **3.2. Extraction procedure optimization**

245 Taking into account the properties of the potential analytes, especially their volatility and 246 considering the advantages of obtaining a final liquid extract (that can be injected several times, even in different chromatographic instruments) purge and trap (P&T) with solvent 247 248 trap elution was selected as extraction technique. This approach requires low economical investment and provides high reproducibility with the advantage of keeping a second 249 extract vial for further analysis. In previous works a similar P&T method with subsequent 250 251 GC-MS analysis was developed in our research group for the determination of volatiles in tomato and melon samples(Beltran, Serrano, López, Peruga, Valcárcel, & Roselló, 2006; 252 253 Fredes, Sales, Barreda, Valcárcel, Roselló, Beltrán, 2016), and it has been used as a starting point for method development, although some changes were considered to improve itsperformance in application to olive oil samples.

256 Efficiency of two different sorbent materials was tested in the P&T step: Tenax TA® SPE 257 and non-porous carbon Envi-Carb cartridges. For this purpose, aliquots of an extra virgin 258 olive oil sample were extracted using Tenax (n=3) and Envi-Carb (n=3) with the same extraction conditions. The comparison of peak areas of some selected analytes in these 259 chromatograms showed that both cartridges provided almost the same performance, as seen 260 previously in melon samples (Fredes et al, 2016). Although Tenax can be a bit more 261 262 efficient for some compounds (20% more area) both sorbents allow to extract quantitively 263 the majority of the volatiles from the sample. - Moreover, compared to Tenax, Envi-Carb cartridges are commercially available (while Tenax cartridges are produced on demand, 264 with longer supply times) and have an analysis cost of almost 3 times lower. Finally, Envi-265 Carb was selected to proceed with further development. 266

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#### 268 **3.3. Metabolomics optimization**

After optimization of extraction and chromatographic procedures, 425 oil samples (including 300 samples for model generation and 125 samples for model validation) were analyzed in several batches, both for extraction (6 samples each extraction batch) and chromatographic analysis (including samples, 10 QC and 10 blank samples in each sequence). Firstly considering a completely unbiased analysis, a Principal Component Analysis (PCA) was applied over the training data matrix corresponding to 300 olive oil samples. Figure 2a illustrates the PCA obtained.

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277	Data mining was carried out using MzMine 2.0 software with the conditions indicated in
278	Figure 1. The first part of the procedure (including cropping, mass detection, building
279	ehromatograms and baseline cut-off) led to a data matrix of areas corresponding to the
280	markers detected in the form of combination of m/z value and retention time (m(m/z in
281	Da))t(time in sec)). This matrix was then filtered to remove too low intensity peaks (could
282	be considered noise) and unpaired markers (which were only present in a low proportion -
283	less than 20% of samples). The last step consisted in a "forced integration" of the
284	considered markers in all samples, in order to eliminate the presence of zeroes in the final
285	data matrix. The forced integrated area was established as the 1 % of the background
286	absolute signal at the considered RT, as this was the normalization step which led to best
287	results. Finally a last filter was applied to eliminate duplicated signals that corresponds to
288	the same marker.
289	Data matrix obtained after applying MzMine 2.0 was exported to a text file (.csv). Using
290	Microsoft Excel absolute areas were converted to relative areas using the area of one of the
291	TPP m/z ions (m/z 327), which was quantified in each sample using TargetLynx® software.
292	The relative areas were corrected with the nearest QC and then pareto scaled to minimize
293	the impact of the peaks with higher standard deviation. In order to carry out the statistical
294	analysis the EzInfo software (U Metrics) was applied as it is easy to use and, from our
295	previous experience, it allows to obtain well founded results (Díaz et al., 2014).
296	Firstly considering a completely unbiased analysis, a Principal Component Analysis (PCA)
297	was applied over the training data matrix corresponding to 300 olive oil samples. Figure 3a
298	illustrates an example of a PCA corresponding to analysis subset of 60 olive oil samples.

299	This PCA analysis seemed to be at least promising, as at first sight a rough distinction
300	between "extra" and "lampante" can be seen, although "virgin" samples appeared in the
301	middle and overlapped the two other groups. Additionally, this step allows to check the
302	correct normalization of the samples by analyzing the behavior of the QC samples, ideally
303	clustered in the center of the Score Plot. This type of analysis is employed for dimension
304	reduction; as it is not guided, the QC samples joining in the center of the plot implies the
305	correct normalization of the batch.

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The next step consisted in performing a Partial Least Squares-Discriminant Analysis (PLS-DA), which considers additional information about groups during classification process. It is a biased PCA which provides a better separation between groups based on some parameters. Thus, a PLS-DA was applied to construct a statistical model to find differences in volatile compounds composition between samples (see **Figure 3b**2b). As it can be seen, the PLS-DA analysis separated the three groups of samples slightly better, though some group overlapping was still present.

313 Finally, in order to get information relative to the relevant compounds allowing sample classification, an Orthogonal Partial Least Squares-Discriminant Analysis (OPLS-DA) was 314 applied to highlight the differences between groups facing two different groups at a time, 315 316 regarding different compound concentration in samples. For this purpose, S-Plot graphics 317 for all two-classes comparisons, Virgin vs Extra, Extra vs Lampante and Virgin vs 318 Lampante, were obtained (Figure 43) and studied looking for endpoints. Theoretically, in an S-Plot the points with a p-corr value between 0.9 and 1 are closely related to the 319 320 compounds characterizing the samples of one group, which constitute important class markers. Thus the most significant ions, *i.e.* the further from the center in the S-Plot and 321

with a p-corr close to 1, were selected as tentative markers. According to this, a total of 15 322 markers (corresponding to combination of an m/z ion and retention time) were selected and 323 324 the corresponding chromatographic peaks, processed with TargetLynx<sup>®</sup> for a better 325 precision of peak areas. Table 1 shows the data for the 15 markers, together with the assigned elemental compositions for the protonated molecule and for the main fragment 326 327 ions. In most cases they are labelled as markers as the high number of possible compounds 328 with the same fragments made unfeasible to unequivocally identify them. We have been able to completely identify some of the markers even with the confirmation through 329 330 injection of reference standards,

331 An example of the elucidation process of a marker can be found in **Figure S14**, with the experimental APCI accurate-mass spectra obtained by GC-QTOF MS for the high and low 332 collision energy functions and with the possible chemical structures proposed for the most 333 abundant fragment ions together with the experimental mass errors (in mDa). The structure 334 was assigned to ethyl-2-methylbutanoate and was then confirmed by the injection of its 335 336 standard. This compound has been already identified by some authors in olive oil samples 337 (Dierkes, Bongartz, Guth, & Hayen, 2012; Kiralan, Ozkan, Koyluoglu, Ugurlu, Bayrak, & Kiritsakis, 2012) and has been reported as related to the fruity, sweet and green apple 338 flavour. 339

Finally, considering the 15 markers selected, a series of linear combinations of these ions were studied in order to achieve the best separation possible between groups, and a model for the classification of the three groups was built. This model was based in a combined double classification, first differentiating between Extra/Virgin oils and Virgin/Lampante oils and then distinguishing among the three different classes.

As an efficiency pre-test, the samples used to build the model were classified using it, 345 comparing the model classification results with the classes assigned by the corresponding 346 347 "PANEL TEST", which was considered as reference. After the first classification, which distinguishes between only two classes, "extra" and "lampante" oils, nearly 90 % of oils 348 were correctly classified, considering that "virgin" oils can be present in both groups. Then, 349 350 a second classification was applied to both groups, allowing a final distinction between the 351 three different classes. After the application of these series of binary classifications, around 70 % of the samples were correctly classified as extra, virgin or lampante. In all cases, the 352 worst classification efficiency corresponded to virgin samples, which are always the most 353 354 difficult to assign.

In order to take the maximum advantage possible from the samples, PLS-DA and OPLS-355 DA analysis considering the defects perceived by the PANEL TEST were created. As it can 356 be seen in Fig S3a, the samples cannot be easily separated by defects with a PLS-DA, as 357 the overlapping of samples with different defects is quite important. From Fig S3b-S3e 358 359 obtained facing samples with each kind of defect with the flawless extras in several OPLS-360 DA analyses, a clearer distinction is observed, which points out the fact that, from an analytical point of view, the extra samples are easier to classify from the rest by the absence 361 of defect rather than the presence of a good attribute. This preliminary approach to the 362 study of the defects shows a promising potential for the refining of the classification 363 364 methodology, and will be further studied in depth with an appropriate design to maximize the outputs of the research, but they won't be analysed in more detail here. 365

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#### 367 3.4. Model Validation

Once the model was created with the 300 training samples, it was validated using the 125 368 blind samples to evaluate its efficiency. Results obtained for the ternary classification for 369 370 the blind samples were sent to the project coordinating centre in order to evaluate the 371 accuracy and the performance of the model. As it can be seen in **Figure 5**, extra samples 372 were always correctly classified as extra or, in some few cases, as virgin, but never as 373 lampante, which would lead them to be discarded with a high economic loss. In addition, there was a low portion of virgin oils which got classified as lampantes and a small quantity 374 of lampantes which got overrated. It is important to highlight that the method is able to 375 distinguish the samples which have clear defects and poor fruity properties, as the match in 376 the first binary classification is around 90%. It is also remarkable that almost none of the 377 378 lampantes (a residual 5%) are overclassified as extras. This can be very useful in discarding 379 clear lampante samples and filtering the amount of samples which have to be tested by the test panel, being an important complement to it. The distinction between lampantes and 380 virgins, as they usually share defects, has shown to be the most difficult, with a 381 considerable number of lampantes classified as virgin. This is also agreement with the 382 preliminary results of the PLSDA and OPLS-DAs obtained considering the defects found 383 in the samples (Fig S3), and the efforts of future works will focus on the refining of the 384 385 methodology from those two perspectives.

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## 387 4. CONCLUSIONS

A simple and efficient methodology has been developed for the analysis of volatile
compounds in olive oil samples, with a prior extraction by P&T technique and subsequent
analysis by GC-(APCI)QTOF MS.

The use of GC-(APCI)QTOF MS has allowed to work in MS<sup>E</sup> mode, which provides 391 information at low and high collision energies which has been proved to be an important 392 393 added benefit in order to elucidate the markers. The developed method has demonstrated to be sensitive enough and capable of obtaining a considerable number of resolved 394 chromatographic peaks for volatile compounds. The use of APCI source with water as 395 396 modifier has favoured the formation of the protonated molecules  $([M+H]^+)$  which, added to the soft in source fragmentation, enhanced the sensitivity for most of the compounds when 397 compared to the traditional EI source. 398

The methodology was evaluated in terms of accuracy, obtaining an overall accuracy near to 70 % on 125 blind samples, discarding only lampante samples and a low percentage of virgin samples, which makes it perfect for being a complementary analysis together with the "PANEL TEST" method.

The metabolomics-based approach allowed the classification of the samples with a considerable performance and a minimal data treatment after the method was built. Among the different software used, the combination of MzMine 2.0 and Ez-Info dropped the best results. MzMine 2.0 allowed the inclusion of a large number of samples in order to build the model, and the addition of new samples to expand the population taken to create it. On the same way, Ez-Info provided reliable results with a really simple user interface and was preferred to other software due to the previous experience working with it.

In order to improve the classification method, a higher number of samples should be
required to include every possible combination of defects and goods. Considering the huge
variety of possible defects and goods than can give a specific classification to the samples,

the overall accuracy of 70 % achieved enable the developed methodology to be animportant complement to the official PANEL TEST.

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## 427 FIGURE CAPTIONS

428 **Figure 1**: Workflow followed for chromatographic data treatment on MzMine 2.0.

429	Figure 2: PCA (a) and PLS-DA (b) figures corresponding to the samples analysed to create
430	the model.

- 431 Comparison between the MS spectrum of  $\alpha$ -farnesene obtained by (A) electron ionization 432 (NIST library), and by APCI-QTOF (MS<sup>E</sup> mode) (B) high energy function and (C) low
- 433 energy function. Figure 3: SPLOTs Extra vs Virgin, Virgin vs Lampante and Extra vs

434	Lampante. Tentative markers are enlightened in red squares and designated by their	
435	corresponding M(m/z)T(Rt). Example of a PCA (a) and PLS DA (b) figures corresponding	
436	to one batch analysis of 60 samples.	
437	Figure 4: GC-(APCI) QTOF MS narrow-window extracted ion chromatogram (mass	
438	window 0.02 Da) showing the detection of a possible marker, tentatively identified as	
439	ethyl-2-methylbutanoate- Experimental APCI accurate-mass spectra for the high collision	
440	energy function and for the low collision energy function are also shown with the chemical	
441	structures proposed for the most abundant fragment ion together with the experimental	
442	mass errors (in mDa). SPLOTs Extra vs Virgin, Virgin vs Lampante and Extra vs	
443	Lampante. Tentative markers are enlightened in red squares and designated by their	
444	corresponding M(m/z)T(Rt).	
445	Figure 5: <u>Confusion matrix showing the Comparison comparison</u> between the objective	C
J	Figure 3. Companyon matrix showing the companyon companyon otween the objective	<u> </u>
446	(up) of getting every blind sample classified as "PANEL TEST" classification and the	
446	(up) of getting every blind sample classified as "PANEL TEST" classification and the	
446 447	(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed	
446 447 448	(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.	
446 447 448 449	<ul> <li>(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.</li> <li>Figure S1: Comparison between the MS spectrum of α-farnesene obtained by (A) electron</li> </ul>	
446 447 448 449 450	<ul> <li>(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.</li> <li>Figure S1: Comparison between the MS spectrum of α-farnesene obtained by (A) electron ionization (NIST library), and by APCI-QTOF (MS<sup>E</sup> mode) (B) high energy function and</li> </ul>	
446 447 448 449 450 451	<ul> <li>(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.</li> <li>Figure S1: Comparison between the MS spectrum of α-farnesene obtained by (A) electron ionization (NIST library), and by APCI-QTOF (MS<sup>E</sup> mode) (B) high energy function and (C) low energy function. GC-(APCI) QTOF MS narrow-window extracted ion</li> </ul>	
446 447 448 449 450 451 452	<ul> <li>(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.</li> <li>Figure S1: Comparison between the MS spectrum of α-farnesene obtained by (A) electron ionization (NIST library), and by APCI-QTOF (MS<sup>E</sup> mode) (B) high energy function and (C) low energy function. GC-(APCI) QTOF MS narrow-window extracted ion chromatogram (mass window 0.02 Da) showing the detection of a possible marker,</li> </ul>	
446 447 448 449 450 451 452 453	<ul> <li>(up) of getting every blind sample classified as "PANEL TEST" classification and the results (down) after processing these unknown samples through the entire developed procedure.</li> <li>Figure S1: Comparison between the MS spectrum of α-farnesene obtained by (A) electron ionization (NIST library), and by APCI-QTOF (MS<sup>E</sup> mode) (B) high energy function and (C) low energy function. GC-(APCI) QTOF MS narrow-window extracted ion ehromatogram (mass window 0.02 Da) showing the detection of a possible marker, tentatively identified as ethyl 2 methylbutanoate Experimental APCI accurate mass</li> </ul>	

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457	Figure S2: Full-scan total ion chromatogram (TIC) obtained by the injection of a QC.		
458			
459	Figure S3: (A) PLS-DA analysis based on the defects determined by the PANEL TEST in		
460	a subset of 150 samples used for the creation of the model. OPLS-DA analysis facing (B)		
461	"Atrojado" defect versus no defect, (C) "Agrio" defect versus no defect, (D) "Moho" defect		
462	versus no defect and (E) "Heno" defect versus no defect. Darkened area highlights the		
463	region of major predominance of the defect.		
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466	Conflict of Interest		
467	The authors declare no conflict of interest		
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