

1 **Geographical discrimination of extra-virgin olive oils from the Italian**
2 **coasts by combining stable isotope data and carotenoid content within**
3 **a multivariate analysis**

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13 **Abstract**

14 We have determined the isotopic composition and the carotenoid contents of 38 extra-virgin olive
15 oils (EVOOs) from seven regions along the Italian coasts, by means of isotope ratio mass
16 spectrometry (IRMS) and resonant Raman spectroscopy (RRS), respectively. The application of
17 linear discriminant analysis to our overall results demonstrated the combination of isotope and
18 carotenoid data is a promising method to discriminate EVOOs from production sites that are impacted
19 by similar geographical and climatic parameters. In particular, this dual approach allowed correct
20 classification of 82% EVOO samples, while separate IRMS and RRS investigations were able to
21 discriminate only samples from Sicily and Latium, respectively.

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24 Keywords: Olive oil, Raman spectroscopy, isotope ratio mass spectrometry, carotenoid content,
25 chemometrics, geographical origin.

1. Introduction

Extra-Virgin Olive Oil (EVOO) is produced by physical extraction from the fruits of the *Olea europaea* tree. The geographical origin of EVOO is one of the most relevant factors that determine its quality and, thus, its commercial value, since the EVOO composition and nutraceutical values may depend on regional differences in climate, soil, agricultural practices and local cultivars. Thus, the European Union (EU) has set regulations for the protection of claims relating to geographic indications and origin designations for EVOOs (Woodcock, Downey, & O'Donnell, 2008), and products from defined high-quality oil-producing regions may use a Protected Designation of Origin (PDO) or Protected Geographical Indication (PGI) on their labels. For these reasons, food processors, retailers, enforcement agencies, and consumers require an independent mechanism to confirm that any given EVOO claiming PDO or PGI status actually complies with all relevant specifications. The development of analytical methodologies for EVOO authentication is, thus, a challenging task, and several experimental techniques have been employed to assure olive oil traceability and authenticity. Chromatographic techniques, such as gas chromatography (Aparicio & Aparicio-Ruiz, 2000), gas chromatography mass spectrometry (Spangenberg & Ogrinc, 2001), and high performance liquid chromatography (Christopoulou, Lazaraki, Komaitis, & Kaselimis, 2004) have been employed to estimate the content of specific compounds (such as fatty acids, sterols, phenolics or hydrocarbons) as a function of olive oil geographical origin. Isotope ratios mass spectrometry (IRMS) has demonstrated that EVOOs from different production sites can be discriminated based on their stable isotopic composition (Angerosa et al., 1999; Camin et al., 2010). Nuclear magnetic resonance (Mannina, Patumi, Proietti, Bassi, & Segre, 2001), solid-phase micro-extraction gas chromatography-mass spectrometry (SPME-GC) (Young, Quinn & Trumble 2012), and spectroscopic techniques, such as near infrared (Casale, Casolino, Oliveri, & Forina, 2010), Fourier-transform infrared (Tapp, Defemez, & Kemsley, 2003) and Raman (Korifi, Le Dreau, Molinet, Artaud, & Dupuy, 2011) spectroscopies have allowed olive oil origin to be determined using fingerprinting approaches.

51 However, when a specific meteorological or geographical factor affects different geographical
52 regions similarly, EVOOs may be difficult to distinguish if analysed using a technique that is sensitive
53 to that parameter. In this case, an improved classification of food products can be achieved by
54 combining the analytical data obtained using different techniques, within a multivariate approach
55 (Casale, Casolino, Oliveri, & Forina, 2010; Chiavaro et al., 2011).

56
57 In particular, we have previously shown that the combination of $^{13}\text{C}/^{12}\text{C}$ and $^{18}\text{O}/^{16}\text{O}$ data, as
58 measured by IRMS, could provide information about the geographical origin of olive oil (Portarena,
59 Gavrichkova, Lauteri, & Brugnoli, 2014; Portarena et al., 2015). However, the water cycle has a
60 significant impact on carbon and oxygen isotope composition in plant tissues, which is mediated by
61 the evaporative demand of the geographical origin, stomatal conductance, and photosynthetic
62 capacity. Thus, stable isotope analysis may fail in discriminating plant materials coming from regions
63 where the meteorological cycle of evaporation, condensation, and precipitation is impacted by similar
64 climate parameters, as occurring along the Italian coasts (Longinelli & Selmo, 2003).

65 Resonant Raman spectroscopy (RRS) is able to excite resonantly the vibrational modes of carotenoid
66 pigment molecules, allowing a comparative quantification of carotenoid content (El-Abassy,
67 Donfack, & Materny, 2009). This may depend on climatic parameters, such as light exposition,
68 distance from the sea, temperature, and amount of precipitations (Issaoui et al., 2010). However, the
69 carotenoid content can be affected also by genetic factors, mainly related to the olive variety
70 (Giuffrida, Salvo, Salvo, La Pera, & Dugo, 2007), agronomical aspects, such as nutrient availability
71 and fruit ripening (Baccouri et al., 2008), and processing and storage conditions (Gimeno, Castellote,
72 Lamuela, De la Torre, & Lopez-Sabater, 2002). Since all these aspects may be specific of the olive
73 growing region, the carotenoid content could be a good candidate as EVOO geographical tracer.

74 In this preliminary work, we selected EVOO samples coming from PDO production areas along the
75 Italian coasts, which is a macro-region characterized by similar climatic and geographical factors,

and analysed isotopic composition and carotenoid content by means of IRMS and RRS, respectively. The geographical discrimination of our EVOO samples based on these two parameters is unsatisfactory if considered separately. However, by combining IRMS and RRS outputs within a multivariate statistical method, the analysis resulted in a prediction model that was able to classify correctly 82% of our EVOO samples.

Thus, the combination of isotope data and carotenoid content is a promising experimental approach to discriminate geographically EVOOs, even if their production sites are impacted by similar geographical and climatic parameters.

2. Materials and methods

2.1 Sampling sites

EVOO samples from 38 different sites located in seven different Italian regions (Apulia, Latium, Liguria, Molise, Sardinia, Sicily, Tuscany) were collected during the 2011 harvest. Olives were harvested and milled by local farmers, and olive oil samples were then given to us by the Consorzio Olivicolo Italiano (UNAPROL). Sampling sites were selected along the Italian coasts, in a belt of 70 km, to reduce the so-called ‘continental effect’ (Angerosa et al., 1999), and they are shown in Fig. SM1 (see Supplementary Material).

The olive growing areas were differentiated by latitude, longitude, altitude and climatic conditions. Geo-climatic information on the sampling sites are reported in Table SM2 (see Supplementary Material). Colleagues from the Institute of Atmospheric Sciences and Climate of National Research Council (CNR-ISAC) provided temperature, total rainfall and relative humidity mean values. Latitude and longitude were taken from Google Earth (www.googleearth.com). The farmers provided data of harvest and altitude, based on their local knowledge.

101 **2.2 Stable isotope analysis**

102 Analyses of $^{13}\text{C}/^{12}\text{C}$ and $^{18}\text{O}/^{16}\text{O}$ isotope content ratio were performed using an isotope ratio mass
103 spectrometer (Isoprime, GV, Cheadle, UK) connected to an elemental analyser (NA1500, Carlo Erba,
104 Milan, Italy) and to a pyrolysis system (Euro Pyr-OH, Euro Vector Instruments & Software, Milan,
105 Italy). Isotopic compositions were scale normalized with IAEA international standards, on two
106 contemporaneously analysed scale anchors, and are expressed as δ notation according to the
107 expression: $\delta = (R_s - R_{\text{std}}) / R_{\text{std}} \times 1000$, where R_s is the isotope ratio of the sample and R_{std} is the
108 isotope ratio of the international standard. In particular NBS-22 fuel oil and IAEA-CH6 sucrose were
109 used for scale normalization of measured $\delta^{13}\text{C}$ values to the VPDB scale; IAEA-601 and IAEA-602
110 were used for scale normalization of measured $\delta^{18}\text{O}$ values to the VSMOW scale.

111 The standard deviation of replicate measurements for each standard and sample was $\pm 0.1\text{‰}$ for $\delta^{13}\text{C}$,
112 and $\pm 0.3\text{‰}$ for $\delta^{18}\text{O}$. Possible changes in analytical conditions within a run were corrected using
113 NBS-22 fuel oil and IAEA-CH6 sucrose for $\delta^{13}\text{C}$ measurements and IAEA-601 (IAEA) and IAEA-
114 CH6 sucrose ($\delta^{18}\text{O} = +36.4\text{‰VSMOW}$) for $\delta^{18}\text{O}$. This latter reference was assigned since 2005
115 (Boschetti & Iacumin, 2005) and accepted in the European TRACE project (contract No. FP6-2003-
116 FOOD-2-A 006942).

117

118 **2.3 Raman spectroscopy analysis**

119 Samples of selected EVOOs were dripped on to glass microscope slides and illuminated by the 514.5
120 nm line of an Argon laser with an excitation power of less than 5 mW on the sample. The laser was
121 focused within the drop using a microscope equipped with a 10x objective. The scattered signal was
122 then recorded at 180° backscattered geometry and dispersed by a single monochromator with a 1800
123 grooves/mm diffraction grating. The spectrometer (Super Labram, Jobin Yvon) was equipped with a
124 liquid nitrogen cooled charge coupled device (CCD) detector with a resolution better than 5 cm^{-1} .
125 Energy calibration was adjusted daily, by using a Si single crystal as a reference sample.

For each EVOO sample, RR spectra in the 900-1800 cm^{-1} range were acquired from four different points on the droplet. The intensity variability among spectra from the same sample was less than 5%. Each spectrum was obtained by averaging five scans of 10 s each. The weak fluorescence background was subtracted from the spectra by an automatic procedure implemented in the LabSpec5 software (Horiba Scientific, Edison, New Jersey). The four spectra representative of each sample were averaged before the statistical analysis.

132

133 **2.4 Statistical treatment of data**

The data were statistically evaluated using Statistica v8 (StatSoft Italia srl, Padua, Italy). First, principal component analysis (PCA) was performed using as input variables the intensities of Raman bands listed in SM3 (see Supplementary Material), properly normalized at the intensity of the peak located at 1440 cm^{-1} (Zou et al., 2009). This allowed us to study the structure of the dataset, and to detect the most discriminating coordinates (principal components, PCs).

Linear discriminant analysis (LDA) classification model was then developed using as input variables the PCs obtained by PCA, together with the carbon and oxygen isotopic values, after suitable variable standardization. Canonical variates (CVs) were obtained from LDA as the vectors minimizing the Wilks' lambda parameter, which is calculated as the sum of squares of the distances between points belonging to the same category divided by the total sum of squares. Values of Wilks' lambda approaching zero are obtained with well resolved categories, whereas overlapped categories make Wilks' lambda to approach one.

To test the predictive discrimination power and the stability of the obtained classification model, this was further cross-validated using standard approaches. In detail, 5 different sets of 7 oils (one for each region, randomly selected) were removed from the dataset. Each time, the model was calculated based on the remaining 31 cases and then validated with all 38 samples.

150

151 **3. Results and discussion**

152

153 **3.1 Isotopic results**

154 Stable isotope data for all the EVOO samples are plotted in Fig.1. Each sample is represented by a
155 symbol in a bidimensional plane with $\delta^{13}\text{C}$ and $\delta^{18}\text{O}$ values as coordinates.

156 The range of variation in $\delta^{13}\text{C}$ was about 2.7‰, (from -30.2‰ to -27.5‰), while a wider range of
157 4.8‰ is found for $\delta^{18}\text{O}$ (from 21.7‰ to 26.5‰). Interestingly, the relatively high $\delta^{18}\text{O}$ and $\delta^{13}\text{C}$
158 values (up to 26.5‰ and -27.5‰, respectively) values (...) obtained for EVOOs from Sicily were
159 due to the markedly hot climate characterizing the region, and to the related plant physiological
160 response. In fact, hot and dry weather conditions may induce partial stomatal closure, with the
161 consequent decrease in intercellular and atmospheric CO_2 concentration ratio and increase in $\delta^{13}\text{C}$
162 and $\delta^{18}\text{O}$. However, the isotopic data ranges found in the present study were smaller than those
163 recently observed by our laboratory on a more heterogeneous set (Portarena, Gavrichkova, Lauteri,
164 & Brugnoli, 2014). As a result, the sample clustering is not clear, likely due to the influence of the
165 so-called “coast effect” on climatic parameters and, hence, on the isotopic composition of plant
166 tissues (Longinelli & Selmo, 2003).

167

168 **3.2 Resonant Raman spectroscopy results**

169 Representative RR spectra collected from the 38 EVOO samples are shown in Fig. 2. All the spectra
170 presented the same Raman features, which are located at 965 cm^{-1} , 1004 cm^{-1} , 1081 cm^{-1} , 1155 cm^{-1} ,
171 1270 cm^{-1} , 1301 cm^{-1} , 1440 cm^{-1} , 1522 cm^{-1} , 1655 cm^{-1} , and 1748 cm^{-1} , in agreement with previous
172 EVOO RRS investigations (El- Abassy, Donfack, & Materny, 2009). The Raman bands are listed in
173 SM3 (see Supplementary Material), together with the corresponding chemical groups, vibrational
174 modes and molecules. In particular, we have attributed the bands at 965 cm^{-1} (C=C bending out-of-
175 plane), 1081 cm^{-1} (C-C stretching), 1301 cm^{-1} (C-H twisting), 1440 cm^{-1} (C-H scissoring), and 1655
176 cm^{-1} (C=C stretching) to free fatty acids (FFAs) (Yang & Irudayaraj, 2001; El- Abassy, Donfack, &

Materny, 2009), and the band at 1748 cm^{-1} to triglycerides (C=O stretching in the ester linkage between a fatty acid and the glycerol) (Weng, Weng, Tzeng, & Chen, 2003). The relative spectral intensity of these bands was similar for all the spectra, indicating that the composition in terms of fatty acids of our EVOO samples were similar. Contrary to this, a strong variability is observed in the spectral intensity of the three main bands due to the resonant excitation of the carotenoid vibrational modes, which are located at 1004 cm^{-1} (C-CH₃ bending), 1155 cm^{-1} (C-C stretching) and 1522 cm^{-1} (C=C stretching) (Merlin, 1985; El- Abassy, Donfack, & Materny, 2009). This indicates that our samples were highly heterogeneous in term of carotenoid content and composition. A separate comment is required for the band at 1270 cm^{-1} : it is generally attributed to the C-H in-plane bending of FFAs (Yang & Irudayaraj, 2001; El- Abassy, Donfack & Materny, 2009), but possibly also to the coupling of the C-H in-plane bending and C-C stretching modes of carotenoids (Koyama et al., 1982; Merlin, 1985; Berezin & Nechaev, 2005).

189

190 **3.3 Principal component analysis of resonant Raman data**

PCA was performed using as input variables the normalized intensities of selected Raman bands, and 9 PCs are obtained. The list of the first 5 PCs and their scores, together with eigenvalues and explained variance, is shown in SM4 (see Supplementary Material). The scores for 38 EVOO samples on the two-dimensional plane defined by the first two PCs are shown in Fig. 3. The percentage of variance explained by PC1 and PC2 was 67% and 19%, respectively.

According to the variable loadings listed in SM4, the descriptors contributing most in PC1 were the spectral intensities of the Raman bands at 1004 cm^{-1} and 1270 cm^{-1} . This confirms that the carotenoid content is the most discriminating factor, and also suggests that the 1270 cm^{-1} band may be linked to carotenoids rather than to FFAs. Differentiation on PC2 axis was explained mainly explained by the intensity of the Raman bands at 1655 cm^{-1} and 1081 cm^{-1} , which are both due to FFA vibrational modes. This indicates that the FFA composition of our EVOO samples was slight different. Indeed, the three main FFAs of EVOO (oleic acid, palmitic acid and linoleic acid) have the same vibrational

203 structure (i.e., vibrational modes at about 1655 cm⁻¹, 1440 cm⁻¹, and 1081 cm⁻¹) but different relative
204 spectral intensities of the peaks (Gelder, Gussem, Vandenabeele, & Moens, 2007). For instance, the
205 intensity ratio between the band at 1655 cm⁻¹ and that at 1440 cm⁻¹ was about 0.7 in the oleic acid
206 spectrum, about 0.1 in the palmitic acid spectrum, and about 2.4 in the linoleic acid spectrum
207 Thus, PCA indicates that our EVOO samples had different total carotenoid contents (major
208 contribution, due to PC1) and different FFA compositions (minor contribution, due to PC2). Both of
209 these differences can be induced by differences in weather conditions during growth and harvesting
210 (Issaoui et al., 2010), genetic factors (Giuffrida, Salvo, Salvo, La Pera, & Dugo, 2007), agronomical
211 practices (Baccouri et al., 2008) and oil extraction system (Gimeno, Castellote, Lamuela, De la Torre,
212 & Lopez-Sabater, 2002). However, the class separation as obtained by PCA was not satisfactory: only
213 a partial separation of Latium and Tuscany samples from the rest of the data was obtained (see Fig.
214 3), which may arise from the peculiar varieties and/or agricultural and extraction practices
215 characterizing these two regions.

216

217 **3.4 LDA of combined IRMS and RRS data**

218 The results obtained by IRMS and RRS separately were combined within a LDA. CVs obtained,
219 together with Wilks' Lambda, p values and eigenvalues, are presented in Table 1. The first two CVs
220 were described mainly by the content in terms of $\delta^{18}\text{O}$ (CV1) and carotenoids (CV2, mostly dependent
221 on PC1). The bidimensional plot for CV1 and CV2 scores is shown in Fig. 4: a clearer discrimination
222 of the regional origin for our samples was obtained, with respect to individual IRMS and RRS
223 analyses. In particular: Sicily, Sardinia and Liguria EVOOs were well separated from the rest of the
224 samples; Latium, Tuscany and Molise data were clustered satisfactorily; Apulia oils are not well
225 grouped.

226 According to this description, the classification model derived by LDA (shown in Table 2) correctly
227 classified 82% of EVOO samples. In particular, correct predictions were obtained for five regions,
228 while low prediction percentages were obtained for Latium (57%) and Apulia (50%); this is probably

229 due to the morphological and climatic inhomogeneity of the sampling sites. To test the predictive
230 discrimination power and the stability of the model, it was cross-validated using five different sets of
231 EVOOs, as described in Section 2.4, and obtained the correct classification for 77 ± 5 % of the
232 unknown samples.

233

234 **4. Conclusions**

235 EVOOs from selected sites along the Italian coast are not well discriminated geographically based on
236 either their carbon and oxygen isotopic composition or carotenoid content, when analysed separately.
237 This might be due to the similarity in geo-climatic conditions for some of the locations. With the aim
238 of enhancing differences among the oils, we explored the potential for improved geographical
239 classification based on the combination of these parameters within a multivariate statistical approach.
240 The proposed model resulted in the correct classification of 82% of samples. Even if this percentage
241 is not very high (based on a small sample size), the model was stable.

242 The inclusion of other analyses could improve the accuracy of this prediction model. For example,
243 based on our results, the different free fatty acids composition of EVOOs could be included using
244 chromatographic techniques, such as gas chromatography - flame ionisation detector (GC-FID) or
245 gas chromatography–mass spectrometry (GC-MS).

246 Similarly, the dependence of isotopic and lipid composition of EVOOs on genetic, agronomic and
247 processing factors needs to be investigated in order to develop more sensitive tools for traceability.

248

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252 execution of IRMS analyses. Thanks are due to Prof. Cannistraro and to the staff of the Biophysics
253 and Nanoscience Centre of the University of Tuscia (Viterbo, Italy) for providing the Raman
254 spectroscopy facilities and support.

255 **Appendix A. Supplementary data**

256 Supplementary data associated with this article can be found, in the online version, at

257 <http://dx.doi.org/10.1016/j.foodchem.2016.07.135>.

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259

260 **Figure captions**

261

262 **Figure 1.** $\delta^{13}\text{C}$ and $\delta^{18}\text{O}$ values of 38 EVOO samples from seven different Italian regions, collected
263 from sites along the Italian coast. Samples from the same geographical region are represented by the
264 same symbol, according to the legend.

265

266 **Figure 2:** Representative RR spectra (one per each sampling site) with indicated the major
267 Raman bands. The spectra are shown after background subtraction and intensity normalization at the
268 1440 cm^{-1} peak. The spectra showing the highest and the lowest carotenoid-related band intensities
269 (which are obtained from a Latium and a Liguria oil, respectively) are drawn as black curves. The
270 other spectra are shown in grey, to demonstrate the variability of the RRS data.

271

272 **Figure 3.** Scatterplot of the scores of 38 EVOO samples on the two-dimensional plane defined by
273 the first two PCs as calculated by PCA of the normalized intensities of selected Raman bands. Each
274 EVOO sample is represented by a symbol, grouped by the region of origin, according to the legend.

275

276 **Figure 4.** Scatterplot of the scores of 38 EVOO samples on the two-dimensional plane defined by the
277 CV1 and CV2 resulting from a LDA of combined IRMS and RRS results (as obtained by PCA), after
278 variable standardization. Each EVOO sample is represented by a symbol, grouped by the region of
279 origin, according to the legend.

280

281

282 **Table 1.** Structure matrix of the model from LDA, together with statistical values of Wilks' Lambda,
 283 p value and eigenvalue calculated for each CV.

	CV1	CV 2	CV 3	CV 4	CV 5
PC1std	-0.548	0.947	-0.209	-0.267	-0.044
PC2std	0.407	0.110	0.850	0.121	0.239
PC3std	-0.360	0.258	0.276	-0.055	-0.198
PC4std	0.028	0.279	-0.005	0.548	0.151
PC5std	0.733	0.172	-0.085	-0.171	0.186
PC6std	0.229	0.501	0.354	-0.401	-0.288
PC7std	0.206	0.208	0.052	0.012	0.140
PC8std	-0.206	0.111	0.501	-0.164	0.020
PC9std	0.186	0.406	-0.340	-0.832	0.182
$\delta^{13}\text{Cstd}$	0.198	-0.196	-0.418	0.475	-0.779
$\delta^{18}\text{Ostd}$	-1.368	0.112	-0.042	0.594	0.295
Wilks'					
Lambda	0.013	0.069	0.193	0.433	0.731
p value	0.000	0.013	0.121	0.496	0.845
Eigenvalue	4.304	1.794	1.249	0.686	0.294

291 **Table 2.** Classification matrix as obtained from LDA.

292

	Percent	Apulia	Latium	Ligury	Molise	Sardinia	Sicily	Tuscany
	Correct	p=0.158	p=0.184	p=0.132	p=0.079	p=0.105	p=0.105	p=0.237
Apulia	50.00	3	2	1	0	0	0	0
Latium	57.14	1	4	0	0	1	0	1
Ligury	100.00	0	0	5	0	0	0	0
Molise	100.00	0	0	0	3	0	0	0
Sardinia	100.00	0	0	0	0	4	0	0
Sicily	100.00	0	0	0	0	0	4	0
Tuscany	88.00	1	0	0	0	0	0	8
Total	81.58	5	6	6	3	5	4	9

293

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Figure 1

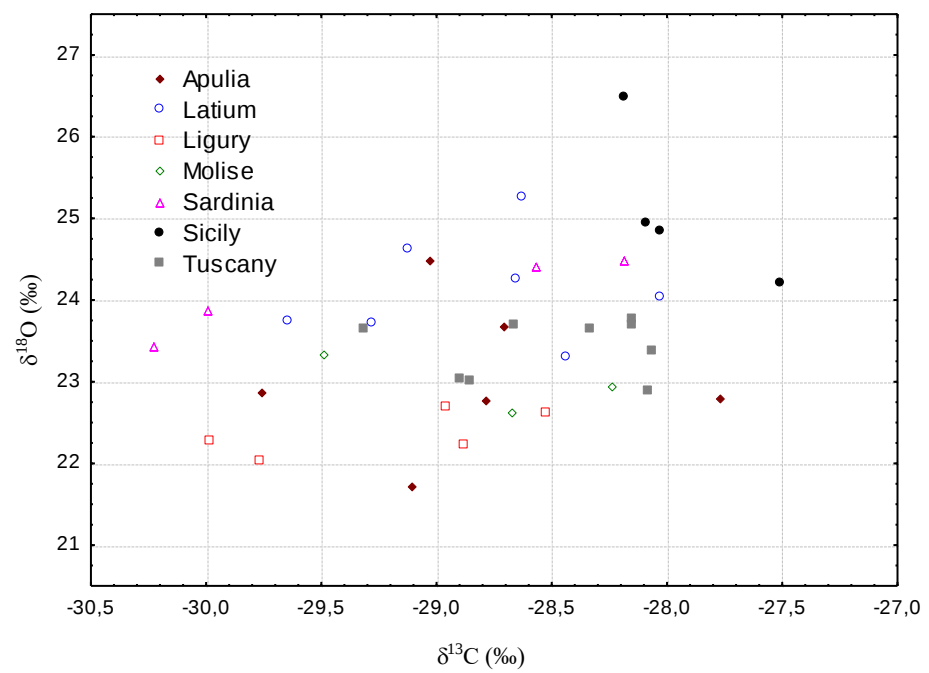


Figure 2

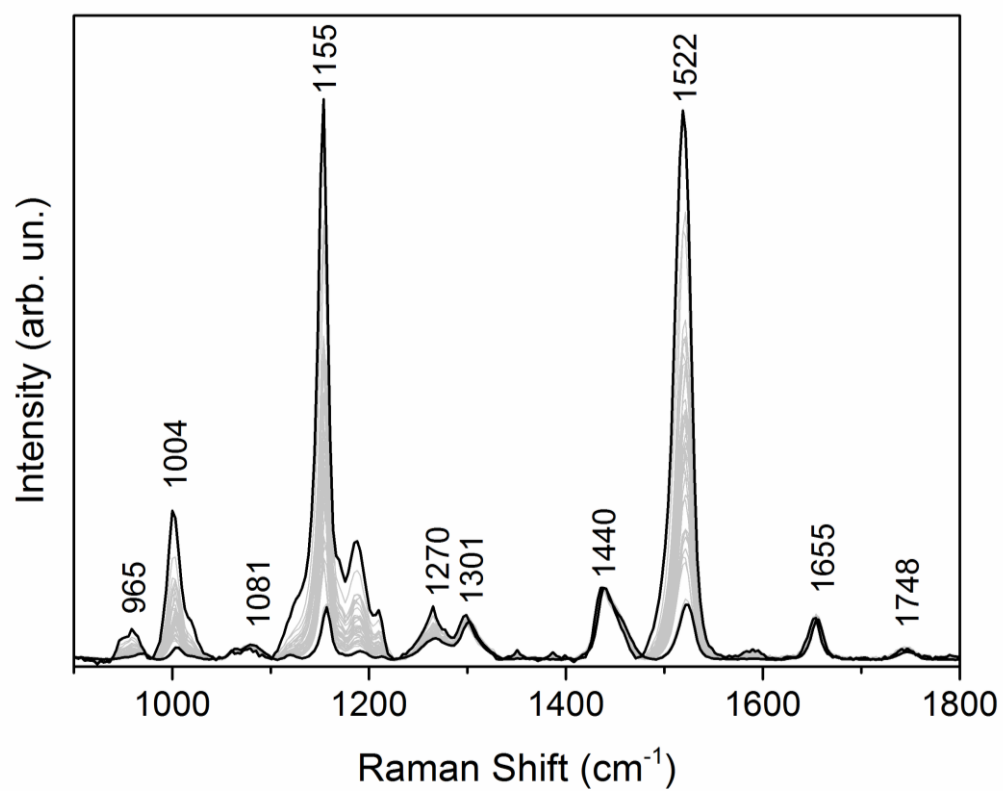


Figure 3

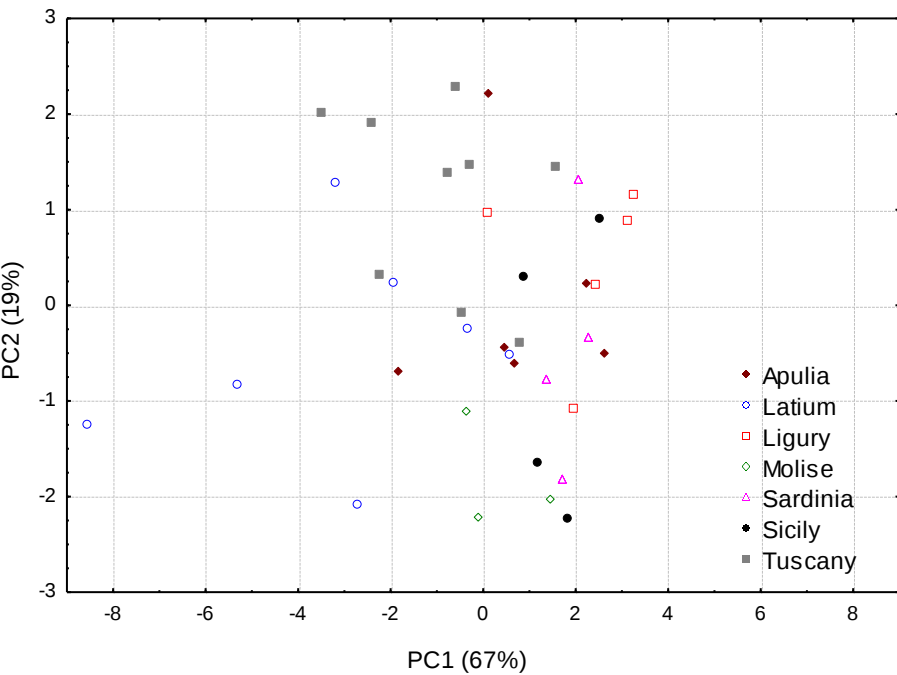


Figure 4

