

1 **Study of phenolic extractability in grape seeds by means of ATR-FTIR**
2 **and Raman spectroscopy.**

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4 Short title: Grape seed phenolic extractability using ATR-FTIR & Raman
5 spectroscopy

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7 Julio Nogales-Bueno^a; Berta Baca-Bocanegra^a, Abigail Rooney^b, José Miguel
8 Hernández-Hierro^a, Hugh J. Byrne^b and Francisco José Heredia^{a*}.

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10 ^a *Food Colour and Quality Laboratory, Department of Nutrition and Food Science,*
11 *Facultad de Farmacia, Universidad de Sevilla, 41012 Sevilla, Spain.*

12 ^b *FOCAS Research Institute, Dublin Institute of Technology, Kevin Street, Dublin 8,*
13 *Ireland*

14 * **Corresponding author:** Francisco José Heredia

15

16 **Phone:** +34 9545 56495

17 **Fax:** +34 9545 56110

18 **E-mail:** heredia@us.es

19

20 **ABSTRACT**

21 Near infrared hyperspectral imaging has been applied to grape seeds in order to select a
22 representative subset of samples according to their spectral features in the 900-1700 nm
23 range. Afterwards, selected grape seeds have been classified according to their total
24 phenol and flavanol extractabilities. In this way, samples were sorted in three different
25 groups identified as low, medium and high extractability levels.

26 In order to establish the chemical structures which can be responsible for the different
27 extractabilities, vibrational spectroscopy has been applied to the non-extracted material
28 after seed extractions. Attenuated total reflectance Fourier transform infrared (ATR-
29 FTIR) and Raman spectra of non-extracted seed material have been recorded and their
30 main spectral features have been linked to extractabilities of flavanolic and total phenolic
31 compounds.

32 The vibrational spectroscopic analysis confirms that grape seed phenolic extractability is
33 influenced by the cell wall composition (polysaccharides, lignins, pectins) and by the
34 degree of esterification of pectins.

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37 **KEYWORDS**

38 Wine, grape seeds, phenolic extractability, ATR-FTIR spectroscopy, Raman
39 spectroscopy, hyperspectral imaging.

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43 1. INTRODUCTION

44 Although grape seeds represent only 0–6% of berry weight, they effect the sensory
45 properties of wine. Depending on varieties, they contain between 20 and 55% of the
46 phenolic compounds of the berry (Ribéreau-Gayon, Dubourdieu, Doneche, Lonvaud,
47 Glories, Maujean, et al., 2006). Phenolic compounds are characterized by having at least
48 one aromatic ring with one or more hydroxyl groups attached. Flavonoids are the most
49 numerous amongst phenolics. The generic structure of the major flavonoids comprises
50 fifteen carbons (C6-C3-C6), with two aromatic rings (C6) connected by a heterocyclic
51 pyran ring of three-carbon bridge (C3). Flavonoids are secondary metabolites that have
52 important and varied roles (Crozier, Clifford, & Ashihara, 2006), including well-known
53 health benefits. They possess ideal structural chemistry for free radical-scavenging
54 activities, and they have been shown to be more effective antioxidants *in vitro* than
55 vitamins E and C, on a molar basis (Rice-Evans, Miller, & Paganga, 1997).

56 The main subclasses of flavonoids present in grapes (*Vitis vinifera* L.) are flavonols,
57 flavanols and anthocyanins, and they are transferred to the wine during the fermentation
58 process. Flavonols and anthocyanins are principally localized in the skin, whereas
59 flavanols are synthesized primarily in seeds and stems (Jackson, 2000). Flavanols
60 constitute the most abundant phenolic family in grape seeds and play a relevant role in
61 the sensory characteristic of red wines. They can increase the wine body, but also the
62 sensations of astringency and bitterness (Waterhouse, 2002). Moreover, flavanols, like
63 the main phenolic families, can increase and stabilize the wine color by means of the
64 copigmentation phenomenon (Boulton, 2001; Gordillo, Rodríguez-Pulido, González-
65 Miret, Quijada-Morín, Rivas-Gonzalo, García-Estévez, et al., 2015; Nogales-Bueno,
66 Baca-Bocanegra, Jara-Palacios, Hernández-Hierro, & Heredia, 2016).

67 The importance of controlling the amount of phenolic compounds that may be extracted
68 from grapes to wine lies in the fact that the phenolic content wine depends mainly on the
69 amount of phenolic compounds released from grape to wine. It is well known that the
70 extractabilities of phenolic compounds from the solid parts of grapes (seeds and skins)
71 depends on the grape maturity. However, in skins, phenolic extractability increases with
72 maturity, whereas, in seeds it decreases. (Ribéreau-Gayon, et al., 2006).

73 The effect of cell wall degradation on the degree of extraction of phenolic compounds
74 from grape skins has already been widely studied. Several studies have been developed
75 using this matrix in order to control and understand the relationship between cell wall
76 composition and the extraction degree of phenolic compounds. An increase in the sugar
77 content (degree of ripening) has been correlated with decreasing amounts of cell wall
78 material, galactose, cellulose and mannose, accompanied by a decrease in the degree of
79 methylation of pectic polysaccharides or pectins (Ortega-Regules, Ros-García, Bautista-
80 Ortín, López-Roca, & Gómez-Plaza, 2008). These factors could be responsible for the
81 different extractabilities of phenolic compounds, bearing in mind that differences on
82 thickness or density of the skin cell-wall could also play a role.

83 Regarding grape seeds, extractability decreases during ripening, from about 25% at the
84 first stages to about 5% in the later (Rodríguez-Pulido, Hernández-Hierro, Nogales-
85 Bueno, Gordillo, González-Miret, & Heredia, 2014). It might be inferred that this
86 phenomenon could be attributed to the presence of the middle integument, which is
87 waterproof and very hard when seeds ripen and lignify, and could prevent the extraction
88 (Cadot, Minana-Castello, & Chevalier, 2006). On the other hand, changes in the cell wall
89 polysaccharide structure could affect the solubility of the flavanols (Bautista-Ortín,
90 Jiménez-Pascual, Busse-Valverde, López-Roca, Ros-García, & Gómez-Plaza, 2013).
91 These authors found that maceration enzymes, normally used in the wine elaboration

92 process to degrade skin cell wall in order to increase color, can also favour the extraction
93 of flavanols from the seeds. These oenological products easily degrade the cell walls of
94 seeds (especially if they are unripe and show low lignification), promoting the release of
95 seed compounds into the must-wine, increasing the wine body, but also the sensations of
96 astringency and bitterness. It is therefore of interest to evaluate how cell wall structure
97 changes among seed samples of different grape varieties or with different levels of
98 phenolic compounds extractability.

99 Near infrared spectroscopy (NIRS) has been used in order to screen total or extractable
100 phenolic compounds in grapes obtaining quite good results (Ferrer-Gallego, Hernández-
101 Hierro, Rivas-Gonzalo, & Escribano-Bailón, 2011; Kemps, Leon, Best, De
102 Baerdemaeker, & De Ketelaere, 2010; Nogales-Bueno, Baca-Bocanegra, Rodríguez-
103 Pulido, Heredia, & Hernández-Hierro, 2015; Nogales-Bueno, Hernández-Hierro,
104 Rodríguez-Pulido, & Heredia, 2014; Rodríguez-Pulido, et al., 2014; Torchio, Río Segade,
105 Giacosa, Gerbi, & Rolle, 2013). However, it is not possible to easily interpret the
106 relationship between cell wall composition and phenolic compounds extractability based
107 on near infrared (NIR) region features. The near infrared region contains absorption bands
108 corresponding to overtones and combinations of fundamental C–H, O–H and N–H
109 vibrations. Additionally, the presence of Fermi resonances can also increase the
110 complexity of the NIR spectra. Therefore, many band assignments can only be tentative
111 or unresolved which limits the usefulness of the near infrared region in structural
112 determination. In comparison, mid infrared spectra exhibit sharp and narrow peaks
113 essentially related to fundamental modes of vibration which can be easily assigned to
114 chemical structures (Bokobza, 1998). Mid infrared spectroscopy has previously been used
115 in order to relate skin cell wall composition to anthocyanin extractability (Femenia,
116 Sánchez, Simal, & Rosselló, 1998; Hernández-Hierro, Quijada-Morín, Martínez-

117 Lapuente, Guadalupe, Ayestarán, Rivas-Gonzalo, et al., 2014). In these studies, mid
118 infrared spectral features could be directly linked to the esterification of pectins, which
119 might be responsible for the different extractabilities of the aforementioned phenolic
120 compounds. Moreover, this spectroscopic technique has been applied to the study and
121 quantitation of phenolic compounds extracted during winemaking (Di Egidio, Sinelli,
122 Giovanelli, Moles, & Casiraghi, 2010; Fragoso, Acena, Guasch, Mestres, & Busto, 2011).
123 However, in these cases, cell wall composition has not been taken into account.

124 Cell wall structure of grape and other fruits has been widely studied by means of
125 vibrational spectroscopy. Vibrational spectroscopy is a subset of spectroscopy which
126 comprises IR and Raman spectroscopy and analyses vibrations within a molecule (or
127 material) (Byrne, Ostrowska, Nawaz, Dorney, Meade, Bonnier, et al., 2014). These two
128 vibrational spectroscopic techniques are complementary. Whereas electric dipole
129 transitions of IR (and UV–visible) absorption require a change of the dipole moment of
130 the material as a result of the transition, Raman scattering requires a change in the
131 polarizability of the bond as a result of the transition. Thus, access to molecular level
132 information by means of two different physical processes can be achieved (Byrne,
133 Sockalingum, & Stone, 2011).

134 Mid infrared spectroscopy has been employed to study cell wall structure of grapes and
135 other fruits (Acebes, Largo-Gosens, Hernández-Altamirano, García-Calvo, Alonso-
136 Simón, & Álvarez, 2014; Fasoli, Dell’Anna, Dal Santo, Balestrini, Sanson, Pezzotti, et
137 al., 2016; Kyomugasho, Christiaens, Shpigelman, Van Loey, & Hendrickx, 2015; Wilson,
138 Smith, Kačuráková, Saunders, Wellner, & Waldron, 2000; Zietsman, Moore, Fangel,
139 Willats, Trygg, & Vivier, 2015). However, studies which employ mid infrared
140 spectroscopy to study the cell wall structure of grapes seeds have not been performed.
141 Raman spectroscopy has also been employed to study cell wall structure in fruits,

142 although, no studies have been found for grapes (Chylińska, Szymańska-Chargot, &
143 Zdunek, 2014; da Silva, Vandenabeele, Edwards, & Cappa de Oliveira, 2008; Gierlinger,
144 Keplinger, & Harrington, 2012; Szymańska-Chargot, Chylińska, Pieczywek, Rösch,
145 Schmitt, Popp, et al., 2016). Therefore, although both mid infrared and Raman
146 spectroscopy have proved to be useful and reliable techniques for the study of the cell
147 wall structure in plant and fruit tissues, they have not yet been applied to the study of
148 grape seeds.

149 In the present study, near infrared hyperspectral imaging has been applied to grape seed
150 samples in order to select a representative subset of samples according to their near
151 infrared spectral features. Afterwards, total phenol and flavanol extractabilities have been
152 obtained for these selected samples and they were sorted into three different groups,
153 identified as low, medium and high extractability levels. Next, Raman and attenuated total
154 reflectance Fourier transform infrared (ATR-FTIR) spectra have been collected for non-
155 extracted material after grape seed extractions. Finally, ATR-FTIR and Raman spectra
156 have been analyzed in order to relate their more important features to phenolic
157 extractability levels in grapes seeds. Moreover, spectral features are also related to the
158 principal sample attributes. To our knowledge, that this is the first time that ATR-FTIR
159 and Raman spectroscopy are jointly applied to grape seed non-extracted material samples
160 in order to relate their more important spectral features to their phenolic extractability
161 levels.

162

163 **2. MATERIAL AND METHODS**

164 *2.1. Samples*

165 *V. vinifera* L. cv. Syrah and Tempranillo red grape samples were collected from two
166 vineyards located in the Condado de Huelva Designation of Origin D.O. (Andalusia,

167 Spain). Grapes were collected on two different dates, when Tempranillo and Syrah
168 vineyards were respectively harvested (August 7 and 11, 2014). A total of two hundred
169 Syrah and Tempranillo grapes were collected, one hundred per variety. In order to achieve
170 representative samples sets, single grapes were collected from the top, middle and bottom
171 of the cluster and from the sunlight and shade side. The samples were refrigerated and
172 immediately transported to the laboratory.

173 *2.2. Sample selection*

174 Grape seeds were removed from the whole grapes and left to dry at room temperature.
175 Then individual near infrared hyperspectral images were recorded for grape seeds
176 belonging to the same grape and the near infrared spectrum was saved (between 950 and
177 1650 nm) as described elsewhere (Rodríguez-Pulido, et al., 2014). Afterwards, sample
178 selection was carried out as described in Nogales-Bueno, et al. (2015). Briefly, an
179 unsupervised pattern recognition technique, principal component analysis (PCA), was
180 used to select representative samples from the spectral data set. Eight principal
181 components were taken into account in order to explain ninety per cent of the spectral
182 variability of original spectral matrix. Mahalanobis distances (H) for each sample were
183 calculated and samples were grouped according to a neighborhood H criterion ($NH \leq$
184 0.6). Thus, 66 groups with different spectral characteristics were created and one sample
185 from every group was selected. These 66 selected samples were used in all the subsequent
186 analyses. Grape seeds were weighed, immediately frozen and stored at -20 °C until
187 analyses were performed.

188 *2.3. Model wine and exhaustive extractions*

189 Two different extractions were carried out: a model wine extraction and an exhaustive
190 extraction.

191 For model wine extraction, grape seeds were immersed in a model wine hydroalcoholic
192 solution (4 g L⁻¹ tartaric acid, 12.5% (v/v) ethanol, adjusted at pH 3.6 with NaOH 0.5 M)
193 for a maceration period of 72 h. Then, supernatants were used into the subsequent
194 analyses.

195 Afterwards, grape seed samples were freeze-dried, macerated in methanol:water 75:25
196 (v/v), sonicated during 15 minutes (JP Selecta, Barcelona, Spain) and centrifuged (830×g,
197 15 minutes). These extractions were repeated twice in order to achieve an exhaustive
198 extraction of phenolic compounds. The methanolic extracts were combined and finally
199 made up to a final volume of 50 mL with methanol.

200 As a by-product of model wine and exhaustive extractions, non-extracted material (NEM)
201 was obtained from each grape seed sample. These NEM samples were freeze-dried and
202 then stored in a desiccator until further use.

203 *2.4. Total phenol and flavanol extractability determinations*

204 Total phenol and flavanol extractabilities were determined for grape seed samples. They
205 were determined for each supernatant coming from the model wine and exhaustive
206 extractions. Then, extractabilities of each sample were evaluated, respectively, as the
207 fractions of total phenols and flavanols extracted by the model wine solution with respect
208 to the exhaustive extraction. Finally, grape seed samples were sorted according to their
209 phenolic extractability levels expressed as percentages.

210 Total phenol contents were determined using the Folin–Ciocalteu method (Singleton &
211 Rossi, 1965). Two hundred microliters of exhaustive or model wine extractions were
212 mixed with 1.5 mL of sodium carbonate (20 % w/v), 500 μL of Folin reagent and made
213 up to 10 mL with ultrapure water.

214 Flavanol contents were determined following a modification of the method of Vivas,
215 Glories, Lagune, Saucier, and Augustin (1994). Ten or twenty microliters of exhaustive

216 or model wine extractions were mixed with 190 or 180 μL of methanol respectively and
217 1 mL of DMACA (4-dimethylaminocinnamaldehyde) reagent. The DMACA reagent was
218 prepared immediately before use, containing 0.1% (w/v) DMACA in a mixture of
219 HCl:methanol (1:10, v/v).

220 Both Folin and DMACA analyses were performed on an Agilent 8453 UV-visible
221 spectrophotometer (Palo Alto, USA), equipped with diode array detection (DAD),
222 measuring absorbance at 765 and 640 nm respectively. The extract volumes were
223 appropriately modified for samples which needed it.

224 *2.5. ATR-FTIR data collection*

225 ATR-FTIR spectra were recorded with the Perkin Elmer (MA, USA) Spotlight 400N
226 Universal Attenuated Total Reflectance (UATR) accessory of the spectrometer, which
227 employs a 9-bounce diamond top-plate for this analysis. Spectral data were the result of
228 32 scans, with a spectral resolution of 1 cm^{-1} and covering the spectral range between 600
229 and 4000 cm^{-1} .

230 All samples were measured with a force gauge of 140 units. NEM from grape seed
231 samples were powders and 6 absorption spectra were collected for each grape. A
232 background spectrum was also recorded and automatically subtracted by the software.

233 *2.6. Raman data collection*

234 A Horiba Jobin-Yvon LabRAM HR800 spectrometer with an external 300 mW diode
235 laser operating at 785 nm as source was used throughout this work. For the measurements,
236 a $\times 100$ objective (MPlanN, Olympus) was employed, providing a spatial resolution of ~ 1
237 μm at the sample. The confocal hole was set at $100\text{ }\mu\text{m}$, the specified setting for confocal
238 operation. The system was spectrally calibrated to the 520.7 cm^{-1} spectral line of silicon
239 and the intensity response function was corrected using the Standard Reference Material
240 (SRM) No. 2243 of the National Institute of Standards, Boulder, Colorado, USA (NIST

241 SRM 2243, 2242, 2241). The LabRAM system is a confocal spectrometer that contains
242 two interchangeable gratings (300 and 900 lines per mm respectively). In the following
243 experiments, the 300 lines per mm grating was used, giving a spectral dispersion of ~ 1.5
244 cm^{-1} per pixel. The detector used was a 16-bit dynamic range Peltier cooled CCD detector.
245 All spectra were recorded over the spectral range between 400 and 3500 cm^{-1} and with a
246 spectral resolution of ~ 0.9 to 1.6 cm^{-1} . Six spectra were collected for each NEM from
247 grape seed samples.

248 2.7. Data analysis

249 K-means cluster analysis of near infrared hyperspectral data was carried out in order to
250 sort grape seed samples according to their phenolic extractability levels (total phenol,
251 flavanol extractabilities). Initial cluster centers were chosen in order to maximize the
252 initial between-cluster distances. Samples were sorted into three groups, low, medium
253 and high phenolic extractability levels. K-means analysis was carried out by means of
254 Statistica v.8.0 software (StatSoft Inc., OK, USA, 2007).

255 Both ATR-FTIR and Raman raw data needed spectral pre-treatments. A multiplicative
256 scatter correction (MSC) was applied to the ATR-FTIR raw data in order to remove the
257 scattering caused mainly by particle size and compaction. Win ISI (v1.50) (Infrasoft
258 International, LLC, Port. Matilda, PA, USA, 2000) software was used for this aim.
259 Baseline correction was applied to Raman raw data. This correction was carried out using
260 MATLAB R2012b (The Mathworks, Natick, MA, USA, 2012) and following the
261 algorithm described elsewhere by Mazet, Carteret, Brie, Idier, and Humbert (2005). This
262 algorithm estimates the background of a spectrum by means of a non-quadratic cost
263 function. Afterwards, this cost function was subtracted from the raw spectra. Asymmetric
264 truncated quadratic was the cost function which gives the best results to estimate

265 background in Raman spectra of NEM from grape seed samples. Moreover, the
266 backgrounds were estimated by a 5-order polynomial and with thresholds of 0.01.
267 Peak heights were calculated for ATR-FTIR and Raman pre-treated spectra and PCA was
268 applied to both whole spectral and peak height matrices. Win ISI (v1.50) software was
269 used for PCA of whole spectral matrices, *whereas PCA for peak heights matrices was*
270 *performed by means of Statistica v.8.0 software (StatSoft Inc., OK, USA, 2007). The*
271 *aforementioned peaks heights for ATR-FTIR and Raman were previously obtained from*
272 *spectra and pre-treated spectra respectively using MATLAB R2012b (The Mathworks,*
273 *Natick, MA, USA, 2012).* Furthermore, univariate analysis of variance (Tukey *post hoc*
274 test) was applied to find differences in the peak heights (dependent variables) among
275 different factors (extractability levels or grape varieties). The statistically significant level
276 was considered at $\alpha = 0.05$. Statistica v.8.0 software was used to develop these analyses.

277 **3. RESULTS AND DISCUSSION**

278 *3.1. Total phenol and flavanol extractability levels*

279 K-means cluster analysis of *near infrared hyperspectral* data sorted grape seed samples in
280 three different groups. The analysis allocated samples into groups according to their total
281 phenols and flavanols extractability levels and, afterwards, these groups were named as
282 low, medium and high extractability levels. Table 1 shows *the* extractability levels of total
283 phenols and flavanols for grape seed samples and the number of samples classified in
284 each cluster by the k-means method.

285 *3.2. ATR-FTIR data*

286 Raw average spectra of NEM from grape seed samples with low, medium and high
287 extractabilities are shown in Figure 1A. MSC pre-treatment was applied to ATR-FTIR
288 spectra and average spectra were calculated for each grape seed NEM sample. Then, the
289 spectra were vector normalized and mean centered. Afterwards, a PCA was applied to

290 this spectral matrix. Overall, the spectral variability explained was 99% using 13 principal
291 components and Mahalanobis distances for each sample were calculated. Samples were
292 ranked in order of their H (Mahalanobis) distance from the mean spectrum of the entire
293 sample set and the $H > 3$ criterion was applied in order to look for spectral outliers. Only
294 one H-outlier was found, and it was not taken into account hereafter. Figure 1B shows
295 the scores of the NEM from grape seed samples in the space defined by the first and
296 second principal components, which described 44.31% (PC1) and 35.72% (PC2) of the
297 variability in the data. In the scores plot, the samples are represented by a color code
298 indicating the different extractability levels. It can be seen that medium extractability
299 samples overlap low and high extractability samples. In fact, if medium extractability
300 samples are removed from this plot, it is possible to find some degree of separation
301 between low and high extractability samples (Figure 1C).

302 Figure 1D shows the MSC pre-treated average spectra of NEM from grape seed samples
303 with low and high extractabilities, which are quite similar. Moreover, Figure 1D shows
304 the loadings of PC1 and PC2. PC1 and PC2 loadings show large variations in several
305 spectral regions which might influence the grape seeds phenolic extractability. For
306 example, in the scores plot, almost all high extractability samples are located in the
307 negative PC2 region (Figure 1C), accordingly, the negative features in the PC2 loading
308 (1250-1700 and 3000-3500 cm^{-1} in Figure 1D) might have significant influence on
309 phenolic compound extractability. The main features presented for grape seed NEM
310 samples are described in Table 2. These features are mainly ascribed to polysaccharides,
311 lignins, lipids, pectins and phenolic compounds (Fasoli, et al., 2016; Foo, 1981; Heredia-
312 Guerrero, Benítez, Domínguez, Bayer, Cingolani, Athanassiou, et al., 2014; Lupoi, Singh,
313 Parthasarathi, Simmons, & Henry, 2015; Szymanska-Chargot & Zdunek, 2013; Wilson,
314 et al., 2000).

315 With the aim of identifying what spectral regions have a major influence on phenolic
316 extractability, a univariate analysis of variance was carried out. Peak heights were
317 measured in MSC pre-treated grape seed NEM spectra. Thirteen peaks were taken into
318 account (Table 2). A univariate analysis of variance, a statistical method used to analyze
319 the differences among group means and their associated procedures, was performed. Peak
320 heights were used as dependent variables, whereas extractability level, shown in Table 1,
321 was used as factor or independent variable. This analysis is described in detail in Table
322 3, which shows the heights (mean and standard error) for all peaks described in Table 2.
323 Moreover, in Table 3, values are marked with letters which indicate the existence or not
324 of statistical differences between extractability levels. No significant differences were
325 found for the peak at 1743 cm^{-1} , ascribed to $(\text{C}=\text{O})_{\text{ester}}$, and for the peaks at 2854 and 2924
326 cm^{-1} , ascribed to CH_2 stretches. Thus, they appear not to have any influence on phenolic
327 compounds extractability. Significant differences ($p < 0.05$) were found for the peaks at
328 781 and 1154 cm^{-1} , but only between medium and high extractability levels. However,
329 no significant differences were found between low and high extractability levels. These
330 two peaks are related to phenolic compounds, and therefore, this fact can show some kind
331 of differentiation between phenolic compounds in medium and high extractability
332 samples, but not between these and low extractability samples. Finally, there are
333 significant differences ($p < 0.05$) in the remaining peaks, 1020 and 3285 cm^{-1} , between
334 the three extractability levels and between low/medium levels and high level for peaks at
335 1235 , 1318 , 1370 , 1441 , 1519 and 1610 cm^{-1} . It is to be noted that almost all these peaks
336 fall within the spectral regions that principal component analysis showed to have
337 significant influence on phenolic compound extractability. Despite the obvious fact that
338 peaks related to the remaining phenolic compounds (1441 and 1519 cm^{-1}) (Foo, 1981;
339 Heredia-Guerrero, et al., 2014) influence phenolic extractability, these results confirm

340 that phenolic extractability is also influenced by the cell wall composition (peaks at 1020,
341 1235, 1318, 1370 and 3285 cm^{-1} ascribed to polysaccharides, lignins and pectins) and by
342 the degree of esterification of pectins (1610 cm^{-1} , non-esterified pectins) (Gao, Fangel,
343 Willats, Vivier, & Moore, 2015; Szymanska-Chargot, et al., 2013; Wilson, et al., 2000).
344 However, this last peak is also ascribed to C=C aromatic stretching in phenolic
345 compounds by other authors (Ricci, Olejar, Parpinello, Kilmartin, & Versari, 2015).
346 In addition, using grape variety as factor, a new analysis of variance was carried out and
347 significant differences were found for peaks at 1020, 1370, 1441 and 3285 cm^{-1} among
348 Tempranillo and Syrah samples. All these peaks also have significant influence on the
349 extractability of phenolic compounds. Therefore, these results show that ATR-FTIR
350 spectral features are linked to grape variety and it can be confirmed that this spectroscopic
351 tool can be used to study the relation between grape variety and phenolic compound
352 extractability of grape seeds.

353 3.3. Raman data

354 In order to confirm or add more information to the above findings, Raman spectra were
355 measured for NEM from grape seeds. A baseline correction procedure was carried out for
356 each grape seed NEM spectrum (Figure 2A) and 5 peaks were selected, those at 1269,
357 1442, 1609, 1655 and 2916 cm^{-1} (Table 4) (Chylińska, et al., 2014; da Silva, et al., 2008;
358 Gierlinger & Schwanninger, 2006; Lupoi, Gjersing, & Davis, 2015; Seidler-Lozykowska,
359 Baranska, Baranski, & Krol, 2010; Szymańska-Chargot, et al., 2016). These peaks were
360 selected because they had a good signal to noise ratio and were present in most samples.
361 Then, PCA was performed, both of whole spectra (data not shown) and peak heights
362 (Lupoi, Singh, et al., 2015). The average peak heights were calculated for each sample
363 and PCA was carried out using these data. Figure 2B shows the scores of the NEM from
364 grape seed samples in the space defined by the first and second principal components,

365 which described 65.39% (PC1) and 23.86% (PC2) of the variability in the data. In the
366 scores plot, the samples are represented by a color code indicating the different
367 extractability levels. No trends were found, as samples are overlapped in this space, and
368 thus, PCA cannot be used to interpret the differences between phenolic extractability
369 levels in this case.

370 Afterwards, one-way univariate analyses of variance were performed. Peak heights were
371 used as dependent variables whereas grape seed variety and extractability levels (Table
372 1) were used as independent variables or factors in the different analyses. Significant
373 difference ($p < 0.05$) among grape varieties was found for the peak at 1655 cm^{-1} . This
374 peak is related to the C=C stretch in **lignins** and fatty acids. Moreover, significant
375 differences were found for the peaks at 1609 , 1655 and 2916 cm^{-1} when extractability
376 levels were taken into account. These bands are respectively due to phenolic compounds,
377 **lignins** and polysaccharides (Gierlinger, et al., 2006; Lupoi, Gjersing, et al., 2015;
378 Seidler-Lozykowska, et al., 2010). The band at 1609 cm^{-1} , assigned to aromatic C=C
379 skeletal stretching, has a strong Raman activity, whereas it only weakly active in infrared
380 spectroscopy. However, the anti-symmetric stretching of COO^- only has infrared activity
381 (Sene, McCann, Wilson, & Grinter, 1994). This allows an assignment of this Raman
382 feature (1609 cm^{-1}) to phenolic compounds and not to pectins. Therefore, Raman
383 spectroscopy enables confirmation of the linkage between cell wall components and
384 phenolic extractability in grape seeds.

385 **4. CONCLUSION**

386 ATR-FTIR and Raman spectroscopy have been proven to be effective and reliable tools
387 to relate the more important spectral features to phenolic extractability levels in grape
388 seed and other sample attributes. ATR-FTIR data analysis confirmed that grape seed
389 phenolic extractability is influenced by the cell wall composition (polysaccharides,

390 lignins and pectins) and by the degree of esterification of pectins, despite the obvious fact
391 that peaks related to the remaining phenolics compounds also influence on phenolic
392 extractability. Moreover, the results obtaining from the infrared spectra have been
393 confirmed by Raman spectroscopy. Peaks linked to cell wall composition have been
394 shown to have significant influence on grape seed extractability.

395 Although FTIR spectroscopy has been applied to grape samples in several studies, to our
396 knowledge, [Raman spectroscopy had not been tested in grape seeds in order to study their](#)
397 [cell wall structure](#). Raman spectroscopy has been shown to be a reliable tool for studying
398 cell wall structure in grape seed tissue. Nonetheless, a comprehensive study should be
399 made in order to improve spectra features, i.e., signal to noise ratio or the background
400 produced by fluorescence.

401 **ABBREVIATIONS USED**

402 ATR-FTIR, attenuated total reflectance Fourier transform infrared; DAD, diode array
403 detector; DMACA, 4-dimethylaminocinnamaldehyde; IR, infrared; H, Mahalanobis
404 distance; MSC, multiplicative scatter correction; NEM, non-extracted material; NH,
405 neighborhood Mahalanobis distance; NIR, near infrared; NIRS, near infrared
406 spectroscopy; PC, principal component; PCA, principal component analysis.

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413 **CONFLICT OF INTEREST**

414 The authors declare no potential conflict of interest.

REFERENCES

- 415
416
417 Acebes, J. L., Largo-Gosens, A., Hernández-Altamirano, M., García-Calvo, L., Alonso-
418 Simón, A., & Álvarez, J. M. (2014). Fourier Transform Mid InfraRed
419 spectroscopy applications for monitoring the structural plasticity of plant cell
420 walls. *Frontiers in Plant Science*, 5.
- 421 Bautista-Ortín, A. B., Jiménez-Pascual, E., Busse-Valverde, N., López-Roca, J. M., Ros-
422 García, J. M., & Gómez-Plaza, E. (2013). Effect of Wine Maceration Enzymes on
423 the Extraction of Grape Seed Proanthocyanidins. *Food and Bioprocess
424 Technology*, 6(8), 2207-2212.
- 425 Bokobza, L. (1998). Near infrared spectroscopy. *Journal of Near Infrared Spectroscopy*,
426 6(1), 3-17.
- 427 Boulton, R. (2001). The Copigmentation of Anthocyanins and Its Role in the Color of
428 Red Wine: A Critical Review. *American Journal of Enology and Viticulture*,
429 52(2), 67-87.
- 430 Byrne, H. J., Ostrowska, M. K., Nawaz, H., Dorney, J., Meade, D. A., Bonnier, F., &
431 Lyng, M. F. (2014). Vibrational Spectroscopy: Disease Diagnostics and Beyond.
432 In M. Baranska (Ed.), *Optical Spectroscopy and Computational Methods in
433 Biology and Medicine*, (pp. 355-399). Dordrecht, Netherlands: Springer
434 Netherlands.
- 435 Byrne, H. J., Sockalingum, G. D., & Stone, N. (2011). Chapter 4 Raman Microscopy:
436 Complement or Competitor? In *Biomedical Applications of Synchrotron Infrared
437 Microspectroscopy: A Practical Approach*, (pp. 105-143). Karlsruhe, Germany:
438 The Royal Society of Chemistry.
- 439 Cadot, Y., Minana-Castello, M. T., & Chevalier, M. (2006). Anatomical, histological, and
440 histochemical changes in grape seeds from *Vitis vinifera* L. cv Cabernet franc

441 during fruit development. *Journal of Agricultural and Food Chemistry*, 54(24),
442 9206-9215.

443 Crozier, A., Clifford, M. N., & Ashihara, H. (2006). *Plant Secondary Metabolites.*
444 *Occurrence, Structure and Role in the Human Diet* Oxford, England: Blackwell
445 Publishing.

446 Chylińska, M., Szymańska-Chargot, M., & Zdunek, A. (2014). Imaging of
447 polysaccharides in the tomato cell wall with Raman microspectroscopy. *Plant*
448 *Methods*, 10(1), 1-9.

449 da Silva, C. E., Vandenabeele, P., Edwards, H. G. M., & Cappa de Oliveira, L. F. (2008).
450 NIR-FT-Raman spectroscopic analytical characterization of the fruits, seeds, and
451 phytotherapeutic oils from rosehips. *Analytical and Bioanalytical Chemistry*,
452 392(7), 1489-1496.

453 Di Egidio, V., Sinelli, N., Giovanelli, G., Moles, A., & Casiraghi, E. (2010). NIR and
454 MIR spectroscopy as rapid methods to monitor red wine fermentation. *European*
455 *Food Research and Technology*, 230(6), 947-955.

456 Fasoli, M., Dell'Anna, R., Dal Santo, S., Balestrini, R., Sanson, A., Pezzotti, M., Monti,
457 F., & Zenoni, S. (2016). Pectins, Hemicelluloses and Celluloses Show Specific
458 Dynamics in the Internal and External Surfaces of Grape Berry Skin During
459 Ripening. *Plant & Cell Physiology*, 57(6), 1332-1349.

460 Femenia, A., Sánchez, E. S., Simal, S., & Rosselló, C. (1998). Effects of Drying
461 Pretreatments on the Cell Wall Composition of Grape Tissues. *Journal of*
462 *Agricultural and Food Chemistry*, 46(1), 271-276.

463 Ferrer-Gallego, R., Hernández-Hierro, J. M., Rivas-Gonzalo, J. C., & Escribano-Bailón,
464 M. T. (2011). Determination of phenolic compounds of grape skins during

465 ripening by NIR spectroscopy. *Lwt-Food Science and Technology*, 44(4), 847-
466 853.

467 Foo, L. Y. (1981). Proanthocyanidins: Gross chemical structures by infrared spectra.
468 *Phytochemistry*, 20(6), 1397-1402.

469 Fragoso, S., Acena, L., Guasch, J., Mestres, M., & Busto, O. (2011). Quantification of
470 Phenolic Compounds during Red Winemaking Using FT-MIR Spectroscopy and
471 PLS-Regression. *Journal of Agricultural and Food Chemistry*, 59(20), 10795-
472 10802.

473 Gao, Y., Fangel, J. U., Willats, W. G. T., Vivier, M. A., & Moore, J. P. (2015). Dissecting
474 the polysaccharide-rich grape cell wall changes during winemaking using
475 combined high-throughput and fractionation methods. *Carbohydrate Polymers*,
476 133, 567-577.

477 Gierlinger, N., Keplinger, T., & Harrington, M. (2012). Imaging of plant cell walls by
478 confocal Raman microscopy. *Nat. Protocols*, 7(9), 1694-1708.

479 Gierlinger, N., & Schwanninger, M. (2006). Chemical Imaging of Poplar Wood Cell
480 Walls by Confocal Raman Microscopy. *Plant Physiology*, 140(4), 1246-1254.

481 Gordillo, B., Rodríguez-Pulido, F. J., González-Miret, M. L., Quijada-Morín, N., Rivas-
482 Gonzalo, J. C., García-Estévez, I., Heredia, F. J., & Escribano-Bailón, M. T.
483 (2015). Application of Differential Colorimetry To Evaluate Anthocyanin–
484 Flavonol–Flavanol Ternary Copigmentation Interactions in Model Solutions.
485 *Journal of Agricultural and Food Chemistry*, 63(35), 7645-7653.

486 Heredia-Guerrero, J. A., Benítez, J. J., Domínguez, E., Bayer, I., Cingolani, R.,
487 Athanassiou, A., & Heredia, A. (2014). Infrared and Raman spectroscopic
488 features of plant cuticles: a review. *Frontiers in Plant Science*, 5.

489 Hernández-Hierro, J. M., Quijada-Morín, N., Martínez-Lapuente, L., Guadalupe, Z.,
490 Ayestarán, B., Rivas-Gonzalo, J. C., & Escribano-Bailón, M. T. (2014).
491 Relationship between skin cell wall composition and anthocyanin extractability
492 of *Vitis vinifera* L. cv. Tempranillo at different grape ripeness degree. *Food*
493 *Chemistry*, *146*(0), 41-47.

494 Jackson, R. S. (2000). Chemical Constituents of Grapes and Wine. In R. S. Jackson (Ed.),
495 *Wine science: principles, practice and perception*, (pp. 232-280). San Diego,
496 California: Academic Press.

497 Kemps, B., Leon, L., Best, S., De Baerdemaeker, J., & De Ketelaere, B. (2010).
498 Assessment of the quality parameters in grapes using VIS/NIR spectroscopy.
499 *Biosystems Engineering*, *105*(4), 507-513.

500 Kyomugasho, C., Christiaens, S., Shpigelman, A., Van Loey, A. M., & Hendrickx, M. E.
501 (2015). FT-IR spectroscopy, a reliable method for routine analysis of the degree
502 of methylesterification of pectin in different fruit- and vegetable-based matrices.
503 *Food Chemistry*, *176*, 82-90.

504 Lupoi, J. S., Gjersing, E., & Davis, M. F. (2015). Evaluating Lignocellulosic Biomass, Its
505 Derivatives, and Downstream Products with Raman Spectroscopy. *Frontiers in*
506 *Bioengineering and Biotechnology*, *3*(50).

507 Lupoi, J. S., Singh, S., Parthasarathi, R., Simmons, B. A., & Henry, R. J. (2015). Recent
508 innovations in analytical methods for the qualitative and quantitative assessment
509 of lignin. *Renewable and Sustainable Energy Reviews*, *49*, 871-906.

510 Mazet, V., Carteret, C., Brie, D., Idier, J., & Humbert, B. (2005). Background removal
511 from spectra by designing and minimising a non-quadratic cost function.
512 *Chemometrics and Intelligent Laboratory Systems*, *76*(2), 121-133.

513 Nogales-Bueno, J., Baca-Bocanegra, B., Jara-Palacios, M. J., Hernández-Hierro, J. M., &
514 Heredia, F. J. (2016). Evaluation of the influence of white grape seed extracts as
515 copigment sources on the anthocyanin extraction from grape skins previously
516 classified by near infrared hyperspectral tools. *Food Chemistry*, In press.

517 Nogales-Bueno, J., Baca-Bocanegra, B., Rodríguez-Pulido, F. J., Heredia, F. J., &
518 Hernández-Hierro, J. M. (2015). Use of near infrared hyperspectral tools for the
519 screening of extractable polyphenols in red grape skins. *Food Chemistry*, *172*,
520 559-564.

521 Nogales-Bueno, J., Hernández-Hierro, J. M., Rodríguez-Pulido, F. J., & Heredia, F. J.
522 (2014). Determination of technological maturity of grapes and total phenolic
523 compounds of grape skins in red and white cultivars during ripening by near
524 infrared hyperspectral image: A preliminary approach. *Food Chemistry*, *152*, 586-
525 591.

526 Ortega-Regules, A., Ros-García, J. M., Bautista-Ortín, A. B., López-Roca, J. M., &
527 Gómez-Plaza, E. (2008). Changes in skin cell wall composition during the
528 maturation of four premium wine grape varieties. *Journal of the Science of Food*
529 *and Agriculture*, *88*(3), 420-428.

530 Ribéreau-Gayon, P., Dubourdieu, D., Doneche, B., Lonvaud, A., Glories, Y., Maujean,
531 A., & Branco, J. M. (2006). *Handbook of Enology, The Microbiology of Wine and*
532 *Vinifications* West Sussex, England: J. Wiley & Sons.

533 Ricci, A., Olejar, K. J., Parpinello, G. P., Kilmartin, P. A., & Versari, A. (2015).
534 [Application of Fourier Transform Infrared \(FTIR\) Spectroscopy in the](#)
535 [Characterization of Tannins. *Applied Spectroscopy Reviews*, *50*\(5\), 407-442.](#)

536 Rice-Evans, C. A., Miller, J., & Paganga, G. (1997). Antioxidant properties of phenolic
537 compounds. *Trends in Plant Science*, *2*(4), 152-159.

538 Rodríguez-Pulido, F. J., Hernández-Hierro, J. M., Nogales-Bueno, J., Gordillo, B.,
539 González-Miret, M. L., & Heredia, F. J. (2014). A novel method for evaluating
540 flavanols in grape seeds by near infrared hyperspectral imaging. *Talanta*, *122*(0),
541 145-150.

542 Seidler-Lozykowska, K., Baranska, M., Baranski, R., & Krol, D. (2010). Raman analysis
543 of caraway (*Carum carvi* L.) single fruits. Evaluation of essential oil content and
544 its composition. *Journal of Agricultural and Food Chemistry*, *58*(9), 5271-5275.

545 Sene, C., McCann, M. C., Wilson, R. H., & Grinter, R. (1994). Fourier-Transform Raman
546 and Fourier-Transform Infrared Spectroscopy (An Investigation of Five Higher
547 Plant Cell Walls and Their Components). *Plant Physiology*, *106*(4), 1623-1631.

548 Singleton, V. L., & Rossi, J. A. (1965). Colorimetry of Total Phenolics with
549 Phosphomolybdic-Phosphotungstic Acid Reagents. *American Journal of Enology
550 and Viticulture*(16), 144-158.

551 Szymańska-Chargot, M., Chylińska, M., Pieczywek, P. M., Rösch, P., Schmitt, M., Popp,
552 J., & Zdunek, A. (2016). Raman imaging of changes in the polysaccharides
553 distribution in the cell wall during apple fruit development and senescence.
554 *Planta*, *243*(4), 935-945.

555 Szymanska-Chargot, M., & Zdunek, A. (2013). Use of FT-IR Spectra and PCA to the
556 Bulk Characterization of Cell Wall Residues of Fruits and Vegetables Along a
557 Fraction Process. *Food Biophys*, *8*(1), 29-42.

558 Torchio, F., Río Segade, S., Giacosa, S., Gerbi, V., & Rolle, L. (2013). Effect of Growing
559 Zone and Vintage on the Prediction of Extractable Flavanols in Winegrape Seeds
560 by a FT-NIR Method. *Journal of Agricultural and Food Chemistry*, *61*(38), 9076-
561 9088.

562 Vivas, N., Glories, Y., Lagune, L., Saucier, C., & Augustin, M. (1994). Estimation du
563 degré de polymérisation des procyanidines du raisin et du vin par la méthode au
564 p-diméthylaminocinnamaldéhyde. *Journal International des Sciences de la Vigne
565 et du Vin*, 28, 319-336.

566 Waterhouse, A. L. (2002). *Wine phenolics* New York, New York: The New York
567 Academy of Sciences.

568 Wilson, R. H., Smith, A. C., Kačuráková, M., Saunders, P. K., Wellner, N., & Waldron,
569 K. W. (2000). The Mechanical Properties and Molecular Dynamics of Plant Cell
570 Wall Polysaccharides Studied by Fourier-Transform Infrared Spectroscopy. *Plant
571 Physiology*, 124(1), 397-406.

572 Zietsman, A. J. J., Moore, J. P., Fangel, J. U., Willats, W. G. T., Trygg, J., & Vivier, M.
573 A. (2015). Following the Compositional Changes of Fresh Grape Skin Cell Walls
574 during the Fermentation Process in the Presence and Absence of Maceration
575 Enzymes. *Journal of Agricultural and Food Chemistry*, 63(10), 2798-2810.

576

577 **FIGURE CAPTIONS**

578 **Figure 1:** (A) Raw average spectra of NEM from grape seed samples with low, medium
579 and high extractabilities. (B) Score plot of the first two principal components after PCA
580 performed on ATR-FTIR spectra recorded from grape seed NEM samples. The individual
581 data points have been color coded according to the results of k-means cluster analysis;
582 low, medium and high phenolic extractability. (C) Score plot without medium
583 extractability samples. (D) PC1 and PC2 loading plot and average MSC pre-treated ATR-
584 FTIR spectra for low and high phenolic extractability grape seed NEM samples.

585 **Figure 2:** (A) Raw Raman spectrum, baseline and corrected spectrum for a grape seed
586 NEM sample. (B) Score plot of the first two principal components after PCA performed
587 on peak heights of Raman spectra recorded from grape seed NEM samples. The
588 individual data points have been color coded according to the results of k-means cluster
589 analysis; low, medium and high phenolic extractability.