Front-face and right-angle fluorescence spectroscopy for monitoring extra virgin olive oil spectrum evolution

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Abstract

Fluorescence spectroscopic techniques were applied in a study in order to monitor the evolution of the spectral pattern of extra virgin olive oil (EVOO) samples exposed to indirect light at room temperature. Detailed information was extracted from three-dimensional front-face (FF) fluorescence spectra with excitation wavelengths ranging from 230 to 646 nm and emission wavelengths ranging from 250 to 698.5 nm. Relevant emission regions were revealed in FF experiments that were useful to study the variability of the characteristic spectral patterns and to develop fast inspection procedures. Such procedures include simultaneous excitation at wavelengths lower than 400 nm, which was proposed and implemented in a rightangle prototype in order to monitor the evolution of EVOO samples exposed to indirect light. Emission signals at local maxima around 400, 434, 464, 513 and 674 nm were found to be relevant. Ongoing research highlights that hyperspectral images provide spectral patterns of the olives, allowing more precise sorting into categories, which would enable the classification into lots of oils with more homogeneous characteristics for subsequent modelling.

Keywords: oil quality, oil storage, oxidation, principal component analysis

INTRODUCTION

Olive oil consumption is spreading all over the world, promoted by the sensorial attributes and health benefits associated with the chemical and nutritional composition of extra virgin olive oil (EVOO) (López-Miranda et al., 2010). Under oxidation-inducing conditions promoted by illumination, a sample of olive oil might contain a plethora of oxidation compounds. Most of the olive oil components involved in the oxidation processes, such as chlorophyll, antioxidant compounds (polyphenols and α -tocopherol) and primary and secondary oxidation products, are fluorescent molecules (Sikorska et al., 2012). Thus, a complex fluorescence spectral pattern is expected to reveal the oxidation status. Fast inspection procedures are of great interest. Hence, identification of the bands of excitation wavelengths that provide information on oxidation evolution is a major issue to develop a fast procedure with simultaneous excitation that gives rise to useful emission signals. The present work shows the preliminary results of a study where detailed regions of excitationemission wavelengths have been extracted from three-dimensional front-face (FF) fluorescence spectra. Simultaneous excitation at the wavelengths of interest has been proposed and implemented in a right-angle (RA) prototype in order to monitor the evolution of EVOO samples exposed to indirect light.

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The hyperspectral imaging system allows integrating spectroscopic and imaging techniques to enable direct identification of different components or quality characteristics and their spatial distribution within the tested samples (Lara et al., 2013; Lleó et al., 2011). Hyperspectral imaging techniques have emerged as a powerful tool in agricultural and food systems, as they overcome the limitations of spectroscopic and vision techniques.

Hyperspectral images could provide detailed spectral patterns useful to characterize olives and to sort them into homogeneous categories before oil extraction.

MATERIALS AND METHODS

Set 1 included two EVOO samples obtained directly from producers (P1 and P2) and five EVOO samples (according to commercial labelling) acquired from retail markets (C1, C2, C3, C6 and C8), and seven olive oil samples from retail markets (OO1-OO7) (Table 1). 3D FF and RA fluorescence spectra were obtained from each sample, and then analysed and compared in order to assess the most suitable configuration of the RA experiments. Set 2 included 13 EVOO samples, according to commercial labelling, obtained from retail markets. Commercial samples (C1-C13) included monovarietal oils, known varietal blends and unknown varietal blends (Table 2). These samples were exposed to indirect light in transparent glass bottles at room temperature for 5 months. RA fluorescence spectra were acquired monthly from each sample from the control month at the beginning of the experiment to the fifth month under treatment. Principal component analysis (PCA) was performed on centred data. Set 3 included olive fruit samples, cultivar 'Picual', sorted visually into group 1 (G1, green and 'pintonas' olives), G2 (purple olives) and G3 (black olives).

Table 1. Characteristics of extra virgin olive oil (EVOO) and olive oil (OO) samples of set 1 analysed by 3D front-face and right-angle fluorescence spectroscopy.

Sample	Cultivar(s)	Origin	K ₂₇₀
EVOO samples P1	Arbequina	Producers	0.11
P2	Picual	Producers	0.12
C1	Arbequina	Retail market	0.12
C2	Hojiblanca	Retail market	0.13
C3	Picual	Retail market	0.14
C6	Blend of Hojiblanca, Arbequina, Picual, Cornicabra	Retail market	0.14
C8	Arbequina	Retail market	0.13
OO samples			
OO1		Retail market	0.28
OO2		Retail market	0.34
OO3		Retail market	0.41
OO4		Retail market	0.41
OO5		Retail market	0.41
OO6		Retail market	0.48
OO7		Retail market	0.50

Table 2. Characteristics of the EVOO samples of set 2 obtained at retail markets analysed by right-angle fluorescence spectroscopy.

Sample	Cultivar(s)	Sample	Cultivar
C1 ¹	Arbequina	C8 ¹	Arbequina
C2 ¹	Hojiblanca	C9	Blend
C3 ¹	Picual	C10	Blend

C4	Blend	C11	Blend
C5	Blend	C12	Blend
C6 ¹	Blend of Hojiblanca, Arbequina, Picual, Cornicabra	C13	Blend
C7	Blend of Hojiblanca, Arbequina, Picual, Cornicabra		

¹Analyzed by 3D FF.

3D FF fluorescence spectra were measured using a spectrofluorometer (LS45; PerkinElmer, Waltham, MA, USA) equipped with a xenon lamp source, excitation and emission monochromators and a front-face sample-cell holder. The excitation wavelength ranged from 230 to 646 nm (step 4 nm) and the emission wavelength ranged from 250 to 698.5 nm (step 0.5 nm). Excitation and emission monochromator slit widths were set at 10 nm. Quartz cuvettes (10×10×45 mm) were filled with the oil samples without prior preparation. For each olive oil sample, a rectangular area was selected that included the region of oxidation products. The excitation band ranged from 274 to 400 nm and the emission band ranged from 375 to 550 nm. The coefficient of variation (CV) of the signal intensity was computed at every pair (ex_i, em_j) among the samples in both the EVOO group and the OO group.

For each excitation wavelength, the excitation spectrum was obtained as the sum of the signal along the selected emission wavelengths. The emission spectrum was obtained as the sum of the signal along the selected excitation wavelengths for each emission wavelength. Univariate and multivariate analyses were developed in order to identify the emission wavelengths with potential information related to K_{270} . The univariate approach consisted of computation of the coefficient of determination R^2 between the signal intensity at each individual wavelength from 275 to 550 nm and the analytical K_{270} values. The multivariate approach computed the b-coefficients of a partial least-square regression (PLSR) for the whole wavelength range. Performance was evaluated through R^2 and the biascorrected standard error of prediction (SEP_c). RA fluorescence spectra were obtained using a photonic multichannel spectrometer (Hamamatsu, Japan) with detection wavelengths ranging from 196.9 to 958.8 nm (step 0.75 nm). A UV-VIS light source (L10290, Hamamatsu, Japan) was used with a deuterium lamp, with spectral range from 200 to 400 nm. Integration time was set to 1400 ms and three repetitions per measurement were averaged.

A prototype with a right-angle sample-cell holder was designed and assembled by LPFTAGRALIA. Two optical filters were coupled to the set-up so as to constrain both the excitation and emission wavelength ranges. The optical filter coupled with the fibre coming from the light source allowed a beam with wavelengths below 400 nm. The other optical filter limited the registered beam to wavelengths higher than 400 nm. Measurements were carried out using quartz cuvettes (10×10×45 mm) directly on the oil samples without prior preparation. The hyperspectral vision system consisted of a push-broom CCD camera (Andor Luca) equipped with a spectrograph Headwall Photonics Hyperspec™ VNIR (400-1000 nm, 189 wavelengths, spectral resolution 3.17 nm). The illumination was provided by two halogens lamps. The platform moved under the camera and the image was scanned line by line according to the movement (push-broom system). The spatial resolution was 260 μ m. Once the raw images were acquired, the corresponding relative reflectance hypercube was computed, containing the relative reflectance spectrum of each pixel of the image with respect to a reference (mean spectrum of a barium sulfate white reference). Spectra were pre-processed by applying standard normal variate and Savitsky Golay techniques.

RESULTS AND DISCUSSION

For each olive oil sample of set 1, a rectangular area was selected that included the region of oxidation products. The excitation band ranged from 274 to 400 nm and the emission band ranged from 375 to 550 nm. Examples are shown in Figure 1. The CV of the signal intensity was computed at every pair (ex_i, em_j) among the samples in the corresponding group. For EVOO samples, the region with the highest CVs appeared within ex_i from 274 to 318 nm and em_j from 380 to 490 nm, i.e., CVs higher than 25% up to 40% (red and dark yellow

in Figure 1). Smaller regions showed CVs around 25% (light yellow in Figure 1) and a large region showed values around 15% (light blue in Figure 1).

Higher CVs were found for the OO samples than for EVOO samples. Most of the points within the selected area showed CVs higher than 30% (Figure 1). Excitation wavelengths around 280, 325 and 345 nm provided wide emission ranges with the highest CVs, around 35%. According to this observation, sources of information on the oxidation conditions probably arise at least from these three excitation spectral regions. Therefore, a more complete evaluation relies on the registration of the signal resulting from wide bands of excitation wavelengths. When fast monitoring is also required, the optimal solution is to implement a simultaneous excitation at those wavelengths. Another approach was considered by simulating a simultaneous excitation by summing the emission signal along the excitation wavelengths. Two spectral analyses, univariate and multivariate, were developed on the emission spectra in order to identify the emission wavelengths with potential information related to K_{270} . OO samples were included so as to increase the range of K_{270} values.

Figure 1. Upper left: example of raw signal from 3D FF spectrum of an EVOO sample. Bottom left: coefficient of variation (CV) (%) for the group of EVOO samples. Upper right: example of raw signal from 3D FF spectrum of an OO sample. Bottom right: CV (%) for the group of OO samples. Note different colour scales.

For the EVOO group, low R^2 values were obtained, below 0.4, over the whole range. Despite that, two relevant wavelengths can be identified as corresponding to spectral peaks, i.e., around 375 and 443.5 nm, the former with the highest R^2 (Figure 2; upper left). As for the OO group, R^2 values higher than 0.7 were obtained at 400 nm and subsequent wavelengths for the whole range. When the whole group was analysed (EVOO and OO), R^2 reached values above 0.9 for all wavelengths, with a maximum of 0.98 around 445 nm (Figure 2; upper left). The high R^2 for all emission wavelengths suggests that an intense oxidation produces a general increase in the signal along the whole spectrum, which is related to the overall increase of the K_{270} value. Signal from secondary oxidation products seems to dominate the whole spectra, probably concealing other signals. The multivariate approach identifies a specific emission band ranging from 375 to 450 nm that provides higher b-coefficients, with a maximum around 410 nm (Figure 2; upper right). R^2 was 0.94 and SEP_c was 0.0364 using one latent variable. In Figure 2 (bottom left), the intensity of the emission signal at 400 nm is depicted versus the measured K_{270} values. In Figure 2 (bottom right), a similar result is obtained when multivariate prediction is illustrated by measured K_{270} values versus predicted K_{270} values for one latent variable. The results obtained are similar to those found in previous studies where

a single excitation wavelength was used for excitation. Kyriakidis and Skarkalis (2000) used 365 nm as the excitation wavelength and found an R^2 of 0.96 between emission at 445 nm and K_{270} . Guzmán et al. (2015) found the best performance by applying PLSR with six latent variables for excitation at 380 nm.

Emission spectra ranged from 300 to 800 nm. The root mean square prediction error was 0.08 and the R^2 was 0.92 obtained with external validation. However, in the present work, a broad band of excitation wavelengths was studied, with the aim of capturing signal arising from a wider variety of compounds. In view of these results, the signal arising from most of the fluorophores related to the oxidation processes could be obtained with excitation at wavelengths below 400 nm. The emission signals would be related to oxidation by-products and chlorophylls. A wide range of excitation would allow capture of different compounds undergoing production and degradation processes. Therefore, subsequent analyses focused on simultaneous excitation of olive oil samples exposed to indirect light in an RA prototype. At control month, spectra were dominated by signals from chlorophyll at around 670 nm and the shoulder at 725 nm (Figure 3). After 2 months, a considerable decrease was observed. Opposite evolution was observed for peaks within the region of oxidation products. Noticeable increases were detected for peaks with local maxima at around 400, 434-437, 464-469 and 510-518 nm (Figure 3). Such peaks were in accordance with those found in FF experiments. PC1 loadings in Figure 3 illustrate the variability. Modelling this evolution with regard to official EVOO samples is the next stage that will be faced.

Figure 2. Upper left: coefficient of determination R^2 between the signal intensity at each individual wavelength from 275 to 550 nm and the analytical K_{270} values. Bottom left: intensity of the sum of emission signal at 400 nm versus measured K_{270} values. Upper right: b-coefficients of a PLSR. Bottom right: measured K_{270} values versus predicted K_{270} values by a PLSR (one latent variable). SEP_c , bias-corrected standard error of prediction.

Spectral patterns were established for each maturity stage according to increasing levels of ripening stage: G1 green olives, G1 'pintonas' olives, G2 purple olives and G3 black olives (Figure 4). Green spectra showed a relevant absorption band at 680 nm, corresponding to high chlorophyll content, and high reflectance at 550 nm, probably related to low anthocyanin content. Average spectra of 'pintonas' olives and purple olives showed similar

patterns. It seems that these two types of spectra belonged to the same ripening level. These spectra also showed an absorption band at 680 nm, with lower intensity in comparison to the green olive spectra. In addition, the intensity of reflectance at 550 nm was lower, probably due to higher anthocyanin content. 'Pintonas' olives therefore showed two types of spectra, two different levels of ripening stage within the surface of the fruit. Hyperspectral images are able to sense these differences, in contrast to classical spectroscopy measurements, reflecting non-homogeneities of ripening within the surface of the fruit. Finally, black spectra showed no absorption band at 680 nm and no reflectance at 550 nm, suggesting total degradation of the chlorophyll and high content of anthocyanins, respectively.

Figure 3. Right-angle fluorescence emission spectra of the commercial EVOO samples. Left: at the control month. Middle: after two months exposed to indirect light. Right: loadings of the first principal component (96.52% explained variance).

Figure 4. Left: example of images of olives samples in false colour, merging three colour planes: blue, 450 nm; green, 550; red, 680 nm. Right: whole set of pre-processed spectra ($n=13,500$); average spectrum of each maturity class is plotted.

CONCLUSIONS

The 3D FF spectra reveal detailed excitation-emission regions. The CV of the signal highlights broad regions related to oxidation processes of olive oil. The emission band from 375 to 550 nm includes signals produced by an excitation band from 274 to 400 nm that shows correlation to the characteristic K_{240} , which is identified with advanced oxidation status. Univariate and multivariate approaches provide correlation with this single parameter. Simultaneous excitation with a wide range of wavelengths below 400 nm in RA experiments is a fast procedure to obtain signals arising from most of the fluorophores related to the complex oxidation processes. The pattern of the spectra acquired using an RA device is in accordance with the pattern of the spectra obtained by integrating the emission signal in FF spectra, showing similar positions of the emission peaks. The RA emission spectral pattern of olive oils exposed to illumination evolves to lower chlorophyll signal and higher signal of oxidation by-products. Ongoing research highlights that hyperspectral images provide spectral patterns of the olives, allowing more precise sorting into categories, which would

enable the classification into lots of oils with more homogeneous characteristics for subsequent modelling.

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